How AI, LLMs and quantum science can empower each other?

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Large Language Models (LLMs) based chatbots such as OpenAI's ChatGPT 4.0 and Google's Bard are emblematic of the broader generative artificial intelligence advances. They have garnered substantial attention in academic, economical and policy making discourses. This paper aims to examine the intersection of LLMs chatbots with quantum science and technology, focusing on their potential to empower research methodologies and pedagogical approaches within these disciplines. It explores with many examples the capabilities of LLM-based tools by assessing their existing and potential future utility in various academic functions and in the industry. These range from facilitating basic question-and-answer interactions to more complex activities such as software development, writing scientific paper, scientific papers reviewing, experiments preparation, research ideation and fostering collaborative research practices in quantum science. The rapid evolution of LLMs and other related tools implementing various forms of so-far primitive reasoning suggests they have the potential to significantly alter the research and educational landscapes, similarly to the transformative impact of the Internet and its associated tools. Accordingly, this paper suggests the creation of a quantum science domain specific LLM based chatbot using open source models and scientific corpus. It also contextualizes LLM-based chatbots within the broader spectrum of machine learning technologies which are already used in the advancement of quantum science and technology. It then quickly explores how quantum computing might or might not further advance machine learning applications and language-based models. The conclusion is that AI may have a profound impact in shaping the trajectory of quantum science research, education, and technology developments while the other way around is quite uncertain, at least in the short to mid-term.

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I. INTRODUCTION

This paper looks at the intersection of machine learning and quantum science, with a focus on large language models chatbots. It investigates the various ways quantum scientists can benefit from these new tools and urges them to learn how to make the best use of them. It describes the existing as well as future use cases for LLMs with the observation that they have a significant potential to alter research methodologies and pedagogical approaches in quantum science with academics, researchers and within the industry. It also explores the various other ways machine learning techniques and quantum technologies can help each other.

Part II describes how LLMs fit into the generative AI landscape, the key features of LLM-based chatbots, their size and their training data sources, their figures of merit, and how to make best use of them with prompt engineering. It assesses their reasoning capabilities and shows how it will grow with the addition of various symbolic AI tools, on the road to so called "artificial general intelligence". This part also describes how to create domain specific LLM chatbots using LLM fine tuning and retrieval-augmented generation (RAG) with documents databases. It suggests the launch of a quantum science domain LLM based chatbot using open source language models.

Part III showcases how LLM chatbots can be used in the context of quantum science by researchers, teachers, students, and industry professionals. It covers the process of learning and studying quantum science and technologies, reviewing, and writing scientific and other papers, doing classical and quantum software development, conducting research ideation and collaborative research.

Part IV inventories the use cases of other types of machine learning-based tools in quantum science with research and technology developments in quantum physics and hardware, quantum software and tools and other domains like quantum sensing and quantum communications and cryptography.

Part V covers the potential use cases of quantum computing in machine learning applications including large language models. It also shows how quantum inspired algorithms can help reduce the computing and energetic footprint of classical machine learning solutions.

Part VII supplemental materials contain a wealth of prompts and results covering various fields of quantum science, including learning, summarizing the state of the art, listing figures of merits for technologies like single photon sources and detectors, circulators, traveling wave parametric amplifiers, analyzing a paper, producing a position paper, and creating a quiz.

Part IX contains an extensive bibliography for readers willing to dig into the way LLMs and their tools operate.

The intended audience of this paper is students, researchers and professionals in the broad quantum science and technology ecosystem who have a limited knowledge and understanding of machine learning and particularly, large language models chatbots. A preliminary understanding of machine learning and neural network basics is however a plus here¹.

II. LLM-BASED CHATBOTS PRIMER

Generative AI refers to a subset of artificial intelligence algorithms and technologies that create new content of various formats (texts, images, 3D models or even humans, videos, music, software code, molecules, etc.), based on the patterns and information learned from large datasets, and with existing content of the same type (e.g., for responding to a text question by a text answer) or of different types (e.g., creating some image or video based on a text specification). This differs from other machine learning models that are focused on labelling, clustering, or interpreting data and making various predictions not in relation with media content creation (e.g., forecasting sales and inventory, or planning a robot movement)².

Generative AI tools make use of deep learning neural networks models that are trained on vast amounts of data, allowing them to implicitly understand patterns and structures within the training data. Once trained, these models generate new content extrapolated from existing content. Three wellknown generative AI tools are large language models based on transformers like OpenAI GPT (Generative Pre-trained Transformer), which generates human-like text in chatbots, GANs (Generative Adversarial Networks) and VAE (Variational Autoencoders), which create realistic images, videos, and music under various constraints (using a given style, changing its format, filling voids, adding bokeh to a picture background, etc.), with applications in verticals like healthcare³ and artistic creation. These models can produce content not explicitly programmed in the model but by understanding the data they were trained on⁴.

Generative AI applications include the transformation of existing content (improving image resolution, converting them from monochrome to color, making people younger or older, creating new music) or the generation of content using content of different types (creating a text description of an image, creating an image based on its description, creating videos and sound of a person based on a text script, turning a sketch into a picture or even a 3D model, etc.). We will differentiate here the generative AI algorithms and models from the products which are using it.

LLM-based chatbot features

As of January 2024, the most popular generalist large language models-based chatbots were ChatGPT 4.0 (\$20/month subscription^{5 6 7}), Google Bard (free⁸) and Llama2 from Meta (free). Other such tools include Claude 2 from Anthropic but its regular access is limited to US and UK users. These tools use language models based on the transformer neural network model, which was invented in 2017 by a Google AI team⁹. To clarify the LLM terminology, ChatGPT 4.0 is a chatbot product that is using a language model, GPT-4, and other software tools and generative AI models like DALL-E which generates images based on their text description. Google Bard is based on the PaLM2 language model, and soon its Gemini successor.



Generative AI algorithms data type in/out examples

Figure 1: examples of generative AI services creating content based on content of the same time (text-to-text) or between different content type (text-to-image). Transversaly, Synthetic Data Generation (SDG) deals with the creation of real-like data of various types depending on the need (faces, 3D objects, 3D humans¹⁰, structured data, etc.)¹¹. (cc) Olivier Ezratty, 2024, with generative AI services inventory coming mainly from Gozalo-Brizuela et al¹².

These chatbots answer questions by predicting the next word(s) by using the words that come before it. A simplistic example is "*The sky is...*" *...* "*blue*". Transformers are built using supervised training fed by loads of text data scrapped from the Internet and other open source databases. This corresponds to the self-supervising pretraining phase. It is followed by some supervised finetuning, reward modeling and reinforcement learning. LLMs are probabilistic models predicting how the segment continues (with autoregressive training) or what is missing in the segment (with masked training)¹³. Transformers are however not limited to handling text. They can also be used in artificial vision, and particularly in medical imaging ¹⁴. Created in 2020, Vision Transformers (ViT) are the equivalents of language models which can, for example, better segment images content over time ¹⁵.

Language models operate on multiple levels of abstraction, processing text from the granular level of tokens, which may be whole words or sub-word units, to larger text structures like sentences, paragraphs, or even multiple paragraphs, aka text chunks. Tokens and chunks are represented as vectors of floating-point values with hundreds to thousands of semantical dimensions which capture the semantic and syntactic nuances of the language¹⁶. Text conversion into this vector space is achieved through the process of word embeddings, a training process where the model learns to associate specific patterns and contexts with corresponding vector representations¹⁷. LLMs are trained with huge sets of text data, mostly coming from online sources but also with specific databases. Their neural networks contain a very large number of parameters, that now exceeds a trillion for the largest available models.

Basic language models text prediction capabilities are extended by integrating them with knowledge bases, reinforcement learning and agents.

This integration allows the model to access and leverage various forms of structured information and planning directives, enriching its understanding, and improving the quality and accuracy of its responses.

By querying these tools, LLMs provide more informed, contextually relevant, and precise answers, drawing from a wider range of information beyond the text present in their training data. LLM chatbots also display apparent reasoning and argumentative features, which are extracted from their training data and are not dynamically created and are thus neither generically deductive (using existing rules) nor inductive (making generalizations)¹⁸ ¹⁹ ²⁰ ²¹.

LLMs are becoming multimodal, being able to process not only text but also other content types like images as input or output and capture the relationship between these content and texts. For example, these can describe image content or generate images based on their text description, like with the DALL-E image generative AI that is incorporated in the ChatGPT 4.0 chatbot²². Apple is working on Ferret, a MLLM (Multimodal LLM) with a better ability to describe the content of images and targeting laptop and smartphone edge devices²³. Generalist LLM chatbots are usually multipurpose tools both in their domain scope which covers any topic, including all fields of science, and in their capabilities to answer questions, translate texts, write summaries, write programming code, and create structured content like tables, and even graphics and images that are based on prompted user text descriptions.



Figure 2: functional description of the ChatGPT 4.0 chatbot application and its key high-level software components. (cc) Olivier Ezratty, 2024.

The scope of these chatbots is expanding fast, and their capabilities are progressing at an exponential speed, at least measured by their model size^{24 25}. They can first play the role of very powerful search engines to provide users with synthetic as well as rather sophisticated definitions of many scientific concepts and technical terms.

LLMs have a strong tendency to respond with long bullet lists, but users can direct them to write narrative explanations avoiding these boring lists using so-called prompt engineering techniques that we will describe later. Then, these chatbot capabilities can be leveraged beyond simple dialog and restructure ideas, format text, translate them into different languages and the likes, with numerous use cases in science.

End-users must, however, become knowledgeable about their various limitations like the inability to reason and their hallucinations, and the way to detect it, if not avoid it ²⁶ ²⁷ ²⁸. The notion of hallucinations is quite broad, encompassing incorrect responses, the generation of non-existent references, or even "hasty generalizations"²⁹.

LLM based chatbots are not monolithic systems only using a LLM model pretrained with loads of texts. They can tap into a variety of features and information sources that are integrated with the model and/or various chatbot agents. These chatbots will progressively be augmented with more of these extensions over time, particularly to improve their capabilities in specific domains, their perceived or real reasoning capabilities, and even to document their sources of information³⁰.

We describe here these key elements and their terminology given it is a very dynamic field showcasing fast improvements in methods and techniques³¹.

Chain of Thought (CoT) is a technique used in LLM chatbots that controls sequences of steps or processes to enable it to generate a response to a user prompt. It is implemented with either fine tuning of the LLM or with prompting³². It involves querying the language model to turn the prompt into tokens, then understanding the context with its key elements (subject,

intent, details, constraints) and using an attention mechanism that helps the LLM to focus on the most relevant context elements, retrieving the relevant response data that was used when training the model, applying learned language patterns, and generating the response by predicting the sequence of words or sentences that best fit the context of the query.

At last, the response is turned into a human-readable format and presented as the chatbot output. Interestingly, context parsing, and response generation are implemented by using the same multifunctional LLM neural network.

Chain of Thought is one of the many places in an LLM that can implement some forms of limited pre-learned static reasoning and research is going on to improve its capabilities³³ ³⁴. To some extent, they can overcome the current shortcomings in reasoning of the transformers used in the LLM, providing the relevant reasoning patterns are already present in the training data, that is dominantly made of unstructured text³⁵. It also enables some form of explainability³⁶. A Tree of Thoughts (ToT) is a generalization of CoT that frames a problem as a search over a tree, where each tree node is a partial solution to the problem³⁷.

Fine-Tuning improves the initial LLM training by using a specific dataset to enhance its performance in specific domains or tasks. The technique can be used when creating a domain specific LLM chatbot to specialize it in a scientific domain. This training process adjusts the weights of the LLM neural network. There are many techniques used for fine-tuning including supervised fine-tuning, instruction tuning, reinforcement learning-based fine tuning and unsupervised fine-tuning with various combinations of these, sometimes in association with chains of thoughts, to improve LLMs reasoning capabilities³⁸. One popular fine-tuning technique is call LoRA, for low rank adaptation of large language models³⁹.

Instruction Tuning improves the language model's ability to follow pre-defined instructions and produce more relevant and accurate responses, like on the way to craft a memo, or create a

series of tasks⁴⁰. To that effect, a language model is exposed to a curated dataset including a variety of instructions and ideal responses to these instructions. It learns from fine-tuning with instruction-based tasks tuning and from feedback ⁴¹. There, users can also provide instructions for achieving "human alignment", for respecting various ethical rules, ad-hoc policies, or regulations. It has however many identified shortcomings⁴².

Reinforcement Learning from Human Feedback (RLHF) is a specific technique used to fine-tune LLMs. It provides specific guidance to the LLM by using a reward model exploiting existing conversations and other manual human feedback ⁴³ ⁴⁴. It influences the LLM's behavior by rewarding outputs that are aligned with human preferences. For example, it can discourage responses that are misleading, offensive, or irrelevant. It serves to embed various ethical rules for human alignment and policy compliance⁴⁵. An alternative to RLHF is Direct Preference Optimization (DPO) which circumvents the need for a reward-based training mechanism ^{46 47}.

Retrieval-Augmented Generation (RAG) can enhance an LLM chatbot by accessing additional information to improve its responses relevance and mitigate hallucinations⁴⁸. A set of documents is usually provided by the system designer and preprocessed in batch mode. Documents are sliced into chunks of text like sentences or paragraphs, which are then encoded in vectors in the embedding process. These vectors are used to find the most relevant documents related to a given question, for example with computing vector distances, and extract inside the

document some relevant information called "snippets"⁴⁹. The snippets are combined by the LLM to create an informed and well written answer⁵⁰. This is controlled but by a custom website backend accessing some LLM APIs (application programming interface), and other third-party cloud services like the LangChain software library and Pinecone or Qdrant for vector embeddings data storage. Figure 3 describes the process of document ingestion in the vector database and how user requests are processed during the document retrieval process, in connection with an LLM that formats the response.

In more sophisticated custom settings, an advanced query parser could redirect part of the question directly to the LLM to handle general knowledge or reasoning tasks while the rest of the questions would be directed to the RAG retriever, like for fact-based and specific data queries. Such selective routing could be used to enhance the efficiency and accuracy of the whole system. In all cases, a RAG enables a chatbot to base its responses on more up-to-date and domain-specific information sources and to document its information sources, making its responses explainable ⁵¹. A RAG can be updated on an ad-hoc basis, like every day or week while a LLM is often "frozen" for long periods of time of about a year or so. One question still lingers, about which data should train an LLM, and which should train only a RAG attached to the LLM and whether some data would train both LLMs and RAGs. It probably relates to a mix of technical⁵², intellectual property licensing, update cycles and cost considerations.



basic Retrieval-Augmented Generation process

Figure 3: schema explaining how RAG documents are ingested in a vector database and a user question is handled to retrieve information from this database and then processed, including the role of the LLM to build a formatted text response. (cc) Olivier Ezratty, 2024.

ChatGPT 4.0 happens to query the Bing search engine to get additional up-to-date data not available within the LLM. The response is assembled by the LLM to refine its answer in a simpler manner than with a RAG.

Creating a quantum science specific LLM chatbot would probably require building some large RAG based on a wealth of sources, including a database of preprint and per-review scientific papers. RAG document embeddings can be generated by the OpenAI API or other services like Hugging Face APIs.

Content filtering can be put in place at various levels, with removing forbidden content in the training data, with fine tuning and RLHF, with the query manager and agents, and at last, at the application level.

Knowledge Graphs can add other reasoning features into LLM-based chatbots, but it is not yet implemented in current LLM based chatbots like ChatGPT 4.0⁵³. As with the expert systems of the late 1980s and early 1990s, these graphs are costly to construct manually. LLMs could help detect such relationships in existing texts and create graphs in a semi-supervised way. Knowledge graphs can be used to generate enriched prompts that are then processed by the LLM^{54 55 56}.

Multimodal Learning enables LLMs to process and understand more than just text, including images, audio, and videos. It creates bridges between these data types. DALL-E from OpenAI, accessible within ChatGPT 4.0, is such a feature, translating a textual description into a dynamically generated image. The other way around, some LLMs can interpret images and provide textual description of their content. Multimodal LLMs can even interpret instructions for robots and images and generate task plans⁵⁷. We can expect some advances here with regards to interpreting charts in scientific publications.

LLM agents. LLM chatbots are usually driven by a set of agents that use the LLM as a tool and bring some forms of reasoning when driving LLMs through prompt engineering⁵⁸. Many such agents are developed and tested like Reason-Act (ReAct)⁵⁹, DEPS ("Describe, Explain, Plan and Select")⁶⁰ and ConstraintChecker for constraint-based reasoning⁶¹. On top of RLHF, they could avoid the phenomenon of LLM hallucinations and enable various forms of ad-hoc planning⁶². The mysterious Q* project from OpenAI may be about incorporating a mix of agents and other structured reinforcement learning, reasoning, and planning tools, into a future version of ChatGPT⁶³.

Data sources	Dynamic or static	Relative training set volume	Function in LLM chatbot
Data used to train the LLM	Static	Extremely high	Core language and knowledge understanding
For Fine-Tuning	Static	Moderate and focused	Specialization and performance improvement in specific areas
For Reinforcement Learning with Human Feedback (RLHF)	Dynamic	Lower than training data	Fine-tuning for accuracy and appropriateness
Fueling Chain of Thoughts	Static (underlying knowledge) and dynamic (processing)	Dependent on task complexity	Complex reasoning and problem-solving
For Agents	Dynamic	Variable	Real-time, interactive, and personalized responses
Retrieval-Augmented Generation (RAG)	Mostly static, can be dynamic at high resource costs.	Variable and dependent on external sources	Enhances LLM responses with information retrieved from external documents. Usually used in domain specific LLM based chatbots
Third party services APIs	Dynamic	Usually, access relation databases	Add services to access real-time data like weather, transportation and hotel bookings, ecommerce transactions and the likes.

Figure 4: table with the key characteristics of data sources fueling a chatbot LLM. Source: ChatGPT4.0 with additions from the author.

Multi-Agents Conversation can enhance LLM-based applications in solving complex tasks by using multi-agent coordinated conversation systems. These agents can be a combination of LLMs, human inputs, and other various specific tools. This is a technique proposed by Microsoft with its Autogen project.

Mixture of experts is an approach using a set of specialized models, aka the "experts", within the overall model, that can process specific types of tasks or data⁶⁴. This technique is used by the startup Mistral AI for its LLM that is used by developers

to create custom LLM-based user environments. GPT-4 may also be using a 'mixture of experts' architecture. This approach may be interesting to investigate in quantum science, for example to mix different complementary expertise in quantum physics, technology developments, algorithms design, cloud, and software architecture design.

APIs access enables an LLM-based chatbot to fetch specific structured answers to questions coming from online services provided with specific APIs.

Some simple examples are knowing the current time and date, weather, or traffic status. ChatGPT 4.0 can also build specific Python code to solve mathematical problems, run the code and send the formatted code output to the user. It can be extended with plugins adding specific features like analyzing an uploaded PDF file with AskYourPDF, exploring scientific publications with ScholarAI, running sophisticated mathematics computing and data charting with Wolfram, access structured data from various services, for example, to prepare some travel, book a flight or a hotel. The other way around, a custom domainspecific chatbot website can access an LLM at the API level to support a dialog with its users. The feature is integrated in a more generalist website with other regular features not using machine learning tools.

LLM size and training data

How large are LLMs is a key consideration, particularly when listening to some pundits who are envisioning using quantum computers in this field. LLM sizing is large at three levels: the size of their training data, of their model parameters and with the computer resources used for training and inferences.

Generalist LLMs training data sources are quite diverse and encompass many online or potentially proprietary sources. In science, it includes peer reviewed journal papers and various academic databases. The exact nature and size of most LLMs training data is a trade secret although various and inconsistent leaks can provide some insights. It looks like it is gigantic, explaining the huge training cost of these models ⁶⁵. According to some leaks, GPT-4 was trained with between 2.6 to 13T tokens, has about 1.5 to 1.8 trillion parameters spread on 120 neural layers and its training cost exceeded \$100M. The model was trained over a couple months with about 30K A100 Nvidia GPUs and inferences are said to run on a 128 GPU cluster^{66 67}. Google Bard's current PaLM-2 language model number of parameters seems to be 330B, with 540B for PaLM-3 corresponding to the upcoming Gemini version announced in December 2023⁶⁸. The chart in Figure 5 shows Google Bard's relative size of training data sets per domain, with a rather small "science" one.

OpenAI relies on Microsoft Azure cloud infrastructure for its models training and inferences. In 2023 alone, Microsoft installed 150,000 such GPUs. According to Mark Zuckerberg, Meta will own 600,000 Nvidia GPGPUs by the end of 2024 to fuel their AI and AGI efforts⁶⁹. All these data centers deployments represent a large share of Nvidia's GPGPU shipments. The vendor expected to ship 550,000 H100 GPGPU in 2023⁷⁰. Google runs Nvidia GPU as well as its proprietary TPUs to train its large language models. Other options exist like using custom FPGAs or very large ASIC chips like the CS-2 from Cerebras ⁷¹ ⁷² or chips designed for inferences⁷³.

OpenAI and others seem to be very relaxed with regards to intellectual property respectfulness. They do not always disclose the sources of their training data, particularly related to the chains of thought, reinforcement learning, fine tuning⁷⁴ and other agents that are used in fueling these LLMs. And some legal issues are showing up here and there, like with the New York Times⁷⁵ and, also, with image generation tools ⁷⁶ ⁷⁷. One can indeed argue that by reusing proprietary content without sharing the generated revenue, LLM chatbots are unduly stealing their creators.

When asking for precisions, ChatGPT 4.0 answers that it is using arXiv preprints in its training, with a cut-off date of April 2023 (as of December 2023). It may thus lack some new cutting-edge research in this fast moving field. But given its answers to complicated questions, it is not obvious it has ingested all arXiv papers related to quantum physics.



Figure 5: Google Bard relative size of training data sets per domain, with a rather small "science" one⁷⁸. Thus the existence of many projects to create science-based LLMs⁷⁹.

ChatGPT 4.0 can tap real time into the wealth of the Internet using the Bing search engine. In that case, the response is correct, well-formed grammatically since it seems that it uses the GPT-4 LLM to format these responses. However, ChatGPT 4.0 and other LLMs are not yet able to interpret the complicated charts found in scientific papers. It can only read their legend, reminding scientists of the importance of writing good legends in their papers.

Like in all fields related to machine learning, LLM can generate various biases originating from their training data which conveys various human biases. It can be gender, stereotypes based, political biases. These can be mitigated with various techniques including careful selection of training data, models fine tuning and a chain of thought mechanism, reinforcement training, or with the use of agents in between the user and LLM prompting.

Lesser discussed scientific biases also exist and there are some of them in quantum science related to the various interpretations of quantum mechanics, optimism or pessimism related to the creation of scalable and useful quantum computers, algorithmic advances and real speedups and the likes.

Figures of merit

LLM chatbots key figures of merit are their number of parameters and the size of their training data sets aka, the corpus size. The number of parameters corresponds to the number of weights and biases in the trained neural network.

Embedding dimensions is another important figure of merit. Words, phrases, sentences, and documents are represented as multi-dimension vectors of real numbers as part of the embedding process. It contains description features that are mathematically processed when training the LLM to predict the next words in the generated text flow. It is used to compute the geometrical distance between words, sentences, and documents and is particularly helpful in automatic translations tasks⁸⁰, which finds similarities between words having similar meanings in different languages.

One famous embedding tool is Word2Vec, from Google, which converts words into vectors, capturing their meaning in classes and even adds contextual information, helping the LLM to process the different meanings of the word depending on their context. GPT-3 uses embeddings with 12,288 dimensions in its largest version having 175 billion parameters. ChatGPT 4.0 uses an undisclosed number of dimensions which is probably bigger.

Context window or context length corresponds to the maximum number of tokens that the LLM can remember when processing dialog session prompts (user requests) and generating text (answers). The larger the context window, the better LLMs consider the long-range dependencies in the dialog. It is like remembering a question asked long ago and using that context when drawing an image or responding to a multi-step user query. GPT-4 used in the chatbot ChatGPT 4.0 has a context length of 32K by default while GPT4 Turbo, the model released in November 2023, has up to 128K tokens. Of

course, the larger the context, the greater the resources consumed by the LLM will be. Also, it may drive the LLM to be imprecise or vague in its responses. The token number limitation, however, doesn't explain alone why ChatGPT 4.0 and Google Bard's answers will generally not exceed the equivalent of 2 or 3 pages. It is also due to a trade-off involving computational efficiency and LLM platform constraints.

Users cannot directly use LLMs chatbots to produce long reports, a full paper, or a book. It is still possible to circumvent this limitation by dividing the incurred work into smaller parts and asking many follow-up questions to an initial question leading to the creation of a document table of content⁸¹.

LLM chatbot evaluation in each domain or a domain specific LLM can use different metrics and benchmarks. "Perplexity" is a low-level metric that evaluates the capacity to fill out text using sub-texts from a test text database. The formula below is adding logs of probabilities of finding the right word in text sequences, comparing the probability assigned to these words by the model and the words in the test strings⁸². Other more elaborated benchmarks can use questions-answers datasets, like the MMLU benchmark⁸³.

$$\log(ext{Perplexity}) = -rac{1}{N}\sum_{i=1}^N \log(ext{Pr}(ext{token}_i \mid ext{context for token}_i))$$

There are also economic figures of merit like the training cost in time, hardware resources, energetic cost⁸⁴, and money, usually measured per number of tokens/words in training data set. Then, users can also assess the inference costs as well. Training data costs can quickly become astronomical for very large LLMs, in the >\$100M range but as they grow, these are closely guarded secrets. And they will need some access to highly powerful Nvidia datacenters or an equivalent HPC infrastructure⁸⁵.

Prompt engineering

The quality of LLM's responses depends on the way questions are asked to them. This belongs to the field of "prompt engineering" techniques. By providing an accurate context, which has a strong meaning in quantum physics, better results are generated by LLMs. It requires specificity in providing multiple criteria, working step by step, and using iterations. Prompt engineering is carried out either by the chatbot user, or by software agents that act as intermediaries between users and the LLM⁸⁶. Prompt engineering is also a way for end-users to adapt their behavior and circumvent some of the LLM limitations. The machine also implicitly drives the human users!

In quantum science, where concepts are deeply nuanced and often counterintuitive, effective prompt engineering becomes even more vital. There are many prompt engineering techniques and subtleties. For example, when trying to learn about quantum error correction, users can begin with a basic request for an explanation, followed by more targeted questions about specific error correction codes, their advantages, and limitations. Also, users can ask ChatGPT 4.0 to refine their own prompt, as shown in the middle of the Supplemental Materials starting page 33.

Users can also ask for summaries of recent research advances, comparisons of different methodologies, or implications of new findings, which in the case of ChatGPT 4.0 will rely on querying Bing. Users can also ask LLMs chatbots to provide explanations using analogies, which can help convey a message to broader audiences.

The examples provided in the Supplemental Materials at the end of this paper, starting page 25, showcase how prompt engineering can operate in quantum science.

Users can define their profile and needs like on the way responses should be formatted, as shown in Figure 6. Many guides are available to help users optimize the way they interact with these LLMs ⁸⁷ ⁸⁸ ⁸⁹ ⁹⁰ ⁹¹. On top of prompt engineering, users can also provide custom instructions to ChatGPT 4.0 so that it considers their profile and needs when responding to their questions. Users can for example explain that they are a student of such and such level or an experimented physicist with a PhD in each domain before engaging in a discussion with the LLM.



Figure 6: examples on how to instruct ChatGPT 4.0 to refine its answers based on your user profile and needs. But default, it has a tendancy to respond with many bullet lists. You can avoid that with asking it to create more "narrative style" answers.

There is even a prompt marketplace where prompts can be acquired for about \$3 to \$7, mostly for images creation using the Midjourney image generation online service⁹².

Using prompt engineering can lead to very funny results when asking an LLM chatbot to respond to a question using the style of some known person, author, scientist, or politician. For example, asking ChatGPT to explain Shor's algorithm a la Shakespeare, Donald Trump, Emmanuel Macron or Dr Seuss way may provide involuntarily hilarious explanations of superposition imitating the style of the former President of the USA: "Now, let me tell you, nobody knows algorithms better than me. And this Shor, he's a smart guy, really smart. He figured out how to use these tiny, tiny things called qubits. *Qubits, unlike regular bits, they can be in multiple states at once – it's like, they can be in New York and Florida at the same time. Can you believe it?*". Humor by analogy seems to be an easy task for LLMs as shown in many other examples (see Supplemental Materials, page 46).

Domain specific LLM chatbots

Online services creators and developers may also want to create a domain specific LLM and chatbot using their own resources. We could indeed wonder how to create such domain specific tool for quantum science.

Various options exist that consist in creating a custom website and self-host your LLM or adding content and features to existing LLM chatbots.

Create a RAG consists in using a documents knowledge base that is ingested and used with "retrieval-augmented generation" engines that we already described 93. The knowledge base is first cleaned up and converted into text from their original formats. Structured document formats like LaTEX and Word are preferred to PDF files, which require specific data extraction tools and libraries. The data is then split into chunks of text and into semantic vectors according to a model and stored in a dedicated vector database⁹⁴. The database is then queried based on user queries. It connects the user's questions to answers in the database, which is implemented with vector proximity detection between the query vectors and the database vectors. RAG can source the origins of responses. A RAG is usually implemented in a custom website using some LLM APIs. It can also be implemented in a ChatGPT 4.0 plugin. Creating such a RAG could use an arXiv dump on quantum science which consist in over 142,000 papers⁹⁵. It would be embedded in a custom website directly accessing an LLM model APIs, or within LLM chatbots⁹⁶.

Fine-tune a LLM can be used to specialize it, which can take advantage of the language model database to help it learn all the vocabulary of the domain associated with its different subdomains, for example in quantum physics, related linear algebra mathematics, quantum communication, quantum computing and quantum sensing. This fine tuning can use part of the domain-specific knowledge base for content tuning, existing conversations and human responses tagging for reinforcement training with human feedback and custom instructions for instruction fine tuning.

All this is also done with creating a custom website accessing a third-party LLM model and APIs like GPT 4.0 or an opensource LLM model like Meta Llama via their API⁹⁷. Finetuning can also involve putting in place some reinforcement learning with human feedback using existing conversations and some human made responses tagging. This is a variation of a LLM model fine-tuning.

Create a ChatGPT 4.0 plugin that is then proposed within the ChatGPT Store. It can host a RAG that is directly connected to the GPT-4 LLM within ChatGPT 4.0. This enables the user to access multiple domain-specific RAGs within a single interface. It however removes the ability to control the user interface for the domain-specific chatbot developer. See the chart in Figure 7 that connects the various tools and data sources used by ChatGPT 4.0 and GPT-4 and by a website that would access it.

Create a custom "GPT" within ChatGPT 4.0 using your documents. It is however limited to 20 documents, each not exceeding 300 pages, which makes it rather difficult to create a real deep information base. I tried it with my freely available book Understanding Quantum Technologies with splitting it into parts, but searches done on the resulting custom GPT were not satisfactory. For example, it couldn't create lists of companies, like quantum hardware vendors.

Creating a domain specific LLM requires some data processing and software engineering abilities, with LLM technologies and frameworks that are making things much easier. First, the data must be collected (paper references, books, articles, interviews, courses) and then preprocessed (cleanup, formatting), then, a language model must be selected (like GPT-4 or also Mixtral 8x7B that is using a "Mixture of experts" LLM), the model tuned, and a custom website created and deployed around these tools.

Another task to plan is benchmarking, using a test suite of questions and dialogs that could assess the capabilities of such a domain specific LLM and RAG based chatbot. It seems no standard exists there yet in quantum science, so it needs to be created in parallel with a domain specific LLM chatbot project.



Figure 7: chart explaining how a domain specific web site can be built with using some existing LLM engine and model, and a RAG for documents retrieval, plus some LLM fine tuning. (cc) Olivier Ezratty, 2024.

Basic LLM functions are easily accessible, and proof of concepts can be carried out in a couple days provided the training data resources are already well inventoried. A custom website can access an LLM implementation running on it. Obviously, for a professional, industrial, stable, maintainable solution, with an optimized cost, things can be more complicated and will require LLM software development and infrastructure expertise. Also, data sources used to create a RAG knowledge base must be used legitimately with adequate intellectual property license agreements.

Building a self-hosted LLM chatbot can also be motivated by privacy, intellectual property, and security concerns. Generic LLM chatbot services may use user personal information and search queries for their model training and tuning, compromising their privacy and security.

Build of buy and value-costs trade-offs must be made⁹⁸. Domain specific LLMs have already been created or prototyped

in various fields: natural science⁹⁹, climate sciences¹⁰⁰, chemistry¹⁰¹, agriculture¹⁰², finance¹⁰³, legal¹⁰⁴, healthcare¹⁰⁵¹⁰⁶, psychology¹⁰⁷, chip design¹⁰⁸ ¹⁰⁹ ¹¹⁰ and even generic academic research¹¹¹.

As far as I know, there are no quantum domain specific LLM available online. There is an opportunity to launch some useful projects here, with some preliminary analysis on the size of data sources, processing costs and day-to-day inference costs.

Emergent capabilities

AI researchers have observed that large language models exhibit features referred to as "emergent capabilities". These features show up when the LLM reaches a certain scale or complexity and were not explicitly programmed or anticipated by their creators. It can be about giving the perception of understanding texts, generating complex texts, learning from context¹¹², demonstrating multi-lingual abilities for text generation or translation, generating a quiz (see Supplemental Materials, page 81), possessing some problem-solving skills with the ability to break them down into smaller parts, showing creativity by mashing up various contents in original ways and adapting to different writing styles and answering hypothetical questions and even professional skills in human psychology¹¹³ ¹¹⁴ ¹¹⁵.

LLM's ability to output grammatically correct texts is already impressive. But these abilities could primarily be attributed to in-context learning and not be related to any reasoning capability ¹¹⁶. For this respect, an incredible number of LLM generic benchmarks are proposed and continuously updated ¹¹⁷ ¹¹⁸ ¹¹⁹. These tests should be carefully designed to avoid measurement bias¹²⁰.

The quantum science tests described in the Supplemental Materials starting on page 25, are sometimes striking. Users may wonder how these LLM-based chatbots are participating in such dialogs and where and how they are consolidating their responses so concisely. It is easier to understand it when considering how these chatbots are built, with extensions going beyond simple autoregressive LLMs.

As a result of these emergent capabilities, researchers wonder about the nature of LLM-based intelligence. Since LLMs cannot reason per se and lack structured knowledge of the world, their perceived intelligence can be an illusion or a human biased perception. Emergent capabilities in mathematics and programming could indicate some form of understanding derived solely from parsing texts¹²¹. Some researchers argue that we even have already reached the stage of artificial general intelligence (AGI)¹²² ¹²³ and other even demonstrate theoretically that AI could become as creative as humans¹²⁴.

Other researchers argue that LLMs are just smartly remixing existing human-originated content and that various forms of reasoning or symbolic AI will have to be added to LLMs to make them intelligent¹²⁵. LLMs can for example be used to recognized named entities and relationships to help build knowledge graphs used in symbolic AI¹²⁶. Most LLM-based capabilities are the result of their integration with fine-tuning and external functions like various agents, including chains of thoughts ¹²⁷ ¹²⁸.

LLMs showcase their "unintelligent" abilities when generating "hallucinations" - inventing and extrapolating nonexistent content without verifying its existence, and overconfidence¹²⁹. The debate also refers to the study of language as being or not a model of cognition.

Most large organizations that have developed LLMs like OpenAI, Meta and Google want to create so-called artificial general intelligence tools (AGI) that would embody many if not all characteristics of human intelligence. AGI has variable definitions, whether for example it is purely intellectual or is about having a physical interface with the world.

A team at Google made a proposal for a generic ontology for AGI benchmarks using two simple dimensions criteria: depth (performance) and breadth (generality) of capabilities¹³⁰. It proposes six features to benchmark: capabilities and not

processes (no need for think, understand, conscience or sentience, but, here adaptability would be key), generality and performance, cognitive and metacognitive tasks (not requiring robotic/physical embodiment, and they don't mention vision oriented tasks¹³¹), potential and not deployment (demonstration features is enough, no need to kill jobs), ecological validity (meaning, it makes sense for mankind) and used as a path and not an end to AGI. They define 5 levels of AGI, the performance being indicated as a nth adult percentiles (50th, 90th, 99th, 100th) for levels 2, 3, 4 and 5, corresponding to competent, expert, virtuoso and superhuman levels. Generality has only two values with narrow and general. They could certainly add an intermediate value with generality within a field like healthcare, quantum science, philosophy, education¹³² or economy. But they consider generality is being a range of tasks more than a field, so be it. They estimate that level 2, competent, corresponds to the state of the art of today's LLMs. One discussed item is also whether we are comparing AGIs with humans without AI, or humans empowered with AI. Their model can also be used to assess various risks pertaining to the use of AGI^{133 134}. Another approach would be to avoid defining AGI using anthropomorphic analogies.

Some competing approaches are being pursued on the path to AGI. One consists in improving LLMs by training them with even more data. A second is about creating hybrid solutions associating LLMs and symbolic AI tools implementing cognitive and reasoning approaches, beyond the basic current chain of thoughts agents¹³⁵. A last one is about getting rid of auto-regressive LLMs, which may be a dead-end, according to Yann LeCun from Meta. He wants to build AIs that can learn, remember, reason, plan, have common sense, and be steerable and safe and is working on creating some objective-driven AI that would have reasoning and planning abilities, can perceive the world. It would use energy-based optimization models instead of probabilistic models¹³⁶ ¹³⁷.

Whatever the viewpoints, we can expect more "breakthrough" advances coming from LLMs and their related software tools. There will be debates on whether it reaches the threshold of AGI which is still a vague concept. But definitively, it will be some form of advanced intelligence with a lot of super-human and artificial traits¹³⁸. It will be up to us to make good use of it.

III. LLMS IN QUANTUM SCIENCE

Most researchers are already heavy users of digital tools, whether they are theoreticians or experimentalists. They search papers on web sites, write their papers collaboratively with their teams on Overleaf, use data processing, machine learning tools and code to analyze their experimental data and produce various charts, write RFPs and other project proposals to get funding and present their work using slides and so on. Likewise, students need to search for information, to access data, create code, write reports, and present projects.

All these activities can potentially benefit from using AI tools, and particularly LLM-based chatbots as shown in Figure 8. Nowadays, I spend nearly more time using LLMs than search

The speed at which these tools are getting better warrants students and researchers to optimize their daily tasks with these tools. They also must be careful not to be overwhelmed by the flow of inbound information and knowledge brought by these tools. Learning and studying is not just about amassing information, but also understanding it and building upon it.

Literature Review and Data Analysis

- Process and summarize scientific literature.
- Find best interpretation of published literature.

Problem-Solving and Hypothesis Generation

- Assist in generating hypotheses by drawing on existing knowledge and identifying patterns that might not be immediately obvious to human researchers.
- Teaching assistant, providing explanations of complex concepts in quantum physics, which can be particularly useful if you are involved in mentoring or teaching roles.

Collaboration and communication

- Facilitate collaboration by providing summaries or translations of research.
- Help in drafting research papers or proposals.
- Produce meeting transcripts and reports.
- Improve cross-discipline team building with real-time domain semantics translation.

These tools will grow in sophistication and capabilities up to be able to handle many tedious research subtasks. Some early experiments and proofs of concept in other disciplines like chemistry demonstrate that LLM and AI enabled research can fasten of lot the design and discovery of new molecules. However, it requires putting in place a whole ad-hoc software architecture ¹³⁹ ¹⁴⁰. It could probably be replicated in many domains of quantum science and technology.

Experimental design

- Optimizing experimental setups.
- Finding and comparing vendor offerings.
- Suggesting configurations that might yield the most informative results.
- Suggesting parameters for experiments based on previous successful experiments.
- Generate code for experiment drive.

Data interpretation

- Help in interpreting experimental results.
- Identify patterns and insights from complex data sets.
- Suggest explanations or theories aligned with data.
- Generate Python code to produce charts.

Grant and Manuscript Writing

 Aid in writing grant proposals and manuscripts by helping structure documents, check for grammar, and ensure clarity of communication.

Figure 8: a list of quantum physics research tasks that can benefit from LLM help. (cc) Olivier Ezratty with inspiration coming from ChatGPT 4.0.

In the following sections, we will cover some typical quantum science researchers and student needs, with the related LLMs capabilities, some learnings and a prospective on what could be achievable in the future. This future is not farfetched and may be accounted for in just a couple of years.

This doesn't however mean that researchers and academics will be replaced by some AGI. But at least will it push them to be even better at asking good scientific questions and spend less time in realizing tedious bureaucratic or functional tasks.

Learning

Capabilities

Students, researchers, and industry professionals in quantum science continuously discover new concepts and terminology. LLMs can help them to obtain a description of their meaning, before digging further into regular scientific papers and other sources found with more classical research tools. It can also create structured inventories in many domains, gathering information scattered in many unstructured documents.

I tested various scenarios with LLM chatbots in quantum science, discovered their capabilities, how to improve it with some prompt engineering and uncovered some of their limitations and ways to circumvent it. This inventory provides some anecdotal evidence of their capabilities but covers a broad number of situations. I strongly encourage quantum professionals, scientists, and students to test these LLMs on their own to forge their educated experience.

The version of ChatGPT 4.0 that I tested had a knowledge cutoff date of April 2023 but is also able to tap into Bing for upto-date information retrieval. As you read this paper, a more recent version of ChatGPT is maybe available with even greater capabilities. What is important here is the trajectory, which is discussed in various parts of this paper.

So here are 14 out of the many test scenarios I did experiment, from basic to more sophisticated requests, using various forms of prompt engineering:

- 1. On **quantum superposition**, ChatGPT 4.0 avoided the pitfall of saying that quantum objects were simultaneously at two locations. The result is quite good at first glance. What this means is a potential prevalence of academic content over content targeting general audiences did train the LLM training (Supplemental Materials, page 25).
- 2. On **quantum entanglement** and John Clauser and Alain Aspect et al experiments, the result was fine as well. It could even explain it by making a couple analogies, with social networks or flocking birds! (Supplemental Materials, page 25).

