

IBM Quantum @Q2B 2023

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Technical Program Director

Algorithms and Partnerships, IBM Quantum



IBM Quantum Summit 2023

The Era of Quantum Utility

New IBM Quantum Processors: Condor, Heron

IBM Quantum System Two

Software: Qiskit 1.0 & Bringing AI to Quantum

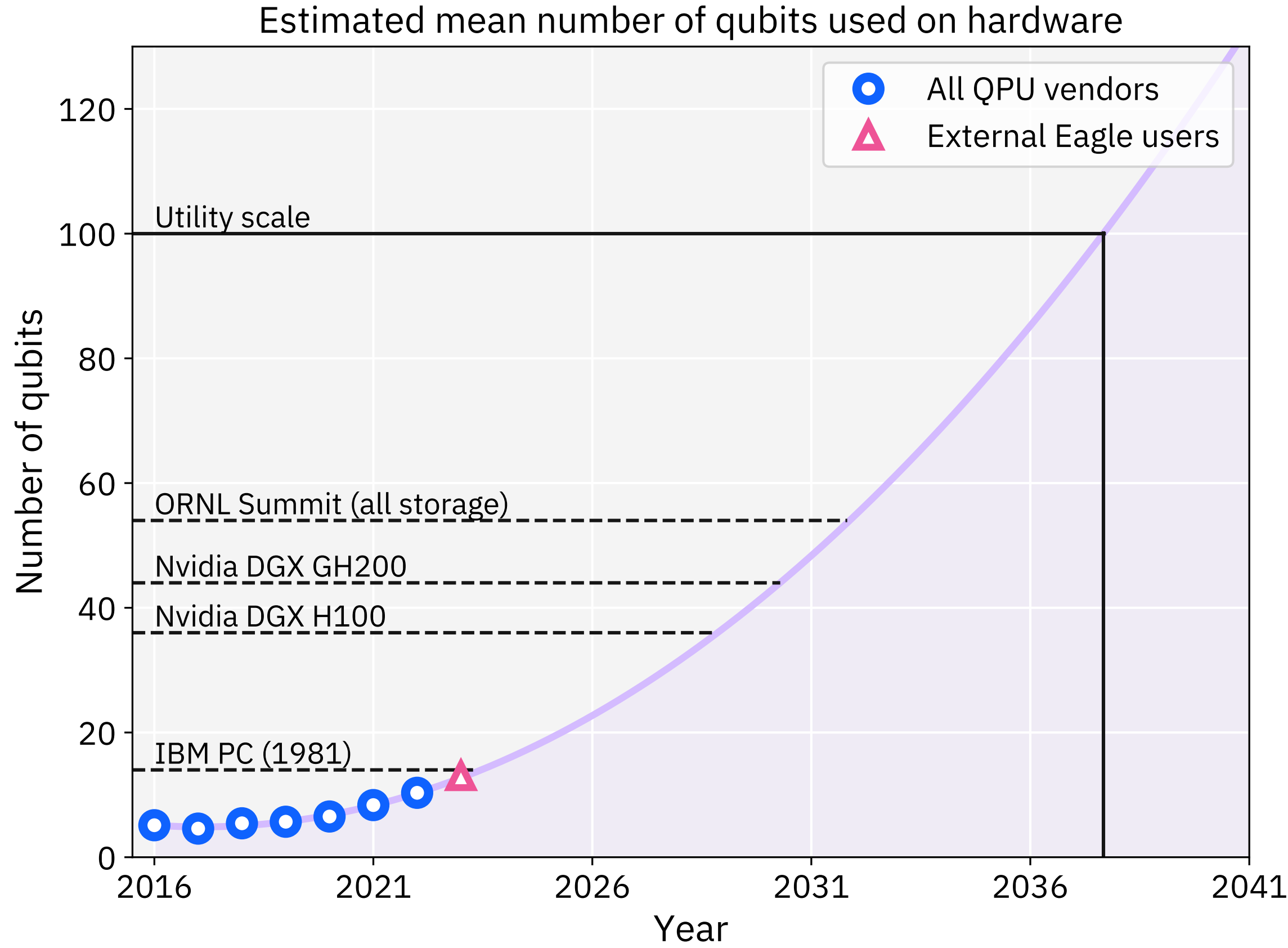
IBM Quantum Development Roadmap Expansion