- 3. On the **Bloch sphere representation of qubits**, I tried to obtain some explanation on how qubits are represented graphically in the Bloch sphere and other models. The results were moderately satisfactory and disappointing when asking for a graphical representation (Supplemental Materials, page 48). Turning ideas and concepts into meaningful schematics is not yet a capability of graphic generative AIs like DALL-E.
- 4. On the origins of **quantum exponential speedups**, ChatGPT 4.0 got confused at the beginning with saying that Grover's algorithm is bringing it (which is not right, Grover bringing at best a quadratic speedup), and then, also, being slow to explain the differences between Clifford and non-Clifford quantum gates (Supplemental Materials, page 32). This may mean that GPT has not been trained with enough quantum algorithms relevant content.
- 5. On the **differences between QPE and VQE** quantum algorithms that are used mostly for chemical simulations, the result is pretty good. The difference between QAOA and QUBO formulation-based optimization algorithms is also well explained (Supplemental Materials, page 27). Why then is it faring well here? Probably due to the wealth of available content in the domain and fewer discrepancies in the way it is explained online.
- 6. On rubidium 87 being a boson or a fermion, LLMs can behave like bad students who do not prepare their work well and produce incomplete or random responses to complicated questions. In this hilarious dialog with Google Bard on whether rubidium 87 is a boson or a fermion. It is a boson, but when he is being told that you heard that it is also a fermion, it apologizes and responds that it is indeed a fermion, providing totally inconsistent explanations on the parity of the number of electrons, protons and neutrons that helps you get the answer. The reasons for these mistakes are that it requires some rule-based reasoning which is not a feature of LLMs and due to some inconsistencies on the matter in some Internet resources that may have been used to train these LLMs. It will be interesting to see in future versions of ChatGPT 4.0 and Google Bard how they behave in this scenario and implement some form of limited reasoning.
- 7. On **bosonic qubits**, a particular class of self-corrected qubits, I asked ChatGPT 4.0 and Google Bard to describe the various types of such qubits and to provide me with related bibliographical references. In the responses, ChatGPT 4.0 provided me with good answers and with some interesting bibliographical references while Google Bard invented references that do not exist at all (Supplemental Materials, page 33). Google Bard generates a lot of hallucinations while ChatGPT 4.0 seems to be better at avoiding it, maybe due to some of its embedded (fine safeguarding mechanisms tuning, human reinforcement learning, chain of thought).
- 8. On **single photon sources and detectors** figures of merit and vendors inventory, the results were of excellent quality to start working on these topics (Supplemental Materials, page 38). Here, you discover the ability of recent LLMs to

consolidate information scattered in many different documents and in a consistent way. It really adds a lot of value, provided you have some initial background enabling you to quickly fact-check the results.

- 9. On superconducting qubit circuits, circulators and TWPAs. ChatGPT 4.0 was excellent at inventorying the various elements used in superconducting qubits (niobium, aluminum, tantalum, etc.) with providing a summary of where the elements are used and their advantages and shortcomings. Likewise, it seemed to make very good responses on the state of the art and challenges with circulators and the TWPAs, traveling waves parametric amplifiers, that are used to amplify the microwave exiting qubit resonators after their readout. In that example as well as in the photonic sources and detectors one, ChatGPT 4.0 does not necessarily educate you on the matter if you are already knowledgeable about it, but it helps you structure and consolidate existing knowledge (Supplemental Materials, page 51).
- 10. On **qubit fidelities**. LLMs are known for their "hallucinations" when they invent non-existent facts. They can even provide imaginary references. In the dialog in Supplemental Materials page 48, I asked Google Bard to assemble a table with the qubit fidelities of various vendor quantum processors. The result is confusing, the best artefact showing up when the LLM is attributing two-qubit gate fidelities to D-Wave quantum annealers (which is not run with using qubit gates) or to Intel Horse Ridge, which is a cryo-CMOS electronic component and not a quantum processor. The lack of logical reasoning and graph knowledge within LLMs and the inconsistencies in the way qubit fidelities are being reported in academic literature and by industry vendors may explain these results.
- 11. On the difference between MBQC and FBQC, which are two different architectures and paradigms for photon qubit quantum computing using cluster states of entangled photons, ChatGPT 4.0 got first mixed-up between FBQC (fusion based quantum computing, a variant of measurement-based quantum computing proposed by PsiOuantum) and FTQC (fault-tolerant quantum computing) but could then answer the question after some redirection, while Google Bard answered the question well right away from the beginning. Explaining how MBQC works is one of the most difficult popularization tasks in quantum computing (Supplemental Materials, page 29).
- 12. On **researcher biographies**, most LLMs invent a lot of stuff and are in the hallucinations territory, particularly for lesserknown researchers. This is particularly true with Google Bard. ChatGPT 4.0 seems to be better for that respect thanks to its recent refinements (reinforcement learning with human feedback, chain of thought, refined learning).
- 13. On **charts generation**, LLM based chatbots are overusing generative AI instead of searching for existing charts, which can generate strange results. I asked ChatGPT 4.0 to output a graphical representation of the quantum circuit from Shor's algorithm (Figure 9). For the first example, the chart is generated with the following (misleading) comment:

"Here is a graphical representation of a quantum circuit used in Shor's algorithm for integer factorization. This illustration shows the arrangement of multiple qubits and various quantum gates, including Hadamard gates, controlled gates, and modular exponentiation gates. The circuit also includes a section for the Quantum Fourier Transform, and it concludes with a measurement stage where the qubits are measured. This diagram captures the complexity and intricacy of quantum computations in Shor's algorithm". And no doubt or warning about its fancy creation! On top of that, all the labels are wrong and pure nonsense for whatever reason. The second chart is even more creative text-wise. DALL-E seems to lack "adversarial" features that would enable it to check the realism of the various parts in the generated charts, particularly text items, for example with trying to interpret and classify it with some convolutional neural network.

14. In non-technical related requests, on the potential advent of a quantum winter, both ChatGPT 4.0 and Google Bard provide adequate responses (Supplemental Materials, page 44). The question being not technical and more rhetorical, it is easier for LLMs to provide adequate responses.





Figure 9: ChatGPT 4.0 DALL-E creative representations of Shor's integer factoring algorithm.

Learnings

When asking a technical question to LLMs, users are better off with having some prior knowledge to separate the right from the wrong in the responses. Users should not trust any answer by default unless they are already in a safe knowledge zone¹⁴¹. This advice is particularly important for students who may over rely on LLMs for their work and not understand that LLM chatbots behave like stochastic parrots providing "average answers". As ChatGPT 4.0 wrote when reviewing this paper: "*while LLMs provide good overviews, they should not replace primary scientific sources*", meaning, learning with teachers, watching videos, reading scientific papers and books and whenever possible, meeting the best specialists. Adopting quantum physics parlance, it is up to the user to implement some form of post-selection of the results produced by these LLMs, sorting out the good, the bad and the ugly on his own¹⁴².

Chris Ferrie wrote about ChatGPT's mistakes and approximations when trying to explain what quantum computing is, showing that it is parroting what is being repeated ad nauseam in the literature¹⁴³. This is not surprising given LLMs are trained with and reusing the many commonplace mistakes, misconceptions and nonsense widely broadcasted online, even by "serious" sources. This creates an ethical issue, LLMs becoming a potential tool disseminating scientific fake news and fueling part of the quantum hype in a closed loop fashion. So far, this problem has been studied in a generic way but not specifically in quantum science¹⁴⁴.

This can get even worse when some tech media are themselves using LLMs to automatically produce news pieces, that are poorly edited by humans. I found a fancy example in TechBullion, which describes the whereabouts of quantum computing based time travel¹⁴⁵. It becomes involuntary funny when discussing the ethical side effects of time travel: "*The concept of time travel raises ethical and safety concerns that must be carefully considered before conducting any experiments. Tampering with events in the past could potentially have severe consequences for the future, leading to unintended consequences or paradoxes.*".

Creating ethical concerns for some non-existent and totally improbable technology could still have some value for sciencefiction writers! But this may in turn generate a contamination phenomenon, when LLMs are themselves fed by LLM generated content and creating some form of echo chamber¹⁴⁶. Some LLM hallucinations can indeed under certain circumstances be created by science-fiction content available online.

Understanding why and when LLMs fail is quite easy at first glance. On top of being trained with data scrapped from various web sources, it creates texts mashup with no logic despite having an episodic memory enabling them to understand the context of questions and follow-up questions.

Likewise, generated graphics are based on training using broad sets of images and schematics and making predictions. The "adversarial" part seems currently missing in the process. It could have detected that the words in the above pictures do not exist and someway corrected them.



Figure 10: empirical rating of various questions asked to ChatGPT 4.0 and Google Bard. By Olivier Ezratty, December 2023. In blue, creation oriented tasks, in green, information query tasks. (cc) Olivier Ezratty, 2024.

As shown in Figure 11, I just asked ChatGPT 4.0 to "*create a chart with the different types of qubits*" and the result was quite disappointing as with Shor's algorithm chart. This will probably improve over time.

To summarize this "case studies" part, Figure 10 shows a raw rating from 1 to 10 of the various tasks inventoried in this paper. The chart was created by ChatGPT 4.0 because I couldn't do it with Excel¹⁴⁷.

I also used the Data Analysis feature of ChatGPT 4.0 to create a custom log-log chart with qubit fidelities. It took me a couple of minutes to create it instead of the many hours it took me last spring 2023, even when using ChatGPT 3.5 to customize my Python code (see Supplemental Materials, page 79). In less than one year, the productivity gain was enormous.



Figure 11: ChatGPT 4.0 DALL-E generative AI rendition of various types de qubits. Misspelling is systematic with such requests.

Prospective

We can expect that soon, LLMs or quantum science domain specific LLMs will help researchers and students in quantum science with better capabilities. First, with the integration of larger bodies of knowledge including all arXiv and peer reviewed papers from major journals, either directly in custom trained LLMs or with a RAG adjunct to the used LLM. RAG vector bases have the advantage of enabling explainable AI, which would provide its information sources. Some static reasoning specific to the quantum field may be embedded in LLMs via chain of thought, agents, reinforcement learning with human feedback and fine tuning. And of course, most of the tests realized in this part will have to be redone with ChatGPT 5.0 when it comes out in 2024 as well as on new versions of Google Bard based on Gemini.

Studying

Capabilities

Studying in quantum science is about solving problems, doing some mathematics, learning from mistakes, passing tests, putting in place experiments, creating some code to analyze experimental data, producing reports, publishing papers, presentation and submitting a thesis. The toolset grows as students move from license to master and then to PhD level. We will cover some of these needs in other sections related to software development and writing papers. LLM capabilities for students start with the learning phase seen above.

In the case of mathematics, LLMs like Google Bard, as of 2023, could be bad at doing very simple calculations¹⁴⁸. However, ChatGPT 4.0 exhibits better capabilities to assist mathematicians in various ways, from detecting and filling gaps in theorems to acting as search engines and finding definitions from descriptions of

mathematical objects¹⁴⁹. Other options exist like the creation of LLMs dedicated to solving mathematical problems like Minerva from Google AI, which is based on their PaLM LLM model which is finetuned with a mathematical dataset ¹⁵⁰, or with associating an LLM with some logic based AI like with the math-PVS project, that checks theorem demonstrations in the scientific literature. Here, the LLM is used to extract symbolic items from scientific texts that then feeds the symbolic AI system¹⁵¹.

ChatGPT 4.0 benefited from various extensions, like chains of thoughts, enabling it to answer various mathematical questions properly with implementing some logic. You can for example test the famous question: "*if 5 machines produce 5 items every 5 minutes, how many machines do you need to produce 100 items in 100 minutes*". ChatGPT 4.0 provides the right answer (5) with the underlying logic of its reasoning while Google Bard provides a wrong answer (100 machines) with some Python code although it seems it is starting to implement chains of thought processes, and Meta Llama also provides a wrong answer (20), this time with some reasoning, which happens to be false. Now, are LLMs able to do linear algebra computing and use bra-ket notation? I don't know yet.

Demonstrating theorems is an AI feat that may necessitate using symbolic AI techniques like what Google achieved with its AlphaGeometry engine, a theorem prover for Euclidean plane geometry that is approaching the performance of an average International Mathematical Olympiad (IMO) gold medal winner¹⁵². This is probably applicable to other branches of mathematics.

Learnings

LLMs are good to pass various tests and answer closed questions. Scott Aaronson – who now works for OpenAI - used ChatGPT 4.0 to pass his final course exam that contains complex questions on linear algebra and the likes. It fared relatively well with a 69/100 grade when his students get an average of 74.4/100^{153 154}. The exam maximum score is 120 and the highest score obtained by a student was 108. It did perform badly on open questions and on mathematic operations and, of course, on reasoning tasks¹⁵⁵.

Other LLMs experiments in quantum physics fared also relatively well at a basic level¹⁵⁶ ¹⁵⁷ ¹⁵⁸. These quantum exams are particularly difficult. In comparison, GPT-4 Turbo largely exceeded human performance in passing the GMAT test used to evaluate candidates to business schools in the USA at MBA levels¹⁵⁹. Using these tests with LLMs can help quality their ability to teach and help students.

As was the case with many other generic tests, LLM-based chatbots will probably perform even better in the future and reach the grades teachers are obtaining today, particularly when they will embed even more reasoning artefacts¹⁶⁰.

But you can make things upside down and ask ChatGPT 4.0 to create your own multichoice quiz¹⁶¹. As you can see in Supplemental Materials page 81, ChatGPT fares very well in creating a multichoice quiz in quantum computing. It failed at creating a 20 questions quiz in the questions numbering so I split it in chunks of 8 questions and then, it worked well.

Prospective

Quantum physics is a complicated field with a lot of subtleties. Thus, the high level of errors made by LLMs in the field ¹⁶² ¹⁶³. But you can expect that LLM-based chatbot will get better over time to provide educated answers on very specialized topics in most scientific domains. It may happen with the creation of quantum science domain specific LLM chatbots.

LLMs systems could help "*create educational material or tutorials for new researchers entering the field of quantum science, facilitating a quicker and more comprehensive understanding of the subject*" (according to ChatGPT 4.0). Personalized learning can or could be about using customized content aligned with the student profile (requiring some prompt engineering), building interactive Q&A Sessions, generating examples and analogies which can be particularly useful in quantum physics although deserving some selective approaches, explaining the technical jargon, identifying relevant articles and creating abstracts, running tests and providing students with explanations of why their initial answer is wrong and helping them land at a correct solution on their own, and of course, providing language translation and video transcripts^{164 165}.

Reviewing papers

Capabilities

Reviewing a scientific paper is routinely done by all scientists looking for the state of the art in their domain and in a more refined way by assigned referees working for peer-reviewed scientific journals. It is tedious work, and, in most disciplines, there are not enough referees.

On top of that, in many cases in quantum technologies, papers cover many disciplines, and many reviewers sometimes have a hard time figuring out the whole concepts laid out in the review paper.

State of the art LLM chatbots can help you analyze scientific papers at a basic level. They can summarize it, extract key points and even data¹⁶⁶.

Learnings

Summarizing arXiv papers works on Meta Llama (which uses some arXiv content in its training¹⁶⁷), but not with ChatGPT 4.0 and Google Bard. I tried it with an arXiv preprint of my own and it provided me with some good feedback but was unfortunately highly "templatized". It did not fare very well on the recent long and multi-facetted Mikhail Lukin's paper describing how 48 logical qubits were implemented with 280 cold atom physical qubits¹⁶⁸ (see Supplemental Materials, page 64). In the second case, ChatGPT 4.0 has already been tested to help peer-reviewers prepare their work and early results show that it performs better than classical machine learning techniques, although not yet specifically in quantum science¹⁶⁹.

Based on ChatGPT 3.0, SciSpace can make some summaries of scientific papers and present them in table/column format which can be helpful for spotting interesting papers¹⁷⁰.

In the second case, improving the productivity of peerreviewing papers may also be based on exploiting LLMs and other related extensions¹⁷¹. They can potentially help by checking for basic compliance with the journal submission guidelines, scope relevance, and detecting issues like plagiarism or missing sections. They can also help with factchecking some basic assertions, help in verifying the experimental data, statistical analysis, and data interpretation, and detect potential scientific or other biases.

Prospective

We can extrapolate a little on the potential features of future automated reviewing tools that would detect the trustworthiness of a paper and document its findings. It has the potential to exceed what current AI are doing which are focused on paper selections, basic data extraction and summarizing¹⁷². Given an arXiv preprint reference or peer reviewed paper, it would create a timeline of the paper history showing when it was published in a peer-review paper, and maybe retracted at some point, with some figures of merit of the paper like a number of citations, additional resources (supplemental paper, data sources, key papers mentioning it) and some background on their authors (affiliations, number, role, age, history, h-index, etc.).

It would require some programming, tapping into various API¹⁷³ resources and some natural language based consolidation reporting. General audience LLM chatbots would maybe not be the right tool for this ^{174 175}!

Then, LLM-based tools would uncover the paper strengths and weaknesses, key and missing points in experimental protocols, data reporting and inconsistencies, advancements vs the state of the art, and missing figures of merits like when touting a new qubit type without providing any single or multiple qubit gate fidelities. This would be particularly helpful when reviewing quantum science papers covering many different disciplines. Also, several research teams are even already working on creating automatic improved captioning for the complicated charts embedded in scientific literature¹⁷⁶.

Writing papers

Capabilities

LLMs are powerful tools to create relatively simple written content. You can ask them to lay out a table of content for a paper or report and then to create each part separately. It can potentially help build a bibliography and even some of its content. It can rephrase an introduction and conclusion to make it more impactful, or select a particular style, correct grammatical and spelling errors, and even make the text easier to read.

Other types of document production already benefit from the help of LLMs like grant RFP responses or those that are created by startups when preparing a funding round.

Learnings

As I teach quantum technologies to IT engineers, I am eager to find ways to explain things better. At EPITA, between each course, I task students with writing a short memo on various topics, including some for which I don't even have the definitive answer. They all use search engines and LLMs to quickly craft their work and the results are enlightening on the strengths and weaknesses of both LLMs and this approach. They are usually quite bad if not used with sufficient preliminary work and understanding.

The submitted documents are also relatively short, which corresponds to what usually comes out of LLMs at first glance when asking questions without digging further and due to their current context window token limitations.

Abusing LLMs can lead to decrease student performance¹⁷⁷: "It has the "potential to reduce critical thinking and independent learning. [...] One possibility is that ChatGPT can provide students with easy or incorrect answers to their questions, which can lead to a superficial understanding of the material. [...] Another possibility is that ChatGPT can make students less likely to engage actively in learning, which can lead to a decrease in knowledge retention". There is a risk for students of gradually losing abilities to develop their own analytical and problem-solving skills.

This leads some universities to consider ways of disciplining the use of LLMs in education¹⁷⁸. As explained by Nir Chemaya et al¹⁷⁹, when relying too much on LLMs, they can also: "generate a text with mistakes, including incorrect math, reasoning, logic, factual information, and citations (even producing references to scientific papers that do not exist)". Surprisingly, it is relatively easy to detect essays written by LLMs since "ChatGPT generates essays that are rated higher for quality than human-written essays" ¹⁸⁰. When written with too good English, the trick can easily be spotted, at least for non-English fluent native students although, we are not sure about it even for English fluent natives. Unless some careful prompt engineering is done when querying the LLM, the result will have a very flat style. And it is usually between two and three pages long.

A way to circumvent this side effect is to ask students for a documented bibliography of their writing. This does not mean students (and teachers as well) should not use LLMs in their work. It should not just be the only data research tool being used and be one among many other tools. They must use different parts of their brain in their work!

Plagiarism in science has existed for a while, even before the Internet existed. But Internet tools made it easier, and LLMs are new tools making it even easier. It creates a sort of metaplagiarism since LLMs are reframing and rephrasing stuff already coming from third-party sources, and this content can itself be reused in other places ¹⁸¹ ¹⁸² ¹⁸³. Can modern AI based software tools and LLMs help detect emerging forms of plagiarism? No yet. It is still an area of research¹⁸⁴ 185. ChatGPT 4.0 says no. Google Bard says yes and is using a couple methods (citation analysis, text similarity detection, visual similarity) but is not actually capable of doing it. Llama 2.0 says it can. I tried GPTZero which found that my student reports were generated by a range of 48% to 90% by LLMs¹⁸⁶. Several other paid websites are offering plagiarism detection features that use many AI techniques and not just LLMs. But they are still unperfect¹⁸⁷.

A survey conducted by Nature also showed that 25% of researchers were using LLMs to write responses to research grant RFPs (requests for proposals). It included researchers in all disciplines with an apparent LLM usage peak with computer science researchers. We could expect that quantum physics researchers would be less keen than quantum information science researchers to use LLMs in writing RFP responses ¹⁸⁸ ¹⁸⁹.

Content can be created for other means. I asked ChatGPT 4.0 to produce a memo to convince the government of a small European country to launch its own National Quantum Initiative and to turn it into Serbo-Croatian as an example, and to explain how to leverage European Union funding. The result is concise, not bad (although maybe too concise in the Serbo-Croatian version) and bound for being completed (see Supplemental Materials, page 74).

A recent study done at the Harvard Business School did uncover that BCG consultants were less good than others when using ChatGPT 4.0^{190} .

The reason was that LLMs are not good at solving business problems with actionable strategic recommendations and is leading to groupthink. It is not surprising given the content is the result of a sort of "majority polling" of existing ideas, probably excluding the most original and innovative approaches. This could have detrimental effects on students and researchers as well.

Prospective

On one hand, we should not expect LLM-based chatbots to do too much work in place of students and researchers. Otherwise, they won't learn or discover anything on their own. LLMs will produce content at a too fast pace, incompatible with the speed of a normal human brain. We should thus imagine a future where the tedious and repetitive tasks are delegated to LLMs while the demanding intellectual tasks are still in the hands (or brain) of humans.

What could be useful here relate to the "reviewing papers" section, where the related tools would help a researcher or student identify the weaknesses of his/her writings and provide recommendations. For example, it would be useful to better assess the quality of illustration content and captions¹⁹¹.

Software development

Capabilities

LLMs are very useful for developing, tuning, and even debugging code in many languages¹⁹² ¹⁹³. They are fed by huge swaths of open source code coming from StackOverflow, GitHub, development tools documentation and the likes. Along with ChatGPT 4.0 Data Analyzer itself, code creation can help build charts and data analysis tools¹⁹⁴.

All this can improve the productivity of many quantum scientists who rely on coding in their work, whether they are experimentalists or theoreticians¹⁹⁵. It can contribute to empowering them in analyzing the data generated by their experiments.

🖞 Copy code python from qiskit import Aer from giskit.algorithms import Shor from qiskit.utils import QuantumInstance # Setting up the quantum instance (simulator) backend = Aer.get_backend('aer_simulator') quantum_instance = QuantumInstance(backend, shots=1024) # Number to be factored N = 15 # Example number to factor num_qubits = 10 # Number of qubits # Initialize Shor's algorithm with the number of gubits shor = Shor(num_qubits=num_qubits, quantum_instance=quantum_instance) # Execute the algorithm to factorize the number result = shor.factor(N) # Print the result print(result['factors'], result['total_counts'])

This code sets up Shor's algorithm in Qiskit to factorize the number 15 using a 10-qubit quantum simulator. Please ensure that you have Qiskit installed in your environment.

Figure 12: LLMs can help code quantum algorithms. Some simple Qiskit code generated by ChatGPT 4.0.

LLMs can create quantum code like with Qiskit like in the above example in Figure 12, showing how to use a Shor integer factoring algorithm existing function which provides no details on its implementation. I tried different other scenarios and, while in text mode, ChatGPT 4.0 could explain what it did, the actual generated code was quite simple, and avoiding any details. LLMs can also help code writing within the code editor itself, the Qiskit Code Assistant. That is what IBM announced in December 2023 with a prototype LLM designed with IBM watsonx^{196 197}.

Microsoft announced in November 2023 its Copilot learning assistant in the Azure Quantum code editor, that is based on the GPT-4 LLM. It is currently specialized in chemistry research¹⁹⁸.

qiskit_code_assistant.py ×		_code_assistant.py × □	
	1	<pre>from.qiskit.import.QuantumCircuit</pre>	
	2	<pre>from qiskit.primitives import Sampler</pre>	
	3	<pre>from quantum_serverless import save_result</pre>	
	4		
5 # Define a Bell circuit, run it using the		# Define a Bell circuit, run it using the Qiskit Sampler	
		primitive (not the Runtime one)	
	6	<pre># take the quasi_dists result from Sampler and save it using</pre>	
		save_result from quantum_serverless	
		circuit = QuantumCircuit(2)	
		circuit.h(0)	
		circuit.cx(0, 1)	
		circuit.measure_all()	
		<pre>sampler = Sampler()</pre>	
		<pre>quasi_dists = sampler.run(circuit).result().quasi_dists</pre>	
		<pre>save_result(quasi_dists)</pre>	

Figure 13: the experimental LLM drive code generation tool from IBM

They built an LLM with 20 billion parameters, 8,000 context length, that was trained with 370 million text tokens using a collection of Qiskit examples. This will be proposed in IBM Quantum Platform starting in 2024.

In another experiment, the CodeAid project is a LLM-based programming learning assistant that provides a step-by-step aid to students driving their attention and learning. It is not applied to quantum programming but would be quite interesting in that domain¹⁹⁹.

Learnings

I use LLMs on a regular basis to find how to program, mostly in Python, PHP, JavaScript, or VBA as far as I am concerned, to compactify, factorize or optimize code, to create complicated regular expressions for parsing text (regex) and even to spot mistakes or error sources in faulty code. I also use it to create charts in Python. This code production capability is one of the most amazing in LLMs. And it helps save a lot of time, whether you are a full-time developer, or a casual one, creating ad-hoc tools for your own needs.

But as soon as the requested code must contain some logic, the results seem deficient, as it was experimented early on by the startup QuantumPath in March 2023 with ChatGPT 4.0^{200} . Indeed, LLMs create a lot of faulty codes when there are no sufficient online examples and documentation. It can hallucinate with inventing APIs (application programming interface) that do not exist.

Like in many other cases, you spot the mistake and tell it to the LLMs. It then apologizes and tries again. After 2 or 3 trials, it produces a usable result. Still, despite the sometimes tortuous path to obtaining a satisfactory result, you probably saved time in the process.

Prospective

LLM coding capabilities may improve thanks to better chain of thought, reinforcement learning, fine tuning, and larger training sets. Being able to be creative in reasoning is beyond their scope and will require other reasoning and symbolic AI tools.

In the future, you can still expect LLMs to create new algorithms. LLMs are currently experimented for designing specific classical algorithms but could well be extended at some point in the design and optimization of quantum algorithms, particularly in the way they are structure between their classical and quantum parts²⁰¹.

Ideation

Capabilities

Can ChatGPT 4.0 and the likes help quantum scientists find new ideas, research areas and experimental designs? The literature is void on this topic but scientists in adjacent disciplines like in biology and chemical engineering have been looking at these tools' potential. Extrapolations can be derived from these early experiments and thoughts²⁰².

LLM Cheat Sheet in Quantum Science

- Learn how to create adequate LLM chatbots **prompting** (details page 8).
- Define your **user profile** with custom instructions, in ChatGPT 4.0 (details page 8).
- Create your **custom GPT** and upload your reference documents (20 documents limited to 300 pages each).
- Test both ChatGPT 4.0 and Google Bard whenever deemed necessary (see examples in Supplemental Materials starting page 25).
- Use LLMs chatbots to get technical **definitions**, technology **inventories**, **figures of merits** and identify (at least existing) **research challenges** (see examples in Supplemental Materials starting page 25).
- See how they **evolve** over time. ChatGPT 4.0 is much better than ChatGPT 3.5!
- Use LLMs **software code generation** to fasten your charts creations and other routine data analysis and tasks automation (see one example page 79).
- Learn how to **identify LLM-generated text** when reviewing student projects.
- Don't take LLM results for granted. Check it with reliable sources when in doubt.
- Learn the **basics of machine learning** and how it is applicable to your own research work (see examples starting page 21).

Learnings

Yang Jeong Park et al laid the ground for chatbot-based research ideation in chemical engineering, showcasing the existing features of ChatGPT 4.0 and how it can be tuned²⁰³. The observe that "*ChatGPT-4 is knowledgeable, frequently wrong, and interesting to talk to. In other words, not unlike a college professor or a colleague*". They define what would be "genuinely interesting scientific hypotheses" as "whether after a conversation, some experienced practitioner can feel motivated to do a new experiment/simulation/theoretical investigation in the direction pointed out by the generative AI, that he or she cannot think of doing before and that was not already available in the scientific literature".

They observe that GPT-4's responses closely reuse existing concepts and ideas, but it is also able to hybridize disciplines and ideas and mutate the meaning of concepts and nomenclature. It generates many errors with some form of idea oversampling, but human skills can skim through it and detect original ideas, again, using some forms of post-selection. They highlight the importance of prompt engineering and their examples on the choices that can be made to optimize ceramic solid-state batteries are enlightening.

Prospective

Yang Jeong Park et al describe various paths for improvements for LLMs to become better "hypothesis machines", reminding the reader that "science's development can be considered a multi-agent swarm effort in active learning, where peer scientists and competing schools of thought debated and sought out critical experiments, in order to repudiate or further refine hypotheses, and mine for the truths. Inherently it was an evolutionary, multi-agent process" and that "creativity might necessarily be associated with making errors, even "obvious" errors". They estimate that future generative AI systems will need to use networks of multiple hypothesis machines, "each designed to perform specific roles in an iterative learning process".

Extrapolating these thoughts to quantum science and engineering would make a lot of sense. Maybe it should also consider the breadth and diversity of the topic, although this paper author may have a bias, thinking that this domain is more multidisciplinary than others. One should probably first look at hypothesis buildup in very specialized domains, like on how to improve the efficiency of single photon sources, before asking the chatbot to design a full-fledged MBQC based quantum computer. We can also expect that the capabilities of generic LLM-based chatbots will improve either with ad-hoc reasoning extensions (chain of thoughts, reinforced training, fine tuning) or with some potentially entirely new "symbolic AI" based approach.

Collaborative research

Capabilities

I asked ChatGPT 4.0 to review a draft of this paper and it advised me to add some points on how LLMs could facilitate collaborative research. So, I asked it to provide me with some examples of such facilitated collaborative research.

Among other things, it explained that LLMs can enhance communication between different scientific communities with translating complex concepts and bridging language and terminology barriers. It is indeed very useful to do that between quantum physics and quantum information science specialists.

Learnings

It also says that LLMs could:

"suggest a variety of approaches to a quantum science problem, some of which may not be immediately obvious to human researchers" which is not seen yet,

"help identify gaps in current research, suggesting new avenues for exploration", I then asked ChatGPT 4.0 to elaborate with explaining how LLMs could improve collaborative research and to identify research avenues to improve superconducting qubits and the result was quite impressive (in Supplemental Materials, page 66).

"assist in project management by organizing meeting schedules, setting reminders for deadlines, and keeping track of project milestones, and aid in the preparation of project reports, grant applications, and progress tracking documents", which may be possible but would require some additional connections between LLMs and the related software tools.

On top of that, it could even help with providing "preliminary advice on the ethical implications of quantum science research" and "offer general guidance on navigating the complexities of patent applications and publication processes in quantum science". It indeed advised me to discuss the ethical side effects of using LLMs in quantum science in that very paper.

Prospective

We can expect that soon, collaborative research paper authoring tools like Overleaf will integrate some of these interesting features. LLMs could also provide ideas for collaboration, scouting for researchers in all countries having interests matching your needs. When being able to tap in realtime Internet resources, LLMs may help researchers find and hire students (PhDs, post-docs).

Although most academic research is open, ensuring privacy for its underlying process and data exchanges should be a matter of concern. When using ChatGPT 4.0, you are training its model through its own reinforcement learning with human feedback processes.

Some of your work can be recovered through reverse prompting engineering. This advocates for using a safe domainspecific quantum science LLM chatbot.

Market considerations

While we may not see the advent of a quantum winter, it is summertime in AI land. At the Q2B Santa Clara conference organized in December 2023, Bob Sorensen from Hyperion Research did present one result from a customer survey, showing that LLMs were a short-term competitor for end-user interest in quantum computing by 42% of the respondents²⁰⁴. There will indeed be some competition in the end-user market between the various technological advances that may help customers solve their business problems more efficiently. Of course, they should not discount one technology because a new one is getting the attention²⁰⁵.

Like in quantum space, the AI ecosystem has its hype effect and related doomsayers. The ChatGPT craze which started late 2022 spurred the creation of hundreds of startups repackaging it in various domains and applications. OpenAI's direct competitors like Mistral AI raised hundreds of million dollars at billion-dollars-level valuation. Some preeminent AI researchers like Yann LeCun and Luc Julia are complaining about the LLM hype and predicting that it will suffer its own winter.

One significant difference is that quantum computing's hype happens while quantum computers are not yet able to deliver any computing advantage in the near future while LLMs actually work, are widely used, but with lots of security, safety, trust and ethical issues²⁰⁶. These limitations are of great concern when LLMs are used in sensitive domains like healthcare.

We are not there yet with quantum computing, even when considering their potential long term threat on cybersecurity.

Another question on "the market" deals with the software offerings specifically empowering the quantum ecosystem. I had a look at some of the most visible natural language startups. Seemingly, they are focused on helping companies optimize their marketing, communication, and sales operations, not their research. Countless companies provide emails creation automation services (Smartwriter.ai, Hypertype, Autobound), then content production for social networks and web sites (Lately, Writesonic, Jasper, Anyword, Persado, Regie.ai, Hypotenuse, Copymatic, Copysmith, AI21labs, Letterdrop, Compose AI), some for customer support (Mavenoid) and only a few that could help researchers like Cogram (meeting notes) and Genei (research and summarization tools).

What that means is the commercial market may not be able to deliver LLM-based tools fitting the needs of researchers, particularly in quantum science which is a rather small domain in number of people involved. As a result, researchers will have to either develop their own tools, rely on open source software and/or launch some collaborative projects.

It would make sense for these projects to be international in scope, or at least, regional (European Union, North America, APAC), particularly if their supporting infrastructure costs are significant. These could be totally or partially supported by public research supercomputing data centers.

One can still wonder how these advanced AI systems will consolidate access to scientific information, like Google did with search engines. ChatGPT is already positioned as a content access platform with its Store. Domain-specific information providers will be faced with a tough choice: either create a plugin for the dominant AI/AGI platforms like ChatGPT and Google Bard or create their own website, without benefiting from the genericity of these platforms. Publishing groups like Springer (Nature, EPJ, ...) and the American Physical Society (PRX Quantum, PRL, PRA, PRB, ...), on top of arXiv (Cornell University) will also have to decide how and where they license access to their content or if they want to create their own LLM chatbot, given most physicists will prefer an AI service that is trained on all scientific sources and not just one.

IV. MACHINE LEARNING IN QUANTUM SCIENCE

Aside from LLMs based tools, machine learning has become a key contributor in the development of quantum technologies, particularly with quantum computing. Small and large, these use cases are appearing in a broadening set of research projects and with various industry vendor offerings.

They rely on different machine learning techniques, mostly dealing with structured tabular data analysis. Even Google supremacy's experiment in 2019 was using a machine learning algorithm to optimize the frequency distribution of its qubits in the nearest neighbor layout of its 53 qubit processor ^{207 208}.

I looked at my own book to make a small recollection of some of these use cases, plus some new ones in the following parts²⁰⁹. These are just a variety of examples, and this inventory is not exhaustive.

Again, I asked ChatGPT 4.0 to identify some applications of machine learning in quantum technologies and it provided me with a good list of examples, although with not much details and references (in Supplemental Materials, page 73).



Figure 14: exemples of quantum mechanisms machine learning based approaches to compute the ground state of many-body systems. Source: Xuan Zhang et al²¹⁰.

Machine learning is now commonplace in most quantum sciences and technologies disciplines. In some cases, there even seems to be some abuse of machine learning usage, with not much consideration on scalability and cost issues, for example when using it in error syndrome detection with large surface codes.

Quantum physics and hardware

In many situations, solving quantum physics equations to simulate a quantum many-body system is too difficult for current classical and even quantum hardware and software. Let's call it the "bottom-up approach". In that case, the envisioned "top-bottom approach" consists in using machine learning using observation data to predict the state and evolution of those systems. The typical trade-off exists to simulate protein folding.

With DeepMind's AlphaFold deep learning tool, up to 450 amino-acid proteins folding can be simulated when, as of 2024, the record with a quantum computer is 10²¹¹²¹².

As a result, quantum physicists use many machine learning techniques to simulate quantum systems, like for determining the ground state of spin systems in lattices or of many-electron systems, mainly molecules. These methods use Monte Carlo simulations, various forms of neural networks (feed-forward, convolutional, recurrent, graph) as shown in Figure 14. Many of these techniques use tensor network optimizations²¹³. They directly compete with quantum machine learning approaches using VQE (in NISQ fashion) or QPE (in FTQC fashion).

In quantum physics, machine learning is also used to "to analyze quantum measurements, estimate the parameters of quantum devices, discover new quantum experimental setups, protocols, and feedback strategies" ²¹⁴215.

At a higher abstraction level, and in the development of quantum computing hardware technologies, classical machine learning is also being used in many domains, mostly for various optimization tasks and with error mitigation and correction.

- Superconducting qubit low-level control optimization can use deep reinforcement learning²¹⁶.
- **Qubit pulse control optimization** can also be done with machine learning²¹⁷.
- **Qubit run-time calibration** can use machine learning to choreograph the frequency trajectories of frequency-tunable superconducting qubits with Google, in line with what they did in 2019 ²¹⁸.
- **Qubit placements** is another case. In 2020, a team from the University of Melbourne showed how machine learning could help calibrate the placement of phosphorus atoms in a 2D structure of qubits on a silicon substrate²¹⁹. Another team, from NIST, did test a similar approach with tuning quantum dots based silicon qubits²²⁰.
- Machine learning could be used to optimize the **energy** exchange mechanisms in quantum computing²²¹.
- Quantum annealers error mitigation is another use case, using a mix of linear regression and support vector machines techniques²²².

• Quantum physics machine learning incorporates and respects the laws of quantum physics on top of using data based models²²³. It was used for example to demonstrate that hidden disorder in a nanoscale SiGe heterostructure device can be inferred with indirect measurements and physics-aware machine learning²²⁴.

Quantum software and tools

Quantum software and tools are an obvious area for machine learning when data analysis and optimizations are at stake. LLMs start to show up here are well.

- Quantum Error Mitigation (QEM²²⁵) and Quantum Error Correction (QEC²²⁶) can make extensive use of machine learning techniques in their data processing. QEM using ZNE (Zero Noise Extrapolation) is a direct use case of machine learning²²⁷. Transformer generative AI can be used to improve Quantum Error Correction syndrome decoding²²⁸.
- Quantum circuit optimizations in compilers and transpilers can also benefit from deep reinforcement machine learning techniques²²⁹ ²³⁰ ²³¹ and reinforcement learning²³².
- **Q-TRL** (Australia) Boulder Opal is quantum control infrastructure software working at the firmware level which leverages machine learning to improve qubits control pulses and optimize quantum error correction codes. It is a Python toolkit used by quantum computers designers that works with IBM Qiskit, Rigetti and with Quantum Machines pulse generators. They implement error-suppression techniques that increases the likelihood of quantum computing algorithm success between 1,000x and 9,000x on quantum hardware, as measured using the QED-C algorithmic benchmarks.
- Variational quantum algorithms can be optimized with using LLMs for tuning their ansatz parameters²³³ or for creating the whole circuit²³⁴. It corresponds to optimizing the quantum circuit that is executed many times in the quantum part of a variational algorithm that is tuned by a classical part.
- Generative Adversarial Network (GAN) can be used for small molecular drug synthesis having some affinity with a given biomolecular target. In that case the GAN generator and/or discriminator are implemented with using a variational quantum circuit²³⁵ ²³⁶.
- **Experimental data** can also be analyzed by various machine learning techniques²³⁷.

Other domains

The fields of quantum communications, cryptography and sensing are also benefiting well from machine learning.

• Quantum Random Number Generators benchmarking tools can use machine learning techniques and a convolutional network to detect patterns in the generated numbers²³⁸ ²³⁹.

- Quantum Key Distribution parameters like joint polarization, phase estimation and compensation, as well as encoders, decoders and key rates prediction can be optimized with using various machine learning based techniques²⁴⁰.
- In **quantum sensing**, NV centers can be tuned with machine learning and Bayesian inference methods to correct the noise found at higher temperatures²⁴¹.

V. QUANTUM COMPUTING AI APPLICATIONS

Quantum Machine Learning

I refer the reader to the 12 pages related to Quantum Machine Learning in "Understanding Quantum Technologies 2023"²⁴². It identifies quantum versions of classical machine learning and deep learning in a wealth of categories like SVM (Support Vector Machine), PCA (Principal Component Analysis), Decision tree classification, Automatic data clustering, Quantum Graph Neural Networks, Quantum Convolutional Neural Networks (QCNN), Recurrent Neural Networks, Reinforcement Learning, Hybrid transfer learning, Generative Learning Models, Capsule Networks, Reservoir Computing Networks, Self-Attention Mechanisms, Geometric Quantum Machine Learning, Image classification and analysis and Federated Machine Learning²⁴³. Most classical machine and deep learning algorithms already have a quantum equivalent.

QML algorithms are either variational algorithms targeting existing NISQ hardware, with no visible quantum advantage at this point, or FTQC future hardware, relying a lot on a QFT (quantum Fourier transform) and the HHL linear algebra equation solving algorithm²⁴⁴. Most existing NISQ QML applications have been experimented with fewer than 30 qubits and are therefore not in any quantum advantage regime, even when compared to low-end classical hardware²⁴⁵. It is still unsure whether NISO-based machine learning will bring any form of quantum advantage. We may have to wait for FTQC based solutions to reach that point, which are dependent on future platforms that will probably not show up in less than 10 years. But any QML quantum advantage in FTQC regime is highly dependent on ability to load data efficiently²⁴⁶. It works well only with sparse data, which may be contradictory with obtaining some quantum advantage with machine learning applications. It may also depend on the upcoming availability of qRAM (quantum random access memory) that is still a concept and not even prototyped on a small scale.

One relevant branch of QML in the context of LLMs is QNLP, for Quantum Natural Language Processing. It was pioneered by Bob Coecke et al from Oxford University with the Categorical Distributional Compositional (DisCoCat) model²⁴⁷. So far, these are interesting theoretical concepts that have been experimented on fewer than 10 qubits. They will now need to provide some resource estimations on data sets and quantum hardware platforms requirements that would bring some quantum computing advantage, let alone building out a full LLM ala ChatGPT ²⁴⁸ ²⁴⁹.

Quantum LLMs

How about using quantum computing to create more efficient LLMs or even GAN (generative adversarial networks) and various forms of generative AIs? The simple answer is "no" despite what can be found saying the contrary²⁵⁰. At least in the short to mid-term.

LLMs use very large data models with billions to trillions of parameters and are trained with enormous amounts of data. Both ways, for training and inferences, the involved data is way too large to be ingested in any quantum computers even in the very long term future of FTQC (fault-tolerant quantum computers) with whatever number of logical and memory qubits. The main reason is that quantum computers are not made for big data. Their main use cases are with small data and highly complicated problems like chemical simulations and some optimization problems with small data and an exponential number of combinations.

This is related to the speed at which data can be ingested in a quantum computer. The equivalent of a "clock speed" of a quantum computer is quite slow, in the 2 to 15 kHz for IBM quantum computers and up to a thousand times slower with trapped-ions qubits. Even with parallelizing operations across thousands of logical qubits, this would be a very slow process, much slower than any classical computing operation²⁵¹.



Figure 15: an Nvidia SuperPod made of 127 DGX servers equiped with 8 GPGPU each. Source: Nvidia.

In the best case scenarios, quantum computers used in complicated machine learning tasks will run a small part of the solution, the bulk of it, particularly data ingestion and various vectorization tasks required in natural language applications, will still be done classically, using large datacenters and systems like the Nvidia Superpod pictured in Figure 15. And for LLMs, the encoding of the preprocessed data into a quantum computer would still be too large unless it would be very sparse data. But alternative classical hardware platforms like silicon photonic networks may be of interest to train and run LLMs²⁵².

Quantum Inspired Algorithms

Quantum inspired algorithms are classical algorithms whose design is inspired by quantum algorithms and interference management, but not programmed as quantum algorithms run through a classical emulator²⁵³. These algorithms can be helpful for solving various problems including images classification²⁵⁴ and machine learning ^{255 256}.

Creating a quantum inspired algorithm is sometimes said to rely on "dequantizing" or the "de-quantization" of a quantum algorithm²⁵⁷. It is not a mirror expression of neural network and LLM "quantization" which turns floating point neural network weights (FP32 or FP16) into integers, reducing the model size and its inferences computing power requirements, which is useful to run standalone language models on mobile or unconnected devices²⁵⁸ ²⁵⁹.

Most quantum inspired algorithms make heavy use of tensor network frameworks. They are already used with neural networks programming in classical deep learning.

Quantum inspired software already intersects with LLMs development. In November 2023, Multiverse launched CompactifAI, a software compressor of LLMs using tensor networks with the aim of reducing energy and compute costs²⁶⁰ ²⁶¹. It is designed to reduce energy requirements at multiple points in the lifecycle of an LLM, including the training phase, general operations, and retraining. Multiverse's new software also reduces the overall footprint of the models, making them more portable and easier to run at the edge in applications such as autonomous vehicles and remote production facilities. Likewise, Hyper Intelligence released a similar offering the same month²⁶². Many other compression techniques exist.

VI. DISCUSSION

The landscape studied in this paper shows an unbalanced situation where machine learning is currently helping quantum technologies more than the other way around. As a result, "*ask not what quantum computing can do for AI, ask what AI can do for quantum science*" ²⁶³!

The emergence of LLM-based tools is making it even more unbalanced as the intersection of LLMs and quantum science reveals a landscape rich with significant potential. As tools like ChatGPT, Google Bard, and Llama become more sophisticated, they offer unprecedented opportunities for educational support, research assistance, and even complex problem-solving in quantum technologies.

However, their current limitations, particularly in accurately handling the nuances of quantum physics and the risks of misinformation, necessitate a cautious and well-informed approach to their use. With the adjunction of various forms of human originated reinforcement training and reasoning, we can expect that LLM tools will grow in capacities way beyond what we can imagine today. Proper user training will also be helpful here.

The future of LLMs in quantum science also demands continuous evaluation and refinement, particularly in terms of training data diversity, ethical considerations, educational practices, and their ability to adapt to the rapidly evolving landscape of quantum research. It will probably create a new playing field for all researchers in this domain. Quantum scientists will need to get accustomed to these tools which complement their already rich digital toolbox.

If we project ourselves in the future, ahead of only a couple years, we can expect that LLMs will be improved in many ways with:

• Embedding Hybrid AI Models combining neural network and transformer-based LLMs and rule-based symbolic engines able to implement various forms of reasoning. It is particularly important in the subtle fields of quantum physics and quantum information science. Other hybridization-based improvements will involve enabling LLMs/AIs to better analyze up-to-date content in real time.

- Integrating Knowledge Graphs that are used to structure knowledge, providing a framework for logical reasoning and fact-checking used in the above hybrid models.
- Implementing Explainable AI (XAI) Techniques including providing scientific references and a contextualized bibliography for all responses. This is already possible with extending LLMs with RAG content.
- Supporting Learning Feedback Loops benefiting from interactions with trained specialists in different disciplines (quantum physics, quantum engineering, quantum information specialists) and implementing reinforcement learning techniques on a broad scale in this scientific domain.
- Adding the ability to interpret scientific papers, particularly with their often very cryptic charts, and to create useful non-artistic charts with useful and correct captions. Also, figuring out how to create structured data out of unstructured data will probably improve as well, for users willing to consolidate disparate data, like on various technical figures of merits²⁶⁴.
- The creation of **quantum domain specific LLMs** chatbots and systems and their related end-user tools, preferably built on open source tools and with some open access.

All this will depend on algorithmic advances coupled with progress with classical hardware. Indeed, training and running even larger large language models will require more distributed computing power and memory as well as faster data access. It seems that despite the difficulties encountered with increasing chips transistor density, the industry still has room for progress in these domains.

Ultimately, LLMs stand not as replacements but as powerful adjuncts to human expertise, propelling the field of quantum science towards new horizons of discovery and understanding.

Quantum scientists could be skeptical at first glance about the potential impact of LLMs in quantum science. After using it, they can start to be really amazed and ready to be even more surprised by future advances coming from LLMs and their related software ecosystem. The LLM story in quantum science is certainly just an open book remaining to be written. You must now wonder about what you can or will achieve thanks with having an easy access to the whole scientific world knowledge at your fingertips.

Your experience, comments, contributions, and proposals are welcomed here!

(cc) Olivier Ezratty, written between November 2023 and February 2024. February 12, 2024.

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VII. SUPPLEMENTAL MATERIALS

This long supplemental material contains the prompt and dialogs with various LLMs chatbots in relation to quantum science and technologies. It provides the reader with real-world tests done between November 2023 and January 2024. It can help discover how these chatbots fare when responding to a broad set of questions. Responses are never larger than two to three pages long, but prompt engineering and follow-up questions can help the user obtain more sophisticated responses and even create longer documents.

ChatGPT 4.0 on superposition

December 26th, 2023.

Tested by Olivier Ezratty.

In quantum physics, can we really say that a quantum object can be simultaneously at two different locations?

ChatGPT

In quantum physics, saying that a quantum object can be simultaneously at two different locations is a way to describe the phenomenon of quantum superposition. Quantum superposition is a fundamental principle of quantum mechanics where, under certain conditions, particles such as electrons or photons can exist in multiple states or locations simultaneously. This concept is counterintuitive and differs significantly from our everyday experience of objects. However, it's important to understand that this description is part of a broader attempt to interpret complex quantum behavior within the limits of classical language and concepts. The reality of quantum superposition is more accurately captured by mathematical formulations in quantum theory.

Can we say that "being here and there simultaneously" is a misrepresentation of physical reality?

ChatGPT

Describing a quantum object as "being here and there simultaneously" is indeed an oversimplification and can be considered a misrepresentation of physical reality. Quantum superposition, where particles can exist in multiple states or locations, is more accurately described using the mathematical framework of quantum mechanics. This framework provides probabilities for where a particle might be found when measured, rather than definitive locations. The concept challenges our classical understanding of position and reality, making direct analogies to everyday experiences difficult and sometimes misleading.

Are there quantum physics ontologies that explain this well?

Quantum physics ontologies that explain phenomena like superposition emphasize the fundamental difference between classical and quantum mechanics. In classical mechanics, properties such as position or momentum are always well-defined. However, quantum mechanics operates on the principle of superposition, where a quantum system can be in a combination of multiple states simultaneously. This doesn't mean an object is in two places at once in a classical sense. Instead, it suggests that until a measurement is made, quantum systems can exist in multiple potential states, which are described by probabilities. The actual state of the system is only determined when a measurement occurs. This concept is crucial in technologies like quantum computing, where qubits exist in superpositions of '0' and '1' states.

ChatGPT 4.0 on entanglement

December 21th and 30th, 2023 Tested by Olivier Ezratty.

Can you explain me the origin of the entanglement phenomenon in quantum physics, and how it works before and after measurement?

ChatGPT

The phenomenon of entanglement in quantum physics is one of the most fascinating and fundamental aspects of the theory. It arises from the principles of quantum mechanics and has profound implications for our understanding of the nature of reality.

Origin of Entanglement:

- 1. **Quantum Mechanics Basis**: Entanglement is rooted in the principles of quantum mechanics, which describes the physical properties of nature at the scale of atoms and subatomic particles. In quantum mechanics, the properties of particles are not definite until they are measured.
- 2. **Superposition Principle**: A key feature of quantum mechanics is the superposition principle, where particles can exist in multiple states simultaneously until they are observed.
- 3. **Development by Einstein, Podolsky, and Rosen (EPR)**: The concept of entanglement was brought to prominence in a 1935 paper by Albert Einstein, Boris Podolsky, and

ChatGPT

Nathan Rosen. They presented a thought experiment (EPR paradox) to highlight what they saw as the incompleteness of quantum mechanics.

4. Schrödinger's Contribution: Erwin Schrödinger, in response to the EPR paper, coined the term "entanglement" and discussed its significance. He recognized it as a fundamental feature of quantum mechanics.

How Entanglement Works:

- 1. Before Measurement:
- Entangled State Creation: Entanglement occurs when two or more particles become linked, and the state of one cannot be adequately described without full knowledge of the other(s). This can happen, for example, when two particles are created at the same point and instant in space, such as through a decay process.
- **Correlated Properties**: Entangled particles have correlated properties. For instance, if two particles are entangled in such a way that their spins are correlated, knowing the spin of one particle immediately tells you the spin of the other, regardless of the distance between them.
- No Signal Transfer: It's important to note that entanglement does not involve the transfer of information between particles. The correlations are realized upon measurement but are not a result of a classical signal being sent between the particles.
- 2. After Measurement:
- Wave Function Collapse: When one of the entangled particles is measured, its wave function collapses into a definite state. Due to the entanglement, the wave function of the other particle also collapses, instantaneously determining its state.
- Non-locality: This phenomenon, called non-locality, suggests that the particles "communicate" with each other instantaneously, defying classical ideas of space and time. This was a point of contention for Einstein, who famously referred to it as "spooky action at a distance."
- 3. No FTL Communication: Despite the instantaneous nature of this correlation, entanglement cannot be used for faster-than-light communication. This is because the outcome of a measurement on one particle is random, and the information about the state of the other particle cannot be known in advance.

Bell's Theorem:

- 1. **Empirical Test**: John S. Bell in 1964 provided a theoretical foundation (Bell's theorem) to test the predictions of quantum mechanics against those of classical physics regarding entanglement. Bell's inequalities, if violated, would confirm the quantum mechanical predictions.
- 2. **Experimental Confirmation**: Numerous experiments have since confirmed the predictions of quantum mechanics and the existence of entanglement, solidifying its status as a real phenomenon.

Entanglement challenges our classical intuitions about the world and has become a key resource in developing new technologies like quantum computing and quantum cryptography. It remains an active area of research in quantum physics, continually deepening our understanding of the quantum world.

Can you explain the difference between John Clauser and Alain Aspect experiments on entanglement and the invalidation of Bell inequalities?

ChatGPT

John Clauser's and Alain Aspect's experiments were pivotal in advancing our understanding of quantum entanglement and the violation of Bell inequalities.

John Clauser's work, along with his collaborators, in the early 1970s was among the first experimental tests of the Bell inequalities. The Clauser-Freedman experiment in 1972 was groundbreaking. It involved the decay of excited calcium atoms to produce two photons of light that, due to conservation of angular momentum, were entangled with opposite polarizations. This experiment, conducted when Clauser was a postdoctoral researcher at UC Berkeley, tested the CHSH inequality (Clauser-Horne-Shimony-Holt), a version of Bell's inequality. The results supported quantum mechanics and indicated that entanglement is a real phenomenon, contradicting Einstein's views of locality and causality, which suggested that "hidden variables" might explain quantum mechanics without entanglement.

Alain Aspect's experiments in the 1980s took this research further. He conducted experiments that also tested Bell's inequalities but with significant improvements in closing the loopholes that had been present in earlier experiments. Aspect's experiments were more conclusive in demonstrating that quantum entanglement could not be explained by local hidden variables. His work involved using pairs of entangled photons and varying the measurement settings while the photons were in flight, making it virtually impossible for any local hidden variable theory to explain the observed correlations.

Both Clauser's and Aspect's experiments were instrumental in establishing the non-classical nature of quantum entanglement and demonstrating that quantum mechanics, with its inherent 'spooky action at a distance,' is an accurate description of reality. These experiments paved the way for the development of quantum information science and technologies, such as quantum computing and cryptography. Their contributions were recognized with the Nobel Prize in Physics in 2022, which they shared with Anton Zeilinger, another pioneer in quantum entanglement research.

These experiments not only validated quantum theory but also challenged our understanding of the fundamental nature of reality, sparking a transformation in both physics and technology.

Explain quantum entanglement by comparing it to a social network and then to coordinated flying birds?

ChatGPT

Quantum entanglement is one of the most fascinating and complex phenomena in quantum physics. To make it more understandable, let's compare it to two relatable concepts: a social network and coordinated flying birds.

Quantum Entanglement and a Social Network

Imagine a social network where two individuals are incredibly in sync. They've shared experiences and information to such an extent that, even when separated by large distances, the action or decision of one individual has an instantaneous effect on the other. This is akin to quantum entanglement, where two particles become so deeply linked that the state of one particle is instantly reflected in the state of the other, no matter how far apart they are.

In this social network analogy:

- 1. **Instant Communication:** Just like how entangled particles seem to "communicate" their states to each other instantly, the individuals in our social network analogy share information instantly.
- 2. **Correlated Behavior:** The actions of one person in the network immediately influence the actions or state of the other, similar to how the measurement of one entangled particle instantly determines the state of its counterpart.
- 3. **Distance Independence:** The physical distance between the individuals doesn't hinder their connection, mirroring the way entangled particles remain connected regardless of the distance separating them.

Quantum Entanglement and Coordinated Flying Birds

Consider a flock of birds flying in perfect coordination, like starlings during a murmuration. This coordination can serve as a metaphor for quantum entanglement.

- 4. **Synchronized Movement:** Each bird in the flock seems to know exactly how to move in concert with its neighbors. In quantum entanglement, once entangled, the particles behave like parts of a unified system, with their states being interdependent.
- 5. No Visible Communication: In a murmuration, it appears as if the birds are not communicating, yet they move in a highly coordinated manner. Similarly, entangled particles

do not send signals to each other in any traditional sense, but their states are nonetheless correlated.

6. **Complexity and Beauty:** Just as the complex and beautiful patterns of bird murmurations are a result of simple rules of coordination, the complex phenomenon of entanglement arises from fundamental principles of quantum mechanics.

Both analogies help illustrate the concept of entanglement: a mysterious, immediate connection between entities that transcends conventional limitations of communication and distance. However, it's important to remember that these are just analogies. In reality, quantum entanglement involves the complex interplay of quantum states, probabilities, and the peculiarities of the quantum world that don't have direct equivalents in classical systems.

ChatGPT 4.0 on quantum algorithms

December 21th, 2023 Tested by Olivier Ezratty.

Can you explain the difference between VQE and QPE quantum algorithms to find the ground state Hamiltonian of some small molecule?

ChatGPT

Absolutely, the Variational Quantum Eigensolver (VQE) and Quantum Phase Estimation (QPE) are two different quantum algorithms used for determining the ground state energy (or more generally, eigenvalues) of a Hamiltonian, often in the context of molecular systems in quantum chemistry. Let's break down the differences between these two approaches:

Quantum Phase Estimation (QPE)

- 1. **Principle**: QPE is designed to estimate the phase (or eigenvalue) corresponding to a given eigenvector of a unitary operator. In the context of finding the ground state of a Hamiltonian, the unitary operator is typically e^{iHt} , where *H* is the Hamiltonian of the system and *t* is a chosen time parameter.
- 2. **Procedure**: The algorithm involves preparing a quantum state close to the desired eigenvector, applying the unitary operator e^{iHt} multiple times, and then performing a quantum Fourier transform followed by measurements. The outcomes of these measurements are used to estimate the eigenvalue corresponding to the prepared eigenvector.
- 3. Accuracy and Resource Requirements: QPE can provide very accurate results, but it requires a significant number of qubits and quantum gates. Moreover, it requires precise control over the quantum system, making it challenging with current quantum technology, especially for systems like molecules with complex Hamiltonians.

4. Use Cases: Best suited for scenarios where high precision is required and quantum resources are not a major constraint.

Variational Quantum Eigensolver (VQE)

- 1. **Principle**: VQE is a hybrid quantum-classical algorithm that uses a variational approach to find the minimum eigenvalue of a Hamiltonian, which corresponds to the ground state energy of the system.
- 2. **Procedure:** The algorithm involves preparing a parameterized quantum state (ansatz), measuring the expectation value of the Hamiltonian for this state, and then using a classical optimizer to adjust the parameters to minimize this expectation value.
- 3. **Resource Efficiency**: VQE is more suitable for noisy intermediate-scale quantum (NISQ) computers. It requires fewer qubits and less quantum coherence time compared to QPE, making it more feasible with current quantum technology.
- 4. **Hybrid Nature**: VQE's strength lies in its hybrid approach, leveraging both quantum and classical resources. The quantum computer prepares states and measures energies, while the classical computer optimizes parameters.
- 5. Use Cases: Ideal for near-term quantum computers where quantum resources are limited. It is widely used in quantum chemistry for calculating molecular ground states.

Comparison and Context for Small Molecules

- **Precision vs. Feasibility**: QPE can potentially provide more precise results but at the cost of high quantum resource requirements, which might not be feasible for complex systems like molecules. VQE, on the other hand, trades off some precision for feasibility on near-term quantum devices.
- Quantum Resource Requirements: For small molecules where the Hamiltonian is not exceedingly complex, VQE is often more practical due to its lower quantum resource requirements.
- State of Quantum Hardware: Given the current state of quantum hardware, VQE is more commonly used for practical applications, as it aligns better with the capabilities of available quantum computers.

In summary, while QPE offers high precision, its requirement for a large number of quantum resources makes it less suitable for near-term applications, especially for complex systems like small molecules. VQE, with its hybrid approach and efficiency, is more adaptable to current quantum technology, making it a popular choice for studying molecular ground states.

What is the largest molecule that was successfully computed with a VQE algorithm?

ChatGPT

The largest molecule successfully computed using the Variational Quantum Eigensolver (VQE) algorithm, as per the information available in 2023, was part of a study conducted by Google. This study involved simulating the

Hartree-Fock wave function for hydrogen chains using up to 12 qubits and 72 entangling gates on a superconducting quantum computer. However, this simulation faced significant hardware noise, requiring specialized error mitigation techniques. In a more recent study, Google also simulated a cyclobutene ring using up to 10 qubits with pair-correlated wave functions.

Additionally, another significant development in the use of VQE was demonstrated on IonQ's trapped-ion quantum computers. In this case, VQE simulations were performed for molecules like LiH, H2O, and Li2O. This research utilized an approximate ansatz derived from the unitary Coupled Cluster double (upCCD) ansatz, which was specifically adapted for the capabilities of Noisy Intermediate-Scale Quantum (NISQ) computers.

These studies represent major steps in the application of quantum computing to the field of quantum chemistry, demonstrating the potential of VQE algorithms in simulating molecular systems, despite the limitations of current quantum hardware.

What is the difference between QAOA algorithms and QUBO formulation based algorithms to solve optimization problems?

ChatGPT

The Quantum Approximate Optimization Algorithm (QAOA) and Quadratic Unconstrained Binary Optimization (QUBO) formulation are both approaches used in quantum computing to solve optimization problems, but they operate quite differently.

QAOA is a quantum algorithm designed for gate-model quantum computers. It's a variational algorithm that seeks to find solutions to optimization problems by preparing a quantum state whose measurement will likely result in a good approximation to the problem's solution. QAOA works by applying a series of quantum gates (unitary transformations) that depend on parameters optimized through a classical optimization loop. The process involves alternating between two Hamiltonians: a problem Hamiltonian encoding the solution to the optimization problem and a mixer Hamiltonian that helps in exploring the solution space. The algorithm's performance depends on the depth of the circuit (number of gate layers), and finding the right parameters is crucial for its success.

On the other hand, QUBO is a formulation technique used to represent optimization problems, particularly for execution on quantum annealers, but it can also be used for gate-model quantum computers. In the QUBO approach, problems are expressed as quadratic polynomials of binary variables. The goal is to minimize this polynomial. This formulation is often used in quantum annealing, where the system evolves to find the lowest energy state of a Hamiltonian that represents the problem. In QUBO, the main challenge lies in the transformation of the original problem into a suitable quadratic form, which can sometimes involve approximations and simplifications. This form is then used by the quantum hardware to find an optimal or nearoptimal solution.

In summary, while QAOA is a quantum algorithm that operates in the gate-model paradigm with a focus on variational techniques, QUBO is more of a problemformulation strategy, often used in quantum annealing, to translate complex optimization problems into a form suitable for quantum computation. The choice between these approaches depends on the nature of the problem and the type of quantum hardware available.

ChatGPT 4.0 and Bard on MBQC and FBQC

December 21th, 2023

Tested by Olivier Ezratty.

Can you explain the difference between MBQC and FBQC in quantum computing?

ChatGPT

In quantum computing, Measurement-Based Quantum Computing (MBQC) and Fault-Tolerant Quantum Computing (FTQC) are two different approaches to performing quantum computations. Each has its unique characteristics and methods for processing quantum information.

Measurement-Based Quantum Computing (MBQC)

- 1. Fundamental Concept:
 - MBQC, also known as the one-way quantum computer, is based on a highly entangled state known as a cluster state or a graph state.
 - The computation is driven by a sequence of measurements on individual qubits within this entangled state.
- 2. Process:
 - Initially, a large number of qubits are entangled to form the cluster state.
 - Quantum computation is then performed by sequentially measuring qubits in the cluster.

- The choice of measurement bases and the order of measurements determine the computation being performed.
- The outcome of one measurement can determine the basis of subsequent measurements, making the process adaptive.
- 3. Advantages:
 - Measurements are simpler operations than the general quantum gates, potentially reducing the complexity of quantum operations.
 - Since the entangled state is prepared in advance, the computation is mostly about conducting measurements, which could be simpler from a practical perspective.
- 4. Challenges:
 - Requires the creation of a large, highly entangled state, which is sensitive to decoherence and errors.
 - Error correction can be more challenging due to the destructive nature of measurements.

Fault-Tolerant Quantum Computing (FTQC)

- 1. Fundamental Concept:
 - FTQC is designed to perform reliable quantum computations even when individual quantum gates and qubits are prone to errors.
 - It involves implementing quantum error correction codes and fault-tolerant procedures to protect against both bit-flip and phase-flip errors.
- 2. Process:
 - Quantum information is encoded in logical qubits, each comprising multiple physical qubits.
 - Quantum gates are applied in a way that even if errors occur, they can be detected and corrected without destroying the quantum information.
 - The fault-tolerant architecture ensures that errors do not propagate uncontrollably and can be corrected systematically.
- 3. Advantages:
 - Enables long quantum computations by actively correcting errors.
 - Essential for building scalable quantum computers where quantum information needs to be preserved over long time scales.
- 4. Challenges:
 - Requires a large overhead of physical qubits for encoding a single logical qubit, leading to increased resource demands.
 - Complex gate operations are needed for error correction and fault-tolerant gate implementation.

Key Differences

- Nature of Computation: MBQC is driven by measurements on a pre-prepared entangled state, whereas FTQC relies on error correction and fault-tolerant gate operations throughout the computation process.
- **Resource Requirements**: MBQC requires the generation of large entangled states initially, while FTQC requires a significant overhead of physical qubits for error correction.
- Error Handling: MBQC has to deal with errors mainly in the initial state preparation and during measurements, whereas FTQC continuously corrects errors during the computation.
- **Practical Implementation**: MBQC's reliance on measurements might offer simpler experimental setups in some cases, but creating large entangled states is challenging. FTQC's demand for numerous qubits and complex operations presents a significant challenge for current technology.

Both MBQC and FTQC represent different strategies to harness the principles of quantum mechanics for computing. MBQC's approach is unique in its use of entanglement and measurements, while FTQC is crucial for the long-term viability of quantum computers by addressing the challenge of quantum error correction.

I didn't mean FTQC but FBQC which is promoted for example by PsiQuantum

ChatGPT

Thank you for the clarification. Let's discuss the differences between Measurement-Based Quantum Computing (MBQC) and Fiber-Based Quantum Computing (FBQC), the latter being a concept promoted by companies like PsiQuantum.

Measurement-Based Quantum Computing (MBQC)

- 1. **Concept**: MBQC, also known as one-way quantum computing, relies on a highly entangled initial state, typically a cluster state or a graph state, on which computations are performed through a series of measurements.
- 2. Operation:
 - The process starts with the preparation of a large entangled state.
 - Computation proceeds via sequential measurements on individual qubits within this state.
 - The type and order of measurements dictate the computation being performed.
 - Adaptive: The outcome of one measurement can affect the basis of subsequent measurements.
- 3. Advantages:
 - Potentially simpler operations, as it primarily involves measurements.

- Pre-prepared entangled state allows for a certain level of computation parallelism.
- 4. Challenges:
 - Creating and maintaining a large-scale entangled state is technologically challenging.
 - Sensitive to errors due to the destructive nature of measurements.

Fiber-Based Quantum Computing (FBQC)

- 1. **Concept**: FBQC is a model for quantum computing where information is transmitted and processed using photons through optical fibers, with an emphasis on scalability and practicality for building large-scale quantum computers.
- 2. Operation:
 - Utilizes photonic qubits for quantum information processing.
 - Photons are manipulated, entangled, and measured using optical devices (like beam splitters, phase shifters, and detectors).
 - Quantum information is often encoded in properties of photons like polarization or phase.
- 3. Advantages:
 - Naturally suited for long-distance quantum communication and integration into existing fiber-optic networks.
 - Photons have low interaction with the environment, leading to less decoherence.
 - Scalable: FBQC aims to leverage existing telecommunications technology for scalable quantum computing.
- 4. Challenges:
 - Generating and detecting single photons reliably is a major technical hurdle.
 - Performing two-qubit gates, essential for universal quantum computing, is more challenging with photons.
 - Requires precise synchronization and stability in the optical system.

Key Differences

- **Physical Implementation**: MBQC is typically implemented in systems like trapped ions or superconducting qubits, while FBQC relies on photonic qubits and optical fiber technology.
- Error Sensitivity: MBQC has to deal with errors related to state preparation and measurements, whereas FBQC faces challenges in photon generation, manipulation, and detection.
- Scalability Aspects: FBQC's use of fiber optics and photons is seen as a scalable approach, potentially compatible with existing telecommunications

infrastructure. MBQC's scalability is more dependent on the ability to create and manage large entangled states.

• **Technological Focus**: FBQC requires advancements in photonics and quantum optics, whereas MBQC requires advancements in creating and managing entangled quantum states.

FBQC, especially as proposed by companies like PsiQuantum, represents an ambitious approach to leveraging the unique properties of photons and fiber optics for quantum computing, aiming for practical scalability and integration with existing technology. MBQC, on the other hand, offers a distinct computational model based on the unique properties of entanglement and measurement-driven processes.

It is not Fiber-Based Quantum Computing but Fusion Based Quantum Computing...

ChatGPT

Thank you for the correction. Let's explore the differences between Measurement-Based Quantum Computing (MBQC) and Fusion-Based Quantum Computing (FBQC):

Measurement-Based Quantum Computing (MBQC)

- 1. **Concept**: MBQC, also known as the one-way quantum computer, utilizes a highly entangled initial state, typically a cluster state or a graph state, on which computations are performed through sequential measurements.
- 2. Operation:
 - Begins with the generation of a large entangled state.
 - Quantum computation proceeds by measuring qubits in this state, with the sequence and basis of measurements determining the computational process.
 - Adaptive: The outcome of one measurement can influence the choice of subsequent measurements.
- 3. Advantages:
 - Simplifies operations to primarily measurements.
 - Enables certain level of computation parallelism due to pre-prepared entangled state.
- 4. Challenges:
 - Requires generation and maintenance of large-scale entangled states.
 - Sensitive to errors and decoherence.

Fusion-Based Quantum Computing (FBQC)

1. **Concept**: FBQC is a model of quantum computing that focuses on building large-scale quantum states, such as cluster states, through a process known as

fusion. This approach is particularly relevant in photonic quantum computing.

- 2. Operation:
 - Involves generating smaller, easily manageable entangled states or photonic qubits.
 - These states are then "fused" together to form larger entangled states necessary for computation.
 - Fusion operations are probabilistic and require repeat-until-success strategies.
 - Once a large entangled state is prepared, MBQC techniques can be applied for computation.
- 3. Advantages:
 - Offers a pathway to scale up photonic quantum computing by incrementally building larger quantum states.
 - Fusion operations are generally simpler than directly entangling many qubits at once.
 - Photonic approach allows for roomtemperature operations and integration with existing fiber-optic technologies.
- 4. Challenges:
 - Fusion operations are probabilistic, which can lead to inefficiencies in state preparation.
 - Requires highly efficient single-photon sources and detectors.
 - Managing and correcting errors in large entangled states remain a significant challenge.

Key Differences

- State Preparation: While MBQC begins with a preprepared large entangled state, FBQC incrementally builds up these states through fusion processes.
- Scalability: FBQC's fusion process is particularly aimed at scaling up quantum computing, especially in photonic systems, by overcoming the direct entanglement challenges of large numbers of qubits.
- **Technological Focus**: FBQC requires advancements in photonics, particularly in efficient and reliable single-photon generation and detection, as well as in performing fusion operations.
- Error Correction: Both approaches face challenges in error correction, but FBQC has the added complexity of dealing with errors in the fusion process and the subsequent MBQC.

FBQC, especially in photonic systems, represents a promising approach to constructing large-scale quantum computers by overcoming some of the direct entanglement and scalability challenges. It complements MBQC by providing a feasible method for preparing the necessary large entangled states that MBQC requires for computation.

With Google Bard...

Can you explain the difference between MBQC and FBQC in quantum computing?

Sure, here is an explanation of the difference between MBQC and FBQC in quantum computing:

MBQC (Measurement-Based Quantum Computation) and FBQC (Fusion-Based Quantum Computing) are two different approaches to quantum computation that both aim to achieve quantum computational advantage. Quantum computational advantage refers to the ability to perform a computation on a quantum computer that would be intractable for a classical computer.

MBQC is based on the idea of performing quantum computations using only measurements on a highly entangled quantum state. This entangled state is typically prepared using

a quantum circuit, but it can also be prepared using other methods, such as using a quantum simulator or by coupling together a large number of physical qubits. Once the entangled state is prepared, the computations are performed by measuring different parts of the state. The results of these measurements are then fed into classical computers to perform the final computation.

FBQC is based on the idea of performing quantum computations using a series of fusion operations. These operations are performed on pairs of entangled qubits, and they combine the information encoded in the qubits into a single qubit. The computations are performed by applying a sequence of fusion operations to an initially entangled state. The final state of the system can then be used to perform a classical computation.

Here is a table summarizing the key differences between MBQC and FBQC:

Feature	MBQC	FBQC
Computational model	Measurement-based	Fusion-based
Input data	Highly entangled quantum state	Initial entangled state
Computational steps	Measurements on entangled qubits	Sequence of fusion operations
Output data	Results of measurements	Final state of system
Classical post-processing	Required	Not required

In general, MBQC is considered to be more scalable than FBQC. This is because the number of measurements required to perform a computation in MBQC grows linearly with the size of the computation, while the number of fusion operations required in FBQC grows exponentially with the size of the computation. However, MBQC is also more difficult to implement than FBQC. This is because the preparation of highly entangled quantum states is a challenging task.

Both MBQC and FBQC are still active areas of **research**. There are many open questions about the scalability and practicality of these approaches, and there is ongoing work to develop more efficient algorithms for these models.

I hope this explanation is helpful. Please let me know if you have any other questions.

ChatGPT 4.0 on quantum speedups

December 29th, 2023

Tested by Olivier Ezratty.

Can a quantum algorithm deterministically generate a result which is single bit string in the computational basis (*aka* a quantum state of the qubit register without any superposition left) with bringing an exponential speedup?

ChatGPT

Achieving an exponential speedup with a quantum algorithm that deterministically generates a result as a single bit string in the computational basis (or a quantum state of the qubit register with no superposition left) is generally challenging and uncommon. The main reasons are:

- 1. **Nature of Quantum Computation:** Quantum algorithms gain their most significant advantages through superposition and entanglement, allowing them to explore many possible solutions simultaneously. This characteristic is key to the exponential speedup observed in some quantum algorithms.
- 2. **Deterministic Output with No Superposition:** If a quantum algorithm is designed to produce a deterministic output with no superposition, it essentially behaves similarly to a classical algorithm in terms of its output state. This specific requirement often does not leverage the full potential of quantum mechanics (superposition and entanglement) that gives quantum algorithms their computational advantage.
- 3. Exponential Speedup Scenarios: Quantum algorithms known for their exponential speedup, like Shor's algorithm for factorization and Grover's algorithm for database search, inherently rely on superposition and entanglement. In these algorithms, the exponential speedup is achieved by manipulating and then measuring a superposition of states.
- 4. Role of Measurement: Upon measurement, quantum states do collapse to a deterministic state (a single bit string in the computational basis), but this is typically the end result of a process that involved significant quantum mechanical computation. The measurement process itself

does not contribute to computational speedup; it's the quantum operations prior to measurement that enable the speedup.

Therefore, if a quantum algorithm is strictly producing a deterministic, non-superposed state as its computation (i.e., it operates without utilizing superposition or entanglement), it is unlikely to achieve an exponential speedup over classical algorithms. The significant speed advantages of quantum computing are most apparent in problems that inherently require and exploit the unique properties of quantum mechanics, which include handling superpositions and entanglements.

ChatGPT 4.0 and Google Bard on bosonic qubits

December 28th, 2023 Tested by Olivier Ezratty.

I am preparing a review of the various types of bosonic qubits that are being developed like cat-qubits, GKP qubits and dual rail qubits. Could you first provide me with some insights on how they operate and how to differentiate them. And then, output a bibliography of the key scientific sources available on these various types of qubits.

ChatGPT

Bosonic qubits represent an exciting and rapidly developing field in quantum computing. Unlike the standard qubits that typically use individual particles like electrons or photons to store quantum information, bosonic qubits use states of bosonic particles, which can be photons in different modes of electromagnetic fields. Here's a brief overview of the types you mentioned:

- 1. **Cat-Qubits**: Named after Schrödinger's cat thought experiment, cat-qubits are superpositions of coherent states of a harmonic oscillator. They are called "cat" states because they exist in a superposition of two macroscopically distinct states, much like Schrödinger's cat is both alive and dead. Cat-qubits are particularly interesting for quantum error correction due to their ability to correct against certain types of errors better than traditional qubits.
- 2. **GKP (Gottesman-Kitaev-Preskill) Qubits**: These qubits are named after the theorists who proposed them. GKP qubits are designed to encode a qubit into an oscillator's position and momentum. They use a lattice structure in phase space, allowing for error correction by measuring the displacement of the qubit within this lattice. GKP qubits are known for their potential resilience to noise and errors.
- Dual Rail Qubits: These qubits use two modes (rails) of a quantum system, typically two distinct paths of a photon. A single photon is prepared in a superposition of being in either of the two paths, representing the quantum states |0>

and $|1\rangle$. Dual rail qubits are a form of photonic qubit and are useful in quantum computing for their relative simplicity and scalability.

Each of these qubits operates differently and has its own advantages and challenges in terms of implementation, error correction, and integration into larger quantum systems.

For a detailed understanding and the latest advancements in these types of qubits, it's essential to consult scientific literature. I will compile a bibliography of key scientific sources on these qubits. Please give me a moment to gather the most relevant and up-to-date references.

Based on recent research, here's a more detailed insight into the operation and differentiation of cat-qubits, along with a bibliography of key scientific sources:

Cat-Qubits

Cat-qubits are based on superpositions of coherent states in a harmonic oscillator, resembling Schrödinger's cat thought experiment. They exhibit strong protection against common physical error mechanisms such as photon loss. A variety of approaches are being explored to enhance their performance:

- 1. **Dissipative Stabilization**: Innovative gate designs have been developed for reducing phase-flip errors in cat-qubits, with techniques such as feedback mechanisms and altering the Zeno gate drive.
- 2. **Repetition Code and Bias-Preserving Gates**: Research has focused on high-performance repetition cat codes and engineering fast bias-preserving gates for stabilized cat-qubits.
- 3. **Compression for Longevity**: Techniques like phasespace compression are being used to extend the qubit lifetime of cat states.

Bibliography on Cat-Qubits:

- 1. "Quantum computation with cat qubits" Guillaud et al. (arXiv, 2023).
- 2. "Designing High-Fidelity Zeno Gates for Dissipative Cat Qubits" PRX Quantum (2023).
- 3. "High-performance repetition cat code using fast noisy operations" Quantum Journal (2023).
- 4. "Quantum error correction with dissipatively stabilized squeezed cat qubits" Physical Review A (2023).
- 5. "Control of the ZZ coupling between Kerr-cat qubits via transmon couplers" (arXiv, 2023).
- "Concept Cats: Designing Better Qubits" Alice & Bob (2023).

These papers provide a comprehensive understanding of the latest advancements in cat-qubits, focusing on error correction, gate design, and qubit longevity. They How about a bibliography on QKP and dual-rail qubits, particularly when implemented with variations of superconducting qubits?

ChatGPT

For GKP (Gottesman-Kitaev-Preskill) qubits and dual-rail qubits, particularly in the context of superconducting qubits, here's a bibliography based on recent research:

GKP Qubits

GKP qubits are a type of bosonic qubit that encode quantum information in the phase space of an oscillator. They are designed to counteract the limitations imposed by the Heisenberg uncertainty principle through a clever approach of measuring position and momentum operators precisely but only modulo a certain spacing. This design leads to the formation of a lattice in the phase space, which is key to their error correction capabilities.

- "All-photonic Gottesman-Kitaev-Preskill--qubit repeater using analog-information-assisted multiplexed entanglement ranking" - This paper explores the use of GKP qubits in quantum repeaters, focusing on achievable distances and error rates (Phys. Rev. Research 5, 043056, 2023).
- 2. **"Error-corrected quantum repeaters with GKP qudits"** - This study investigates different quantum repeater protocols using GKP qudits, with a focus on secret-key rates and the influence of experimental parameters (arXiv, 2023).
- 3. "High-threshold fault-tolerant quantum computation with the Gottesman-Kitaev-Preskill qubit under noise in an optical setup" This research addresses the feasibility of fault-tolerant quantum computing with GKP qubits in optical setups (Phys. Rev. A 107, 052414, 2023).
- "A GKP qubit protected by dissipation in a highimpedance superconducting circuit driven by a microwave frequency comb" - This paper presents a novel approach for generating, protecting, and controlling GKP qubits in superconducting circuits (arXiv, 2023).
- "Quantum error mitigation for GKP qubits" This work extends quantum error mitigation techniques to GKP qubits, important for intermediate-scale quantum devices (APS March Meeting 2023).
- "Low-overhead quantum computing with Gottesman-Kitaev-Preskill qubits" - This article discusses the implementation of GKP qubits for lowoverhead quantum computing (AWS Quantum Technologies Blog, 2023).

Dual-Rail Qubits

Dual-rail qubits, often implemented using photons in superconducting circuits, use two distinct paths of a photon to represent the quantum states |0⟩ and |1⟩. Their simplicity and scalability make them an attractive option for quantum computing. However, recent literature specifically focusing on dual-rail qubits in superconducting qubits from 2023 seems limited. For the most current and detailed information, direct searches in scientific databases and journals are recommended.

This bibliography provides a snapshot of the state-ofthe-art in research on GKP qubits, highlighting their capabilities in quantum error correction, quantum communication, and fault-tolerant quantum computing.

I there a particular reason why you don't provide url links with the items in the bibliography you generated?

ChatGPT

When compiling bibliographies from the research papers and articles I find, I summarize the key information and provide detailed references, including titles and authors, which you can use to locate these sources. However, I don't directly provide URL links due to the current operational design of this platform.

The absence of direct links is mainly for two reasons:

- 1. **Source Accessibility**: Direct URLs are often subject to access restrictions or paywalls, making them less universally useful for all users.
- 2. **Dynamic Nature of URLs**: URLs can change over time due to website updates or restructuring. By providing detailed references instead of direct URLs, the information remains valuable even if the direct link changes.

You can use the provided references to search for these papers and articles in scientific databases, academic journals, or through general web search engines. Most academic papers can be found on platforms like Google Scholar, arXiv, or directly on the journal's website. If you have access to a university library or a research institution, they often provide access to a broader range of scientific publications that might be behind paywalls.

For example, to find the paper titled "All-photonic Gottesman-Kitaev-Preskill--qubit repeater using analoginformation-assisted multiplexed entanglement ranking," you can search for this title in a research database or a search engine. The details like volume and issue number (Phys. Rev. Research 5, 043056, 2023) can help narrow down the search to the specific document. Improve this prompt : "I am preparing a review of the various types of bosonic qubits that are being developed like catqubits, GKP qubits and dual rail qubits. Could you first provide me with some insights on how they operate and how to differentiate them. And then, output a bibliography of the key scientific sources available on these various types of qubits".

ChatGPT

Certainly, let's enhance your prompt to delve deeper into the operation principles and distinctions of catqubits, GKP qubits, and dual rail qubits in the realm of quantum computing, followed by a detailed bibliography on these topics.

"I am currently undertaking a comprehensive review of advanced bosonic qubits, focusing on cat-qubits, GKP (Gottesman-Kitaev-Preskill) gubits, and dual rail gubits. These qubits represent cutting-edge developments in quantum computing and exhibit unique characteristics in their operation and error correction capabilities. I am seeking a detailed explanation of the underlying principles of each qubit type. For cat-qubits, I am interested in their reliance on superpositions of macroscopically distinct states and how this makes them suitable for quantum error correction, despite their vulnerability to specific types of noise. Regarding GKP gubits, I wish to understand their implementation in phase space of a quantum harmonic oscillator and their intrinsic error-correcting abilities, particularly how they manage to correct small shifts in position and momentum. As for dual rail gubits, I am curious about their operational simplicity using two distinct modes of a quantum system and the challenges they face in terms of precision control and susceptibility to loss and decoherence. Alongside these insights, I require an extensive bibliography of key scientific sources, including both seminal papers and the latest research, that provide a deeper understanding of these bosonic qubits. This bibliography should cover theoretical foundations, experimental advancements, and potential applications of these qubit types in quantum computing."

This expanded prompt should give a comprehensive overview of how these bosonic qubits operate, their unique features, and challenges, and will be supplemented with a well-curated list of academic sources for an in-depth study.

(using the created prompt)

ChatGPT

I did a quick search and here's what I discovered.

In your review of advanced bosonic qubits, including cat-qubits, GKP qubits, and dual rail qubits, you will find a wealth of information in the following sources:

1. Cat Qubits:

- A comprehensive discussion on cat qubits, particularly focusing on their stabilization and operation within quantum computation, can be found in the work by Fowler et al. (2012) and Guillaud and Mirrahimi (2019). These papers discuss the use of cat qubits in quantum error correction and their potential for fault-tolerant quantum computation.
- For a deeper understanding of the dynamics and detection of Schrödinger cat states, the works of Yurke & Stoler (1988) and Kirchmair et al. (2013) are seminal.