<https://www.youtube.com/watch?v=De2IlWji8Ck>



The era of quantum utility

Quantum state of play

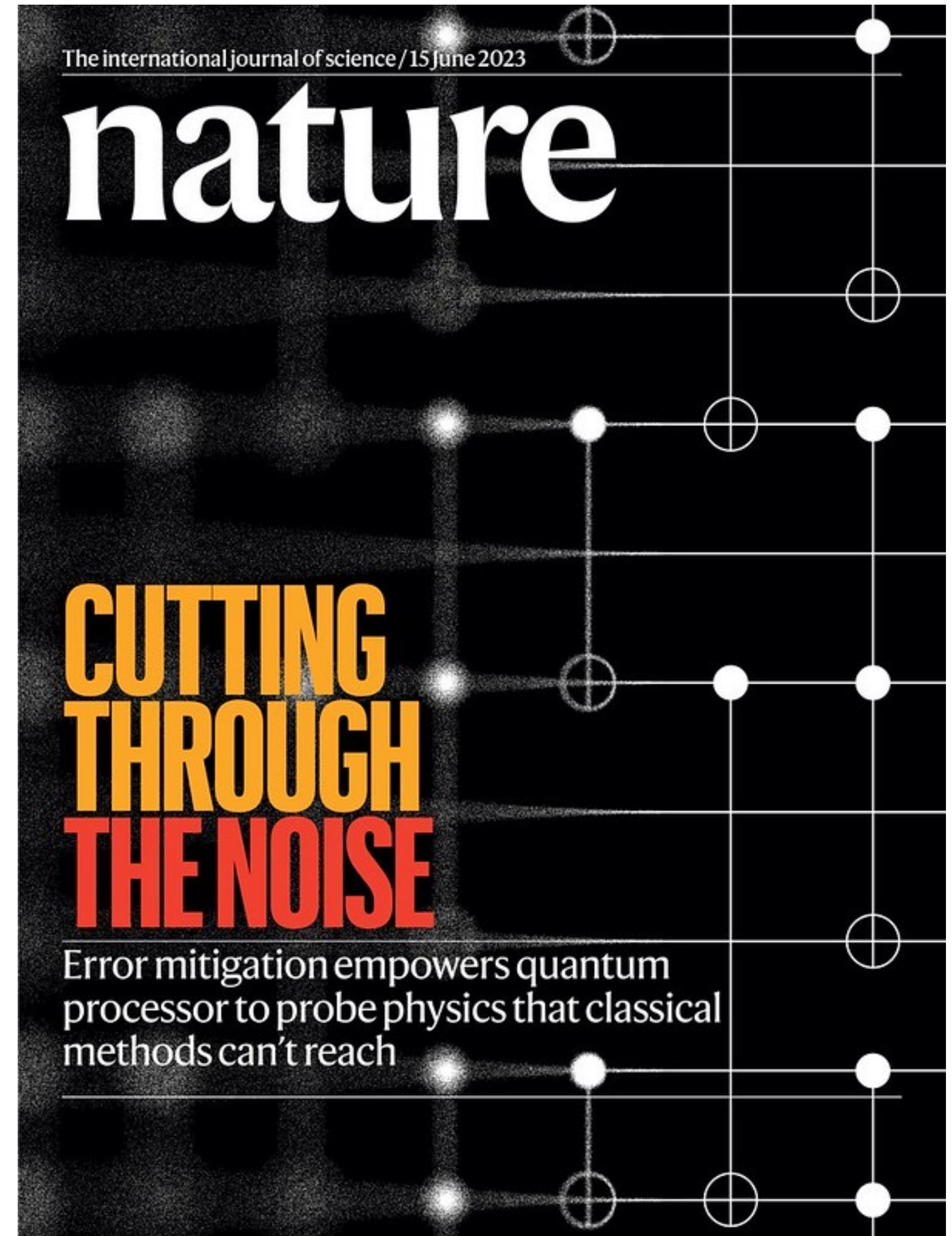


Data for all vendors taken from: arXiv:2307.16130

We need a *disruptive change* to unlock the potential of quantum computation.