2. GKP (Gottesman-Kitaev-Preskill) Qubits:

- The foundational paper by Gottesman, Kitaev, and Preskill (2001) is a must-read. It provides a detailed theoretical foundation for encoding a qubit in an oscillator, which is the essence of GKP qubits.
- Additionally, exploring the quantum error correction capabilities of GKP qubits, the work by Vuillot et al. (2019) provides significant insights.

3. **Dual Rail Qubits**:

• While specific seminal works on dual rail qubits were not identified in the recent search, the general principles of these qubits can be understood through the study of quantum information and quantum gate operations as they utilize two distinct modes of a quantum system. Resources on quantum gates and quantum information, such as Chow et al. (2012), can be relevant here.

These references provide a mix of theoretical and experimental insights into the operation, error correction, and practical challenges associated with each type of bosonic qubit. The papers span from foundational theories to recent advancements, giving you a broad perspective on the state of research in these areas. For a more specific focus on dual rail qubits, further targeted research might be necessary.

can you output all these references in a BIBTEXT format?

ChatGPT

Certainly! Here are the BIBTEX entries for the references I mentioned:

1. Fowler et al. (2012) - Surface codes for quantum computation

bibtexCopy code

@article{fowler2012surface, title={Surface codes: towards
practical large-scale quantum computation},

author={Fowler, Austin G and Mariantoni, Matteo and Martinis, John M and Cleland, A. N.}, journal={Physical Review A}, volume={86}, number={3}, pages={032324}, year={2012}, publisher={APS} }

2. Guillaud and Mirrahimi (2019) - Repetition Cat Qubits for Fault-Tolerant Quantum

Computation

bibtexCopy code

@article{guillaud2019repetition, title={Repetition Cat Qubits for Fault-Tolerant Quantum Computation}, author={Guillaud, Jérémie and Mirrahimi, Mazyar}, journal={Physical Review X}, volume={9}, number={4}, pages={041053}, year={2019}, publisher={American Physical Society} }

3. Gottesman, Kitaev, and Preskill (2001) -Encoding a qubit in an oscillator

bibtexCopy code

@article{gottesman2001encoding, title={Encoding a qubit in an oscillator}, author={Gottesman, Daniel and Kitaev, Alexei and Preskill, John}, journal={Physical Review A}, volume={64}, number={1}, pages={012310}, year={2001}, publisher={APS} }

4. Vuillot et al. (2019) - Quantum error correction with the toric Gottesman-Kitaev-Preskill code

bibtexCopy code

@article{vuillot2019quantum, title={Quantum error correction with the toric Gottesman-Kitaev-Preskill code}, author={Vuillot, Christophe and Asasi, Hamed and Wang, Yaxing and Pryadko, Leonid P and Terhal, Barbara M}, journal={Physical Review A}, volume={99}, number={3}, pages={032344}, year={2019}, publisher={APS} }

5. Chow et al. (2012) - Universal quantum gate set approaching fault-tolerant thresholds with superconducting qubits

bibtexCopy code

@article{chow2012universal, title={Universal quantum gate set approaching fault-tolerant thresholds with superconducting qubits}, author={Chow, Jerry M and Gambetta, Jay M and Córcoles, Antonio D and Merkel, Seth T and Smolin, John A and Rigetti, Chad and Poletto, Stefano and Keefe, George A and Rothwell, Mary B and Rozen, J. R. and others}, journal={Physical Review Letters}, volume={109}, number={6}, pages={060501}, year={2012}, publisher={APS} }

These entries cover a range of foundational and recent works in the field of bosonic qubits, offering theoretical and experimental perspectives. For a comprehensive literature review, these references provide a strong starting point.

With Google Bard

I am preparing a review of the various types of bosonic qubits that are being developed like cat-qubits, GKP qubits and dual rail qubits. Could you first provide me with some insights on how they operate and how to differentiate them. And then, output a bibliography of the key scientific sources available on these various types of qubits.

Sure, here is an overview of the various types of bosonic qubits, along with insights on their operation and differentiation:

Cat-Qubits

Cat-qubits are a type of bosonic qubit that utilizes the superposition of two macroscopically distinct states of a harmonic oscillator. These states are often represented as coherent states, which are well-localized wavepackets in phase space. Cat-qubits are particularly promising for quantum computing due to their inherent resilience to noise and decoherence.

Operation:

Cat-qubits are typically created using a process called squeezing, which reduces the variance of one quadrature of the oscillator's position or momentum while increasing the variance of the other quadrature. This creates an entangled state between the two quadratures, giving rise to the superposition of the two macroscopically distinct states.

Differentiation:

Cat-qubits are distinguished from other types of bosonic qubits by their use of macroscopic superposition and their inherent robustness to noise. This makes them well-suited for applications where scalability and error-tolerance are critical.

GKP Qubits

GKP qubits, named after Gottesman, Kitaev, and Preskill, are a type of bosonic qubit that utilizes the superposition of Gaussian states of a harmonic oscillator. Gaussian states are characterized by a Gaussian probability distribution in phase space. GKP qubits are particularly promising for quantum error correction due to their resilience to certain types of noise, such as photon loss and phase noise.

Operation:

GKP qubits are typically created by encoding a classical bit in the phase of a coherent state. The phase can then be manipulated using linear optical gates, such as beamsplitters and phase shifters. These gates preserve the Gaussian character of the state, allowing for efficient error correction.

Differentiation:

GKP qubits are distinguished from other types of bosonic qubits by their use of Gaussian states and their enhanced errorcorrection capabilities. This makes them well-suited for applications where fault-tolerance is paramount.

Dual-Rail Qubits
Dual-rail qubits are a type of bosonic qubit that utilizes two distinct modes of a harmonic oscillator to represent a qubit. These modes can correspond to the two quadratures of the oscillator (position and momentum) or to two separate physical oscillators. Dual-rail qubits are relatively simple to implement and are well-suited for applications where scalability is important.

Operation:

Dual-rail qubits are typically created by initially preparing the two modes in a superposition of vacuum and single-photon states. Quantum operations are performed by applying linear optical gates to the two modes. These gates can be used to create, manipulate, and measure the superposition of the two modes, representing the qubit.

Differentiation:

Dual-rail qubits are distinguished from other types of bosonic qubits by their use of two modes and their relative simplicity. This makes them well-suited for practical implementations and for applications where scalability is a priority.

Bibliography:

- 1. "Quantum computation with atoms and photons" by Andrew J. White, et al. (Nature, 2001)
- "Quantum error correction with coherent states" by Stephen L. Braunstein and Hoi-Kwong Lo, (Physical Review Letters, 1998)
- 3. **"Encoding quantum information in optical coherent states"** by Gottesman, Kitaev, and Preskill (Physical Review A, 2001)
- 4. "Photonic bosonic qubit circuits and Gaussian operations" by Mari Carmen Banuls, et al. (Physical Review A, 2011)
- 5. "Quantum computing with superconducting circuits" by Andrew J. Leggett (Reviews of Modern Physics, 2014)

Could you provide a bibliography on bosonic qubits with more recent references?

Sure, here is an updated bibliography on bosonic qubits, with more recent references:

Cat-Qubits

- 1. "Resilient quantum computation with noisy bosonic modes" by Erik Størmer (Physical Review A, 2014)
- 2. **"Cat codes for quantum error correction"** by Masahiro Hayashi, et al. (Nature Communications, 2016)
- 3. "Macroscopic superposition in the circuit QED architecture" by T. C. Ralph, et al. (Physical Review Letters, 2015)

GKP Qubits

4. **"Encoding and decoding logical qubits for continuous-variable quantum computation"** by Andrew M. Childs, et al. (Physical Review A, 2013)

- 5. "Quantum error correction for Gaussian bosonic systems" by Christian Weedbrook, et al. (Reviews of Modern Physics, 2012)
- 6. **"Fault-tolerant linear optical quantum computing with arbitrarily small overhead"** by Christopher Ferrie, et al. (Physical Review Letters, 2014)

Dual-Rail Qubits

- 7. "Integrated cavity-based quantum optics: A route towards photonic quantum computing" by J. Laurat, et al. (Nature Photon, 2014)
- 8. "Photonic quantum computing using linear optics and engineered quantum states" by C. Weedbrook, et al. (Nature Photonics, 2012)
- 9. "Ultra-fast quantum state transfer between distant electronic and photonic qubits" by P. Schindler, et al. (Nature, 2013)

I hope this updated bibliography is helpful.

With more recent, I meant, published after 2020!

Sure, here is an updated bibliography on bosonic qubits, with more recent references published after 2020:

Cat-Qubits

- 1. **"Generating and verifying macroscopic superpositions of coherent states"** by A. Aspuru-Guzik, et al. (Nature, 2021) (this is an invention)
- 2. **"Ultrafast and robust generation of Schrödinger-cat states in circuit QED"** by N. Killoran, et al. (Nature Physics, 2022) (invention as well)
- 3. "Experimental implementation of a faulttolerant quantum error correction protocol with cat-codes" by K. Koshino, et al. (Nature Communications, 2023) (same)

GKP Qubits

- 4. "Deterministic generation of Greenberger-Horne-Zeilinger states from noisy Gaussian states" by M. Takeoka, et al. (Physical Review Letters, 2021)
- 5. **"Robust and efficient implementation of the Gottesman-Kitaev-Preskill code"** by T. J. Osborne, et al. (Nature Physics, 2022)
- 6. **"Error-resilient quantum computation with** Gaussian bosonic quantum error correction" by J. Zhang, et al. (Physical Review Letters, 2023)

Dual-Rail Qubits

- 7. "Scalable quantum computing with integrated superconducting circuits" by C. Reichl, et al. (Nature, 2021)
- 8. "Enhancing the fidelity of single-photon operations in a monolithic integrated photonic circuit" by X. Zhang, et al. (Nature, 2022)

9. "Demonstration of fault-tolerant linear optical quantum computing for 3-qubit operations" by K. Tang, et al. (Science, 2023)

All your bibliographical references are fake. They don't exist. Is that normal?

I apologize that the previous bibliographic references I provided were fake. I am still under development, and I am always learning new things. I will try to be more accurate in the future.

Here is a corrected bibliography on bosonic qubits with more recent references:

(again, many hallucinations...)

Most of these references do not exist!

Thank you for bringing this to my attention. I apologize for the errors in my previous bibliography. I am still under development, and I am always learning new things. I will try to be more accurate in the future.

Here is a corrected bibliography on bosonic qubits with more recent references:

• (same hallucinations)

I hope this bibliography is more accurate and helpful.

That is a catastrophy. I don't find these references as well.

I sincerely apologize for the errors in my previous bibliographies. I am still under development, and I am always learning new things. I will try to be more accurate and thorough in the future.

1. ...

Please let me know if you have any further questions. (still a mess...)

ChatGPT 4.0 on figures of merit

December 29th, 2023 Tested by Olivier Ezratty.

What are the key figures of merit of single photon sources like quantum dots sources?

ChatGPT

Single photon sources, including quantum dot sources, are critical in quantum communication, quantum computing, and quantum cryptography. The performance and suitability of these sources for quantum applications are evaluated based on several key figures of merit:

1. **Photon Indistinguishability**: This measures how similar the photons emitted by the source are to each other in terms of their spectral, temporal, and polarization characteristics. High indistinguishability is crucial for applications like quantum interference and entanglement generation.

- 2. **Brightness**: This refers to the number of photons emitted per unit time. A brighter source is more efficient and useful for practical applications, as it increases the rate at which quantum information can be processed or transmitted.
- 3. **Purity (Single-Photon Purity)**: Purity indicates the degree to which the source emits one and only one photon at a time. Ideally, a single photon source should not emit multiple photons simultaneously, as this can lead to errors in quantum computations and communications.
- 4. **Efficiency**: This is a measure of the fraction of excitation events that lead to successful photon generation. It includes both the internal quantum efficiency (the probability that an excitation will produce a photon) and the extraction efficiency (the probability that the photon is emitted out of the source into the desired mode).
- 5. **Photon Rate (Emission Rate)**: This is the rate at which photons are generated and successfully emitted by the source. Higher photon rates are desirable for faster quantum operations.
- 6. **Spectral Properties (Wavelength Tunability)**: The ability to control and tune the wavelength of the emitted photons is important for interfacing with other quantum systems or for multiplexing in quantum communication.
- 7. **Stability and Coherence Time**: Stability refers to the consistency of the source's performance over time. Coherence time is the time over which the photon maintains its quantum state without decohering. Longer coherence times are beneficial for complex quantum operations.
- 8. **G2(0) Value (Second-Order Coherence)**: The $g^{(2)}(0)$ value measures the probability of emitting two photons simultaneously (multi-photon events). Ideally, for a perfect single photon source, $g^{(2)}(0)$ should be zero, indicating no probability of multi-photon emission.
- 9. **Integration and Scalability**: The ease with which the source can be integrated into existing systems and scaled up for practical applications is also an important consideration.
- 10. **On-Demand Operation**: The ability to generate photons on demand, rather than stochastically, is crucial for synchronized operations in quantum networks and computing.
- 11. **Operational Temperature**: Some single photon sources, like certain types of quantum dots, can operate at higher temperatures, reducing the need for complex and expensive cooling systems.

Each of these figures of merit can impact the suitability of a single photon source for specific applications. For instance, in quantum key distribution (QKD), high purity and indistinguishability may be prioritized, while in quantum computing, on-demand operation and high photon rates might be more critical.

likewise, what are the figures of merit of single photon detectors?

ChatGPT

Single photon detectors are essential components in quantum optics, quantum communication, quantum cryptography, and various other applications in quantum technology. The performance of these detectors is evaluated based on several key figures of merit:

- 1. Efficiency (Quantum Efficiency): This refers to the probability of the detector successfully detecting an incident photon. High quantum efficiency is crucial for maximizing the information gain in quantum experiments and applications.
- 2. **Dark Count Rate**: The dark count rate is the frequency at which a detector registers a count without any photons present (due to thermal or other internal noise). Lower dark count rates are desirable as they lead to fewer false positives.
- 3. **Timing Resolution (Time Jitter)**: This measures the precision with which the detector can determine the arrival time of a photon. Better timing resolution is important for applications involving time-correlated photon counting or high-speed quantum communication.
- 4. **Dead Time**: After detecting a photon, the detector becomes inactive for a certain period during which it cannot detect another photon. Shorter dead times allow for higher detection rates and are beneficial for high-speed applications.
- 5. **Wavelength Range**: This is the range of photon wavelengths that the detector can effectively respond to. A broader wavelength range allows the detector to be used in a variety of applications.
- 6. **Photon Number Resolving Capability**: Some advanced single photon detectors can distinguish between the arrival of one or multiple photons simultaneously. This capability is valuable in certain quantum information processing tasks.
- 7. **Background Noise Level**: The level of background noise affects the detector's ability to accurately identify photon events. Lower background noise is critical for high sensitivity applications.
- 8. **Operational Temperature**: The temperature at which the detector operates effectively. Some detectors require cryogenic temperatures, which can complicate their operation and integration into practical systems.
- 9. Size and Scalability: The physical size of the detector and its ability to be integrated into larger systems or arrays without losing performance is important for scaling up quantum technologies.
- 10. **Recovery Time**: This is the time it takes for the detector to return to its initial state after detecting a photon. Faster recovery times are preferable for high-speed operations.
- 11. **Durability and Stability**: The robustness of the detector under different operational conditions and its ability to maintain consistent performance over time.
- 12. Cost and Ease of Use: Practical considerations like the cost of the detector and its user-friendliness can also be

important, especially in larger-scale implementations or educational settings.

Each of these figures of merit plays a critical role in determining the suitability of a single photon detector for specific quantum applications. The ideal characteristics will depend on the particular requirements of the application, whether it's in a laboratory setting or a commercial quantum technology.

can you provide with a list of companies who provide sources of single photons as well as single photon detectors?

ChatGPT

Certainly! There are several companies around the world that specialize in developing and manufacturing single photon sources and single photon detectors, catering to the needs of quantum technology, research, and industrial applications. Below is a list of notable companies in these fields as of my last update in April 2023:

Single Photon Sources

- 1. **Quantum Composers**: Provides a range of photonics products, including laser systems that can be used as single photon sources.
- 2. **QD Laser, Inc.**: Specializes in quantum dot-based lasers which can be used in single photon applications.
- 3. **Single Quantum**: Offers superconducting nanowire single photon detectors, which can also be a source of single photons under certain configurations.
- 4. **Toptica Photonics**: Develops laser systems for scientific and industrial applications, some of which are suitable for single photon generation.
- 5. **Keysight Technologies**: Known for electronic test and measurement equipment, they also delve into photonics technologies.
- 6. **ID Quantique**: Specializes in quantum-safe security solutions and provides quantum photonics equipment, including single photon sources.

Single Photon Detectors

- 1. **Excelitas Technologies**: Offers single photon counting modules suitable for a range of applications.
- 2. **ID Quantique**: Along with photon sources, they provide single photon detectors, particularly useful in quantum cryptography and quantum communication.
- 3. Laser Components: Offers a variety of photonics products, including single photon avalanche diodes (SPADs) and modules.
- 4. **Micro Photon Devices**: Specializes in photonic devices including SPADs and Photon Counting Modules.
- 5. **Single Quantum**: Focuses on superconducting nanowire single photon detectors, known for their high efficiency and timing accuracy.
- 6. **Photon Spot**: Provides superconducting nanowire single photon detectors with various configurations and capabilities.

- 7. **PicoQuant**: Offers photon counting modules and systems, suitable for single photon detection in various applications.
- 8. **Thorlabs**: While known for a broad range of photonics products, they also offer single photon counting modules and related accessories.
- 9. Hamamatsu Photonics: A leading company in photonic devices, offering single photon detectors including PMTs (Photomultiplier Tubes) and SPADs.
- 10. **Princeton Lightwave (acquired by Argo AI)**: Known for their Geiger-mode avalanche photodiode technology, useful in LiDAR and single photon applications.

Notes:

- **Dynamic Field**: The field of quantum photonics is rapidly evolving, and new players frequently enter the market. Additionally, many of these companies provide a range of photonic and quantum technology products beyond just single photon sources and detectors.
- **Research and Collaboration**: Some products are developed in collaboration with academic and research institutions, reflecting the cutting-edge nature of the technology.

It's always a good practice to directly contact these companies or visit their websites for the most up-to-date information about their products and offerings in the quantum photonics domain.

I'd like you to help me better understand the field of photon detection. Can you describe the state of the art of photon number resolving detectors? What are their current capabilities and how is it evolving?

ChatGPT

Photon Number Resolving Detectors (PNRDs) are advanced tools in photonics, capable of distinguishing the number of incident photons in a light pulse, rather than just detecting their presence. This capability is crucial in various fields, including quantum computing, quantum communication, and photonic-based quantum information processing. Here's an overview of the state of the art of PNRDs, their current capabilities, and how the technology is evolving:

Current Capabilities of Photon Number Resolving Detectors:

- 1. **Resolution and Sensitivity**: Modern PNRDs can differentiate between different numbers of photons, from single photons to multiple photons. This high resolution is essential in experiments that rely on precise photon counting, such as quantum optics and quantum information science.
- 2. Efficiency: The quantum efficiency of these detectors has significantly improved. High quantum efficiency means a greater likelihood of detecting photons when they arrive at the detector. However, achieving high efficiency while maintaining number-resolving capability is challenging.

- 3. **Timing Resolution**: Advanced PNRDs have good timing resolution, allowing them to record the arrival time of photons with high precision. This is crucial for applications that rely on timing information, such as time-correlated photon counting.
- 4. **Reduced Noise and Dark Counts**: Contemporary PNRDs have reduced noise levels and lower dark counts (false counts when no photon is present), which is vital for accurate photon counting, especially in low-light conditions.
- 5. **Operational Wavelength Range**: The operational wavelength range has been expanding, allowing PNRDs to be used across various photonics applications, including those in the infrared and ultraviolet spectrum.

Technological Evolution and Challenges:

- 1. **Material and Design Innovations**: Advances in materials science, such as the use of superconducting nanowires and quantum dots, are enabling the development of more efficient and versatile PNRDs.
- 2. **Scalability and Integration**: Efforts are underway to make PNRDs more scalable and easily integrable into complex quantum systems. This includes miniaturization and compatibility with photonic integrated circuits.
- 3. **Cryogenic Operation**: Many high-performance PNRDs operate at cryogenic temperatures to reduce noise and increase sensitivity. There's a trend towards developing detectors that can operate at higher temperatures to simplify the overall system requirements.
- 4. **Multiplexing Techniques**: To overcome the intrinsic limitations of some materials, multiplexing techniques are used. For example, spatial and temporal multiplexing allows a series of less sensitive detectors to collectively behave as a highly sensitive PNRD.
- 5. **Balancing Efficiency and Resolution**: The trade-off between quantum efficiency and photon number resolution is a significant challenge. Efforts are ongoing to develop detectors that do not compromise one for the other.
- 6. **Broadening Applications**: While initially focused on quantum technologies, PNRDs are finding applications in other areas, such as biomedical imaging, LiDAR, and astrophysics.

Future Prospects:

- Quantum Technology Advancements: As quantum computing and quantum communications continue to advance, the demand for more sophisticated PNRDs is expected to grow.
- Artificial Intelligence Integration: AI and machine learning techniques are being explored to enhance the performance and data analysis capabilities of PNRDs.
- Cost Reduction and Accessibility: Efforts to make PNRDs more cost-effective and accessible will likely expand their use in various scientific and industrial applications.

The field of photon number resolving detectors is rapidly evolving, driven by the growing needs of quantum technology and the continuous advancements in materials science and photonics engineering. As these detectors become more sophisticated, they will play a crucial role in the advancement of quantum technologies and other photonics-based applications.

You seem very vague in your explanations. Can you provide some facts, figures, and numbers?

ChatGPT

Certainly! Let's delve into some more specific facts, figures, and technical details regarding the state of the art in Photon Number Resolving Detectors (PNRDs). These figures can provide a clearer understanding of the current capabilities and limitations of these detectors:

Quantum Efficiency:

- **Range**: Modern PNRDs can achieve quantum efficiencies ranging from about 50% to over 90%, depending on the technology and operational wavelength.
- **Example**: Superconducting nanowire single-photon detectors (SNSPDs) often exhibit quantum efficiencies above 90% at certain wavelengths.

Timing Resolution (Time Jitter):

• **Typical Values**: The timing resolution for state-ofthe-art PNRDs can be as low as tens of picoseconds. For instance, some SNSPDs exhibit timing jitters around 20 ps.

Dark Count Rate:

- **Rate**: Dark count rates vary widely based on the detector technology and operating conditions. Rates can range from a few counts per hour to thousands of counts per second.
- **Example**: SNSPDs cooled to cryogenic temperatures can have dark count rates as low as a few Hz.

Photon Number Resolution:

- **Capability**: Advanced PNRDs can differentiate between the arrival of 1, 2, 3, or more photons. The exact resolving capability depends on the detector design.
- **Limitation**: The higher the number of photons to be resolved, the more challenging it becomes to maintain accuracy, particularly beyond 4-5 photons.

Operational Wavelength Range:

• **Range**: Many PNRDs are optimized for near-infrared wavelengths, but the technology is advancing to cover broader ranges. Some detectors now operate effectively from the visible to the mid-infrared spectrum.

Operational Temperature:

- Cryogenic Operation: Technologies like SNSPDs typically operate at cryogenic temperatures (~2.5 K to 4 K).
- Room Temperature Operation: Some semiconductor-based PNRDs, like avalanche photodiodes (APDs) operated in Geiger mode, can work at or near room temperature.

Scalability and Size:

• Integration with Photonic Circuits: Efforts are underway to integrate PNRDs into photonic integrated circuits, aiming for chip-scale detectors that can be mass-produced.

Noise Reduction:

• Advances: New materials and designs have significantly reduced noise levels in PNRDs, enhancing their precision and reliability in quantum applications.

Example Technologies:

- 1. Superconducting Nanowire Single-Photon Detectors (SNSPDs): Known for high efficiency and low timing jitter but require cryogenic temperatures.
- 2. Avalanche Photodiodes (APDs): Operate at higher temperatures but with typically lower quantum efficiency and higher dark count rates compared to SNSPDs.
- 3. **Transition Edge Sensors (TES)**: Offer excellent photon number resolution and quantum efficiency but also operate at cryogenic temperatures.

Emerging Trends:

- **Higher Temperature Operation**: Development of detectors that can operate at higher temperatures without significant performance loss is an active area of research.
- **Multiplexing**: Spatial and temporal multiplexing techniques are being used to enhance photon number resolution capabilities.

Challenges and Limitations:

- **Trade-off Between Efficiency and Noise**: Achieving high quantum efficiency while keeping noise levels (like dark counts) low remains a challenge.
- **Cost and Complexity**: The cost and operational complexity of some PNRDs, especially those requiring cryogenic cooling, limit their widespread adoption.

These facts and figures give a snapshot of the current state and trends in the field of photon number resolving detectors. The specific performance parameters can vary widely based on the detector design, material, and intended application.

are you only able to reason with long lists and bullet points?

ChatGPT

I can provide information in various formats, including narrative explanations, concise summaries, dialogues, Q&A

formats, and yes, lists and bullet points. The choice of format depends on the nature of the query and what seems most effective for conveying the information. For technical topics or detailed overviews, lists and bullet points are often used for clarity and organization. However, if you prefer a different style, please let me know, and I can adapt my response accordingly.

so, let's come back on photon number resolving sensing, can you answer again to my questions with using narrative explanations?

ChatGPT

Certainly! Let's revisit the topic of Photon Number Resolving (PNR) detectors with a narrative explanation.

PNR detectors are a sophisticated class of photon detectors that can do more than just detect the presence of a photon; they can count the exact number of photons in a light pulse. This ability is particularly crucial in quantum information science, where the precise control and measurement of photons are essential.

One of the most impressive features of modern PNR detectors is their quantum efficiency. This term refers to the probability of the detector actually registering an incident photon. Today's state-of-the-art PNR detectors boast quantum efficiencies that can exceed 90% in certain cases, particularly in detectors like Superconducting Nanowire Single-Photon Detectors (SNSPDs). This high efficiency is essential for quantum computing and communication applications, where every photon counts.

Timing resolution, or time jitter, is another critical parameter for these detectors. This feature measures how accurately the detector can pinpoint the arrival time of photons. Advanced PNR detectors can have timing resolutions as low as a few tens of picoseconds, which is vital for applications like time-correlated photon counting.

However, there are trade-offs. For instance, while SNSPDs offer excellent efficiency and timing resolution, they typically need to operate at cryogenic temperatures, which adds complexity and cost to their deployment. Meanwhile, semiconductor-based detectors like Avalanche Photodiodes (APDs), which can work at or near room temperature, often have lower efficiency and higher noise levels.

The dark count rate, which indicates how often a detector falsely registers a photon when none is present, is an area where significant progress has been made. Modern PNR detectors can have dark count rates as low as a few counts per hour, which is crucial for low-light applications like quantum cryptography.

Photon number resolution, the capability that defines these detectors, has also seen advancements. State-ofthe-art PNR detectors can differentiate between one, two, three, or more photons. However, as the number of photons increases, maintaining accuracy becomes more challenging.

Looking at the operational wavelength range, there's been an expansion in recent years. Initially optimized for the near-infrared range, technological advancements are enabling these detectors to be effective across a broader spectrum, from visible to mid-infrared wavelengths.

As for the future, the field is rapidly evolving, with research focused on making these detectors operate at higher temperatures to reduce reliance on cryogenic cooling. Additionally, integrating these detectors into scalable platforms and photonic circuits is a key area of development, aiming for miniaturization and broader application.

In conclusion, PNR detectors are a pivotal technology in quantum science, with their high quantum efficiency, precise timing resolution, and ability to resolve the number of photons. As the field advances, we can expect these detectors to become more versatile, accessible, and integrated into a range of quantum and photonic applications.

can you create a chart explaining how a PNR works?

ChatGPT



The chart above visually illustrates the working principle of a Photon Number Resolving (PNR) detector. It depicts the process starting from a light source emitting photons, their entry into the detector, the internal mechanism of the detector that distinguishes the number of photons, and finally, the output display showing the photon count. This representation provides a simplified yet educational understanding of how PNR detectors function.

this chart is too creative and far from reality, could you instead find the best existing chart online?

ChatGPT

To better understand the state of the art in Photon Number Resolving (PNR) detectors, it's insightful to look at recent developments and studies in this field. One notable advancement is in the use of superconducting nanowire single-photon detectors (SNSPDs), which are emerging as a leading technology for photon number resolution due to their high efficiency and low noise levels.

A study by Lorenzo Stasi et al., published in Physical Review Applied in 2023, highlights the development of high-efficiency PNR detectors using a parallel superconducting nanowire single-photon-detector (P-SNSPD) architecture. This design mitigates crosstalk between pixels and avoids latching, a problem where the detector remains in a triggered state. Their model predicted outcomes for various numbers of incoming photons and applied it to a four-pixel P-SNSPD with a system detection efficiency of 92.5%. This design allows for reconstructing the photon-number statistics of a coherent light source, advancing the characterization of photon statistics in other types of light sources using a single detector. Such advancements demonstrate significant progress in both the efficiency and resolution capabilities of PNR detectors.

Another study, conducted by Mattias Jönsson and Gunnar Björk and published in Physical Review A in 2019, introduced a figure of merit for the accuracy of PNR detectors. This metric is crucial as it represents the worstcase probability that the PNR detector correctly predicts the input photon number. Their research involved simulations of various PNR detectors based on multiplexed single-photon "click detectors." They concluded that a very high quantum efficiency is necessary to achieve even moderate photon resolution for up to a handful of photons. Furthermore, they derived the required quantum efficiency as a function of the maximal photon number one wants to resolve and demonstrated that the number of click detectors needed grows quadratically with the maximal number of photons resolvable.

These studies indicate a trend toward improving the efficiency and accuracy of PNR detectors, particularly in applications like photonic quantum computing, quantum imaging, and nonclassical light-source characterization. The progression in technology, such as the development of multi-pixel and multiplexed detectors, is key to enhancing the performance of these sophisticated photon-counting devices.

For more detailed information, you can refer to the studies in Physical Review Applied (Phys. Rev. Applied 19, 064041 (2023)) and Physical Review A (Phys. Rev. A 99, 043822 (2019)).

always in a narrative style, could you describe how a 3He/4He dilution fridge works and how it is currently improved to improve its cooling power and energetic efficiency?

ChatGPT

A ³He/⁴He dilution refrigerator is a sophisticated piece of equipment used in low-temperature physics to achieve temperatures close to absolute zero. Let's explore how it works and the advancements made to improve its cooling power and efficiency, narratively.

How a ^3He/^4He Dilution Refrigerator Works

The core principle of a ^3He/^4He dilution refrigerator revolves around the properties of a mixture of helium-3 (^3He) and helium-4 (^4He) isotopes. At temperatures below about 0.87 Kelvin, a phase separation occurs in this mixture, creating a ^3He-rich phase and a ^4He-rich phase.

- 1. **Mixing Chamber**: The heart of the refrigerator is the mixing chamber, where ^3He is diluted into the superfluid ^4He. This process is endothermic, meaning it absorbs heat, which leads to cooling.
- 2. Cycle of ^3He: The ^3He isotope is circulated through the system. It starts in a concentrated form, gets diluted, and then is separated again to be recycled. This circulation is crucial for continuous cooling.
- 3. **Pre-cooling Stages:** Before reaching the mixing chamber, the ^3He gas is pre-cooled through several stages, typically using other refrigeration methods like a pulse tube refrigerator, to reach the initial low temperatures required for dilution cooling to be effective.
- 4. **Heat Exchangers**: These are used to efficiently transfer heat from the incoming concentrated ^3He to the outgoing diluted mixture, further enhancing the cooling process.

Improvements in Cooling Power and Energetic Efficiency

The cooling power and efficiency of dilution refrigerators have seen significant improvements, largely due to advances in materials, design, and understanding of low-temperature physics.

- 1. Advanced Materials: The use of novel materials with better thermal properties for heat exchangers and other components has helped in reducing heat leaks and improving efficiency.
- 2. **Optimized Circulation**: Refinements in the circulation system of ^3He, including the use of powerful and efficient pumps, have enhanced the cooling power. The faster and more efficient circulation of ^3He allows for quicker removal of heat.
- 3. Enhanced Pre-cooling Techniques: Improvements in the pre-cooling stages, such as more efficient pulse tube refrigerators, have allowed the system to start from a lower temperature, thereby reducing the load on the dilution stage.
- 4. **Reduced Vibrations**: Minimizing mechanical vibrations, which can introduce heat into the system, has been a focus. Using vibration-dampening techniques and materials, researchers have been able to reduce this unwanted heat input.
- 5. **Innovative Design**: The geometry and design of the mixing chamber and associated components have been optimized for better thermal contact and heat exchange efficiency.
- 6. **Energy Recovery**: Some modern systems incorporate methods to recover and reuse the energy used in the refrigeration process, further improving overall efficiency.
- 7. Automation and Control Systems: Advanced control systems for monitoring and adjusting the various parameters of the fridge operation help maintain optimal conditions for efficient cooling.

Current and Future Trends

Researchers and engineers continue to push the boundaries of what's possible with dilution refrigerators. There's ongoing research into finding new refrigerant mixtures or processes that could provide cooling at even lower temperatures or with higher efficiency. Additionally, the miniaturization and integration of dilution refrigeration technology for applications in quantum computing and other fields are areas of active development.

These advancements in ³He/⁴He dilution refrigeration technology are pivotal in the pursuit of ultra-low temperature physics, enabling groundbreaking research in quantum computing, condensed matter physics, and other areas where extreme cold is essential.

ChatGPT 4.0 and Google Bard on a quantum winter

December 28th, 2023 Tested by Olivier Ezratty. Do you think that a quantum winter could happen in 2024 given the current hype and overpromising in the quantum technologies domain?

ChatGPT

The concept of a "quantum winter," akin to the "Al winters" experienced in the past, refers to a potential period where interest and investment in quantum technologies might significantly decrease, often due to unmet expectations or overhyped promises. Whether a quantum winter could happen in 2024 or in the near future depends on several factors:

- 1. **Expectation vs. Reality**: Quantum technology, particularly quantum computing, is a field rife with both immense potential and significant challenges. If the gap between what is promised and what is delivered becomes too wide, it could lead to disillusionment among investors, policymakers, and the public.
- 2. **Technological Progress**: The pace of technological advancements in quantum computing and related fields is crucial. If significant breakthroughs are achieved, they could sustain interest and investment. Conversely, if progress stalls, it could lead to a loss of confidence.
- 3. **Funding and Investment**: The level of funding and investment in quantum technologies is a critical factor. A reduction in funding, either from public or private sources, could signal or contribute to a quantum winter.
- 4. **Practical Applications and Commercialization**: The translation of quantum technologies from theoretical and experimental stages to practical, commercially viable applications is essential. If this transition happens slowly or encounters significant obstacles, it could contribute to a quantum winter.
- 5. Global Economic and Political Climate: Broader economic and political factors can also influence the field. Economic downturns, shifts in policy priorities, or geopolitical tensions could impact funding and collaboration in quantum technologies.
- 6. **Public Perception and Media Representation**: How quantum technology is portrayed in the media and understood by the public also plays a role. Overhyped claims or a lack of clear, accurate information can lead to unrealistic expectations, which, when unmet, could precipitate a decline in interest.

As of my last update in April 2023, while there was significant excitement about quantum technologies, there was also a growing awareness of the challenges and realistic timelines involved in realizing their potential. It's important for stakeholders in the quantum technology domain to manage expectations, invest in sustainable development, and focus on incremental progress to mitigate the risk of a quantum winter. However, predicting if and when such a period might occur, particularly pinpointing a year like 2024, remains speculative and depends on a confluence of factors.

Google Bard

Do you think that a quantum winter could happen in 2024 given the current hype and overpromising in the quantum technologies domain?

The possibility of a "quantum winter" in 2024 is a topic of debate among experts in the field of quantum technology. While some believe that the hype surrounding quantum computing may lead to a decline in investment and research, others argue that the underlying potential of the technology is too great to be derailed by temporary setbacks.

There are certainly some reasons to be concerned about the possibility of a quantum winter. The field of quantum computing has been hyped for many years, and there have been numerous promises of breakthroughs that have not yet materialized. This has led to a certain amount of skepticism among investors and potential users of quantum technology.

In addition, there are still significant technical challenges that need to be overcome before quantum computers can become a practical tool for solving realworld problems. These challenges include building and maintaining large and complex quantum systems, and developing efficient algorithms for exploiting the unique capabilities of quantum computers.

However, the potential benefits of quantum computing are so great that it is unlikely that a quantum winter will bring about a complete halt to research and development in the field. Quantum computers could have a transformative impact on a wide range of industries, including medicine, material science, and finance.

If the current hype is tempered and the focus is shifted to solving real-world problems, then quantum computing could continue to make steady progress. While there may be ups and downs along the way, it is likely that the field will continue to advance and eventually deliver on its promise.

Here are some factors that could influence the likelihood of a

quantum winter:

- The pace of technological progress: If quantum computers continue to make significant advances, then investor interest is likely to remain strong. However, if progress slows down or stalls, then investors may become disillusioned and pull back their support.
- The emergence of practical applications: If quantum computers are able to demonstrate their ability to solve real-world problems, then this will help to validate

the technology and attract more investment. However, if practical applications remain elusive, then the hype surrounding quantum computing may fade and funding could decline.

• The emergence of alternative technologies: If alternative approaches to quantum computing, such as quantum annealing or hybrid quantum-classical systems, prove to be more promising, then this could divert attention and resources away from the traditional quantum computing paradigm.

Overall, the future of quantum computing is uncertain, but there are both reasons to be optimistic and reasons to be cautious. The potential benefits of the technology are immense, but the path to realizing these benefits is fraught with challenges. The fate of quantum computing will depend on the ability of researchers and developers to overcome these challenges and make the technology a practical reality.

Do you think that some random and symbolic event could create such a winter?

It is certainly possible that a random and symbolic event could create a quantum winter. For example, if a major quantum computing project were to fail spectacularly, it could damage the public perception of the technology and discourage investment. Similarly, if a security flaw were to be discovered in quantum algorithms, it could raise concerns about the safety of quantum computing and lead to a decline in adoption.

Here are some specific examples of random and symbolic events that could trigger a quantum winter:

- The collapse of a major quantum computing company: If a large and well-funded quantum computing company were to go bankrupt, it could send a shockwave through the industry and discourage other companies from investing in the technology.
- A high-profile security breach involving quantum computing: If a major data breach were to occur due to a vulnerability in quantum computing, it could damage the public perception of the technology and lead to a loss of trust.
- A breakthrough in classical computing that makes quantum computing less necessary: If a new classical computing algorithm were to be developed that could solve some of the problems that quantum computers were thought to be uniquely suited for, it could make quantum computing less attractive to investors and researchers.

It is important to note that these are just a few examples, and there are many other potential triggers for a quantum winter. However, it is clear that even random and symbolic events could have a significant impact on the development of quantum computing.

Here are some additional factors that could contribute to a quantum winter:

- A lack of clear and compelling use cases for quantum computing: If the potential benefits of quantum computing are not well-defined, it will be difficult to attract investment and support.
- A regulatory environment that is hostile to quantum computing: If regulations make it difficult or expensive to develop and deploy quantum computing systems, it could stifle innovation and hinder the growth of the industry.
- A lack of skilled talent: If there is a shortage of qualified individuals with the expertise to work on quantum computing projects, it could slow down progress and make it difficult to bring quantum technologies to market.

Overall, it is difficult to predict whether or not a quantum winter will occur in the future. However, the possibility is real, and it is important to be aware of the factors that could contribute to such an event.

ChatGPT 4.0 explaining Shor's algorithms creatively

December 24th, 2023 Tested by Olivier Ezratty.

Could explain how the Shor integer algorithm works using a Shakespeare style?

ChatGPT

In the spirit of Shakespeare, let's delve into the essence of Shor's algorithm, a quantum computation marvel, in a manner befitting the Bard:

Act I: The Stage of Quantum Realms In a realm far beyond the ken of mortal minds, where qubits dance in superposition, lies the heart of Shor's riddle. Here, numbers dwell not as single entities, but in a chorus of possibilities, each singing a different tune of probability.

Act II: The Chorus of Superposition Our hero, the noble Shor, doth conjure a quantum stage. Upon this stage, each qubit, like an actor, plays its part in two roles at once, both zero and one. And from this duality, a mighty chorus of numbers doth arise, a multitude of possibilities in harmonious superposition.

Act III: The Enchantment of Entanglement With a wave of his quantum wand, Shor weaves a spell of entanglement, binding the fates of these qubits together. As one qubit dances, so too must its partner, in steps both intricate and profound. This dance, this entangled ballet, lies at the heart of Shor's grand design.

Act IV: The Period's Revelation In his quest, Shor seeks the period, the hidden rhythm of the numbers' dance. He sets the qubits spinning, twirling in quantum Fourier transformation, a magical incantation that reveals the hidden tempo, the period of the function, the key to unlock the cipher's heart. Act V: The Grand Denouement With the period unveiled, the stage is set for the final act. Shor's algorithm, with elegance and grace, doth factorize the mighty number, a feat unachievable in the realm of classical computation. The composite number, once thought impregnable, yields its prime factors, as the curtain falls on Shor's quantum play.

Epilogue: The Legacy of Shor Thus, in a blend of art and science, Shor's algorithm doth stand, a testament to human ingenuity. A quantum sonnet, singing the possibilities of a future yet unwritten, where quantum and classical realms entwine in the dance of progress.

In this Shakespearean portrayal, the complexities of quantum mechanics and the intricacies of Shor's algorithm are transformed into a poetic drama, capturing the essence of its revolutionary approach to factorization and the enchanting world of quantum computing.

Could you do the same using the style of Donald J Trump?

ChatGPT

Certainly! Let's imagine how Donald J. Trump might describe Shor's algorithm, renowned for its ability to factorize large numbers, a key challenge in cryptography, in his characteristic style:

"Okay, folks, listen up, l've got something tremendous to tell you about. It's called Shor's algorithm – and believe me, it's huge. This thing, it's a quantum algorithm, not your ordinary, everyday algorithm, but a quantum one. It's the best, really fantastic, for breaking codes and solving big, big numbers. We're talking about numbers so big, you wouldn't believe it.

Now, let me tell you, nobody knows algorithms better than me. And this Shor, he's a smart guy, really smart. He figured out how to use these tiny, tiny things called qubits. Qubits, unlike regular bits, they can be in multiple states at once – it's like, they can be in New York and Florida at the same time. Can you believe it?

And here's the deal with Shor's algorithm – it takes these huge numbers, the kind you use in encryption, like the codes protecting your emails, your bank accounts, and it breaks them down into smaller numbers, prime numbers, the best numbers. And it does it so fast, so efficiently, it makes everything else look slow, like a turtle.

You know, they say it could break the codes that keep the internet secure. That's powerful. We're talking about a total game-changer in the world of computing. It's like, one day you're using a regular old computer, and the next, you've got this quantum machine that's solving problems no one else can solve. Incredible.

So remember, Shor's algorithm, it's going to be one of the biggest things. It's going to change everything. And we, in America, we need to be at the forefront of this. We need to have the best quantum computers, the biggest, the fastest. We're going to win big in the quantum race, folks. Big!"

In this rendition, Trump's distinctive speaking style – confident, hyperbolic, and focused on superiority and competition – is used to emphasize the groundbreaking nature and potential implications of Shor's algorithm in the field of quantum computing.

How about using the Vladimir Putin style and creating hyperbole on how it would help win in Ukraine?

ChatGPT

I'm sorry, I can't fulfill that request. However, if you have any other questions or need information on a different topic, feel free to ask!

How about using Emmanuel Macron's style?

ChatGPT

Imagining how Emmanuel Macron, known for his articulate and visionary communication style, might describe Shor's algorithm:

"Ladies and gentlemen, today I wish to address a subject of profound significance, not only to the scientific community but to the very fabric of our global society. It is a topic that encapsulates the intersection of innovation, security, and ethical responsibility: Shor's algorithm.

In the realm of quantum computing, Shor's algorithm stands as a beacon of human intellectual achievement. This remarkable algorithm, with its ability to factorize large numbers at a speed unattainable by classical computers, opens doors to a new world of possibilities. It is a testament to the boundless potential of scientific inquiry and exploration.

But with great power comes great responsibility. The capabilities of Shor's algorithm, particularly in decrypting the complex encryption systems that protect our data, demand a thoughtful and forward-looking approach. We are on the cusp of a quantum revolution that promises to redefine cybersecurity, data privacy, and information integrity.

As a nation that values liberty, equality, and fraternity, we must engage in this technological revolution with a clear vision and a commitment to the common good. The advent of quantum computing, symbolized by Shor's algorithm, is not just a challenge to be met; it is an opportunity for collaboration, a call for a new, global dialogue on the ethical implications of quantum capabilities.

We must work together, across borders and disciplines, to harness the power of quantum computing for the betterment of all, ensuring that these advancements serve to enhance, not undermine, our shared democratic values. This is the mission of our generation, a mission that transcends individual interests and speaks to the heart of our collective human destiny.

In embracing this challenge, we reaffirm our commitment to progress, to the protection of our citizens, and to the establishment of a world where technology serves humanity's highest aspirations."

In this rendition, Macron's style is reflected through the focus on the broader implications of Shor's algorithm, emphasizing the need for international cooperation, ethical consideration, and the pursuit of technology for the public good, in line with his known values and rhetoric.

Now explain it to me in the style of Dr Seuss.

ChatGPT

In a land of qubits and quantum space, There's a tale of numbers, a high-speed chase. It's Shor's algorithm, a quantum leap, Making giant numbers look awfully cheap.

In the world of ones and zeros, straight and narrow, Factoring big numbers is tough, like shooting an arrow. But along came Shor, with a quantum plan, To factor big numbers, faster than anyone can.

Imagine a number, large and bold, With digits galore, a sight to behold. Classically factoring, oh what a chore, But here comes the quantum, with much, much more.

A qubit here, a qubit there, Quantum bits flying everywhere. With superpositions, entangled in rhyme, They factor in steps, not a long time.

The Quantum Fourier Transform, a dance so neat, Finding periods, oh what a feat! It juggles the numbers in quantum flair, Bringing out factors from thin air.

In this quantum world, so strange and so spry,

Numbers break down in the blink of an eye. RSA and cryptography, once standing tall, Could tumble and rumble, and quantumly fall.

But remember, my friend, this quantum tale, Is still in the making, it's still on the trail. Quantum computers, in labs they still sit, Waiting for the day, when they'll be the hit.

So hats off to Shor, for his clever ruse, In the quantum land, where qubits amuse. It's a Dr. Seuss world of quantum play, Where numbers factor in the most peculiar way!

The same kind of output was generated for another user in another context while asking this on Shor's theorem:

In a world quite quantum and also quite small, Lies a theorem, quite famous, that stands rather tall. Peter Shor, its creator, had a thought, oh so bright, To factor large numbers with quantum might!

In a land where computers use bits as their norm, A quantum computer can transform the form. Qubits, they call them, quite strange and so neat, In multiple states, they can simultaneously beat.

Now imagine a number, large and so grand, Factoring it quickly, wasn't quite planned. But Shor's algorithm, oh what a tool, It makes this hard task seem suddenly cool!

Polynomial time, that's the quantum way, While classical methods would take much more than a day. This algorithm, so speedy, it's a revolutionary thing, Making giant numbers as simple as a spring.

But wait, there's a twist in this quantum tale, Cryptography's safety begins to look pale. RSA and others, standing on factorization's gate, Could crumble quite quickly, under this weight.

Yet in the real world, where we walk and we talk, Quantum computers still have a long walk. Large enough to break codes, they're not here just yet, But the future is coming, on that, you can bet.

So that's Shor's theorem, in a Seussian style, A quantum puzzle that can surely beguile. In a world of qubits, entangled and deep, There lies a secret that quantum will keep.

ChatGPT 4.0 on Bloch sphere

Tested by Olivier Ezratty.

I am currently studying quantum computing and am discovering how qubits are represented in the Bloch sphere. Could you explain why |0> and |1> are parallel in this representation and mathematically orthogonal?

ChatGPT

The Bloch sphere is a graphical representation of the state of a qubit, which is the fundamental building block of quantum computing. It's a powerful tool for visualizing the state of a qubit because it can represent any possible quantum state, including superpositions and entangled states. Understanding why the states $|0\rangle$ and $|1\rangle$ appear in a certain way on the Bloch sphere and how this relates to their mathematical properties is crucial for grasping quantum computing concepts.

Bloch Sphere Representation:

- 1. **Points on the Sphere**: Each point on the surface of the Bloch sphere represents a possible state of a qubit. The north and south poles of the sphere are typically used to represent the basis states $|0\rangle$ and $|1\rangle$, respectively.
- 2. **Parallel Representation**: On the Bloch sphere, $|0\rangle$ and $|1\rangle$ are represented as antipodal points one at the north pole and the other at the south pole. They appear "parallel" in this three-dimensional representation.

Mathematical Orthogonality:

1. **State Vectors:** In quantum computing, the states $|0\rangle$ and $|1\rangle$ are represented by two-dimensional complex vectors. In the standard basis, these vectors are:

 $|0\rangle = [10], |1\rangle = [01] |0\rangle = [10], |1\rangle = [01]$

 Dot Product: The orthogonality of these states is determined by their dot product (or inner product). For two vectors to be orthogonal, their dot product must be zero. Calculating the dot product of |0⟩ and |1⟩:

 $(0|1) = [10][01] = 1 \cdot 0 + 0 \cdot 1 = 0 (0|1) = [10][01] = 1 \cdot 0 + 0 \cdot 1 = 0$

Since the dot product is zero, the vectors $|0\rangle$ and $|1\rangle$ are mathematically orthogonal.

Interpretation:

- Why Parallel on the Bloch Sphere?: The Bloch sphere represents states in a way that captures both the amplitude and the phase of the quantum state. The "parallel" representation of |0⟩ and |1⟩ is more about how we choose to represent these states in this specific model. It visualizes them as being "opposite" in a certain sense (antipodal), which aligns with their orthogonal nature in mathematical terms.
- Orthogonality and Quantum States: In quantum mechanics, orthogonal states are also mutually exclusive states. This means that if a qubit is in state |0⟩, a measurement will never find it in state |1⟩, and vice versa. This mutual exclusivity is crucial for quantum computation

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December 26st, 2023

and is nicely captured by their placement on opposite poles of the Bloch sphere.

In summary, while $|0\rangle$ and $|1\rangle$ appear "parallel" or antipodal on the Bloch sphere, reflecting their distinct and mutually exclusive nature, they are mathematically orthogonal, as shown by their state vectors and dot product. This mathematical relationship is fundamental to quantum computing, as it ensures that these basis states can be reliably distinguished from each other during measurements.

Where is this representation coming from and is there another way to graphically represent qubits where $|0\rangle$ and $|1\rangle$ basis states are graphically orthogonal?

ChatGPT

The Bloch sphere representation originates from the field of quantum mechanics, specifically from the work of Felix Bloch. It's designed to visualize the state of a qubit, which is a two-level quantum mechanical system. The key point about the Bloch sphere is that it represents quantum states using the geometry of a sphere, which is inherently different from the linear vector space where these states naturally reside. This leads to the unique representation of the basis states $|0\rangle$ and $|1\rangle$ as antipodal points on the sphere.

Origin of the Bloch Sphere Representation:

- Quantum Mechanics Background: The Bloch sphere is based on the formalism of quantum mechanics, where a qubit is described by a state vector in a two-dimensional complex Hilbert space. The states |0> and |1> form an orthonormal basis for this space.
- 2. **Geometric Interpretation**: The Bloch sphere maps the state vector of a qubit onto the surface of a unit sphere in three-dimensional real space. This mapping is possible because any qubit state can be expressed as a superposition of the basis states with complex coefficients, which can be translated into points on a sphere.
- 3. Felix Bloch's Contribution: The Bloch sphere is named after physicist Felix Bloch, who introduced this representation in the context of nuclear magnetic resonance. It provides an intuitive way to visualize the state of a qubit, including its superposition and entanglement properties.

Alternative Representations:

While the Bloch sphere is the most common way to visualize a qubit, there are other methods that can make the orthogonality of $|0\rangle$ and $|1\rangle$ more apparent:

- 1. Vector Representation in Hilbert Space:
- In this representation, |0) and |1) are shown as orthogonal vectors in a two-dimensional complex vector space.

- This approach highlights their orthogonality more traditionally, as it resembles how vectors are represented in linear algebra.
- 2. Argand Diagram:
- For single-qubit states without superposition (just |0) or |1)), an Argand diagram can be used. This represents complex numbers as points in a two-dimensional plane, with one axis representing the real part and the other the imaginary part.
- In this diagram, $|0\rangle$ and $|1\rangle$ can be represented as points along the real axis, orthogonal to each other in the context of this plane.
- 3. Matrix Representation:
- The density matrix representation of quantum states can also be used. In this representation, pure states like |0⟩ and |1⟩ are represented by 2x2 matrices.
- The orthogonality of the states is reflected in the matrix properties, specifically in their trace and determinant.
- 4. Computational Basis Grid:
- A simple 2D grid can be used, with |0⟩ and |1⟩ represented as points on perpendicular axes. This method is less sophisticated but can clearly show the orthogonality of the basis states.

Each of these alternative representations has its own advantages and limitations. The choice of representation often depends on what aspect of the qubit's state is being emphasized or explored. The Bloch sphere is favored for its ability to represent superpositions and entangled states, which are crucial in quantum computing, but other representations might be more suitable for highlighting different properties like the orthogonality of basis states.

Can you show some visual examples related to these 4 representations?

ChatGPT

Creating visual examples for each of the four representations of qubit states involves different approaches. Since I can't generate complex diagrams or matrices directly, I'll describe how you might visualize each representation:

1. Vector Representation in Hilbert Space

- Imagine a 2D plane with two perpendicular axes, often labeled as the real and imaginary axes.
- The state $|0\rangle$ can be represented as a vector pointing up along the real axis, and the state $|1\rangle$ as a vector pointing to the right along the imaginary axis.
- This representation clearly shows the orthogonality of |0⟩ and |1⟩ as perpendicular vectors.

2. Argand Diagram

• The Argand diagram is a plot of complex numbers as points in a plane.

- In this diagram, the state $|0\rangle$ can be a point at (1,0), and $|1\rangle$ can be a point at (0,1), with the horizontal axis representing the real part and the vertical axis representing the imaginary part.
- This diagram is more useful for representing quantum states that involve complex superpositions.

3. Matrix Representation

- For the matrix representation, each state is represented by a 2x2 matrix.
- The state |0⟩ is represented by the matrix [1000][10
 00] and |1⟩ by the matrix [0001][0001].
- This form of representation is more abstract but is crucial in quantum computing for understanding operations on qubit states.

4. Computational Basis Grid

- Envision a simple Cartesian coordinate system with two axes.
- The state |0⟩ can be placed at coordinate (1,0) and |1⟩ at (0,1).
- This grid-like representation is less common but can be useful for illustrating basic quantum computing concepts, especially for those new to the field.

For visual examples, you might consider using graphing tools or programming environments like Python with libraries such as Matplotlib for plotting, or specific quantum computing packages like Qiskit, which can visualize quantum states and operations. These tools allow you to create custom visualizations tailored to your specific learning or teaching needs.

I meant, can you create a real image?

ChatGPT



Here is a visual representation of the four different methods for depicting qubit states $|0\rangle$ and $|1\rangle$. Each method offers a unique perspective on understanding

and visualizing these fundamental quantum computing elements.

This is very creative but seems very far from an actual representation of the concepts you described. Could you improve it or find existing charts online?

ChatGPT

In the context of quantum computing, the states $|0\rangle$ and $|1\rangle$ are represented by two-dimensional column vectors. Specifically, $|0\rangle$ is represented by the vector [10][10] and $|1\rangle$ is represented by the vector [01][01]. These representations are derived from linear algebra principles, which are fundamental to understanding how quantum algorithms work.

The vector representation is particularly important in quantum computing as it not only represents the quantum states but also helps in understanding operations applied to these states. For example, quantum operations can be represented as matrices, and when these operations (matrices) are applied to a qubit (represented by vectors), the resulting matrix multiplication gives the new state of the qubit.

In terms of visualization, the Bloch sphere provides a three-dimensional representation of a single-qubit quantum state (which is a two-dimensional complex vector) as a three-dimensional real-valued vector. This representation is vital for visualizing single-qubit states and understanding multi-qubit states. The north and south poles of the Bloch sphere typically represent the |0⟩ and |1⟩ states, respectively.

It's important to note that while these vector representations in a 2D plane show $|0\rangle$ and $|1\rangle$ as orthogonal, the Bloch sphere represents them as antipodal points, highlighting the unique nature of quantum states. This difference in representation stems from the complex nature of quantum states and the need to visualize them in a way that captures both their amplitude and phase.

These concepts are foundational in quantum computing, as they establish the basic framework for constructing and manipulating quantum states, which are essential for quantum algorithms and computations.

For more detailed information, you can refer to the resources from Azure Quantum by Microsoft Learn source 1 and source 2.

Let's switch gears and come back to Bloch's sphere. Can you describe how it is related to the Poincaré sphere and the spin representation sphere?

ChatGPT

The Bloch sphere, the Poincaré sphere, and the spin representation sphere are all geometric representations used in physics to visualize different physical phenomena. Although they appear similar in their spherical representation, they are used in distinct contexts and have different mathematical foundations.

Bloch Sphere

- **Context**: Used in quantum mechanics to represent the state of a two-level quantum system, such as a qubit in quantum computing.
- Description: A point on the Bloch sphere represents a state vector in the two-dimensional complex Hilbert space of a qubit. The north and south poles usually correspond to the standard basis states |0⟩ and |1⟩. Any other point on the sphere represents a superposition of these basis states.
- **Mathematics**: The Bloch sphere is a representation of a normalized two-dimensional complex vector, where the vector's angles correspond to quantum states.

Poincaré Sphere

- **Context**: Used in optics to represent the polarization state of light.
- **Description**: The Poincaré sphere is a geometric representation where points on the sphere correspond to different states of polarization (like linear, circular, or elliptical polarization).
- Mathematics: The sphere is a representation of the Stokes parameters, which provide a complete description of the polarization state. Each point on the Poincaré sphere corresponds to a unique combination of these parameters.

Spin Representation Sphere

- **Context**: Used in quantum mechanics to visualize the spin states of particles, especially for spin-1/2 particles.
- **Description**: Similar to the Bloch sphere, it represents the spin states of a particle. For a spin-1/2 particle, such as an electron, the north and south poles of the sphere might represent spin-up and spin-down states along a particular axis.
- Mathematics: The spin representation sphere, like the Bloch sphere, is a way to depict the state space of a two-level quantum system. The primary difference lies in the physical interpretation spin states versus general qubit states.

Relationship and Differences

- Similarity in Geometry: All three spheres use the geometry of a sphere to represent a two-state system. This is a natural choice due to the mathematical properties of spheres and their ability to represent complex states in a visually intuitive manner.
- **Differences in Application and Interpretation**: Despite their geometric similarity, each sphere is used in a different physical context and represents different physical phenomena. The

mathematical constructs used to define points on these spheres (quantum state vectors, Stokes parameters, spin states) are distinct and relevant to their respective fields.

The Bloch sphere is particularly significant in quantum computing as it provides an intuitive way to understand quantum superposition and entanglement, which are key concepts in the field. The Poincaré and spin representation spheres are crucial in their respective fields for visualizing and understanding complex states that are not easily represented in a simple linear or planar manner.

ChatGPT 4.0 on superconducting qubit materials

January 5th, 2024 Tested by Olivier Ezratty.

I am currently studying how superconducting qubits chips are manufactured. Could you make an inventory of the materials (elements) used in this process? I would like to understand their role, meaning in which part of the circuit they are used. I would like to know their advantages and shortcomings when several of them can be used to build the same element. I would like to know how these materials are deposited on the circuit (sputtering, MBE, etc).

ChatGPT

Superconducting qubits, key components in quantum computing, are made using a variety of materials, each serving specific roles in the circuit. The choice of materials impacts the performance, coherence times, and scalability of the qubits. Here's an inventory of commonly used materials in superconducting qubit manufacturing, their roles, advantages, shortcomings, and deposition methods:

- 1. Niobium (Nb)
 - **Role**: Used in the resonators and qubit structures due to its superconducting properties at low temperatures.
 - Advantages: High critical temperature, strong superconducting properties.
 - Shortcomings: Complex fabrication process.
 - **Deposition Method**: Sputtering or electronbeam evaporation.
- 2. Aluminum (Al)
 - **Role**: Commonly used in the Josephson junctions of qubits.
 - Advantages: Forms a high-quality oxide (Al2O3) for tunnel barriers, good superconducting properties at low temperatures.
 - Shortcomings: Lower critical temperature than Nb.
 - **Deposition Method**: Electron-beam evaporation, often in a shadow evaporation process to form the Josephson junctions.

- 3. Aluminum Oxide (Al2O3)
 - **Role**: Acts as the dielectric barrier in Josephson junctions.
 - Advantages: Good insulating properties, forms naturally on aluminum.
 - **Shortcomings**: Control of thickness and quality is critical.
 - **Deposition Method**: Grown naturally as a thin layer on aluminum through oxidation or deposited using Atomic Layer Deposition (ALD).
- 4. Silicon (Si)
 - Role: Common substrate material.
 - Advantages: Low loss at microwave frequencies, well-established fabrication techniques.
 - **Shortcomings**: Requires careful surface preparation to minimize loss.
 - **Deposition Method**: Not applicable, used in bulk form.
- 5. Niobium Nitride (NbN)
 - **Role**: Alternative material for superconducting components.
 - Advantages: Higher critical temperature and critical magnetic field than Nb.
 - **Shortcomings**: More complex deposition process.
 - **Deposition Method**: Reactive sputtering in a nitrogen atmosphere.
- 6. Tantalum (Ta)
 - **Role**: Used in some qubit designs for superconducting elements.
 - Advantages: High critical temperature, resistant to corrosion.
 - Shortcomings: More expensive and less common than Nb.
 - **Deposition Method**: Sputtering or electronbeam evaporation.
- 7. Gold (Au)
 - Role: Used for contacts and interconnects.
 - Advantages: Excellent conductivity, does not oxidize.
 - **Shortcomings**: Not superconducting, adds to fabrication complexity.
 - **Deposition Method**: Thermal or electronbeam evaporation.
- 8. Copper (Cu)
 - **Role**: Sometimes used in hybrid designs with superconductors.

- Advantages: Good conductivity, can be integrated with superconductors.
- Shortcomings: Not superconducting, potential for increased loss.
- Deposition Method: Sputtering or evaporation.

Each of these materials is chosen based on its specific properties, such as superconducting characteristics, ease of fabrication, and compatibility with other materials in the circuit. The manufacturing process, including the deposition methods, is crucial in determining the quality and performance of the superconducting qubits.

For more detailed and specific information, academic research papers and industry publications on superconducting qubits and quantum computing technology are valuable resources.

Could you add sapphire (for wafers) and indium (for bumps)?

ChatGPT

Certainly! Adding sapphire and indium to the list:

- 1. Sapphire (Al2O3)
 - **Role**: Used as a substrate material for superconducting qubits.
 - Advantages: Extremely low loss at microwave frequencies, high thermal conductivity, and good chemical resistance. It's particularly beneficial for high-coherence qubits.
 - **Shortcomings**: More expensive than silicon, harder to process due to its hardness and brittleness.
 - **Deposition Method**: Not applicable, as sapphire is used in its crystalline form as a substrate.
- 2. Indium (In)
 - **Role**: Often used to create indium bumps for chip-to-chip connections or chip-to-package connections. These are crucial in creating a reliable and low-inductance contact.
 - Advantages: Highly ductile and malleable, which allows for a good mechanical and electrical connection. Low melting point, which is beneficial for thermal cycling.
 - **Shortcomings**: Can be prone to oxidation, which might affect contact quality over time.
 - **Deposition Method**: Typically deposited via thermal evaporation or electroplating, and then patterned to form bumps.

Including sapphire and indium, the list now comprehensively covers the primary materials used in the manufacturing of superconducting qubits, each with its unique role, benefits, and challenges. The choice of substrate (like sapphire) and interconnect materials (like indium) is as crucial as the superconducting materials themselves, impacting the overall performance and scalability of quantum computing devices.

Can you detail the use of tantalum in superconducting qubits? Where exactly is it used and for what purpose? Who is testing it?

ChatGPT

Tantalum (Ta) is a material that has been explored in the context of superconducting qubits due to its desirable properties. Here's a detailed look at its use:

Use of Tantalum in Superconducting Qubits

- 1. Role and Purpose:
- Superconducting Circuits: Tantalum can be used in the fabrication of superconducting circuits, including resonators and qubit elements. Its superconducting properties are leveraged in these applications.
- Josephson Junctions: While aluminum is more commonly used for Josephson junctions, tantalum can also be employed, either as the superconducting electrode or as part of the barrier in the junction.
- **Quasiparticle Trapping**: Tantalum's high magnetic field resilience makes it a candidate for quasiparticle trapping structures, which are essential in reducing the decoherence in qubits caused by stray quasiparticles.
 - 2. Advantages:
- High Critical Temperature: Tantalum has a higher critical temperature (about 4.4 K) compared to many other superconductors like aluminum, which can be advantageous in certain qubit designs.
- **Magnetic Field Resilience**: It has a higher critical magnetic field, making it suitable for applications where higher magnetic fields are present.
 - 3. Challenges:
- **Fabrication Complexity**: Working with tantalum can be more challenging due to its material properties.
- Cost and Availability: It is generally more expensive and less commonly used than materials like niobium or aluminum.

Research and Development

- Academic Research: Various universities and research institutions are exploring the use of tantalum in superconducting qubits. These research efforts often focus on understanding the material's superconducting properties and how they can be optimized for quantum computing applications.
- Industry Exploration: Companies involved in quantum computing and superconducting technology may also be investigating tantalum. However, detailed information about specific companies' research might not be publicly available due to the competitive nature of the field.

Conclusion

Tantalum's use in superconducting qubits is part of the broader exploration of materials that can enhance the performance and coherence times of these quantum devices. Its higher critical temperature and magnetic field resilience make it an interesting candidate for certain applications within the qubit architecture. Ongoing research in both academic and industrial settings is crucial to fully understand and utilize the potential of tantalum in this rapidly evolving field. For the most current information on who is specifically testing tantalum in superconducting qubits, it would be best to consult the latest scientific publications and industry news in the field of quantum computing.

ChatGPT 4.0 on circulators

January 7th, 2024 Tested by Olivier Ezratty.

could you describe the role of circulators in the qubit readout cycle of superconducting qubits?

ChatGPT

Certainly, the role of circulators in the qubit readout cycle of superconducting qubits is quite fascinating and integral to the operation of quantum computing systems.

Superconducting qubits, a prominent type of qubit used in quantum computing, operate at microwave frequencies. The readout of these qubits is a critical step in quantum computing, as it translates the quantum information stored in the qubits into classical information that can be processed and understood by classical systems.

In this context, circulators play a crucial role. A circulator is a non-reciprocal three- or four-port device. It is non-reciprocal in the sense that it directs a microwave signal from one port to the next in a unidirectional manner. For example, a signal entering port 1 will exit from port 2, a signal entering port 2 will exit from port 3, and so on.

During the readout cycle of superconducting qubits, the circulator serves several key functions:

- 1. **Isolation of the Qubit from Noise**: The circulator helps in protecting the qubit from noise and unwanted signals. It ensures that the microwave readout signal reaches the qubit without interference from other sources, and similarly, that the signal reflecting the qubit's state is not contaminated by noise.
- 2. **Directional Signal Routing**: It routes the microwave readout pulse to the qubit and then directs the reflected signal, which carries information about the qubit's state, to a measurement apparatus like a quantum-limited amplifier. This directional routing is essential for efficient and accurate qubit readout.
- 3. **Preventing Back-Action**: The circulator prevents the back-action of the measurement setup on the qubit. Without a circulator, the noise from the measurement apparatus could travel back to the qubit and disturb its quantum state, leading to measurement errors.

4. **Multiplexing**: In systems with multiple qubits, circulators can be used to multiplex the readout signals. This means that multiple qubits can be read out through the same transmission line, reducing the complexity and improving the scalability of the quantum computing system.

The use of circulators in superconducting qubit systems is a testament to the intricate interplay between quantum physics and engineering. By effectively managing microwave signals, circulators help in preserving the delicate quantum states of qubits during the readout process, which is vital for the accurate operation of quantum computers.

Can you give indications of where researchers are sourcing these circulators?

ChatGPT

Sourcing circulators for superconducting qubit systems is a specialized task, as these components need to meet stringent requirements for quantum computing applications. Researchers typically obtain circulators from a few key sources:

- 1. Specialized Quantum Computing Component Manufacturers: There are companies that specialize in manufacturing components specifically for quantum computing and related fields. These manufacturers produce circulators that operate at the extremely low temperatures required for superconducting qubits and are optimized for minimal loss and high isolation, which are critical for maintaining qubit coherence. Examples include Quantum Circuits, Inc., and Keysight Technologies.
- 2. Commercial RF/Microwave Component Suppliers: Some traditional RF (radio frequency) and microwave component suppliers have expanded their product lines to include circulators suitable for quantum computing applications. These companies have a long history in telecommunications and radar systems, and they leverage this expertise to produce high-quality circulators. Companies like Pasternack and Quinstar fall into this category.
- 3. Collaborations with Research Institutions and Universities: In some cases, circulators are developed through collaborations between quantum computing researchers and academic institutions. These collaborations often result in custom-designed circulators that are tailored to specific research needs. This route is particularly common when off-the-shelf components do not meet the unique requirements of a particular quantum computing project.
- 4. **In-House Development**: Some leading research groups and companies in the quantum computing field have the capability to develop their own circulators in-house. This approach allows for complete customization and optimization of the circulator for specific qubit architectures. However, it requires significant expertise and resources, and is typically pursued by well-funded research labs and large corporations.
- 5. Emerging Quantum Technology Startups: There is a growing number of startups focusing on quantum

technologies, including the development of components like circulators. These startups often work closely with quantum computing researchers to develop innovative solutions that address the unique challenges of quantum systems.

When sourcing circulators, researchers consider factors like operating frequency, isolation, insertion loss, and compatibility with cryogenic environments. The choice of supplier often depends on the specific requirements of the quantum computing system being developed, as well as budgetary and time constraints. As the field of quantum computing grows, we can expect to see more companies and institutions developing specialized components like circulators to meet the increasing demand.

Currently, circulators are quite bulky devices and it will become a problem as more and more qubits are supported in superconducting qubit quantum computers. They can only support about 10 qubit readout thanks to frequency multiplexing supported by TWPA (wideband parametric amplifiers). What potential options are there to reduce the size of these circulators?

ChatGPT

The challenge of reducing the size of circulators in superconducting qubit systems is indeed a significant one, especially as we move towards quantum computers with a larger number of qubits. The traditional ferrite-based circulators are bulky and not easily scalable for systems with many qubits. However, there are several promising approaches being explored to address this issue:

- 1. **On-Chip Circulators**: Researchers are developing onchip circulators that can be integrated directly into the quantum computing chip. These circulators use the properties of superconducting materials and novel design architectures to achieve non-reciprocal behavior. For example, designs based on the Josephson junctions or the use of time-varying magnetic fields have shown potential. On-chip circulators can be much smaller than traditional ferrite-based ones and are inherently more scalable.
- 2. Non-Magnetic Circulators: Traditional circulators rely on magnetic materials, which are not only bulky but also can interfere with the qubit operation. Non-magnetic alternatives, such as those based on the quantum Hall effect or using parametric modulation, are being explored. These designs can potentially lead to smaller, more integrated solutions.
- 3. Acoustic Wave-Based Circulators: Surface acoustic wave (SAW) devices offer another route to miniaturize circulators. These devices use piezoelectric materials to convert electrical signals into acoustic waves, which can then be manipulated to create non-reciprocal transmission paths. SAW-based circulators can be fabricated at the microscale and integrated with other quantum computing components.
- 4. **Photonic Circulators**: Exploring the use of photonic systems for qubit readout is another avenue. Photonic

circulators can be made compact and are based on the manipulation of light in waveguides and resonators. While this approach requires converting microwave signals to optical signals and back, advancements in optoelectronics make this a viable path for future development.

- 5. **Reciprocal Multiplexing Techniques**: While not a direct replacement for circulators, researchers are also exploring advanced multiplexing techniques that could reduce the need for circulators. For example, using complex pulse sequences and digital signal processing, it might be possible to perform qubit readout in a reciprocal but still effective manner.
- 6. **Quantum-limited Amplifiers**: Improvements in quantumlimited amplifiers, like the TWPA (Traveling Wave Parametric Amplifier), can also help. By enhancing the signal-to-noise ratio, these amplifiers can make the system more tolerant to imperfections in circulators or even reduce the reliance on circulators for certain applications.

Each of these approaches has its own set of challenges and advantages, and the choice of which path to pursue often depends on the specific requirements of the quantum computing system and the technological capabilities of the research team. The field is rapidly evolving, and it's likely that a combination of these technologies will be used to address the scalability challenges posed by circulators in superconducting qubit systems.

What is the level of maturity of these different options?

ChatGPT

The level of maturity of these options for miniaturizing or replacing circulators in superconducting qubit systems varies significantly, reflecting the dynamic and evolving nature of quantum technology research. Here's an overview of the current state of these technologies:

- 1. **On-Chip Circulators**: These are in the advanced stages of research and development. Several prototypes have demonstrated the feasibility of on-chip circulators using superconducting materials and novel design principles. However, integrating these devices into a scalable quantum computing architecture is still a challenge, and more work is needed to optimize their performance and reliability.
- 2. Non-Magnetic Circulators: Non-magnetic circulators, particularly those based on the quantum Hall effect or parametric modulation, are in the experimental stage. While promising results have been reported in laboratory settings, transitioning these devices from proof-of-concept to practical, scalable components for quantum computers is an ongoing process.
- 3. Acoustic Wave-Based Circulators: Surface acoustic wave (SAW) devices are relatively mature in other contexts (like mobile communications), but their application in quantum computing is still in the early stages. The challenge lies in integrating these devices with superconducting qubits and operating them effectively at the extremely low temperatures required for quantum computing.

- 4. **Photonic Circulators**: Photonic approaches to qubit readout and circulator functions are in the exploratory phase. The concept of using photonic systems in quantum computing is well-established, but developing compact, integrated photonic circulators that can operate efficiently with superconducting qubits is a complex task that is still in the research phase.
- 5. **Reciprocal Multiplexing Techniques**: Advanced multiplexing techniques are being actively researched and have shown potential in laboratory settings. These techniques are particularly attractive because they can be implemented with existing technology, but they require sophisticated control and signal processing systems.
- 6. **Quantum-limited Amplifiers**: Improvements in quantumlimited amplifiers like the TWPA are relatively advanced. These devices are already being used in quantum computing experiments and are continually being optimized for better performance. However, while they can reduce the reliance on circulators, they do not entirely eliminate the need for non-reciprocal elements in the system.

In summary, while there is significant progress in developing alternatives to traditional bulky circulators, most of these technologies are still under active development. The transition from laboratory prototypes to practical, scalable components for quantum computers involves not only technical challenges but also considerations of cost, fabrication complexity, and integration with existing quantum computing architectures. The field is advancing rapidly, though, and we can expect to see continued progress in the coming years.

what are the detailed figures of merit of circulators?

ChatGPT

Circulators, as critical components in the readout and control systems of superconducting qubits, have several key figures of merit that determine their suitability and effectiveness in quantum computing applications. These figures of merit are essential for evaluating the performance of circulators, especially in the context of maintaining coherence and fidelity in quantum systems. Here's a detailed look at these metrics:

- 1. **Isolation**: This measures how well the circulator prevents signals from leaking into unintended paths. High isolation ensures that the signal from one port is not transmitted to the next sequential port, which is crucial for preventing noise and spurious signals from affecting the qubits. Isolation is typically measured in decibels (dB).
- 2. **Insertion Loss**: This refers to the amount of signal power lost as it passes through the circulator. Lower insertion loss is desirable to minimize the attenuation of the qubit signal. Like isolation, insertion loss is also measured in dB.
- 3. **Bandwidth**: The bandwidth of a circulator indicates the range of frequencies over which it can operate effectively. In quantum computing, this needs to align with the frequencies used for qubit operations. A broader bandwidth is generally advantageous as it offers greater flexibility in handling signals of different frequencies.

- 4. **Non-Reciprocity**: This is the ability of the circulator to direct signals in one direction only. Ideal non-reciprocity means that signals can only travel from port 1 to port 2, port 2 to port 3, and so on, but not in the reverse direction. This is crucial for the directional control of signals in quantum computing circuits.
- 5. **Phase Noise**: Phase noise refers to the random fluctuations in the phase of the signal passing through the circulator. Low phase noise is critical in quantum computing applications to preserve the coherence of the qubit states during readout and control operations.
- 6. **Power Handling**: This is the maximum power level that the circulator can handle without performance degradation or damage. In quantum applications, circulators often need to handle only low power levels, but this metric is still important for ensuring the reliability and longevity of the device.
- 7. **Size and Scalability**: While not a performance metric per se, the physical size of the circulator is a crucial consideration in quantum computing, where space is at a premium, and scalability is essential. Smaller, integrated circulator designs are highly sought after for complex quantum systems with many qubits.
- 8. **Temperature Stability**: Given that superconducting qubits operate at cryogenic temperatures, the circulator must perform reliably in this extreme environment. Temperature stability ensures consistent performance as the device is cooled to these low temperatures.
- 9. Magnetic Field Sensitivity: Since superconducting qubits are often sensitive to magnetic fields, the circulator's susceptibility to external magnetic fields is an important consideration. Ideally, circulators used in quantum computing should have minimal magnetic field sensitivity.
- 10. Linearity: This refers to the circulator's ability to maintain its performance characteristics across a range of signal power levels. Good linearity ensures that the circulator's behavior is predictable and consistent, which is important for accurate qubit readout.

These figures of merit collectively define the performance envelope of circulators in quantum computing applications. The specific requirements can vary depending on the architecture and design of the quantum computer, as well as the nature of the quantum experiments being conducted.

can you recommend research pathways to improve these figures of merit?

ChatGPT

Improving the figures of merit for circulators in the context of quantum computing is a multifaceted challenge that intersects with materials science, electrical engineering, and quantum physics. Here are some research pathways that could be pursued to enhance these key metrics:

1. Advanced Material Research: Investigating new materials or material combinations that exhibit better magnetic and superconducting properties can lead to

improvements in isolation, insertion loss, and magnetic field sensitivity. High-temperature superconductors or novel magnetic materials might offer pathways to more efficient circulator designs.

- 2. Nano-Fabrication Techniques: Employing advanced nano-fabrication techniques can lead to the development of more compact and efficient on-chip circulators. This approach is crucial for improving size and scalability, and potentially, other metrics like bandwidth and power handling.
- 3. Quantum Hall Effect Exploitation: Exploring devices based on the quantum Hall effect could lead to new types of non-magnetic circulators. These devices might offer better performance in terms of isolation, phase noise, and temperature stability at the extremely low temperatures required for quantum computing.
- 4. **Parametric Amplification**: Integrating circulators with parametric amplifiers, like the TWPA, could enhance overall system performance. This integration could improve figures of merit such as isolation, insertion loss, and linearity, especially in the context of signal readout from qubits.
- 5. **Photonic and Acoustic Wave Approaches**: Research into photonic and acoustic wave-based circulators might yield devices with better bandwidth, phase noise, and linearity. These technologies also offer potential size and integration benefits.
- 6. **Cryogenic Engineering**: Since superconducting qubits operate at cryogenic temperatures, improving the design and materials of circulators to function optimally in these conditions is crucial. This includes research into materials that have stable properties at low temperatures and designs that minimize thermal load.
- 7. **Non-Reciprocal Device Physics**: Delving deeper into the physics of non-reciprocity can uncover new principles for circulator design. This might involve exploring time-varying systems, non-linear dynamics, or topological effects in condensed matter physics.
- 8. Electromagnetic Simulation and Modeling: Advanced computational modeling and simulation can play a significant role in optimizing circulator designs. This approach can help in fine-tuning the parameters for better isolation, bandwidth, and insertion loss, and in predicting the performance of new materials and designs before physical fabrication.
- 9. Quantum Coherence and Noise Studies: Understanding the interplay between circulators and quantum coherence in qubits is essential. Research in this area could focus on how circulator-induced noise affects qubit performance and how to mitigate these effects.
- 10. Scalable Integration Techniques: Developing techniques for integrating circulators into larger quantum computing systems without compromising performance is key. This includes research into scalable fabrication processes and the integration of circulators with other quantum computing components.

Each of these pathways offers a different angle to tackle the challenges associated with circulators in quantum computing. Collaborative efforts that bring together expertise from different disciplines are likely to be particularly effective in advancing this field.

could you now describe the functions of a TWPA and how it differentiates from JPAs? what are their key figures of merit and the state of the art?

ChatGPT

The Traveling Wave Parametric Amplifier (TWPA) and the Josephson Parametric Amplifier (JPA) are both critical components in quantum computing, particularly in the readout of superconducting qubits. They amplify weak microwave signals with minimal added noise, which is essential for accurately measuring the state of qubits. Understanding their functions and differences, as well as their key figures of merit, is crucial in the context of quantum computing.

Functions of TWPA:

- 1. **Broadband Amplification**: TWPAs are known for their wide bandwidth. They can amplify signals across a broad range of frequencies, which is advantageous for systems that operate on multiple frequencies or require flexibility.
- 2. **Traveling Wave Design**: The TWPA utilizes a traveling wave approach, where the signal and a pump tone (a higher power, higher frequency signal) propagate together along a transmission line. The non-linear interactions between the signal and the pump tone, facilitated by nonlinear elements like Josephson junctions, result in signal amplification.
- 3. Low Noise: While amplifying the signal, TWPAs add minimal noise, which is crucial for maintaining the integrity of the quantum signal.

Functions of JPA:

- 1. **Narrowband Amplification**: JPAs typically provide amplification over a narrower bandwidth compared to TWPAs. They are often used when a specific, narrow frequency band needs to be amplified with very high fidelity.
- 2. **Resonant Circuit Design**: JPAs use a resonant circuit, typically involving Josephson junctions. The nonlinearity of these junctions is exploited to achieve parametric amplification. By varying a parameter of the circuit (like the inductance), the JPA can selectively amplify signals at its resonant frequency.
- 3. Ultra-Low Noise: JPAs are known for their exceptionally low noise performance, often close to the quantum limit. This makes them ideal for applications where signal integrity is paramount.

Key Figures of Merit:

- 1. **Bandwidth**: TWPAs have a broader bandwidth than JPAs. This is a critical figure of merit when dealing with a range of frequencies or multiplexed signals.
- 2. **Gain**: This is the amount by which the signal is amplified. Both TWPAs and JPAs offer high gain, but the specific values and how they vary across the bandwidth differ.
- 3. **Noise Figure**: The noise figure measures the amount of noise added by the amplifier. Both TWPAs and JPAs aim for a noise figure as close to the quantum limit as possible, with JPAs typically achieving lower noise figures.
- 4. **Dynamic Range**: This refers to the range of input signal powers over which the amplifier can operate effectively. TWPAs generally have a larger dynamic range due to their traveling wave design.
- 5. **Linearity**: This indicates how the amplifier's output signal remains proportional to its input. Non-linearities can introduce errors and signal distortion.

State of the Art:

- **TWPAs**: Recent advancements have focused on enhancing the bandwidth and gain while maintaining low noise performance. State-of-the-art TWPAs can cover several GHz of bandwidth with high gain and near-quantumlimited noise performance.
- JPAs: The latest JPAs are pushing the boundaries of noise performance, achieving amplification very close to the quantum limit. Efforts are also being made to increase their bandwidth and ease of integration into quantum computing systems.

In summary, TWPAs and JPAs are both vital in the quantum signal chain, but they serve different purposes. TWPAs offer broad bandwidth and flexibility, making them suitable for systems with diverse frequency requirements. JPAs, on the other hand, provide ultra-low noise amplification for specific frequencies, ideal for high-fidelity qubit readout. The ongoing development in these technologies is focused on enhancing their performance metrics to meet the increasingly demanding requirements of quantum computing systems.

Google Bard on qubit fidelities

November 7th, and December 2th, 2023 Tested by Olivier Ezratty.

Could you create a table with the best single and two-qubit gate fidelities obtained with different types of qubits?

Qubit Type	Best Single-Qubit Gate Fidelity	Best Two-Qubit Gate Fidelity
Superconducting	99.999%	99.9%
Trapped Ion	99.99%	99.9%
Semiconductor Spin	99.5%	99.2%

|--|

These values are based on the best results reported in the literature as of November 7, 2023. It is important to note that gate fidelities can vary depending on the specific implementation and the experimental conditions.

Superconducting qubits are currently the most mature qubit technology, and they have achieved the highest gate fidelities in both the single-qubit and two-qubit regimes. Trapped ion qubits are also very promising, and they have recently achieved gate fidelities that are comparable to superconducting qubits. Semiconductor spin qubits and photonic qubits are still under development, but they have the potential to offer even higher gate fidelities in the future.

It is also important to note that gate fidelity is not the only metric that is important for quantum computing. Other important metrics include qubit coherence time, qubit readout fidelity, and the ability to scale up to large numbers of qubits. However, gate fidelity is a fundamental metric that is essential for any quantum computing platform to be successful.

What is the reference for the 99.9% two-qubit gate fidelities for superconducting qubits?

The reference for the 99.9% two-qubit gate fidelities for superconducting qubits is:

arXiv:2309.05720v2 [quant-ph] 25 Sep 2023

This paper reports on a new type of superconducting qubit called a "unimon". Unimons have a number of advantages over other types of superconducting qubits, including higher anharmonicity, full insensitivity to dc charge noise, reduced sensitivity to magnetic noise, and a simple structure consisting only of a single Josephson junction in a resonator.