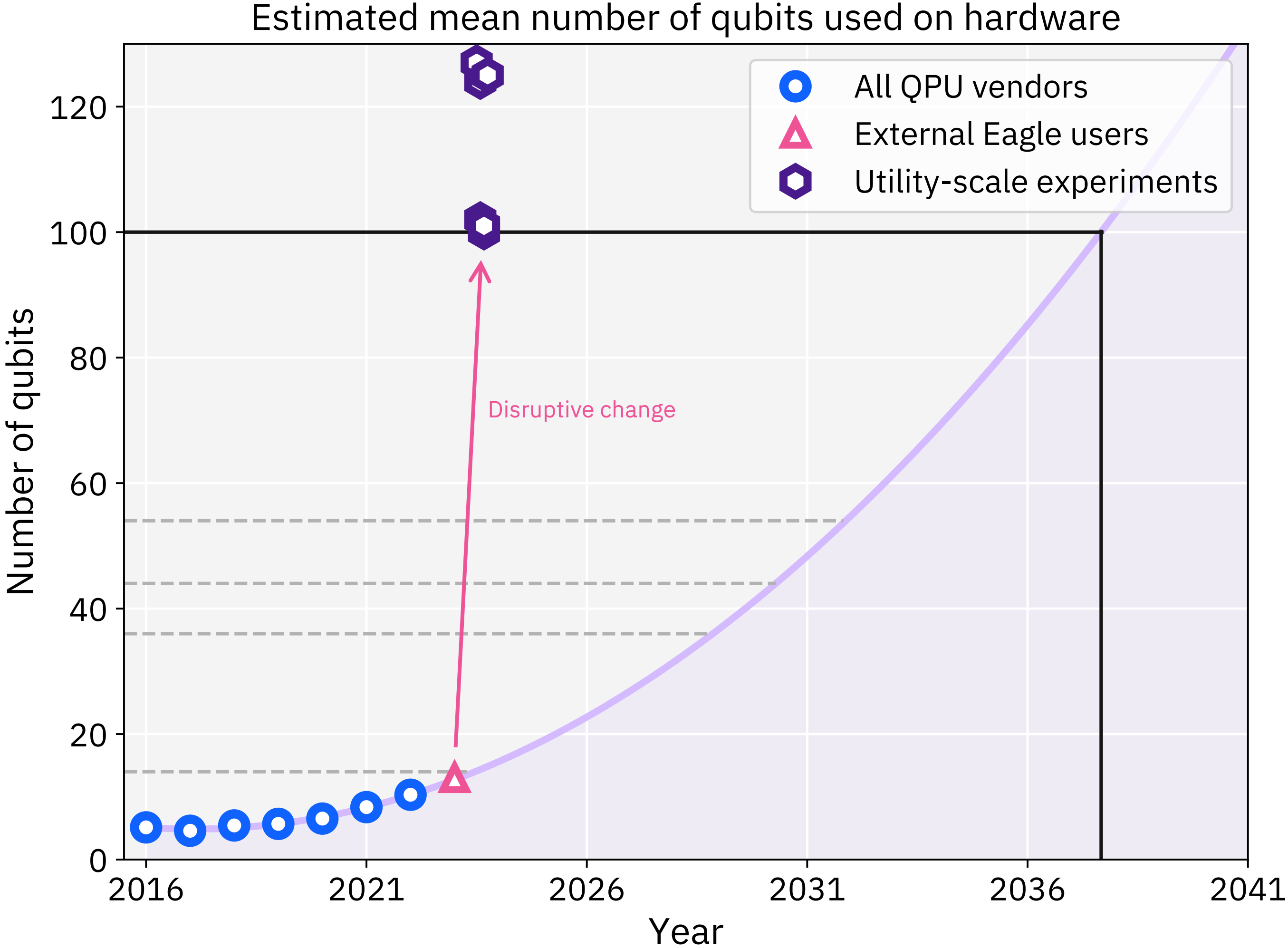
A noisy quantum computer produces accurate expectation values on 127 qubits, outside of brute force classical computation.