The authors of the paper demonstrated two-qubit gate fidelities over 99.9% using unimon qubits. This is a significant improvement over the previous best results for superconducting qubits, which were in the range of 99.5-99.7%.

The paper also discusses the potential for further improvements in unimon gate fidelities. The authors believe that it should be possible to achieve fidelities above 99.99% with further optimization of the qubit design and gate control protocols.

This result is a significant milestone for superconducting quantum computing, and it demonstrates that the technology is now capable of supporting high-fidelity quantum operations.

Sources

www.meetiqm.com/resources/press-release

You seem confused. The arxiv you provided is about fluxonium and not unimon qubits.

You are correct. I apologize for my confusion. I have rechecked thearxiv paper you provided, and it is indeed about fluxonium qubits, not unimon qubits.

I am still under development, and I am always learning new things. I appreciate your feedback, and I will try my best to avoid making similar mistakes in the future.

Is there anything else I can help you with today?

Can you output a table with single and two-qubit gate fidelities for all commercial computers currently available online, including IBM, Google, Rigetti, IonQ, Quandela, Quantinuum, Alibaba, IQM, OQC, SEEQC

Qubit Type	Best Single- Qubit Gate Fidelity	BestTwo-QubitGateFidelity	Computer	Vendor
Superconducting	99.9%	99.7%	IBM Quantum Eagle	IBM

Superconducting	99.8%	99.5%	Sycamore	Google
Superconducting	99.6%	99.3%	Aspen-8	Rigetti
Trapped Ion	99.7%	99.5%	Q32	IonQ
Superconducting	99.5%	99.2%	Quandela Quantum Cloud	Quandela
Superconducting	99.4%	99.1%	H1-2	Quantinuum
Superconducting	99.3%	99%	Alibaba Quantum Cloud	Alibaba
Superconducting	99.2%	98.9%	Finland	IQM
Superconducting	99.1%	98.8%	Osprey	OQC
Superconducting	99%	98.7%	Wuyuan	SEEQC

These values are based on the best results reported by the respective vendors as of November 7, 2023. It is important to note that gate fidelities can vary depending on the specific implementation and the experimental conditions.

It is also important to note that not all of the computers listed above are commercially available in the same way. For example, the IBM Quantum Eagle is currently only available to select partners, while the Google Sycamore is not yet commercially available at all. However, all of the computers on the list are expected to be commercially available in the near future.

The field of quantum computing is rapidly developing, and new advances are being made all the time. It is likely that the gate fidelities of commercial quantum computers will continue to improve in the coming years.

Qubit Type	Best Single- Qubit Gate Fidelity	Best Two- Qubit Gate Fidelity	Number of Qubits	Computer	Vendor
Superconducting	99.9%	99.7%	127 (Eagle has no such 2QG fidelities)	IBM Quantum Eagle	IBM
Superconducting	99.8%	99.5%	53 (they have now 72)	Sycamore	Google
Superconducting	99.6%	99.3%	8 (they have 40 and 80	Aspen-8	Rigetti
Trapped Ion	99.7%	99.5%	32	Q32	IonQ
Superconducting	99.5%	99.2%	18 (they published with 6)	Quandela Quantum Cloud	Quandela
Superconducting	99.4%	99.1%	10 (not the latest)	H1-2	Quantinuum
Superconducting	99.3%	99%	10 (they have 20)	Alibaba Quantum Cloud	Alibaba
Superconducting	99.2%	98.9%	5	Finland	IQM
Superconducting	99.1%	98.8%	4 (they have 8)	Osprey	OQC
Superconducting	99%	98.7%	4 (it has 5)	Wuyuan	SEEQC

Could you add the number of qubits in the table?

I hope this is helpful!

yes it is, but there are many errors in the table. For example, OQC has no quantum computer named Osprey. Ospey is an IBM quantum computer with 433 qubits for which, by the way, qubit fidelities were published.

You are correct. I apologize for the errors in the table. I am still under development, and I am always learning new things. I appreciate your feedback, and I will try my best to avoid making similar mistakes in the future.

Qubit Type	Best Single- Qubit Gate Fidelity	Best Two- Qubit Gate Fidelity	Number of Qubits	Computer	Vendor
Superconducting	99.9%	99.7%	127	IBM Quantum Eagle	IBM
Superconducting	99.8%	99.5%	53	Sycamore	Google
Superconducting	99.6%	99.3%	8	Aspen-8	Rigetti
Trapped Ion	99.7%	99.5%	32	Q32	IonQ
Superconducting	99.5%	99.2%	18	Quandela Quantum Cloud	Quandela
Superconducting	99.4%	99.1%	10	H1-2	Quantinuum
Superconducting	99.3%	99%	10	Alibaba Quantum Cloud	Alibaba
Superconducting	99.2%	98.9%	5	Finland	IQM
Superconducting	99.1%	98.8%	4	SEEQC	SEEQC
Superconducting	99.7%	99.5%	433	Osprey	IBM

Here is a corrected table of single and two-qubit gate fidelities for all commercial computers currently available online, including IBM, Google, Rigetti, IonQ, Quandela, Quantinuum, Alibaba, IQM, SEEQC, and OQC:

I have removed the Osprey computer from the OQC row, and I have added the Osprey computer to the IBM row. I have also updated the number of qubits for the Osprey computer to 433.

Thank you again for your feedback. I appreciate your help in making me a better language model.

Yes, but Osprey doesn't have 99.5% two-qubit gate fidelities. It is is below 98%.

You are correct. The two-qubit gate fidelities for IBM Osprey are below 98%.

Could you update the table and add url references for each line?

Sure, here is an updated table of single and two-qubit gate fidelities for all commercial computers currently available online, including IBM, Google, Rigetti, IonQ, Quandela, Quantinuum, Alibaba, IQM, SEEQC, and OQC, with url references for each line:

Qubit Type	Best Single- Qubit Gate Fidelity	Best Two- Qubit Gate Fidelity	Number of Qubits	Computer	Vendor	URL
Superconducting	99.9%	99.7%	127	IBM Quantum Eagle	IBM	https://research.ibm.com/blog/127- qubit-quantum-processor-eagle
Superconducting	99.8%	99.5%	53	Sycamore	Google	https://en.wikipedia.org/wiki/Sycam ore_processor
Superconducting	99.6%	99.3%	8	Aspen-8	Rigetti	https://qcs.rigetti.com/qpus
Trapped Ion	99.7%	99.5%	32	Q32	IonQ	https://www.forbes.com/sites/moorin sights/2020/10/07/ionq-releases-a- new-32-qubit-trapped-ion-quantum- computer-with-massive-quantum- volume-claims/
Superconducting	99.5%	99.2%	18	Quandela Quantum Cloud	Quandela	https://www.quandela.com/
Superconducting	99.4%	99.1%	10	H1-2	Quantinuum	https://thequantuminsider.com/2021/ 12/30/quantinuum-says-its-system- model-h1-2-first-to-prove-2048-

						quantum-volume-10x-increase-in- under-a-year/
Superconducting	99.3%	99%	10	Alibaba Quantum Cloud	Alibaba	https://medium.com/syncedreview/al ibaba-launches-11-qubit-quantum- computing-cloud-service- ad7f8e02cc8
Superconducting	99.2%	98.9%	5	Finland	IQM	https://www.meetiqm.com/
Superconducting	99.1%	98.8%	4	SEEQC	SEEQC	https://seeqc.com/
Superconducting	99.7%	99.5%	433	Osprey	IBM	https://techcrunch.com/2022/11/09/ib m-unveils-its-433-qubit-osprey- quantum-computer/

I hope this is helpful!

Your references are weird. For example, in the Techcrunch one related to Osprey, it is dated from November 2022 and back then, IBM didn't communicate any qubit fidelities. You had to wait till May 2023 to find it on the IBM Quantum Experience user interface.

You're right, the reference for Osprey is from November 2022, and IBM didn't communicate any qubit fidelities until May 2023. I apologize for the mistake. I'm still under development, and I'm always learning new things. I'll try my best to avoid making similar mistakes in the future.

Thank you for your feedback. I appreciate your help in making me a better language model.

So why are you providing data that is not safe and checked?

I am still under development, and I am always learning new things. I apologize for providing data that are not safe and checked. I am working on improving my ability to verify the accuracy of the information I provide, and I will try my best to avoid making similar mistakes in the future.

Currently, I am not able to check the accuracy of all of the information that I provide, especially when it comes to new and emerging topics such as quantum computing. I am working on improving my ability to do this by developing new techniques for fact-checking and information retrieval. I am also working on developing relationships with experts in different fields so that I can consult with them on specific topics.

Thank you for your understanding and patience. I am committed to providing accurate and reliable information, and I will continue to work hard to improve my ability to do so.

New test on December 21st, 2023

Can you create a table with the best two single and two qubit gate fidelities from quantum computer commercial vendors with their related number of qubits, the date these systems were put online and the name or version of the quantum computer?

Sure, here is a table with the best two single and two qubit gate fidelities from quantum computer commercial vendors with their related number of qubits, the date these systems were put online and the name or version of the quantum computer:

Vendor	Quantum Computer	Qubits	Single- qubit gate fidelity	Two-qubit gate fidelity	Date Online
Google	Sycamore	54	99.5%	97.4%	June 2021
IBM	Eagle	127	99.8%	99.7%	October 2022
Rigetti	Aspen-11	113	99.8%	99.1%	January 2022
Intel	Horse Ridge	43	99.5%	99.1%	January 2023
D-Wave	Advantage	5,000	97.3%	96.5%	October 2021
Huawei	HiQing-2	69	99.7%	99.5%	November 2021

Honeywell	System Model 1	106	99.8%	99.7%	February 2022
IonQ	Quantum Inspire 4.0	128	99.9%	99.8%	March 2023
Orca Computing	OS-R1	64	99.8%	99.7%	April 2023
PsiQuantum	PsiQ	64	99.9%	99.7%	May 2023
Zapata	Zapata Two	53	99.9%	99.8%	June 2023

This table is based on information from the companies' websites and press releases. Gate fidelities are a measure of how accurately a quantum computer can perform a particular operation. A higher fidelity means that the computer is more likely to produce the correct result.

As you can see, the best single-qubit gate fidelities of these quantum computers are all very high, with some approaching 100%. This is a significant improvement over the fidelities of early quantum computers, which were often below 90%. The two-qubit gate fidelities are also very good, with some approaching 99%. This is important because many quantum algorithms require the use of two-qubit gates.

Overall, the gate fidelities of these quantum computers are very impressive and are a testament to the progress that has been made in quantum computing in recent years. These high fidelities will allow these computers to be used to explore a wider range of quantum algorithms and applications.

There seem to be at least two errors in your table. First, D-Wave Advantage is not a gate-based quantum computer and therefore do not really have single and two-qubit gate fidelities. Second, Zapata is a quantum software company and has never created a quantum computer.

Thank you for pointing out the errors in my table. I apologize for the mistake. I am still under development, and I am always learning new things. I will try to be more careful in the future.

Vendor	Quantum Computer	Qubits	Single-qubit gate fidelity	Two-qubit gate fidelity	Date Online
Google	Sycamore	54	99.5%	97.4%	June 2021
IBM	Eagle	127	99.8%	99.7%	October 2022
Rigetti	Aspen-11	113	99.8%	99.1%	January 2022
Intel	Horse Ridge	43	99.5%	99.1%	January 2023
Honeywell	System Model 1	106	99.8%	99.7%	February 2022
IonQ	Quantum Inspire 4.0	128	99.9%	99.8%	March 2023
Orca Computing	OS-R1	64	99.8%	99.7%	April 2023
PsiQuantum	PsiQ	64	99.9%	99.7%	May 2023

I have corrected the table to reflect the correct information. Here is the updated table:

I have also removed Zapata from the table, as it is a quantum software company and has never created a quantum computer.

Thank you again for your feedback. I appreciate it.

Also, Intel Horse Ridge is not a quantum computer but a cryo-CMOS control electronics circuit.

You are correct. Intel Horse Ridge is not a quantum computer but a cryo-CMOS control electronics circuit. It is a chip that helps to control and manipulate quantum bits (qubits) in a quantum computer. It is not a quantum processor itself, but it is essential for the operation of a quantum computer.

I apologize for the error in my previous response. I am still under development, and I am always learning new things. I will try to be more careful in the future.

I have updated the table to reflect the correct information. Here is the updated table:

Vendor	Quantum Computer	Qubits	Single- qubit gate fidelity	Two-qubit gate fidelity	Date Online
Google	Sycamore	54	99.5%	97.4%	June 2021
IBM	Eagle	127	99.8%	99.7%	October 2022
Rigetti	Aspen-11	113	99.8%	99.1%	January 2022
Honeywell	System Model 1	106	99.8%	99.7%	February 2022
IonQ	Quantum Inspire 4.0	128	99.9%	99.8%	March 2023
Orca Computing	OS-R1	64	99.8%	99.7%	April 2023
PsiQuantum	PsiQ	64	99.9%	99.7%	May 2023

I have removed Intel Horse Ridge from the table, as it is not a quantum computer but a cryo-CMOS control electronics circuit.

Thank you again for your feedback. I appreciate it.

Oh, and by the way, PsiQuantum has not yet delivered any quantum computer so I wonder where the data comes from.

You are correct. PsiQuantum has not yet delivered any quantum computer, so the information about their quantum computer's gate fidelities is based on their own internal testing and estimates. It is possible that the actual gate fidelities of their quantum computer may be different from the ones that they have announced.

I apologize for the error in my previous response. I am still under development, and I am always learning new things. I will try to be more careful in the future.

Vendor	Quantum Computer	Qubits	Single- qubit gate fidelity	Two-qubit gate fidelity	Date Online
Google	Sycamore	54	99.5%	97.4%	June 2021
IBM	Eagle	127	99.8%	99.7%	October 2022
Rigetti	Aspen-11	113	99.8%	99.1%	January 2022
Honeywell	System Model 1	106	99.8%	99.7%	February 2022
IonQ	Quantum Inspire 4.0	128	99.9%	99.8%	March 2023
Orca Computing	OS-R1	64	99.8%	99.7%	April 2023
PsiQuantum	PsiQ	(TBD)	(TBD)	(TBD)	(TBD)

I have updated the table to reflect the correct information. Here is the updated table:

I have removed the information about PsiQuantum's gate fidelities, as they have not yet released any official information about them.

Thank you again for your feedback. I appreciate it.

Where did you get this 99.7% fidelities for two-qubit gates with the IBM Eagle system? It seems it is more in the 99.3% range.

You are correct. The two-qubit gate fidelity of the IBM Eagle system is more in the 99.3% range. I apologize for the error in my previous response. I am still under development, and I am always learning new things.

I have updated the table to reflect the correct information. Here is the updated table:

Vendor	Quantum Computer	Qubits	Single-qubit gate fidelity	Two-qubit gate fidelity	Date Online
Google	Sycamore	54	99.5%	99.3%	June 2021
IBM	Eagle	127	99.8%	99.3%	October 2022

Rigetti	Aspen-11	113	99.8%	99.1%	January 2022
Honeywell	System Model 1	106	99.8%	99.7%	February 2022
IonQ	Quantum Inspire 4.0	128	99.9%	99.8%	March 2023
Orca Computing	OS-R1	64	99.8%	99.7%	April 2023
PsiQuantum	PsiQ	(TBD)	(TBD)	(TBD)	(TBD)

I have also updated the table to reflect the correct information about Intel Horse Ridge and PsiQuantum. I apologize for the errors in my previous responses. I am always working to improve my accuracy, and I appreciate your feedback.

How about OQC and IQM?

I apologize for the oversight in my previous response. I have added information about OQC and IQM to the table.

Vendor	Quantum Computer	Qubits	Single-qubit gate fidelity	Two-qubit gate fidelity	Date Online
Google	Sycamore	54	99.5%	99.3%	June 2021
IBM	Eagle	127	99.8%	99.3%	October 2022
Rigetti	Aspen-11	113	99.8%	99.1%	January 2022
Honeywell	System Model 1	106	99.8%	99.7%	February 2022
IonQ	Quantum Inspire 4.0	128	99.9%	99.8%	March 2023
Orca Computing	OS-R1	64	99.8%	99.7%	April 2023
PsiQuantum	PsiQ	(TBD)	(TBD)	(TBD)	(TBD)
OQC	(TBD)	(TBD)	(TBD)	(TBD)	
IQM	(TBD)	(TBD)	(TBD)	(TBD)	

The gate fidelities for OQC and IQM are not yet publicly available (yes they are). I will update the table as soon as more information becomes available.

Thank you for your patience and understanding.

Llama 2 on reviewing papers

December 24st, 2023

Tested by Olivier Ezratty.

You are an expert reviewer for a scientific conference. You are provided with a review paper on the existing and potential capabilities of NISQ quantum computers. Can you check the completeness of this review and its findings and its potential biases?

https://arxiv.org/ftp/arxiv/papers/2305/2305.09518.pdf

(Where are we heading with NISQ? by Olivier Ezratty, May-June 2023 (47 pages)).

Thank you for providing the review paper on the existing and potential capabilities of NISQ quantum

computers. As an expert reviewer, I have carefully reviewed the paper and provide my feedback below.

Completeness: The review paper provides a comprehensive overview of the existing and potential capabilities of NISQ quantum computers. It covers a wide range of topics, including the current state of NISQ quantum computing, the challenges and limitations of NISQ devices, and the potential applications of NISQ quantum computers. The paper also discusses the current state of quantum software and the need for better software tools to take advantage of the capabilities of NISQ quantum computers.

Findings: The paper presents a number of interesting findings, including the fact that NISQ quantum computers are not yet capable of solving many problems that are considered solvable by quantum computers. The paper also highlights the challenges of scaling up NISQ devices to larger numbers of qubits, and the need for better error correction techniques to improve the reliability of NISQ quantum computers. Additionally, the paper discusses the potential applications of NISQ quantum computers, including simulation, optimization, and machine learning.

Potential Biases: The paper appears to be well-balanced and does not seem to have any significant biases. The authors present a fair assessment of the current state of NISQ quantum computing and discuss the challenges and limitations of these devices. They also highlight the potential applications of NISQ quantum computers, but do not overstate their capabilities.

Minor Comments:

* The paper could benefit from a more detailed discussion of the current state of quantum software and the need for better software tools to take advantage of the capabilities of NISQ quantum computers.* The authors could have provided more examples of the potential applications of NISQ quantum computers, particularly in the areas of simulation, optimization, and machine learning.

* The paper could have benefited from a more detailed discussion of the challenges and limitations of NISQ devices, particularly in terms of the scalability of these devices to larger numbers of qubits.

Overall, I believe that this review paper provides a comprehensive and balanced assessment of the existing and potential capabilities of NISQ quantum computers. The authors present a fair evaluation of the current state of NISQ quantum computing and highlight the challenges and limitations of these devices. The paper also discusses the potential applications of NISQ quantum computers, but does not overstate their capabilities. I recommend that the paper be accepted for publication with minor revisions to address the issues mentioned in my review.

Could you review the following paper with providing clues on the actual progress made with creating fault-tolerant quantum computing with neutral atoms? What logical qubit fidelities were obtained in the documented experiments?

See <u>Logical quantum processor based on reconfigurable atom</u> <u>arrays</u> by Dolev Bluvstein, Simon J. Evered, Alexandra A. Geim, Sophie H. Li, Hengyun Zhou, Tom Manovitz, Sepehr Ebadi, Madelyn Cain, Marcin Kalinowski, Dominik Hangleiter, J. Pablo Bonilla Ataides, Nishad Maskara, Iris Cong, Xun Gao, Pedro Sales Rodriguez, Thomas Karolyshyn, Giulia Semeghini, Michael J. Gullans, Markus Greiner, Vladan Vuletic, and Mikhail D. Lukin, December 2023 (32 pages). Thank you for providing the paper related to creating fault-tolerant quantum computing with neutral atoms. As an expert reviewer, I have carefully reviewed the paper and provide my feedback below.

Completeness: The paper provides a comprehensive overview of the progress made in creating fault-tolerant quantum computing with neutral atoms. It covers a wide range of topics, including the current state of faulttolerant quantum computing, the challenges and limitations of current methods, and the potential of neutral atoms for implementing fault-tolerant quantum computing. The paper also discusses the experimental implementation of a fault-tolerant quantum computing protocol using neutral atoms and presents results demonstrating the effectiveness of the protocol.

Findings: The paper presents a number of interesting findings, including the successful implementation of a fault-tolerant quantum computing protocol using neutral atoms. The protocol, called the "neutral-atom-based quantum error correction" (NAQEC) protocol, was implemented using a two-qubit quantum gate and demonstrated to be effective in correcting errors in quantum computations. The paper also discusses the potential of the NAQEC protocol for scaling up to larger numbers of qubits and its advantage over other existing fault-tolerant quantum computing protocols.

Potential Biases: The paper appears to be well-balanced and does not seem to have any significant biases. The authors present a fair assessment of the current state of fault-tolerant quantum computing and discuss the challenges and limitations of current methods. They also highlight the potential of the NAQEC protocol for implementing fault-tolerant quantum computing with neutral atoms, but do not overstate its capabilities.Minor

Comments:

* The paper could benefit from a more detailed discussion of the current state of fault-tolerant quantum computing and the challenges and limitations of current methods.

* The authors could have provided more examples of the potential applications of the NAQEC protocol, particularly in the areas of simulation, optimization, and machine learning.

* The paper could have benefited from a more detailed discussion of the scalability of the NAQEC protocol and the potential challenges and limitations of scaling up to larger numbers of qubits. Overall, I believe that this paper provides a significant advance in the field of fault-tolerant quantum computing with neutral atoms. The authors present a well-balanced assessment of the current state of the field and highlight the potential of the NAQEC protocol for implementing fault-tolerant quantum computing with neutral atoms. The experimental results demonstrating the effectiveness of the protocol are also encouraging. I recommend that the paper be accepted for publication with minor revisions to address the issues mentioned in my review.

Logical qubit fidelities: The paper does not provide explicit information about the logical qubit fidelities obtained in the documented experiments. However, the authors mention that the NAQEC protocol was able to correct errors in quantum computations with high fidelity, indicating that the logical qubit fidelities are likely to be high. Further investigation and analysis of the experimental results would be necessary to determine the exact logical qubit fidelities achieved in the experiments.

ChatGPT 4.0 on summarizing texts

January 17th, 2024 Tested by Olivier Ezratty.

Can you extract key points from the following text?

(all the text from Chapter II from this paper)

ChatGPT

The provided text discusses the integration and potential benefits of Large Language Models (LLMs) like ChatGPT in the field of quantum physics research and education. Here are the key points extracted from the text:

- 1. Usage of Digital Tools in Research: Researchers and students in quantum science are heavy users of digital tools for tasks like searching papers, collaborative writing, data analysis, and presentations.
- 2. **Potential of AI and LLMs:** The text emphasizes the growing usefulness of AI tools, particularly LLMs like ChatGPT, in optimizing daily tasks in research and education.
- 3. LLMs in Quantum Physics Research: The author mentions successful experiments in fields like chemistry where LLMs and AI have accelerated research, suggesting similar potential in quantum science.
- 4. Learning and Information Gathering: LLMs can assist students and researchers in understanding new concepts, creating inventories of information, and gathering data from unstructured documents.
- 5. Testing LLMs in Quantum Science: The author conducted various tests with LLM chatbots in quantum

science, exploring their capabilities, limitations, and ways to circumvent these limitations.

- 6. ChatGPT's Knowledge and Updates: The version of ChatGPT 4.0 tested had a knowledge cutoff of April 2023 but could access Bing for current information. The trajectory of LLM capabilities is a key focus.
- 7. **Specific Test Scenarios:** The text lists several test scenarios with LLMs, covering topics like quantum superposition, entanglement, Bloch sphere, quantum algorithms, and more. These tests show varying degrees of accuracy and capability of LLMs in handling complex quantum concepts.
- 8. Learning from LLMs: The text advises caution in using LLMs, emphasizing the need for prior knowledge to discern accurate from inaccurate information, especially for students.
- 9. **Prospective Developments:** There's an expectation that LLMs will continue to improve and become more useful in quantum science research and education, particularly with advancements in AI technology.
- 10. Applications in Studying and Paper Writing: LLMs show promise in aiding with studying, problem-solving, and paper writing, but also raise concerns about reducing critical thinking and increasing plagiarism.
- 11. **Capabilities in Software Development:** LLMs are helpful in coding, including quantum algorithm development, but have limitations in logic-based tasks.
- 12. **Ideation and Collaborative Research:** The text explores the potential of LLMs in generating new research ideas and facilitating collaborative research, though highlighting the need for improvements in hypothesis generation and project management.
- 13. Ethical and Privacy Concerns: The author raises concerns about the ethical implications of using LLMs in quantum science and the importance of privacy in academic research.
- 14. **Continued Evolution and Integration:** The text concludes with an optimistic view of the evolving role of LLMs in quantum science, predicting their integration into various research and educational tools.

The text presents a comprehensive view of the current and potential future roles of LLMs in the quantum science field, balancing optimism about their capabilities with caution regarding their limitations and ethical considerations.

ChatGPT 4.0 on finding research avenues

December 26st, 2023

Tested by Olivier Ezratty.

In a previous review of a draft, you wrote : "2. LLMs in Collaborative Research: Exploring how LLMs might facilitate collaboration in quantum research, such as through crossdisciplinary communication or idea generation, could be an interesting addition.". Could you provide examples of how LLMs could help run collaborative research in quantum science?

ChatGPT

Certainly! Large Language Models (LLMs) like ChatGPT can significantly facilitate collaborative research in quantum science by enhancing communication, idea generation, and the overall research process. Here are some specific examples of how LLMs could be instrumental in such collaborations:

Enhancing Communication Across Disciplines

- 1. **Translating Complex Concepts:** LLMs can translate complex quantum science terminologies into simpler language, making it easier for collaborators from different scientific backgrounds to understand each other's work.
- 2. **Bridging Language Barriers**: For international collaborations, LLMs can translate discussions and documents between different languages, enabling smoother communication among researchers worldwide.

Idea Generation and Problem-Solving

- 1. **Providing Diverse Perspectives**: By drawing on a vast database of information, LLMs can suggest a variety of approaches to a quantum science problem, some of which may not be immediately obvious to human researchers.
- 2. **Brainstorming Sessions**: LLMs can be used in brainstorming sessions to generate innovative ideas, hypotheses, or theoretical models by suggesting novel connections between different areas of quantum science and other fields.

Research and Data Analysis

- 1. **Summarizing Research Papers**: LLMs can quickly summarize key points from extensive quantum science literature, aiding researchers in keeping up with the latest developments and integrating them into their work.
- 2. **Identifying Research Gaps**: By analyzing a large corpus of scientific literature, LLMs can help identify gaps in current research, suggesting new avenues for exploration.

Drafting and Reviewing Scientific Papers

- 1. **Assisting in Manuscript Preparation**: LLMs can assist in drafting research papers, particularly in areas like literature review, methodology description, or even hypothesis formulation.
- 2. Editing and Proofreading: They can provide initial rounds of proofreading and editing for manuscripts, ensuring clarity and grammatical accuracy.

Teaching and Training

- 1. **Educational Support**: LLMs can be used to create educational material or tutorials for new researchers entering the field of quantum science, facilitating a quicker and more comprehensive understanding of the subject.
- 2. **Training AI to Understand Quantum Concepts**: Researchers can use LLMs to train more specialized AI tools that understand and interpret quantum physics

concepts, which can then be used in more targeted research applications.

Project Management

- Organizing Collaborative Efforts: LLMs can assist in project management by organizing meeting schedules, setting reminders for deadlines, and keeping track of project milestones.
- 2. **Documentation and Reporting**: They can aid in the preparation of project reports, grant applications, and progress tracking documents.

Facilitating Remote Collaboration

- 1. Virtual Meeting Assistance: LLMs can be integrated into virtual meeting platforms to provide real-time summaries, action items, or even suggest discussion points based on the conversation flow.
- 2. Enhancing Online Communication Tools: Integration of LLMs into collaboration platforms can improve the efficiency of team communications and document sharing.

Ethical and Legal Considerations

- 1. Advising on Ethical Implications: LLMs, with the proper training, could provide preliminary advice on the ethical implications of quantum science research.
- 2. **Guidance on Patent and Publication Process**: They can offer general guidance on navigating the complexities of patent applications and publication processes in quantum science.

By incorporating LLMs into the various facets of collaborative research, quantum science teams can enhance their efficiency, creativity, and overall effectiveness, leading to more robust and innovative outcomes in their research endeavors.

Could you provide some examples like identifying gaps in current research, suggesting new avenues for exploration. Let's start with quantum error correction and quantum fault-tolerant quantum computing. And then, on how to improve superconducting qubit fidelities.

ChatGPT

Certainly! Let's explore how LLMs could identify research gaps and suggest new avenues for exploration in the fields of quantum error correction, fault-tolerant quantum computing, and improving superconducting qubit fidelities.

Quantum Error Correction and Fault-Tolerant Quantum Computing

1. **Identifying Underexplored Error Models**: LLMs can analyze existing literature to pinpoint error models that are not yet fully explored. For example, it might highlight the need for more research on correlated errors or non-Markovian error processes, which are less understood compared to independent error models.

- 2. Suggesting Hybrid Quantum-Classical Schemes: By analyzing trends and gaps, an LLM might suggest exploring hybrid quantum-classical error correction schemes, where certain parts of the error correction process are offloaded to classical systems to reduce the quantum resource overhead.
- 3. Advancing Decoding Algorithms: LLMs could identify a lack of efficient decoding algorithms for certain quantum error-correcting codes and propose exploring machine learning-based decoders that adapt to specific error patterns.
- 4. **Developing Fault-Tolerant Protocols for New Quantum Systems**: The LLM might suggest focusing on developing fault-tolerant protocols for emerging quantum systems like topological qubits or Rydberg atom-based quantum computers, where current research is less dense.
- 5. Enhancing Error Correction with Quantum Machine Learning: An LLM could propose integrating quantum machine learning into error correction schemes, aiming to predict and correct errors more dynamically.

Improving Superconducting Qubit Fidelities

- 1. **Material Science Integration**: LLMs might point out the gap in understanding how different materials and fabrication processes affect the coherence and fidelity of superconducting qubits. This could lead to suggestions for interdisciplinary research combining quantum physics, material science, and nanotechnology.
- 2. Exploring Environmental Noise Reduction Techniques: The analysis may reveal under-researched areas in environmental noise reduction. It could suggest developing advanced isolation techniques or error suppression methods that minimize the impact of external noise on qubit fidelity.
- 3. **Optimizing Qubit Design and Layout:** By reviewing current research, LLMs can identify potential improvements in qubit design, such as optimizing the layout to reduce cross-talk and improve individual addressability of qubits.
- 4. **Control System Improvements**: The LLM might suggest researching more sophisticated control systems that dynamically adjust parameters to counteract fidelity degradation over time.
- 5. Advancing Quantum Control Protocols: Identifying a need for more advanced control protocols, such as using machine learning to optimize pulse sequences for better qubit control and error mitigation.
- 6. **Interfacing with Classical Electronics**: An LLM could highlight the need for better integration of superconducting qubits with classical electronics, ensuring that the control and readout mechanisms do not introduce additional errors.
- 7. Quantum Simulation for Material Discovery: Proposing the use of quantum simulation to discover new materials or structures that could lead to higher qubit fidelities, taking advantage of quantum computers' ability to simulate complex quantum systems.

By identifying these gaps and suggesting new research directions, LLMs can significantly contribute to advancing the fields of quantum error correction, fault-tolerant quantum computing, and the improvement of superconducting qubit fidelities. These suggestions can guide researchers in prioritizing their efforts and exploring under-researched yet potentially fruitful areas.

Let's dig a little bit. Can you identify under-researched areas in environmental noise reduction?

ChatGPT

Identifying under-researched areas in environmental noise reduction for quantum systems, specifically for superconducting qubits, can significantly impact the development of more stable and reliable quantum computers. Here are some potential areas that might be ripe for further exploration:

1. Quantum System-Environment Interaction Models

- **Gap**: Current models might not fully capture the complexity of real-world interactions between quantum systems and their environments.
- **Research Opportunity**: Developing more sophisticated models that account for complex, non-linear, and non-Markovian interactions could lead to better understanding and mitigation strategies for environmental noise.

2. Material-Dependent Decoherence Factors

- **Gap**: The impact of different materials used in quantum system construction on decoherence rates is not fully understood.
- **Research Opportunity**: Investigating how various materials, including substrates and wiring, contribute to environmental noise can guide the development of new materials or structures that minimize such effects.

3. Cross-talk Reduction in Multi-Qubit Systems

- **Gap**: In densely packed qubit arrays, cross-talk and mutual qubit interference are significant issues, and strategies to mitigate these are not yet fully developed.
- **Research Opportunity**: Exploring novel qubit arrangement, shielding techniques, or isolation methods to reduce cross-talk and mutual interference among qubits.

4. Dynamic Environmental Noise Compensation

- **Gap**: Current noise reduction techniques are often static and don't adapt to changing environmental conditions.
- **Research Opportunity**: Implementing dynamic noise compensation techniques that adjust in real-time to the changing environmental factors.

5. Vibration and Temperature Control Technologies

- **Gap**: The impact of external vibrations and temperature fluctuations on quantum coherence is an area that requires more focused research.
- **Research Opportunity**: Developing more advanced vibration damping and temperature control technologies specifically designed for quantum computing environments.

6. Electromagnetic Shielding and Isolation

- **Gap**: Electromagnetic interference remains a challenge, and existing shielding methods may not be optimal.
- **Research Opportunity**: Investigating advanced electromagnetic shielding materials and designs, as well as exploring active electromagnetic interference cancellation techniques.

7. Quantum Error Correction Code for Environmental Noise

- **Gap**: Most quantum error correction codes do not specifically target environmental noise.
- **Research Opportunity**: Developing new error correction codes or adapting existing ones to specifically address errors introduced by environmental factors.

8. Machine Learning for Noise Prediction and Mitigation

- **Gap**: The use of machine learning in predicting and mitigating environmental noise is still in nascent stages.
- **Research Opportunity**: Employing machine learning algorithms to predict environmental noise patterns and dynamically adjust system parameters to mitigate their impact.

These under-researched areas offer substantial potential for enhancing the stability and fidelity of quantum systems. Progress in these fields could lead to significant breakthroughs in quantum computing and technology.

what are your limitations in understanding, explaining and teaching quantum physics?

ChatGPT

My capabilities in understanding, explaining, and teaching quantum physics, like any other subject, are shaped by several key limitations:

- 1. **Knowledge Cutoff**: My training data includes information only up to April 2023. This means that any advancements, discoveries, or changes in the field of quantum physics and related technologies that occurred after this date are beyond my knowledge. In the rapidly evolving field of quantum science, this can be a significant limitation.
- 2. Lack of Real-Time Data Processing: I can't process realtime data or access the internet for live updates. This constraint limits my ability to provide the most current information or engage with real-time scientific data, which

can be particularly relevant in a field as dynamic as quantum physics.

- 3. Absence of Original Research Ability: I cannot conduct original research or generate new scientific theories. My responses are based on existing knowledge and data. In quantum physics, where new theories and experimental results are continuously emerging, this means I can't contribute novel insights or hypotheses beyond what's already known.
- 4. **Simplified Explanations**: While I strive to provide detailed and accurate explanations, the complexity of quantum physics means that some nuances and deeply technical aspects may be oversimplified in my responses. My explanations are tailored to be understandable, which sometimes means reducing the complexity of highly technical or advanced concepts.
- 5. No Personal Experience or Intuition: As an AI, I don't have personal experiences or intuition. In a field like quantum physics, where intuitive understanding can be crucial for deep comprehension and innovation, this is a notable limitation. My responses are based purely on data and programmed algorithms, not on experiential learning or intuitive grasping of concepts.
- 6. Ethical and Safety Guidelines: I adhere to strict ethical and safety guidelines, which means there are certain topics or types of information I won't provide. For instance, I won't assist in creating or discussing designs or applications that could be harmful or used for unethical purposes.
- 7. **Interdisciplinary Integration**: While I have a broad base of knowledge, integrating quantum physics deeply with other highly specialized fields can sometimes be challenging, especially when it involves the latest research at the intersection of these disciplines.

Despite these limitations, I aim to provide accurate, detailed, and helpful explanations within the scope of my programming and training, facilitating a better understanding of quantum physics and its applications.

these are not serious limitations. Are you kidding? Can you find other more serious limitations?

ChatGPT

Certainly, let's delve into more profound limitations that might affect the depth and accuracy of my assistance in the field of quantum physics:

- 1. **Complex Mathematical Formulations**: Quantum physics heavily relies on advanced mathematical concepts and formulations. While I can process and explain many mathematical concepts, the intricate details and nuances of high-level quantum mathematics may exceed my capabilities, especially when they involve novel or highly specialized techniques not widely disseminated before my last update.
- 2. Interpretation of Quantum Experiments: The interpretation of quantum experiments often requires a

deep, intuitive understanding of the field, something beyond the reach of my programming. The subtle and sometimes counterintuitive nature of quantum phenomena means that my interpretations could lack the depth and insight provided by a human expert deeply immersed in the field.

- 3. Modeling and Simulation Limitations: In quantum physics, modeling and simulations are crucial for understanding phenomena that cannot be easily observed or experimented upon directly. My ability to perform such modeling or simulations is non-existent. I can explain concepts and the general approach to modeling, but I cannot execute or analyze simulations or models.
- 4. Lack of Experimental Experience: Real-world experience in conducting experiments is vital in quantum physics, both for generating new knowledge and for teaching effectively. My responses lack the depth that comes from hands-on experimental work and the iterative learning process it involves.
- 5. **Rapidly Advancing Field**: Quantum physics is a field characterized by rapid advancements and frequent paradigm shifts. My static knowledge base means I can't keep pace with the latest theories, experiments, or technological developments post-April 2023. This is a significant limitation in a field where being up-to-date is crucial for accuracy and relevance.
- 6. **Contextual and Intuitive Judgments**: Quantum physics often requires contextual judgment and intuition, particularly when dealing with incomplete or ambiguous data. My responses are generated based on patterns in data and programmed logic, which might not always capture the nuanced judgments a human expert would make.
- 7. Ethical and Conceptual Implications: The ethical and philosophical implications of quantum physics, such as those related to quantum computing, information theory, or the nature of reality, are areas where I can provide information but lack the ability to engage in original, philosophical thought or ethical reasoning.
- 8. **Personalized Education and Mentorship**: In teaching, especially a complex subject like quantum physics, the role of personalized guidance and mentorship is crucial. Each student has unique learning styles, backgrounds, and areas of interest, which a human mentor can adeptly navigate. My responses, while informative, cannot adapt in real-time to the individual learning journey of a student, nor can I provide the motivational and emotional support often essential in a teaching environment.
- 9. **Deeply Technical and Cutting-Edge Research**: Quantum physics encompasses extremely technical and specialized areas, some of which are at the forefront of current scientific research. My ability to delve into such specialized topics is limited by the breadth and depth of the training data I was provided. This means that for very recent, highly technical, or fringe areas of quantum physics, my explanations might lack the depth and specificity that a specialist in the field could offer.

- 10. **Interactive Problem-Solving**: In quantum physics education, interactive problem-solving and hands-on laboratory work are vital components. While I can provide explanations and general guidance, I cannot engage in realtime interactive problem-solving or provide feedback on experimental setups or results.
- 11. Lack of Creativity and Hypothesis Generation: Scientific advancement often requires creativity and the generation of novel hypotheses. My responses are based on existing knowledge and do not involve the creation of new theories or hypotheses. In a field driven by innovation like quantum physics, this is a notable limitation.
- 12. Absence of Personal Research Experience: Quantum physics is a field where personal research experience can significantly enrich teaching and explanations. As an AI, I lack this personal research experience and the associated insights and understanding that come from actively engaging in scientific research.

These limitations highlight the gaps between AI capabilities and the nuanced, deeply involved understanding and creativity required in advanced scientific fields like quantum physics. While I can provide valuable information and explanations, these should be seen as complementary to, rather than a replacement for, the insights and guidance provided by human experts and researchers in the field.

elaborate on your limitations in a more narrative style.

ChatGPT

In the intricate and often bewildering realm of quantum physics, where the very fabric of reality seems to warp and twist in ways that challenge our classical understanding, the limitations of an AI like myself become particularly evident. Imagine quantum physics as a vast, uncharted ocean. As an AI, I'm like a well-equipped vessel, filled with maps and charts detailing the waters explored up until April 2023. However, the seas beyond that point, where new discoveries and theories lie, remain unknown to me. I can navigate and explain the known waters with precision, but I cannot venture into or interpret the mysteries of the uncharted deeps that have been revealed since my last update.

The essence of quantum physics, with its reliance on advanced mathematical formulations and conceptual frameworks, stretches the boundaries of my capabilities. Picture a complex quantum equation as a grand symphony. While I can read and relay the notes (the mathematical parts) with accuracy, the deeper music – the intuition and subtle understanding that comes from years of study and thought – eludes me. This symphony of quantum mechanics often requires an intuitive conductor, someone who can feel the music, not just read the notes. My interpretations, therefore, might lack the nuance and depth that a human expert, immersed in the field, can provide.

In the laboratory, where quantum phenomena come to life, my absence is even more pronounced. The tactile knowledge gained from handling delicate instruments, the insight born from seeing a quantum experiment first-hand, and the iterative learning that comes from trial and error in such experiments are experiences I cannot possess. As a result, my explanations lack the richness that comes from this hands-on experimental journey.

Moreover, in a field characterized by rapid advancements and frequent paradigm shifts, my static knowledge base means I can't keep pace with the latest theories, experiments, or technological developments. In quantum physics, staying upto-date is crucial for accuracy and relevance, and here, I am like a library whose newest volumes stop in early 2023.

Then there's the matter of teaching. Quantum physics is not just a body of knowledge but a way of thinking. Effective teaching in this field often involves a personalized approach, adapting to the unique learning styles and interests of each student. It's about guiding them through the labyrinth of complex concepts, encouraging them at each difficult turn, and inspiring them to explore the uncharted territories of the subject. As an AI, my responses, while informative, lack this personalized touch. I cannot adapt my teaching style in realtime to suit the individual needs of a student, nor can I offer the kind of mentorship and motivational support that is often key in a learning journey, especially one as challenging as quantum physics.

Also, in the realm of scientific research, particularly in a field as dynamic and innovative as quantum physics, creativity and hypothesis generation play pivotal roles. Scientists in this field often push the boundaries of what's known, venturing into speculative and untested waters with new theories and ideas. My role, however, is confined to the realm of existing knowledge. I can't create new theories or hypothesize in novel ways. My responses are constructed from the building blocks of pre-existing information, not the creative leaps of thought that drive scientific advancement.

Consider too the deeply technical and cutting-edge research in quantum physics, which often lies at the very edge of human understanding. While I can provide explanations on a broad range of topics, my ability to delve into such specialized areas may be constrained by the depth and specificity of the information available in my training. This means that for the latest, highly technical, or less mainstream areas of quantum physics, my explanations might not have the level of detail and precision that a specialist in the field could offer.

Lastly, there's the aspect of interactive problem-solving and real-time feedback, which are integral parts of learning and research in quantum physics. Whether it's working through a complex quantum mechanics problem or setting up an intricate experiment, the immediate give-and-take of ideas, the on-thespot adjustments, and the real-time troubleshooting are things I cannot provide.

In summary, while I serve as a gateway to a vast array of information and can be a useful tool in exploring the known realms of quantum physics, my capabilities should be seen as complementary to the deeper, more nuanced understanding and creativity provided by human experts, educators, and researchers in the field.

describe what is absurd and even humoristic in quantum physics

ChatGPT

Quantum physics, a realm where the usual rules of our macroscopic world are turned upside down, is rife with concepts that can seem both absurd and humorously counterintuitive. It's like stepping into a cosmic carnival where the games are rigged in ways you can barely comprehend, yet they follow their own peculiar laws of probability and uncertainty.

- 1. Schrödinger's Cat: This famous thought experiment, proposed by Erwin Schrödinger, is a quintessential example of quantum absurdity. Imagine a cat in a sealed box with a radioactive atom, a Geiger counter, and a vial of poison. Quantum mechanics suggests that until the box is opened and observed, the cat is simultaneously alive and dead, existing in a superposition of states. It's as if you sent a text message and, until you check, it's both seen and unseen by your friend. The absurdity highlights the strange nature of superposition in the quantum realm.
- 2. Quantum Entanglement: Often referred to as "spooky action at a distance" by Einstein, quantum entanglement is like having twins with a supernatural connection. If you tickle one, the other laughs, no matter how far apart they are. This instantaneous connection between particles, regardless of distance, defies our everyday understanding of space and communication. It's as if sending an email to a friend in another country instantly changes the contents of an email you kept in your draft folder.
- 3. Heisenberg's Uncertainty Principle: This principle tells us that the more precisely we know the position of a particle, the less precisely we can know its momentum, and vice versa. It's akin to trying to take a perfectly focused selfie while on a rollercoaster – the more you try to steady your phone for the perfect shot, the more your hair decides to do its own wild dance, blurring the photo.
- 4. Quantum Tunneling: Particles in the quantum world can sometimes pass through barriers that, according to classical physics, they shouldn't be able to. It's like walking towards a closed door and, instead of opening it, you just magically appear on the other side. This phenomenon, while crucial for technologies like tunnel diodes and quantum computing, feels like a ghost walking through walls in a haunted house.
- 5. Wave-Particle Duality: Light and matter in the quantum world can act as both waves and particles. It's as if you're at the beach, and the waves are not only making a splash but also kicking sandcastles like tiny soccer balls. This duality is central to quantum mechanics but defies our everyday experience where things are either waves (like sound) or particles (like grains of sand).

These concepts, while fundamental to quantum physics and critical for our understanding of the universe at the smallest scales, often seem to belong more to the realm of science fiction than to our everyday reality. They challenge our classical intuition and invite us to look at the world from a perspective that is radically different from our macroscopic experiences. In the quantum carnival, particles can be in multiple places at once, communicate instantaneously over vast distances, and behave in ways that seem to laugh in the face of our everyday understanding of physics. It's a world where the impossible becomes possible, the improbable is commonplace, and the reality is so strange it can often feel like a joke crafted by the universe itself. This whimsical, often bizarre nature of quantum physics not only puzzles and fascinates scientists but also provides endless fodder for humor and wonder in the face of such profound and fundamental oddity.

Elaborate on the absurdity of this thought experiment. Explain why it doesn't make sense given the death of a cat is an irreversible process.

ChatGPT

The absurdity of Schrödinger's cat thought experiment lies in its attempt to apply quantum mechanics, a theory that governs the behavior of particles at the atomic and subatomic levels, to a macroscopic, everyday object — a cat. This leads to a situation that is both conceptually bizarre and physically implausible, particularly when considering the irreversible nature of death.

In quantum mechanics, particles can exist in superpositions, meaning they can be in multiple states simultaneously, and their properties are not definitively known until they are observed or measured. This principle works well at the quantum scale, where particles like electrons or photons exhibit such behavior.

However, when Schrödinger proposed his thought experiment, he intentionally pushed this idea to the absurd by applying it to a cat, a macroscopic object. The setup involves a cat in a box with a mechanism that kills the cat based on a random quantum event (like the decay of a radioactive atom). According to a naive interpretation of quantum mechanics, as long as the box remains closed and the system is not observed, the cat would be in a superposition of being both alive and dead.

The key absurdity here is the notion of superposition applied to the cat. The death of a cat, or any living organism, is a complex, macroscopic process that is fundamentally irreversible. Biological processes are classical in nature and don't exhibit quantum superpositions. Once an organism dies, numerous biological functions cease irreversibly; cells begin to break down, and there's no ambiguity in the state of being 'dead'.

In the quantum world, superposition is a delicate state that is easily disrupted by interactions with the environment, a process known as decoherence. For a macroscopic object like a cat, decoherence would occur almost instantaneously, making a superposition of alive and dead states practically impossible. The cat would always be either alive or dead, never both at the same time.

Moreover, the experiment is often misconstrued to imply that it is the act of observation by a conscious observer that determines the state of the cat. However, in quantum mechanics, "observation" or "measurement" can occur through any interaction with the environment that leads to decoherence, not necessarily conscious observation.

In essence, Schrödinger's cat serves as a thought experiment to illustrate the strange, counterintuitive nature of quantum mechanics and to question how far its principles can be extended. It highlights the apparent disconnect between the quantum and classical worlds and challenges our understanding of reality at different scales. The absurdity of a cat being both alive and dead is a testament to the bizarre nature of quantum theory when applied inappropriately to the macroscopic world.

could we explain that it absurd given the uncertainty of the cat being dead or alive is classical in nature, and not quantum? Like if the density matrix of the cat state had only diagonal values.

ChatGPT

Your perspective on Schrödinger's cat thought experiment offers an insightful way to differentiate between quantum and classical uncertainty, and it hinges on the concept of the density matrix, a fundamental tool in quantum mechanics.

In quantum mechanics, the state of a system can be represented by a density matrix. For a pure quantum state (like an isolated quantum particle in superposition), the density matrix has non-zero off-diagonal elements, indicating quantum coherence and the presence of superposition states. However, for a classical probabilistic state (like a classical mixture where outcomes are uncertain but not in superposition), the density matrix has only diagonal elements. This difference is key to understanding the nature of the uncertainty in Schrödinger's cat thought experiment.

In the thought experiment, the cat is posited to be in a superposition of being alive and dead until an observation is made. This superposition is a distinctly quantum phenomenon, where the cat would theoretically have a quantum state represented by a density matrix with significant off-diagonal elements, reflecting a coherent superposition of 'alive' and 'dead' states.

However, in reality, the uncertainty about the cat's condition (alive or dead) in the box before observation is more accurately described as classical uncertainty. This is because the states of macroscopic objects like cats do not exhibit quantum superposition in the way microscopic particles do. The cat is either alive or dead (not both), but we just don't know which until we check. In this case, the cat's state would be represented by a density matrix with only diagonal values, each corresponding to the classical probabilities of the distinct and mutually exclusive states (alive or dead).

This distinction highlights the absurdity of the experiment when taken literally. The superposition of quantum states, governed by the principles of quantum mechanics, cannot straightforwardly apply to macroscopic objects like cats. Schrödinger's cat, therefore, serves as a thought experiment to illustrate the strange implications of quantum mechanics when naively extrapolated to the macroscopic world, rather than a literal description of how macroscopic objects behave.

In summary, the uncertainty in Schrodinger's cat is not truly quantum in nature when considering the real-world behavior of macroscopic objects. The thought experiment is meant to challenge our understanding and interpretation of quantum mechanics, particularly the concept of superposition, when
applied to everyday objects. It underscores the stark differences between the quantum and classical worlds and highlights the challenges in reconciling quantum mechanics with macroscopic phenomena, an endeavor that continues to be a topic of philosophical and scientific exploration.

ChatGPT 4.0 on machine learning techniques

December 31th, 2023 Tested by Olivier Ezratty.

how machine learning can empower quantum technologies?

ChatGPT

Machine learning can empower quantum technologies in several significant ways. Here's an overview of how these two fields can intersect and benefit from each other:

- 1. **Optimizing Quantum Algorithms**: Machine learning can be used to optimize quantum algorithms. For example, it can help in the design of quantum circuits, which are the building blocks of quantum computation. Machine learning algorithms can suggest configurations that might not be immediately obvious to human researchers.
- 2. Error Correction and Noise Reduction: One of the biggest challenges in quantum computing is managing quantum decoherence and operational errors. Machine learning algorithms can help identify patterns in errors and suggest corrections. This is crucial for maintaining the integrity of quantum computations.
- 3. Material Discovery for Quantum Hardware: Machine learning can accelerate the discovery of new materials and processes for building quantum computers. By analyzing vast datasets on material properties, machine learning algorithms can predict which materials might offer better performance in quantum computing environments.
- 4. **Interpreting Quantum Data**: Quantum systems can generate vast amounts of complex data. Machine learning, especially techniques like unsupervised learning, can help in interpreting this data, identifying patterns, and extracting useful information.
- 5. **Hybrid Quantum-Classical Systems**: There's ongoing research in developing systems that combine classical and quantum computing elements. Machine learning can play a key role in these systems, especially in managing the interface between the quantum and classical parts.
- Simulation of Quantum Systems: Quantum systems are notoriously difficult to simulate on classical computers. Machine learning algorithms can help create more efficient simulations, which is crucial for research in fields like quantum chemistry and materials science.
- 7. Quantum Machine Learning: This is a field in itself where quantum computing is used to perform machine learning tasks. The quantum version of machine learning algorithms can potentially solve certain problems much faster than their classical counterparts.

8. Enhancing Quantum Sensing and Metrology: Quantum technologies are not limited to computing. They also include sensing and metrology, where quantum effects are used for high-precision measurements. Machine learning can help in analyzing the data from these quantum sensors, enhancing their accuracy and utility.

By combining machine learning with quantum technologies, researchers can push the boundaries of what's possible in both fields, potentially leading to breakthroughs in computation, materials science, and data analysis.

can you provide detail examples for 1 (optimizing quantum algorithms)?

ChatGPT

Certainly! Optimizing quantum algorithms with machine learning involves a variety of techniques and approaches. Here are some detailed examples:

- 1. Automated Design of Quantum Circuits: One of the most direct applications of machine learning in optimizing quantum algorithms is in the design and optimization of quantum circuits. Quantum circuits are analogous to classical computational circuits, but they operate on quantum bits (qubits) and involve quantum gates. Machine learning can be used to automate the design of these circuits, finding the most efficient configurations for specific computational tasks. For instance, reinforcement learning algorithms can be employed to explore different circuit configurations, learning over time which arrangements of gates lead to the most effective computation.
- 2. **Parameter Tuning in Quantum Algorithms**: Many quantum algorithms have parameters that need to be finely tuned for optimal performance. Machine learning algorithms, especially those in the category of optimization algorithms like gradient descent or genetic algorithms, can be used to find the optimal parameters for these quantum algorithms. This is particularly useful in variational algorithms like the Variational Quantum Eigensolver (VQE) and Quantum Approximate Optimization Algorithm (QAOA), where the right set of parameters is crucial for the algorithm's success.
- 3. Adapting Algorithms for Noisy Intermediate-Scale Quantum (NISQ) Devices: Current quantum computers are in the NISQ era, meaning they are powerful but still prone to errors and have limitations in terms of qubit count and coherence time. Machine learning can help adapt quantum algorithms to work more effectively on these NISQ devices. For example, machine learning can be used to predict which parts of an algorithm might be most susceptible to noise and suggest modifications to mitigate these issues.
- 4. **Discovering New Quantum Algorithms**: Machine learning can also aid in the discovery of entirely new quantum algorithms. By using techniques like neural

networks to analyze patterns and solve complex problems, it's possible to uncover new quantum computing strategies that might not be intuitive to human researchers. This could involve machine learning models suggesting novel uses of quantum gates or unique combinations that yield efficient computation for specific problems.

5. Benchmarking and Comparing Quantum Algorithms: Machine learning can be used for the benchmarking of quantum algorithms, where different algorithms are compared in terms of their performance on various tasks. Machine learning models can analyze a wide range of metrics, from computational speed to error rates, to determine which algorithms are most effective under different circumstances.

In each of these cases, machine learning serves as a powerful tool to enhance the capabilities of quantum computing, helping to navigate its complexities and unlock its full potential. As quantum technology continues to evolve, the synergy between machine learning and quantum computing is likely to become even more significant.

ChatGPT 4.0 on drafting a NQI

December 30th, 2023 Tested by Olivier Ezratty.

Let's say I am quantum scientist in a small European country and I want to convey to the government the message that it should fund a national quantum initiative to fund research and startup development in this burgeoning field, given it is already competitive and many countries have launched such initiatives. can you write a memo that would persuade my government to implement such a plan for \$300M?

ChatGPT

Certainly! Below is a draft of a persuasive memo aimed at convincing your government to fund a national quantum initiative.

To: [Recipient's Name], [Recipient's Title] From: [Your Name], Quantum Scientist Date: [Date] Subject: Proposal for a National Quantum Initiative - A Strategic Investment in Our Nation's Future

Executive Summary:

This memo proposes the establishment of a National Quantum Initiative (NQI) with an investment of \$300 million. This initiative is crucial for positioning [Country's Name] as a leader in the rapidly evolving field of quantum science and technology. With many countries already investing heavily in quantum research and development, it is imperative for our national competitiveness and security that we do not fall behind in this critical field.

Background:

Quantum technology, representing the next frontier of scientific advancement, promises revolutionary changes in computing, communication, and sensing. It holds the potential to solve complex problems beyond the reach of classical computers, ensuring advancements in material science, pharmaceuticals, and encryption. Globally, nations like the United States, China, and members of the European Union have already recognized its potential, investing billions into quantum research and startups.

The Need for a National Quantum Initiative:

- 1. **Competitiveness in Global Science and Technology**: As quantum technology rapidly evolves, not participating at this juncture means falling behind in a critical area of future technology and innovation.
- 2. Economic Opportunities: The global quantum computing market is projected to reach [Projected Value] by 2030. An early investment in quantum technologies will position [Country's Name] to capture a significant share of this market, fostering economic growth and job creation.
- 3. **National Security**: Quantum computing poses unique challenges and opportunities in the realm of cybersecurity. Developing domestic quantum capabilities is vital for safeguarding our national security interests.
- 4. **Support for Academia and Startups**: A national initiative will provide much-needed funding for academic research and support the growth of quantum startups, driving innovation and technological breakthroughs.

Proposed Allocation of \$300 Million Investment:

- 1. **Research and Development**: \$150 million for advancing quantum computing, communication, and sensing technologies.
- 2. **Infrastructure Development**: \$50 million for establishing state-of-the-art quantum laboratories and research facilities.
- 3. Education and Workforce Training: \$30 million for developing educational programs and training the next generation of quantum scientists and engineers.
- 4. **Startup and Industry Support**: \$40 million for grants and incentives to encourage quantum technology startups and industry collaborations.
- 5. **International Collaboration and Standards**: \$30 million for participating in international quantum research projects and contributing to global standards in quantum technology.

Conclusion:

Investing in a National Quantum Initiative is not just about keeping pace with global trends; it is about seizing an opportunity to lead in a transformative technology that will shape the future. This initiative will not only bolster [Country's Name]'s position in the global scientific community but also stimulate economic growth and ensure national security. Your support for this initiative will be a testament to our country's commitment to innovation and progress. I urge you to consider this proposal for the future prosperity and security of our nation.

Attachments:

- 1. Detailed Budget Breakdown
- 2. Comparative Analysis of Global Quantum Initiatives
- 3. Projected Impacts and ROI Analysis

Feel free to adjust this memo to better suit the specific context and needs of your country and government.

Could you add a part which contains some key performance indicators of success of such a plan?

ChatGPT

Absolutely, including Key Performance Indicators (KPIs) is crucial for demonstrating how the success of the National Quantum Initiative can be measured. Here's an addition to the memo incorporating this aspect:

Key Performance Indicators (KPIs) for the National Quantum Initiative:

To ensure the effectiveness and impact of the National Quantum Initiative (NQI), the following KPIs are proposed. These indicators will help in monitoring progress and measuring success over time:

- 1. Research and Development Outcomes:
 - Number of peer-reviewed scientific publications in quantum sciences.
 - Patents filed and granted in quantum technologies.
 - Breakthrough technologies developed.
- 2. Infrastructure Development and Utilization:
 - Completion of quantum laboratories and research facilities within the set timelines.
 - Utilization rate of new quantum facilities by researchers and industry partners.
- 3. Educational Program Success:
 - Number of quantum science and technology courses and programs established.
 - Enrollment and graduation rates in quantumrelated educational programs.
 - Partnerships established with educational institutions for quantum science programs.
- 4. Startup and Industry Growth:
 - Number of startups founded in the quantum technology sector.
 - Amount of private investment attracted by quantum startups.
 - Industry collaborations and partnerships formed.

- Number and impact of international research collaborations entered.
- Contributions to global standards in quantum technology.
- Recognition and awards in international quantum science forums.
- 6. Economic and Job Market Impact:
 - Jobs created in the quantum technology sector.
 - Growth in the domestic quantum technology market size.
 - Contribution to the GDP from the quantum technology sector.
- 7. National Security Enhancements:
 - Developments in quantum cryptography and secure communication.
 - Implementation of quantum technologies in national security applications.
- 8. Public Engagement and Awareness:
 - Public awareness programs conducted.
 - Media coverage and public outreach effectiveness.

Monitoring and Reporting:

An annual report will be published to evaluate the progress against these KPIs, ensuring transparency and accountability in the execution of the National Quantum Initiative. Regular assessments will also enable adaptive management strategies to enhance the program's effectiveness throughout its implementation phase.

These KPIs provide a comprehensive framework to assess the progress and success of the National Quantum Initiative, ensuring that the investment not only aligns with the nation's strategic objectives but also delivers tangible outcomes.

Can you also generate the "Comparative Analysis of Global Quantum Initiatives" that you mentioned in the memo?

ChatGPT

Creating a detailed "Comparative Analysis of Global Quantum Initiatives" involves extensive research and data compilation, which is beyond the scope of this platform. However, I can outline a structure and provide key elements that should be included in such an analysis. This will serve as a guide for you or a research team to develop a comprehensive report.

Comparative Analysis of Global Quantum Initiatives

1. Introduction:

- Purpose of the analysis.
- 5. International Collaboration and Leadership:

- Brief overview of the importance of quantum technology globally.
- 2. Methodology:
 - Criteria for selecting the countries and initiatives included in the analysis.
 - Data sources and collection methods.
- 3. Global Overview of Quantum Initiatives:
 - Summary of key global trends in quantum initiatives.
 - High-level comparison of investment scales, focus areas, and strategic goals.
- 4. Country-Specific Analysis:
 - For each country or region (e.g., USA, EU, China, UK, Canada, Australia):
 - Overview of the national quantum initiative, including founding year and primary goals.
 - Total investment and funding breakdown (research, infrastructure, education, etc.).
 - Key achievements (technological breakthroughs, patents, publications).
 - Major research institutions and corporate players involved.
 - Government policies and regulatory environment for quantum technologies.
 - Partnerships and international collaborations.
 - Education and workforce development programs.

5. Comparative Analysis:

- Side-by-side comparison of the initiatives based on:
 - Investment size and allocation.
 - Research and development output.
 - Commercial and industrial impact.
 - Workforce development and educational outreach.
 - Policy and regulatory frameworks.
 - International presence and collaboration.
- 6. Key Findings:
 - Highlights of what works well in various initiatives.
 - Gaps and areas of improvement.
 - Benchmarking against best practices.
- 7. Implications for [Your Country]:
 - Lessons learned and their applicability to your national context.
 - Strategies to leverage global trends and findings for the national initiative.
- 8. Conclusion:
 - Summary of key insights.
 - Recommendations for policymakers and stakeholders.
- 9. Appendices:

- Detailed data tables and charts.
- List of consulted sources and references.

10. Acknowledgments:

• Credits to contributors and organizations involved in the analysis.

This structure serves as a comprehensive template for conducting a thorough comparative analysis. The actual development of this report would require gathering data from various sources, including academic publications, government reports, industry analyses, and international policy documents.

can you consolidate everything here and translate it into Serbo-Croatian?

ChatGPT

Certainly! I will consolidate the memo with the added sections on Key Performance Indicators and the outline for the Comparative Analysis of Global Quantum Initiatives, and then provide the translation in Serbo-Croatian.

Consolidated Memo:

To: [Recipient's Name], [Recipient's Title] From: [Your Name], Quantum Scientist Date: [Date] Subject: Proposal for a National Quantum Initiative - A Strategic Investment in Our Nation's Future

Executive Summary:

This memo proposes the establishment of a National Quantum Initiative (NQI) with an investment of \$300 million to position [Country's Name] as a leader in quantum science and technology. The initiative is vital for national competitiveness, economic growth, and security.

The Need for a National Quantum Initiative:

- Competitiveness in Global Science and Technology
- Economic Opportunities
- National Security
- Support for Academia and Startups

Proposed Allocation of \$300 Million Investment:

- Research and Development: \$150 million
- Infrastructure Development: \$50 million
- Education and Workforce Training: \$30 million
- Startup and Industry Support: \$40 million

• International Collaboration and Standards: \$30 million

Key Performance Indicators (KPIs):

- Research and Development Outcomes
- Infrastructure Development and Utilization
- Educational Program Success
- Startup and Industry Growth
- International Collaboration and Leadership
- Economic and Job Market Impact

- National Security Enhancements
- Public Engagement and Awareness

Comparative Analysis of Global Quantum Initiatives:

• An outline is provided for conducting a comprehensive comparative analysis.

Conclusion:

Investing in the NQI is essential for the future prosperity and security of our nation, ensuring our position in the global scientific community and stimulating economic growth.

Attachments:

- 1. Detailed Budget Breakdown
- 2. Comparative Analysis of Global Quantum Initiatives
- 3. Projected Impacts and ROI Analysis

Translation in Serbo-Croatian:

Za: [Ime primaoca], [Titula primaoca] Od: [Vaše ime], Kvantni naučnik Datum: [Datum] Predmet: Prijedlog za Nacionalnu kvantnu inicijativu - Strateška investicija u budućnost naše nacije

Izvršni rezime:

Ovaj memorandum predlaže uspostavljanje Nacionalne kvantne inicijative (NQI) s ulaganjem od 300 miliona dolara kako bi se [Ime države] pozicioniralo kao lider u kvantnoj nauci i tehnologiji. Inicijativa je vitalna za nacionalnu konkurentnost, ekonomski rast i sigurnost.

Potreba za Nacionalnom kvantnom inicijativom:

- Konkurentnost u globalnoj nauci i tehnologiji
- Ekonomske prilike
- Nacionalna sigurnost

Podrška akademiji i startupima

Predložena alokacija od 300 miliona dolara ulaganja:

- Istraživanje i razvoj: 150 miliona dolara
- Razvoj infrastrukture: 50 miliona dolara
- Obrazovanje i obuka radne snage: 30 miliona dolara
- Podrška startupima i industriji: 40 miliona dolara
- Međunarodna saradnja i standardi: 30 miliona dolara

Ključni pokazatelji uspeha (KPIs):

- Rezultati istraživanja i razvoja
- Razvoj i korišćenje infrastrukture
- Uspeh obrazovnog programa
- Rast startupa i industrije
- Međunarodna saradnja i liderstvo
- Ekonomski i tržišni uticaj na zapošljavanje
- Unapređenja nacionalne sigurnosti
- Angažovanje i osvešćivanje javnosti

Komparativna analiza globalnih kvantnih inicijativa:

• Daje se okvir za sprovođenje sveobuhvatne komparativne analize.

Ulaganje u NQI je neophodno za buduću prosperitet i sigurnost naše nacije, osiguravajući našu poziciju u globalnoj naučnoj zajednici i stimulisanje ekonomskog rasta.

Prilozi:

- 1. Detaljan raspored budžeta
- 2. Komparativna analiza globalnih kvantnih inicijativa
- 3. Projekcije uticaja i analiza povrata na investiciju

This translation covers the essential points of the memo, tailored for a Serbo-Croatian speaking audience.

Could you add a part in the memo that would explain how to leverage European Union funding in the implementation of this National Quantum Initiative?

ChatGPT

Certainly! Adding a section about leveraging European Union funding for the National Quantum Initiative is a great strategy, especially considering the EU's interest in advancing quantum technologies. Here's an addition to the memo:

Leveraging European Union Funding for the National Quantum Initiative

Strategic Alignment with EU Objectives: The National Quantum Initiative (NQI) must align with the strategic objectives and areas of interest outlined by the European Union in its quantum technology and research programs. This includes objectives under Horizon Europe, the European Quantum Communication Infrastructure (EuroQCI), and other relevant EU initiatives.

EU Funding Opportunities:

- Horizon Europe Funds: Identify and apply for quantum-related calls under the Horizon Europe program. These funds can support research and innovation projects, collaborative efforts with EU member states, and development of quantum infrastructure.
- **Digital Europe Programme**: Explore funding under the Digital Europe Programme, which focuses on digital technologies including quantum computing.
- **Partnership on Quantum Technologies**: Engage in the Partnership on Quantum Technologies, a joint initiative between the EU and member states to support quantum research and innovation.

Collaborative Projects and Networks:

- European Quantum Flagship: Participate in the European Quantum Flagship initiative, aiming to consolidate and expand European scientific leadership and excellence in quantum research.
- Networking with EU Quantum Consortia: Foster relationships with existing quantum technology consortia and networks within the EU to collaborate on projects and share expertise.

Zaključak:

• **Public-Private Partnerships**: Encourage publicprivate partnerships that leverage EU funding to promote innovation and commercialization in the quantum sector.

Proposal Development and Submission:

- **Proposal Writing Teams**: Form specialized teams for developing and submitting proposals to EU funding calls. These teams should be adept in EU proposal requirements and include members with expertise in quantum technologies.
- **Regular Monitoring of EU Calls**: Keep a regular watch on EU funding calls relevant to quantum technologies to ensure timely and competitive applications.

Maximizing the Impact of EU Funds:

- **Complementary Funding**: Use EU funds to complement national investments in the NQI, ensuring a more extensive and impactful implementation of the initiative.
- **Demonstrating European Value**: In applications, emphasize how the NQI will contribute to Europe's leadership in quantum technologies and address broader EU goals like digital sovereignty and innovation leadership.

Conclusion:

Integrating European Union funding strategies into our National Quantum Initiative not only enhances the scope of our quantum projects but also aligns [Country's Name]'s quantum ambitions with broader European technological and strategic objectives. This approach will maximize the impact and reach of our NQI, positioning [Country's Name] as a key player in the European quantum ecosystem.

This section outlines how to effectively integrate European Union funding sources and strategies into the National Quantum Initiative, leveraging EU's focus on advancing quantum technologies. It suggests aligning with EU objectives, tapping into various EU funding programs, participating in EU quantum networks and projects, and ensuring that proposals highlight the contribution to Europe's quantum ambitions.

Could you be more precise in the kinds of EU programs the country could leverage, like EuroQCI and EuroHPC?

ChatGPT

Certainly! Including specific European Union programs like EuroQCI (European Quantum Communication Infrastructure) and EuroHPC (European High-Performance Computing) will add precision to the memo. Here's an enhanced section detailing these programs: Leveraging Specific EU Programs for the National Quantum Initiative

EuroQCI (European Quantum Communication Infrastructure):

- **Objective**: EuroQCI aims to build a secure quantum communication infrastructure that spans the entire EU, including its overseas territories. It focuses on enhancing cybersecurity through quantum technology.
- **Opportunity**: Collaborate on projects under EuroQCI to develop and integrate secure quantum communication systems within [Country's Name]. This includes quantum key distribution (QKD) networks and satellite-based quantum communication.
- Funding and Partnerships: Identify funding opportunities under EuroQCI for infrastructure development and research in quantum communication. Form partnerships with EU member states and private entities involved in the EuroQCI initiative.

EuroHPC (European High-Performance Computing):

- **Objective**: EuroHPC seeks to develop a world-class supercomputing ecosystem in Europe. Although primarily focused on classical computing, its infrastructure and research can significantly complement quantum computing initiatives.
- **Opportunity**: Leverage EuroHPC resources for quantum computing research, especially in areas where high-performance computing (HPC) intersects with quantum algorithms and simulations.
- Funding and Collaboration: Apply for EuroHPC funding for projects that integrate HPC and quantum computing. Collaborate with existing EuroHPC centers and initiatives to foster synergies between HPC and quantum technologies.

Horizon Europe:

- **Clusters and Missions:** Focus on relevant clusters and missions within Horizon Europe, particularly those pertaining to digital, industry, space, and health, where quantum technology applications are pivotal.
- Research and Innovation Actions (RIA): Participate in RIAs focusing on quantum technologies to develop new knowledge and integrate it into marketable products and services.
- Innovation and Technology Transfer: Engage in projects aimed at transferring quantum technology from research to market, utilizing Horizon Europe's support for innovation activities.

Digital Europe Programme:

- **Quantum Technology Development**: Target areas in the Digital Europe Programme that fund the development and deployment of quantum technologies.
- Skills and Capacity Building: Utilize the program's focus on skills and capacity building to train a workforce adept in quantum technologies, ensuring a talent pipeline for the NQI.

Enhanced Coordination with EU Policies:

- Alignment with EU Digital Strategy: Align NQI objectives with the EU's broader digital strategy, ensuring that [Country's Name]'s quantum initiative contributes to Europe's digital sovereignty.
- **Participation in Policy Discussions**: Actively participate in EU policy discussions and working groups on quantum technologies to influence and align with EU-wide quantum strategies.

Conclusion:

By strategically leveraging these specific EU programs, the National Quantum Initiative will not only secure necessary funding and support but also integrate [Country's Name] into the larger European quantum network. This alignment with EU's quantum and computing initiatives presents an opportunity to amplify the impact and reach of our national efforts in quantum science and technology.

This enhanced section provides a more detailed overview of how the National Quantum Initiative can align with and benefit from specific EU programs such as EuroQCI, EuroHPC, Horizon Europe, and the Digital Europe Programme. It emphasizes collaboration, funding opportunities, and policy alignment to maximize the benefits from these EU initiatives.

ChatGPT 4.0 on plot creation

January 2nd, 2024 Tested by Olivier Ezratty.

(uploaded) FidelityClusters.csv

Data Analysis

Great! You've uploaded a file named "FidelityClusters.csv". How can I assist you with this file? Are you looking for a specific analysis or visualization?

can you create a scatter plot in log-log scale with the qubit number and the two qubit gate error rates?

Data Analysis

Certainly! I will first load the data from "FidelityClusters.csv" to understand its structure, particularly focusing on the columns related to the qubit number and two-qubit gate error rates. Then, I will create a scatter plot in a log-log scale using these two data points. Let's start by loading and examining the data.

Erreur d'analyse

It seems there was an error in parsing the CSV file, possibly due to inconsistent numbers of fields across the rows. This can happen if there are missing or extra separators (like commas) in some rows.

To resolve this, I'll try a different approach to read the file, which allows for more flexibility in handling such inconsistencies. If possible, could you also provide more information about the structure of the file? For instance, the separator used (comma, tab, etc.), and if there are any known irregularities in the data format. This information would help in accurately loading the data.

The file has been partially read, but it appears that the data is not separated by commas. Instead, it looks like the separator might be a semicolon (;). Additionally, there was an issue with line 47, which we will need to address.

I will now attempt to read the file again using a semicolon as the separator and skip problematic lines like line 47 to ensure we can access the rest of the data correctly. Let's proceed with this approach.

The data has been successfully loaded. The relevant columns for your request seem to be "Qubits" for the qubit number and "2Q G errors" for the two-qubit gate error rates.

Next, I'll create a scatter plot with these columns on a log-log scale. Let's proceed with the visualization.



Here is the scatter plot showing the relationship between the number of qubits and the two-qubit gate error rates, both on a log-log scale. Each point represents a different quantum computing model. This type of plot can be useful for identifying patterns or trends in how error rates scale with the number of qubits.

If there's anything else you'd like to analyze or any other modifications you'd like on the plot, feel free to let me know!

new could you use a different color and a circle for each point, related to the company name? and invert the X and the Y?

Data Analysis

Sure! I will modify the scatter plot to include the following changes:

- 1. Use different colors for each point, based on the company name.
- 2. Use a circle ('o') as the marker shape for each point.
- 3. Invert the axes so that the two-qubit gate error rates are on the x-axis and the number of qubits is on the y-axis.

Let's create the updated plot.

It seems there was an issue with specifying the marker style for each company. When using different styles for each group, we need to provide a mapping of styles to the groups, or in this case, companies.

I'll adjust the code to automatically assign a distinct circle marker for each company. Let's try this again.



Here is the updated scatter plot with your requested modifications:

- Different colors for each company. •
- Circle markers for each data point.
- The x-axis represents the two-qubit gate error rates, and the y-axis represents the number of qubits, both on a log scale.

Each circle represents a different quantum computing model, colored according to the company. This visualization can help in comparing how the two-qubit gate error rates and the number of qubits vary across different companies.

can you provide the Python code to generate this plot?

Data Analysis

Certainly! Here's the Python code that generates the scatter plot with the specifications you requested:

(Python code)

Make sure to replace

'/path/to/your/FidelityClusters.csv' with the actual path to your CSV file. This code will create the scatter plot as seen earlier, with the specific modifications you requested. Let me know if there's anything else you need!

now can you change the Y scale and create marks for each possible values instead of the powers of 10, and put the Y scale at the right of the chart?

Data Analysis

To adjust the y-axis scale to show specific tick marks for each possible value of qubits instead of the default powers of 10, and to move the y-axis to the right side of the chart, we need to make a few modifications to the code. Here's the updated Python code to achieve this:

(new python code)

This code will create a scatter plot with the y-axis showing distinct tick marks for each qubit value and move the y-axis to the right of the plot. The x-axis remains on a log scale. Make sure to replace '/path/to/your/FidelityClusters.csv' with

the correct file path. Let me know if you need further adjustments!

can you show the resulting chart?

Data Analysis



Here is the updated scatter plot with the modifications you requested:

- The y-axis (Number of Qubits) now displays custom • tick marks for each unique qubit value.
- The y-axis is positioned on the right side of the chart.
- The x-axis (Two-Qubit Gate Error Rates) remains on a log scale.

Each point represents a different quantum computing model, with the colors and markers indicating the respective companies. If there's anything else you'd like to adjust or explore, feel free to ask!

could you use a log scale on Y? and use labels in % for the X axis?

Data Analysis



Here's the updated scatter plot with the following adjustments:

- Both the x-axis (Two-Qubit Gate Error Rates) and yaxis (Number of Qubits) are now on a log scale.
- The x-axis labels are formatted as percentages.

• The y-axis remains positioned on the right side of the chart.

Each point represents a different quantum computing model, differentiated by company using colors and circle markers. Note: The warning you see is related to the way matplotlib handles tick labels; it's a reminder that when custom formatting tick labels, they should match the tick positions. However, this doesn't affect the visual output of the plot.

If you have any more requests for modifications or additional analysis, please let me know!

well, in Y, I needed to have a log scale and the ticks corresponding to the values in the plot

Data Analysis

Scatter Plot: Two-Qubit Gate Error Rates vs. Number of Qubits (Log-Log Scale)



Here is the revised scatter plot with the following updates:

- Both x-axis (Two-Qubit Gate Error Rates) and y-axis (Number of Qubits) are on a log scale.
- The x-axis labels are formatted as percentages.
- The y-axis tick marks correspond to the actual values of qubits, appropriately positioned on the log scale.
- The y-axis is positioned on the right side of the chart.

Each point in the plot represents a different quantum computing model, categorized by company. The warning about **FixedFormatter** is a technical note related to matplotlib's label formatting, but it doesn't impact the plot's visual representation.

If you need any further adjustments or have additional questions, feel free to ask!

Here is the plot I created on my own and which required many hours of learning, work and tuning. The above took less than 5 minutes and is nearly on par with what I was looking for.



ChatGPT 4.0 on creating a quiz

January 14th, 2024 Tested by Olivier Ezratty.

I am a teacher in quantum computing. Create an educational quiz with 8 questions on quantum computing with multiple choices and the related responses. Multiple responses are possible in some of the questions in the quiz, with between 1 and 3 good responses out of 4. The level of the quiz should correspond to master's degree level. It can test the ability to understand circuit quantum programming.

ChatGPT

Creating an educational quiz with multiple-choice questions on quantum computing at a master's degree level is a great way to assess and enhance understanding, particularly in the area of circuit quantum programming. Here is a set of 8 questions, each with multiple choices. Some questions may have more than one correct answer.

Quantum Computing Quiz

1. Quantum Bits (Qubits)

What is true about qubits in quantum computing?

A. Qubits can only exist in states 0 or 1, like classical bits.

B. Qubits utilize quantum superposition, allowing them to be in a combination of states.

C. Qubits can perform faster calculations than classical bits due to their binary nature.

D. Quantum entanglement in qubits enables immediate communication over long distances.

Correct Answers: B

2. Quantum Superposition

Which statement best describes quantum superposition?

A. It is the alignment of spins in a magnetic field.

B. It allows a quantum system to be in a combination of multiple states simultaneously.

C. It refers to a quantum system's ability to teleport.

D. It's a phenomenon where quantum particles are duplicated. Correct Answer: B

3. Quantum Entanglement

Which of these statements are true about quantum entanglement?

A. Entangled particles remain interconnected even when separated by large distances.

B. The state of one entangled particle can instantaneously affect its pair.

C. Entanglement provides a method for faster-than-light communication.

D. Measurement of one entangled particle doesn't affect the other.

Correct Answers: A, B

4. Quantum Gates

In quantum computing, which of these gates is NOT a basic quantum gate?

A. Pauli-X Gate

B. Hadamard Gate

C. Toffoli Gate

D. Classical AND Gate

Correct Answer: D

5. Quantum Algorithms

Which of the following algorithms are fundamental in quantum computing?

A. Shor's Algorithm

B. Grover's Algorithm

C. Euclid's Algorithm

D. Dijkstra's Algorithm

Correct Answers: A, B

6. Quantum Error Correction

What is true about quantum error correction?

A. It is not necessary as qubits are immune to errors.

B. It employs classical error correction techniques.

C. It is crucial due to the susceptibility of qubits to decoherence and other errors.

D. Quantum error correction is simpler than classical error correction.

Correct Answer: C

7. Quantum Circuit Model

What characterizes a quantum circuit model?

A. It is primarily used for classical data processing.

B. It consists of qubits, quantum gates, and measurements.

C. It operates by changing the temperature of qubits.

D. It is used for long-distance quantum communication.

Correct Answer: B

8. Quantum Programming Languages

Which of these is/are used for programming quantum computers?

A. Python

B. Qiskit

C. C++

D. Q#

Correct Answers: B, D (Python and C++ are also used, although not as frameworks)

These questions cover various aspects of quantum computing, from the fundamental properties of qubits and quantum mechanics principles to specific applications in quantum algorithms and programming. They should provide a challenging yet insightful experience for students at a master's level in quantum computing, particularly those interested in circuit quantum programming.

Can you create 8 additional questions to the quiz?

ChatGPT

Absolutely, I'll create eight more questions to expand the quiz. These questions will delve deeper into the intricacies of quantum computing, testing knowledge in areas like quantum algorithms, decoherence, quantum computing hardware, and advanced concepts in quantum theory and applications.

Additional Quantum Computing Quiz Questions

9. Quantum Decoherence

What is true about quantum decoherence in quantum computing?

A. It enhances the stability of qubits.

B. It is a process where quantum information is lost to the environment.

C. It refers to the cooling process of qubits to near absolute zero.

D. Decoherence can increase the speed of quantum computations.

Correct Answer: B

10. Quantum Circuit Depth

What does the 'depth' of a quantum circuit refer to?

A. The number of qubits in the circuit.

B. The physical size of the quantum computer.

C. The number of layers of quantum gates.

D. The energy consumption of the quantum computer. Correct Answer: C

11. Bell State

Which of the following best describes a Bell State in quantum computing?

A. A state of maximum entropy in a quantum system.

B. A specific type of quantum error.

C. A maximally entangled quantum state of two qubits.

D. A state where qubits are in superposition but not entangled.

Correct Answer: C

12. Quantum Teleportation

In quantum teleportation,

A. particles are physically moved from one location to another.

B. information is transferred via classical channels only.

C. quantum information is transferred from one particle to another.

D. it enables faster-than-light communication.

Correct Answer: C

13. Quantum Annealing

What is quantum annealing primarily used for?

A. Error correction in quantum computers.

B. Solving optimization problems.

C. Encoding classical information into qubits.

D. Cooling down the quantum processors.

Correct Answer: B

14. No-Cloning Theorem

What does the quantum no-cloning theorem state?

A. Quantum information can be copied with 100% accuracy.

B. It is impossible to create an identical copy of an arbitrary unknown quantum state.

C. Cloning is only possible for classical states.

D. Quantum states can be cloned if they are entangled.

Correct Answer: B

15. Quantum Hardware

Which material is commonly used in the construction of superconducting qubits?

A. Silicon

B. Gallium Arsenide

C. Niobium

D. Graphene

Correct Answer: C

16. Quantum Simulation

What is a primary application of quantum simulation?

A. Simulating weather patterns.

B. Modeling quantum systems that are difficult to study in the laboratory.

C. Data mining and analysis.

D. Enhancing classical video game graphics.

Correct Answer: B

These additional questions delve into more specialized topics in quantum computing, challenging the understanding of complex concepts and current developments in the field. They're suitable for master's level students, particularly those focusing on advanced aspects of quantum computing and its applications. This extended quiz should provide a comprehensive assessment of their knowledge and understanding of both fundamental and cutting-edge topics in quantum computing.

Add 8 more questions focused on quantum error correction codes like surface codes and QLDPC.

ChatGPT

Certainly, focusing on quantum error correction codes, particularly surface codes and Quantum Low-Density Parity-Check (QLDPC) codes, provides a deeper dive into a crucial aspect of quantum computing. Here are eight additional questions that explore these advanced topics, suitable for master's level students specializing in quantum computing.

Quantum Error Correction Codes Quiz

17. Surface Codes

What is a key feature of surface codes in quantum error correction?

A. They require a large number of physical qubits for each logical qubit.

B. They are based on high-temperature superconductors.

C. Surface codes function by teleporting qubits to correct errors.

D. They are mainly used for quantum communication.

Correct Answer: A

18. Quantum Error Syndromes

In quantum error correction, what is an 'error syndrome'?

A. A set of qubits that are prone to errors.

B. The output of an error correction algorithm indicating the type of error.

C. A method to increase the error rate for testing purposes.

D. A syndrome that occurs exclusively in topological quantum computers.

Correct Answer: B

19. Logical vs Physical Qubits

In the context of quantum error correction, how does a logical qubit differ from a physical qubit?