<https://www.nature.com/articles/s41586-023-06096-3>



IBM Quantum systems and Qiskit are bringing a disruptive change.

Data for all vendors taken from: arXiv:2307.16130



Multiple utility-scale experiments within last 6 months (more to come)

Evidence for the utility of quantum computing before fault tolerance

127 qubits / 2880 CX gates

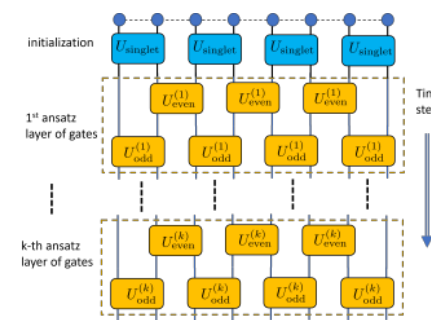
Nature, 618, 500 (2023)



Simulating large-size quantum spin chains on cloud-based superconducting quantum computers

102 qubits / 3186 CX gates

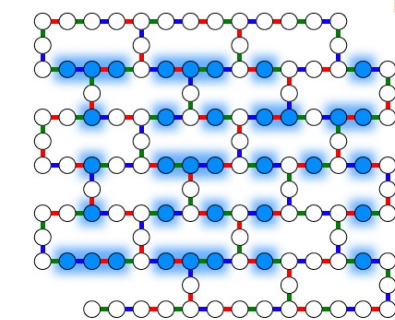
arXiv:2207.09994



Uncovering Local Integrability in Quantum Many-Body Dynamics

124 qubits / 2641 CX gates

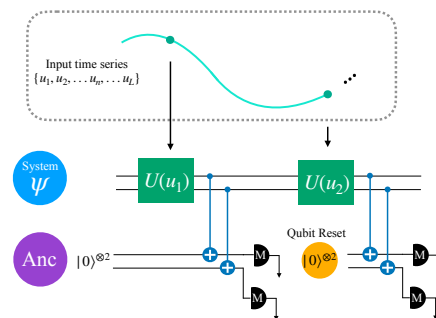
arXiv:2307.07552



Quantum reservoir computing with repeated measurements on superconducting devices

120 qubits / 49470 gates + meas.

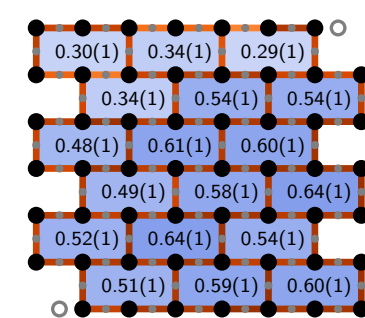
arXiv:2310.06706



Realizing the Nishimori transition across the error threshold for constant-depth quantum circuits

125 qubits / 429 gates + meas.

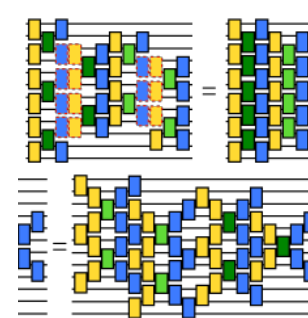
arXiv:2309.02863



Scalable Circuits for Preparing Ground States on Digital Quantum Computers: The Schwinger Model Vacuum on 100 Qubits

100 qubits / 788 CX gates

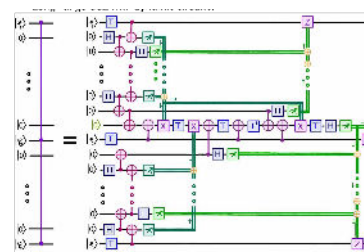
arXiv:2308.04481



Efficient Long-Range Entanglement using Dynamic Circuits

101 qubits / 504 gates + meas.

arXiv:2308.13065



Condor

Pushing the limits of scale & yield

1,121

Superconducting qubits

50%

Increase in qubit density

1 mile +

Of flex cabling



Condor unblocked the
road to scaling.

We now need to focus on gate
depth and quality.