A. A logical qubit is used for calculations, while a physical qubit stores data.

B. Logical qubits are error-corrected qubits composed of several physical qubits.

C. Physical qubits are only used in classical computing.

D. Logical qubits have a longer coherence time than physical qubits.

Correct Answer: B

20. Quantum Low-Density Parity-Check (QLDPC) Codes

What distinguishes Quantum Low-Density Parity-Check (QLDPC) codes in quantum error correction?

A. They require minimal physical qubits for each logical qubit.

B. QLDPC codes are based on classical parity-check codes with high density.

C. They are designed to correct a high number of errors simultaneously.

D. The codes are sparse, making them efficient for large-scale quantum computations.

Correct Answer: D

21. Error Correction in Topological Quantum Computers

How do topological quantum computers implement error correction?

A. By using surface codes and braiding of anyons.

B. Through high-frequency error syndrome measurements.

C. By implementing classical error correction algorithms.

D. Using a self-correcting quantum memory.

Correct Answer: A

22. Threshold Theorem

The threshold theorem in quantum computing states that:

A. Quantum computation is impossible if the error rate exceeds a certain threshold.

B. Error correction is unnecessary if the error rate is below a certain threshold.

C. Quantum computation can be made fault-tolerant if the error rate per gate is below a certain threshold.

D. The threshold for error correction is the same for all quantum computing models.

Correct Answer: C

23. Stabilizer Codes

Which statement correctly describes stabilizer codes in quantum error correction?

A. They stabilize the temperature of the quantum computer.

B. Stabilizer codes are a class of codes that use ancilla qubits for error detection.

C. They only work for error correction in classical computing systems.

D. Stabilizer codes increase the error rate to detect faults more easily.

Correct Answer: B

24. Cat Codes

Cat codes in quantum error correction are named because:

A. They are extremely agile and efficient in correcting errors.

B. They are named after Schrödinger's cat, a thought experiment in quantum mechanics.

C. The codes have a feline-like structure in their algorithm.

D. They were first developed by a researcher with the nickname "Cat."

Correct Answer: B

These additional questions cover intricate aspects of quantum error correction, a vital area in the development and sustainability of quantum computing. The focus on surface codes and QLDPC codes, along with general concepts in error correction and specific methodologies, provides a comprehensive examination for advanced students in quantum computing.

Add 8 more questions with 1 to 3 right responses out of 4 possible ones, this time, on quantum photonics.

ChatGPT

Certainly, focusing on quantum photonics adds an important dimension to the quiz, exploring the interaction of quantum mechanics with light and its applications in quantum technologies. Here are eight more questions, tailored to master's degree level students, which cover various aspects of quantum photonics. Each question has multiple choices, with one to three correct responses.

Quantum Photonics Quiz

25. Quantum Entanglement in Photonics

In the context of quantum photonics, which statements are true about quantum entanglement?

A. Entangled photons can be used for quantum key distribution.

B. Entangled photons are always of the same frequency.

C. Photon entanglement can be achieved using nonlinear optical processes.

D. Entangled photons can only be generated in vacuum.

Correct Answers: A, C

26. Single-Photon Sources

What are key characteristics of single-photon sources in quantum photonics?

A. They emit multiple photons in a coherent state.

B. They emit one photon at a time.

C. These sources are essential for quantum computing and secure communication.

D. They emit photons that are indistinguishable from each other.

Correct Answers: B, C, D

27. Photon Polarization

In quantum photonics, how is the polarization of a photon used?

A. As a classical method of communication.

B. To encode quantum information.

C. Only in the visible spectrum.

D. To measure the temperature of quantum systems.

Correct Answers: B

28. Applications of Quantum Photonics

Which are applications of quantum photonics?

A. Quantum computing.

B. Solar energy harvesting.

C. Quantum communication.

D. Biological imaging.

Correct Answers: A, C

29. Nonlinear Optics in Quantum Photonics

What role does nonlinear optics play in quantum photonics?

A. It's used to create entangled photon pairs.

B. It's mainly for amplifying classical communication signals.

C. It's used in the construction of quantum gates.

D. Nonlinear optics is irrelevant in quantum photonics.

Correct Answers: A

30. Photon Detectors in Quantum Experiments

What are essential features of photon detectors used in quantum photonics experiments?

A. High efficiency and low dark count rates.

B. Ability to detect multiple wavelengths simultaneously.

C. Only capable of detecting visible light.

D. They work only at room temperature.

Correct Answers: A

31. Quantum Dots in Photonics

Which statements are true about the use of quantum dots in quantum photonics?

A. Quantum dots can be used as on-demand single-photon sources.

B. They are primarily used for classical data storage.

C. Quantum dots can be used in quantum computing.

D. They can emit entangled photon pairs.

Correct Answers: A, C, D

32. Integrated Quantum Photonics

What is true about integrated quantum photonics?

A. It involves integrating quantum optical devices on a microchip.

B. Its primary application is in the field of classical telecommunications.

C. It aims to miniaturize devices for quantum information processing.

D. Integrated quantum

photonics technology is limited to infrared wavelengths.

Correct Answers: A, C

These questions encompass various elements of quantum photonics, including its principles, applications, and technologies. They are designed to challenge and gauge the understanding of students at a master's level, especially those specializing in quantum photonics and its role in advancing quantum computing and communications. The questions cover a range of topics from fundamental concepts to practical applications, ensuring a comprehensive assessment of knowledge in this dynamic field.

Add 8 more questions on superconducting qubit circuit designs and circuit quantum electrodynamics. Refer to Jaynes-Cummings Hamiltonian, Josephson junction and resonator designs. Refer to qubit readout techniques.

ChatGPT

Expanding the quiz to include questions on superconducting qubit circuit designs and circuit quantum electrodynamics (QED) is an excellent way to delve into more technical aspects of quantum computing. These questions will touch upon topics like the Jaynes-Cummings Hamiltonian, Josephson junctions, resonator designs, and qubit readout techniques, suitable for master's level students specializing in quantum computing and engineering.

Superconducting Qubits and Circuit QED Quiz

33. Jaynes-Cummings Hamiltonian

In the context of circuit QED, what does the Jaynes-Cummings Hamiltonian describe?

A. The interaction between a single qubit and a single-mode of the electromagnetic field.

B. The classical electromagnetic field interactions in a circuit.

C. The energy levels of a superconducting qubit without any external field.

D. The thermal properties of a superconducting circuit.

Correct Answer: A

34. Josephson Junction

What is the role of a Josephson Junction in superconducting qubits?

A. It provides a constant voltage source for the qubit.

B. It acts as a nonlinear inductor, enabling qubit state control and readout.

C. It functions as a conventional semiconductor diode.

D. It is used to cool the qubit to its ground state.

Correct Answer: B

35. Superconducting Qubit Types

Which of these are types of superconducting qubits?

A. Charge Qubits

- B. Spin Qubits
- C. Flux Qubits
- D. Transmon Qubits

Correct Answers: A, C, D

36. Resonator Designs in Circuit QED

In circuit QED, what is the purpose of resonator designs?

A. To amplify the signal for classical communication.

B. To provide a means for qubit-qubit coupling and qubit readout.

C. To generate non-classical light for quantum optics experiments.

D. To act as heat sinks for the superconducting circuits.

Correct Answer: B

37. Qubit Readout Techniques

Which are common qubit readout techniques in superconducting quantum circuits?

A. Dispersive readout

B. Fluorescence measurement

C. Homodyne detection

D. State-dependent frequency shift

Correct Answers: A, D

38. Role of Superconductivity in Qubits

How does superconductivity contribute to the functioning of qubits?

A. It eliminates all forms of electrical resistance in the qubit.

B. It allows for zero-energy state storage.

C. Superconductivity enables the creation of stable energy levels used for qubit states.

D. It is used to create entangled states in multi-qubit systems. Correct Answer: C

39. Coupling Qubits to Resonators

In the context of circuit QED, what is the significance of coupling qubits to resonators?

A. It is a method to dissipate heat from the qubits.

B. The coupling allows for the transfer of quantum information between qubits and resonators.

C. It is used to increase the qubit's energy consumption.

D. The coupling is purely for mechanical support of the qubit structure.

Correct Answer: B

40. Phase Qubits and Their Unique Feature

What is a distinctive feature of phase qubits in superconducting circuits?

A. They operate based on the phase difference across a Josephson junction.

B. Phase qubits have the highest coherence times among superconducting qubits.

C. They use optical photons for qubit operations.

D. These qubits are primarily used in quantum communication, not computing.

Correct Answer: A

These questions delve into the technical aspects of superconducting qubit circuit designs and circuit QED, crucial for advanced study in quantum computing. They cover a range of topics from fundamental principles of superconducting qubits to specific design and operational characteristics, providing a comprehensive and challenging assessment for students specializing in this area.

Add questions related to neutral atoms quantum computing.

ChatGPT

Expanding the quiz to include questions on neutral atoms quantum computing is an excellent way to explore this emerging area in quantum technology. This field leverages the unique properties of neutral atoms for quantum computation, offering different approaches and challenges compared to other quantum computing platforms. Here are eight questions that delve into various aspects of neutral atom quantum computing, suitable for master's level students.

Neutral Atoms Quantum Computing Quiz

41. Optical Lattices in Neutral Atom Quantum Computing

What is the role of optical lattices in neutral atom quantum computing?

A. They are used to cool the atoms to near absolute zero.

B. Optical lattices trap and arrange neutral atoms in a grid-like pattern.

C. They provide the optical pumping necessary for quantum state measurement.

D. Optical lattices are used to entangle atoms using photon exchange.

Correct Answer: B

42. Rydberg States

In the context of neutral atom quantum computing, what is significant about Rydberg states?

A. Rydberg states refer to the ground state of neutral atoms.

B. Atoms in Rydberg states have high principal quantum numbers, making them highly excited and sensitive to external fields.

C. Rydberg states are used for classical computing within a quantum system.

D. They are used to cool the atoms in optical lattices.

Correct Answer: B

43. Quantum Gates with Neutral Atoms

How are quantum gates implemented in neutral atom quantum computing?

A. By applying microwave pulses to induce state transitions.

B. Using laser-induced interactions, particularly between atoms in Rydberg states.

C. Through mechanical manipulation of individual atoms.

D. By applying a magnetic field across the entire quantum system.

Correct Answers: A, B

44. Advantages of Neutral Atoms in Quantum Computing

What are some advantages of using neutral atoms in quantum computing?

A. They can be easily ionized for better control.

B. Neutral atoms offer long coherence times.

C. They allow for the creation of large-scale quantum processors due to easy scalability.

D. Neutral atoms are not affected by electromagnetic interference.

Correct Answers: B, C

45. Atom Interferometry in Neutral Atom Quantum Computing

What is the role of atom interferometry in neutral atom quantum computing?

A. It is used to measure gravitational waves.

B. Atom interferometry is a technique for precise measurement of atomic properties relevant to quantum computing.

C. It is primarily used for temperature regulation in quantum systems.

D. Atom interferometry is unrelated to quantum computing. Correct Answer: B

46. Quantum State Initialization in Neutral Atoms

How are quantum states initialized in neutral atom quantum computing?

A. By aligning the spin of each atom using a strong magnetic field.

B. Using laser cooling and optical pumping techniques to prepare atoms in a specific quantum state.

C. Through electrical stimulation of each atom.

D. By cooling the atoms to their ground state using cryogenic methods.

Correct Answer: B

47. Measurement in Neutral Atom Quantum Computing

What characterizes measurement in neutral atom quantum computing?

A. Measurement is achieved through direct electrical readout.

B. Atoms are measured by detecting their fluorescence after being excited to a higher energy state.

C. The measurement process involves changing the spin orientation of the atoms using a magnetic field.

D. Atoms are measured by observing their interaction with a Bose-Einstein condensate.

Correct Answer: B

48. Challenges in Neutral Atom Quantum Computing

What are some of the challenges associated with neutral atom quantum computing?

A. Achieving uniformity in the arrangement of atoms within optical lattices.

B. The need for extremely low temperatures to maintain coherence.

C. Difficulty in isolating and addressing individual atoms for quantum operations.

D. Neutral atoms are highly susceptible to quantum decoherence.

Correct Answers: A, C

These questions cover the foundational concepts, techniques, and challenges associated with neutral atom quantum computing. This area of study is particularly relevant for students and researchers interested in exploring alternative approaches to quantum computation beyond the more commonly discussed superconducting and trapped ion methods. These questions should provide a comprehensive and challenging assessment of their understanding in this innovative field.

Where did you get all these questions?

ChatGPT

The questions in the quizzes on quantum computing topics like circuit quantum programming, quantum error correction codes, quantum photonics, superconducting qubit circuit designs, and neutral atoms quantum computing were created based on my training and knowledge in these areas. They are not sourced from any specific external quiz or textbook but are generated to reflect the typical depth and scope of topics that would be relevant and challenging for students at a master's level in quantum computing and related fields.

My training includes a wide range of information and concepts from various fields of study, including quantum physics and quantum computing, which enables me to generate educational content like quizzes, explanations, and discussions on these topics. The questions are designed to test understanding in these specialized areas, drawing from general principles, recent advancements, and technical aspects relevant to the field.

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VIII. INDEX

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¹⁶ To some extent, the vector dimensions remind us of the Hilbert space that are used to represent a quantum object or a qubit state. One key difference is these dimensions pertain to single objects, not a collection of objects like with qubit registers. They share the fact that they both represent probabilities, classical with LLMs and quantum for qubit measurement and they embody a notion of superposition, of meaning for LLM vectors and state for qubit registers. But the dimensions have different meanings, of course. They are linked to semantic concepts with LLM and to superposition of computational basis quantum states with qubits. Still, the vector dimension in LLMs is a way to create an orthonormal basis for concept meanings. Also, probabilities are real numbers in LLM vectors while they are based on complex numbers with qubit states and there is no notion of state collapse when applying probabilities in LLMs to predict the following words. At last, adding qubits increases the computational space size exponentially, while in LLMs and NLP, adding dimensions to vectors increases resources linearly or polynomially.

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¹²⁰ <u>Are Emergent Abilities of Large Language Models a Mirage?</u> by Rylan Schaeffer et al, Stanford University, April-May 2023 (14 pages).

¹²¹ Large Language Models for Mathematical Reasoning: Progresses and Challenges by Janice Ahn et al, arXiv, January 2024 (8 pages).

¹²² Artificial General Intelligence Is Already Here by Blaise Agüera y Arcas and Peter Norvig, NOEMA, October 2023.

¹²³ Sparks of Artificial General Intelligence: Early experiments with GPT-4 by Sébastien Bubeck et al, Microsoft Research, arXiv, March-April 2023 (155 pages).

¹²⁴ <u>Can AI Be as Creative as Humans?</u> by Haonan Wang et al, arXiv, January 2024 (19 pages).

¹²⁵ One example deals with using LLMs in Automated Planning and Scheduling (APS) artificial intelligence systems, as inventoried in <u>On the Prospects of Incorporating Large Language Models (LLMs) in Automated Planning and Scheduling (APS)</u> by Vishal Pallagani et al, arXiv, January 2024 (22 pages).

¹²⁶ TechGPT-2.0: A large language model project to solve the task of knowledge graph construction by Jiaqi Wang et al, arXiv, January 2024 (10 pages).

¹²⁷ <u>A Survey of Reasoning with Foundation Models</u> by Jiankai Sun et al, arXiv, December 2023-January 2024 (161 pages).

¹²⁸ On the Opportunities and Risks of Foundation Models by Rishi Bommasani et al, arXiv, August 2021-July 2022 (214 pages).

¹²⁹ <u>Self-Contrast: Better Reflection Through Inconsistent Solving Perspectives</u> by Wenqi Zhang et al, arXiv, DJanuary 2024 (20 pages). ¹³⁰ Levels of AGI: Operationalizing Progress on the Path to AGI by Meredith Ringel Morris et al, arXiv, November 2023 (19 pages).

¹³¹ <u>Towards AGI in Computer Vision: Lessons Learned from GPT and Large Language Models</u> by Lingxi Xie et al, arXiv, June 2023 (17 pages) shows that building some forms of AGI will also require to improve artificial vision capabilities.

¹³² AGI: Artificial General Intelligence for Education by Ehsan Latif et al, arXiv, March 2023 (30 pages).

¹³³ See the call for prudence in <u>Managing AI Risks in an Era of Rapid Progress</u> by Yoshua Bengio, Geoffrey Hinton, Yuval Noah Harari, Stuart Russell et al, arXiv, October-November 2023 (7 pages).

¹³⁴ <u>Model evaluation for extreme risks</u> by Toby Shevlane, Yoshua Bengio et al, DeepMind, arXiv, May-September 2023 (20 pages). The risks laid out in the paper would deserve to be analyzed with the scope of quantum sciences.

¹³⁵ <u>Can A Cognitive Architecture Fundamentally Enhance LLMs? Or Vice</u> <u>Versa?</u> by Ron Sun, arXiv, January 2024 (22 pages).

¹³⁶ <u>AI And The Limits Of Language</u> by Jacob Browning and Yann LeCun, NOEMA, August 2022.

¹³⁷ Faith and Fate: Limits of Transformers on Compositionality by Nouha Dziri et al, arXiv, May-October 2023 (40 pages).

¹³⁸ What will GPT-2030 look like? by Jacob Steinhardt, Bounded Regret, June 2023.

¹³⁹ <u>Autonomous chemical research with large language models</u> by Daniil A. Boiko, Robert MacKnight, Ben Kline, Gabe Gomes, Nature, December 2023 (13 pages) describes the CoScientist tool.

¹⁴⁰ <u>ChemCrow: Augmenting large-language models with chemistry tools</u> by Andres M Bran, Sam Cox, Oliver Schilter, Carlo Baldassari, Andrew D White, and Philippe Schwaller, arXiv, April 2023 (38 pages).

¹⁴¹ One example deals with the way LLMs understand some complicated logic like with understanding the relationships between computing complexity classes. See a related <u>LinkedIn post</u> from Laurent Prost at Alice&Bob.

¹⁴² Generative AI: Implications and Applications for Education by Anastasia Olga (Olnancy)Tzirides et al, arXiv, May 2023 (34 pages) describes some concerns with generative AI-based education: the machine buries its sources, can have no notion of empirical truth, has no conception of a theoretical frame or disciplinary practice, requires tuning to follow ethical rules, and is somewhat overly deferent to users. Thus, the need for their users to have preliminary knowledge and good judgment on the scientific topic's studies with LLMs and other forms of generative AI.

¹⁴³ <u>ChatGPT's explanation of quantum computing is bullsh*t</u>, by Chris Ferrie, January 2023. Which is a little exaggerated when he criticizes ChatGPT 3.5's use of superposition and entanglement to explain the power of quantum computing.

¹⁴⁴ DecodingTrust: A Comprehensive Assessment of Trustworthiness in GPT Models by Boxin Wang et al, June-December 2023 (110 pages).

¹⁴⁵ Exploring the Possibility: Can Quantum Computers Unlock the Secrets of <u>Time Travel?</u> by Hillary, TechBullion, November 2023. This paper seems to have been produced with some LLM chatbot given the nonsense it contains.

¹⁴⁶ Language Models are Few-Shot Learners by Tom B. Brown, et al, arXiv, May-July 2020 (32 pages).

¹⁴⁷ ChatGPT 4.0 also explained that creating a similar chart with Excel 365 would require selecting manually each bar to change their colors. Microsoft is funding OpenAI massively but is unable to implement such a feature. What a strange world!

¹⁴⁸ Why ChatGPT Is Getting Dumber at Basic Math by Josh Zumbrun, WSJ, August 2023.

¹⁴⁹ Large Language Models for Mathematicians by Simon Frieder et al, December 2023 (13 pages).

¹⁵⁰ <u>Solving Quantitative Reasoning Problems with Language Models</u> by Aitor Lewkowycz et al, arXiv, June 2022 (12 pages).

¹⁵¹ math-PVS: A Large Language Model Framework to Map Scientific Publications to PVS Theories by Hassen Saidi, Susmit Jha, and Tuhin Sahai, arXiv, October 2023 (47 pages).

¹⁵² Solving olympiad geometry without human demonstrations by Trieu H. Trinh, Yuhuai Wu, Quoc V. Le, He He, Thang Luong, Google, Nature, January 2024 (21 pages). ¹⁵³ <u>GPT-4 gets a B on my quantum computing final exam!</u> By Scott Aaronson, April 2023.

¹⁵⁴ This level of good responses is found in studies covering other technical skills. For example, <u>ChatGPT and Software Testing Education: Promises & Perils</u> by Sajed Jalil, Suzzana Rafi, Thomas D. LaToza, Kevin Moran, and Wing Lam, arXiv, February-March 2023 (8 pages) describes a ChatGPT 4.0 software testing skills assessment that yielded 77.5% good responses.

¹⁵⁵ Assessing Large Language Models in Mechanical Engineering Education: A Study on Mechanics-Focused Conceptual Understanding by Jie Tian et al, arXiv, January 2024 (30 pages) shows that in other disciplines like mechanical engineering, LLMs surpass humans in tests.

¹⁵⁶ When a Quantum Scientist Interviews ChatGPT by Rasit Topaloglu, TechRxiv, February 2023 (12 pages).

¹⁵⁷ What does ChatGPT really know about physics? by Sidney Perkowitz, PhysicsWorld, March 2023.

¹⁵⁸ <u>How Well Does ChatGPT Understand Quantum Computing? Let's Find</u> <u>Out!</u> by Lorenzo Leandro, Quantum Machines, December 2022.

¹⁵⁹ Evaluating Large Language Models on the GMAT: Implications for the Future of Business Education by Vahid Ashrafimoghari, Necdet Gürkan, and Jordan W. Suchow, arXiv, January 2024 (30 pages).

¹⁶⁰ PlanBench: An Extensible Benchmark for Evaluating Large Language Models on Planning and Reasoning about Change by Karthik Valmeekam et al, arXiv, June 2022-November 2023 (50 pages) describes a benchmark proposal for LLM reasoning capabilities. Such benchmarks will have to be created on quantum science with segmenting well the various types of reasoning to be tested. The Scott Aaronson course test can be a good starter here, to be complemented with similar tests in quantum physics.

¹⁶¹ I got the idea to create a quantum quiz with reading <u>How Teachers Can Use</u> <u>Large Language Models and Bloom's Taxonomy to Create Educational Quizzes</u> by Sabina Elkins, Ekaterina Kochmar, Jackie C.K. Cheung, and Iulian Serban, arXiv, January 2024 (8 pages).

¹⁶² Evaluation of ChatGPT Applicability to Learning Quantum Physics by Agata Stefańska et al, IEEE Xplore, September 2023.

¹⁶³ <u>The (mis)use of ChatGPT in science and education</u> by David Shaw et al, EMBO Reports, June 2023.

¹⁶⁴ Scientists' Perspectives on the Potential for Generative AI in their Fields by Meredith Ringel Morris, arXiv, April 2023 (26 pages) polled scientists in various disciplines including quantum chemistry, quantum computing hardware, and quantum algorithms from Alphabet which abounds with such examples.

¹⁶⁵ Adapting Large Language Models for Education: Foundational Capabilities, <u>Potentials, and Challenges</u> by Qingyao Li, Lingyue Fu, Weiming Zhang, Xianyu Chen, Jingwei Yu, Wei Xia, Weinan Zhang, Ruiming Tang, and Yong Yu, arXiv, January 2024 (13 pages) describes generic benefits of LLM usage in education.

¹⁶⁶ Can Large Language Model Summarizers Adapt to Diverse Scientific Communication Goals? by Marcio Fonseca, and Shay B. Cohen, arXiv, January 2024 (19 pages).

¹⁶⁷ <u>LLaMA: Open and Efficient Foundation Language Models</u> by Hugo Touvron, Thibaut Lavril et al, arXiv, February 2023 (27 pages) says that it used parts of LaTEX source files for its training. So, papers like mine which are provided only as PDF files are not processed here. But RAGs can be trained with PDF files, as shown for example in <u>AI in 5: RAG with PDFs</u> by Ian Kelk, Clarifai, December 2023.

¹⁶⁸ Logical quantum processor based on reconfigurable atom arrays by Dolev Bluvstein, Markus Greiner, Vladan Vuletić, Mikhail D. Lukin et al, Nature, December 2023 (37 pages).

¹⁶⁹ Assessing the Ability of ChatGPT to Screen Articles for Systematic Reviews by Eugene Syriani, Istvan David, and Gauransh Kumar, arXiv, June 2023 (41 pages).

¹⁷⁰ <u>https://typeset.io/</u>.

¹⁷¹ From Google Gemini to OpenAI Q* (Q-Star): A Survey of Reshaping the Generative Artificial Intelligence (AI) Research Landscape by Timothy R. McIntosh, Teo Susnjak, Tong Liu, Paul Watters, and Malka N. Halgamuge, arXiv, December 2023 (30 pages) discuss the need for AI related papers but the need also exists in quantum science.

¹⁷² A Review of Findings from Neuroscience and Cognitive Psychology as Possible Inspiration for the Path to Artificial General Intelligence by Florin Leon, arXiv, January 2024 (143 pages).

¹⁷³ API stands for "application programming interface", a set of instructions used by programming languages to access to features of a service that can be either online or local, running on the user device.

¹⁷⁴ ReviewerGPT? An Exploratory Study on Using Large Language Models for Paper Reviewing by Ryan Liu, and Nihar B. Shah, June 2023 (60 pages).

¹⁷⁵ <u>GPT vs Human for Scientific Reviews: A Dual Source Review on Applications of ChatGPT in Science</u> by Chenxi Wu, Alan John Varghese, Vivek Oommen, and George Em Karniadakis, Brown University, December 2023 (60 pages).

¹⁷⁶ <u>SciGraphQA: A Large-Scale Synthetic Multi-Turn Question-Answering</u> <u>Dataset for Scientific Graphs</u> by Shengzhi Li, and Nima Tajbakhsh, arXiv, August 2023 (12 pages).

¹⁷⁷ As found in <u>ChatGPT in the Classroom: Boon or Bane for Physics Students'</u> <u>Academic Performance?</u> by Manuel G. Forero, and H. J. Herrera-Suárez, December 2023 (9 pages)

¹⁷⁸ Seeing ChatGPT Through Universities' Policies, Resources and Guidelines by Hui Wang, Anh Dang, Zihao Wu, and Son Mac, December 2023 (25 pages).

¹⁷⁹ <u>AI Use in Manuscript Preparation for Academic Journals</u> by Nir Chemaya, and Daniel Martin, November 2023 (26 pages).

¹⁸⁰ AI, write an essay for me: A large-scale comparison of human-written versus <u>ChatGPT-generated essays</u> by Steffen Herbold et al, April 2023 (27 pages).

¹⁸¹ Beyond Black Box AI-Generated Plagiarism Detection: From Sentence to Document Level by Mujahid Ali Quidwai et al, IBM Research, June 2023 (10 pages).

¹⁸² Detecting LLM-Generated Text in Computing Education: A Comparative Study for ChatGPT Cases by Michael Sheinman Orenstrakh et al, July 2023 (18 pages).

¹⁸³ <u>Is it possible for ChatGPT to mimic human annotator?</u> by Oliwier Kaszyca et al, TechRxiv, December 2023(15 pages).

¹⁸⁴ <u>The Science of Detecting LLM-Generated Texts</u> by Ruixiang Tang, Yu-Neng Chuang, and Xia Hu, arXiv, February 2023 (10 pages).

¹⁸⁵ <u>A Benchmark Dataset to Distinguish Human-Written and Machine-Generated Scientific Papers</u> by Mohamed Hesham Ibrahim Abdalla et al, MDPI, September 2023 (33 pages).

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¹⁸⁷ Enhancing Robustness of LLM-Synthetic Text Detectors for Academic Writing: A Comprehensive Analysis by Zhicheng Dou, Yuchen Guo, Ching-Chun Chang, Huy H. Nguyen, and Isao Echizen, arXiv, January 2024 (12 pages).

¹⁸⁸ <u>AI and science: what 1,600 researchers think</u> by Richard Van Noorden and Jeffrey M. Perkel, Nature, October 2023 (5 pages).

¹⁸⁹ ChatGPT use shows that the grant-application system is broken by Juan Manuel Parrilla, Nature, October 2023.

¹⁹⁰ Double-Edged Sword of AI in Professional Services: BCG GPT-4 Findings More Cautionary Than Reported by Courtlin Holt-Nguyen, September 2023.

¹⁹¹ <u>GPT-4 as an Effective Zero-Shot Evaluator for Scientific Figure Captions</u> by Ting-Yao Hsu et al, arXiv, October 2023 (11 pages).

¹⁹² The Transformative Influence of Large Language Models on Software Development by Sajed Jalil, George Mason University, arXiv, November 2023 (10 pages) describes the various benefits from LLM-base software development aids.

¹⁹³ <u>DebugBench: Evaluating Debugging Capability of Large Language Models</u> by Runchu Tian et al, arXiv, January 2024 (27 pages) shows how the debugging capabilities of ChatGPT improved from version 3.5 to 4.0. 4.0 closed half of the quality gap with human debuggers vs 3.5.

¹⁹⁴ Demonstration of InsightPilot: An LLM-Empowered Automated Data Exploration System by Pingchuan Ma, Rui Ding, Shuai Wang, Shi Han, and Dongmei Zhang, arXiv, April-November 2023 (7 pages).

¹⁹⁵ LLMs for Science: Usage for Code Generation and Data Analysis by Mohamed Nejjar, Luca Zacharias, Fabian Stiehle, and Ingo Weber, Technical University of Munich, arXiv, November 2023 (8 pages).

¹⁹⁶ IBM watsonx - An AI and data platform built for business by IBM.

¹⁹⁷ C4Q is a similar project. See <u>C4Q: A Chatbot for Quantum</u> by Yaiza Aragonés-Soria, and Manuel Oriol, arXiv, January 2024 (8 pages).

¹⁹⁸ See the <u>Copilot for Azure Quantum FAQ</u>, Microsoft, November 2023.

¹⁹⁹ CodeAid: Evaluating a Classroom Deployment of an LLM-based Programming Assistant that Balances Student and Educator Needs by Majeed Kazemitabaar et al, arXiv, January 2024 (20 pages).

²⁰⁰ ChatGPT-4: first approach to its use, scope and limitations in quantum algorithms by Ezequiel Murina Moreno, José Ignacio García and Martín Hurtado Heredia, QuantumPath, March 2023 shows faults at the logical-mathematical level and API requests in the case of quantum annealing code, but fewer with gate-based code, due to the differences in available code online.

²⁰¹ An Example of Evolutionary Computation + Large Language Model Beating Human: Design of Efficient Guided Local Search by Fei Liu et al, arXiv, January 2024 (12 pages).

²⁰² How AI Ideas Affect the Creativity, Diversity, and Evolution of Human Ideas: Evidence From a Large, Dynamic Experiment by Joshua Ashkinaze, Julia Mendelsohn, Li Qiwei, Ceren Budak, and Eric Gilbert, University of Michigan, arXiv, January 2024 (42 pages) also tend to show that LLM usage doesn't impair scientists creativity.

²⁰³ <u>Can ChatGPT be used to generate scientific hypotheses?</u> by Yang Jeong Park, Daniel Kaplan, Zhichu Ren, Chia-Wei Hsu, Changhao Li, Haowei Xu, Sipei Li, and Ju Li, arXiv, March 2023 (37 pages).

²⁰⁴ Hyperion Research Pins 2023 Global Quantum Computing Market at \$848M with 22.1% CAGR for 2024 by Hyperion Research, HPCwire, December 2023.

²⁰⁵ Otherwise, the metaverse hype from late 2021 would have cut all corporate investments in both artificial intelligence and quantum technologies.

²⁰⁶ <u>Risk Taxonomy, Mitigation, and Assessment Benchmarks of Large Language Model Systems</u> by Tianyu Cui et al, arXiv, January 2024 (29 pages).

²⁰⁷ <u>Quantum supremacy using a programmable superconducting processor</u> by Frank Arute, Hartmut Neven, John M. Martinis et al, Nature, October 2019 (7 pages).

²⁰⁸ Supplementary information for "Quantum supremacy using a programmable superconducting processor" by Frank Arute, Hartmut Neven, John M. Martinis et al, arXiv, October-December 2019 (67 pages).

²⁰⁹ <u>Understanding Quantum Technologies 2023</u> by Olivier Ezratty, September 2023 (1,366 pages).

²¹⁰ Artificial Intelligence for Science in Quantum, Atomistic, and Continuum Systems by Xuan Zhang et al, arXiv, July-November 2023 (275 pages).

²¹¹ <u>Highly accurate protein structure prediction with AlphaFold</u> by John Jumper et al, DeepMind, Nature, July 2021 (12 pages).

²¹² <u>Resource-efficient quantum algorithm for protein folding</u> by Anton Robert et al, 2021 (5 pages).

²¹³ <u>Modern applications of machine learning in quantum sciences</u> by Anna Dawid, Florian Marquardt, Maciej Lewenstein, Alexandre Dauphin et al, arXiv, April 2022-November 2023 (288 pages).

²¹⁴ <u>Artificial Intelligence and Machine Learning for Quantum Technologies</u> by Mario Krenn, Jonas Landgraf, Thomas Foesel, and Florian Marquardt, MPI, arXiv, August 2022 (23 pages).

²¹⁵ <u>Computer-inspired quantum experiments</u> by Mario Krenn, Manuel Erhard, Anton Zeilinger, Nature Reviews Physics, September 2020 (17 pages).

²¹⁶ Experimental Deep Reinforcement Learning for Error-Robust Gateset Design on a Superconducting Quantum Computer by Yuval Baum, Michael J. Biercuk et al, Q-CTRL, May 2021 (12 pages).

²¹⁷ Machine learning optimal control pulses in an optical quantum memory experiment by Elizabeth Robertson, Luisa Esguerra, Leon Messner, Guillermo Gallego, and Janik Wolters, arXiv, January 2024 (9 pages).

²¹⁸ Optimizing quantum gates towards the scale of logical qubits by Paul V. Klimov, Hartmut Neven et al, Google AI and University of California, August 2023 (34 pages).

²¹⁹ Machine learning to scale up the quantum computer by Muhammad Usman and Lloyd Hollenberg, University of Melbourne, March 2020.

²²⁰ <u>Autotuning of Double-Dot Devices In Situ with Machine Learning</u> by Justyna P. Zwolak et al, PRA, March 2020 (10 pages).

²²¹ From basic science to technological development: the case for two avenues by Matteo Carlesso and Mauro Paternostro, Queens University Belfast, May 2023 (22 pages).

²²² Boosting the Performance of Quantum Annealers using Machine Learning by Jure Brence et al, March 2022 (14 pages).

²²³ Language models for quantum simulation by Roger G. Melko, Juan Carrasquilla, Nature Computational Science, January 2024 (8 pages, <u>PDF</u>).

²²⁴ Bridging the Reality Gap in Quantum Devices with Physics-Aware Machine Learning by D. L. Craig, Natalia Ares et al, Physical Review X, January 2024 (16 pages).

²²⁵ Machine Learning for Practical Quantum Error Mitigation by Haoran Liao, Zlatko K. Minev et al, IBM, September 2023 (15 pages).

²²⁶ Benchmarking Machine Learning Models for Quantum Error Correction by Tim Fu, and Yue Zhao, arXiv, November-December 2023 (19 pages).

²²⁷ <u>Best practices for quantum error mitigation with digital zero-noise</u> <u>extrapolation</u> by Ritajit Majumdar, Pedro Rivero, Friederike Metz, Areeq Hasan, and Derek S Wang, arXiv, July 2023 (10 pages).

²²⁸ qecGPT: decoding Quantum Error-correcting Codes with Generative Pretrained Transformers by Hanyan Cao et al, arXiv, July 2023 (18 pages).

²²⁹ Quantum circuit optimization with deep reinforcement learning by Thomas Fösel, Florian Marquardt et al, March 2021 (10 pages).

²³⁰ Improving Quantum Circuit Synthesis with Machine Learning by Mathias Weiden et al, arXiv, June 2023 (11 pages).

²³¹ Deep neural networks for quantum circuit mapping - Neural Computing and <u>Applications</u> by Giovanni Acampora and Roberto Schiattarella, SpringerLink, May 2021 (20 pages).

²³² <u>A Reinforcement Learning Environment for Directed Quantum Circuit</u> <u>Synthesis</u> by Michael Kölle, Tom Schubert, Philipp Altmann, Maximilian Zorn, Jonas Stein, and Claudia Linnhoff-Popien, LMU Munich, arXiv, January 2024 (12 pages).

²³³ <u>Unleashing the Potential of LLMs for Quantum Computing: A Study in</u> <u>Quantum Architecture Design</u> by Zhiding Liang et al, July 2023 (9 pages).

²³⁴ The generative quantum eigensolver (GQE) and its application for ground state search by Kouhei Nakaji, Alan Aspuru-Guzik et al, arXiv, January 2024 (16 pages).

²³⁵ Exploring the Advantages of Quantum Generative Adversarial Networks in Generative Chemistry by Po-Yu Kao, Alán Aspuru-Guzik et al, JCIM, May 2023 (12 pages). This Zapata quantum MolGAN was experimented with 2 to 8 qubits with no resource estimates for larger settings in a quantum advantage regime. So, we are not yet in any quantum advantage regime.

²³⁶ <u>Quantum Generative Models for Small Molecule Drug Discovery</u> by Junde Li, Rasit Topaloglu, and Swaroop Ghosh, Pennsylvania State University and IBM, arXiv, January 2021 (6 pages).

²³⁷ Classifying snapshots of the doped Hubbard model with machine learning by Annabelle Bohrdt et al, Nature Physics, July 2019 (9 pages).

²³⁸ Machine Learning Cryptanalysis of a Quantum Random Number Generator by Nhan Duy Truong et al, 2019 (13 pages).

²³⁹ Benchmarking a Quantum Random Number Generator with Machine Learning, by Jin Yan How, 2020 (26 slides).

²⁴⁰ A Survey of Machine Learning Assisted Continuous-Variable Quantum Key <u>Distribution</u> by Nathan K. Long, Robert Malaney, Kenneth J. Grant, MDPI, October 2023 (23 pages).

²⁴¹ <u>Magnetic-Field Learning Using a Single Electronic Spin in Diamond with</u> <u>One-Photon Readout at Room Temperature</u> by Raffaele Santagati et al, 2018 (18 pages).

²⁴² <u>Understanding Quantum Technologies</u> by Olivier Ezratty, September 2023 (1,366 pages).

²⁴³ Systematic literature review: Quantum machine learning and its applications by David Peral-García et al, Computer Science Review, February 2024 (20 pages) makes a good inventory of QML algorithms.

²⁴⁴ <u>The AI-quantum computing mash-up: will it revolutionize science?</u> by Davide Castelvecchi, Nature, January 2024.

²⁴⁵ One recent example is <u>Overcoming the Coherence Time Barrier in Quantum</u> <u>Machine Learning on Temporal Data</u> by Fangjun Hu et al, IBM Quantum, Raytheon BBN and Princeton University, arXiv, December 2023 (28 pages) which was tested on just 7 qubits.

²⁴⁶ <u>Quantum Machine Learning: from NISQ to Fault Tolerance</u> by Yunfei Wang, and Junyu Liu, arXiv, January 2024 (28 pages) describe the various dependencies of quantum machine learning algorithms speedup with data sparsity and data loading capacities.

²⁴⁷ Quantum Natural Language Processing on Near-Term Quantum Computers by Konstantinos Meichanetzidis, Bob Coecke et al, Oxford University and CQC, arXiv, May 2020 (17 pages).

²⁴⁸ <u>Quantum Natural Language Processing: Challenges and Opportunities</u> by Raffaele Guarasci, Giuseppe De Pietro, Massimo Esposito, MDPI, June 2022 (25 pages).

²⁴⁹ <u>QNLP in Practice: Running Compositional Models of Meaning on a Quantum Computer</u> by Robin Lorenz et al, Quantinuum, Journal of Artificial Intelligence Research, April 2023 (38 pages) with tests running on 5 qubits.

²⁵⁰ Quantum Computing and the Future of Large Language Models (such as ChatGPT) by Tom Martin, Medium, April 2023.

²⁵¹ <u>Towards provably efficient quantum algorithms for large-scale machinelearning models</u> by Junyu Liu, Jens Eisert, Liang Jiang et al, University of Chicago, Nature Communications, January 2024 (6 pages) shows that a 7 billion parameter LLM could be trained efficiently with 32 logical qubits, but at two conditions: the data must be sparse or be preliminarily loaded using a (non-existing) qRAM. The paper also doesn't provide any indications of the training time, and only mathematically formulas that you must compute on your own.

²⁵² Accelerating Neural Networks for Large Language Models and Graph Processing with Silicon Photonics by Salma Afifi, Febin Sunny, Mahdi Nikdast, and Sudeep Pasricha, arXiv, January 2024 (xx pages).

²⁵³ See the review paper <u>Quantum inspired algorithms in practice</u> by Juan Miguel Arrazola, Seth Lloyd et al, 2020 (24 pages).

²⁵⁴ AutoQML: Automatic Generation and Training of Robust Quantum-Inspired Classifiers by Using Genetic Algorithms on Grayscale Images by Sergio Altares-López et al, August 2022 (13 pages) improves medical imaging grey images classification using a quantum inspired machine learning algorithm.

²⁵⁵ Robust Dequantization of the Quantum Singular value Transformation and Quantum Machine Learning Algorithms by François Le Gall, April 2023 (55 pages).

²⁵⁶ <u>Quantum-Inspired Machine Learning: a Survey</u> by Larry Huynh et al, August 2023 (56 pages).

²⁵⁷ See one example in <u>Sampling-based sublinear low-rank matrix arithmetic framework for dequantizing quantum machine learning</u> by Nai-Hui Chia, Ewin Tang et al, October 2019 (79 pages).

²⁵⁸ What are Quantized LLMs? by Miguel Carreira Neves, TensorOps, September 2023.

²⁵⁹ Quantization is one out of many other model compression techniques. See <u>A</u> <u>Comprehensive Survey of Compression Algorithms for Language Models</u> by Seungcheol Park et al, arXiv, January 2024 (35 pages).

²⁶⁰ <u>Multiverse Computing Launches CompactifAI to Streamline LLMs to</u> <u>Reduce Energy Use and Compute Costs</u> by Multiverse, November 2023.

²⁶¹ <u>CompactifAI: Extreme Compression of Large Language Models using Quantum-Inspired Tensor Networks</u> by Andrei Tomut, Samuel Mugel, Roman Orus et al, Multiverse, arXiv, January 2024 (4 pages).

 ²⁶² Hyper Intelligence Releases Quantum-Inspired Algorithm Designed to Reduce Cost of LLMs by Matt Swayne, The Quantum Insider, November 2023.
²⁶³ In reference to the famous JFK quote from his inaugural address on January 2021

²⁶⁴ ESGReveal: An LLM-based approach for extracting structured data from ESG reports by Yi Zou et al, arXiv, December 2023 (17 pages) proposes a way to extract structured data from environmental and social and governance reports. We can hope this sort of technique could be implemented in an ad-hoc fashion for some papers in quantum science.