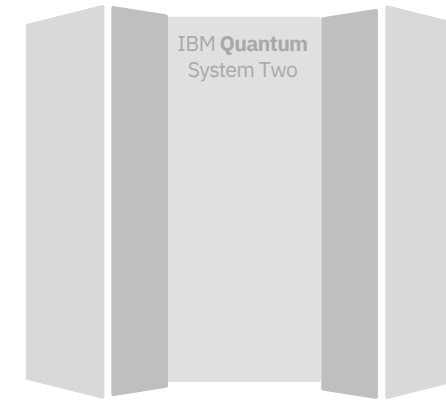
IBM Quantum Heron: A Four-Year Journey

133-qubit count
tunable coupler architecture



IBM Quantum Platform:

Where users come to do work



Qiskit + Systems = Work

Qiskit

1.0

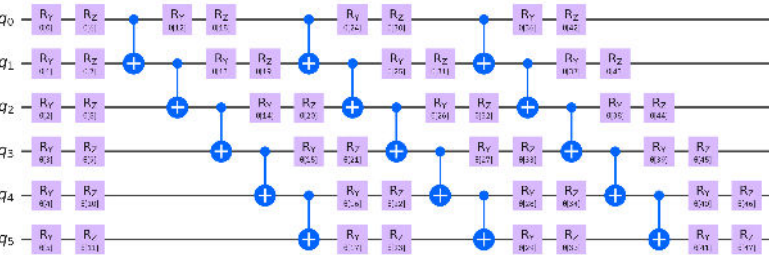
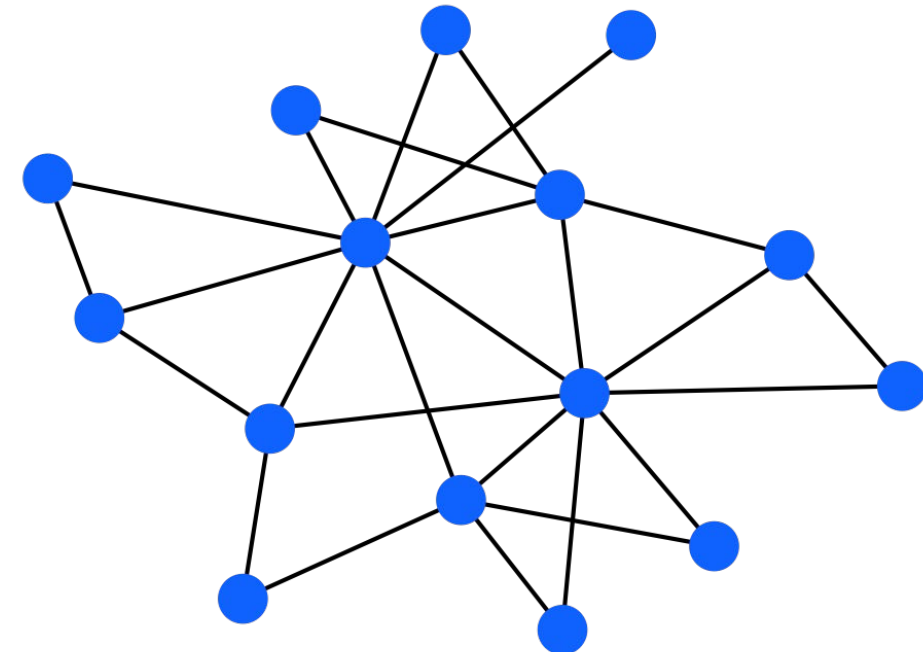
Now with increased performance,
stability, and reliability.

Qiskit Patterns

The anatomy of a quantum algorithm

Step 1

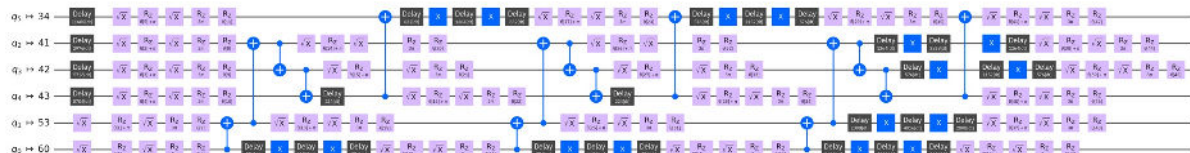
Map quantum circuits and operators.



Step 2

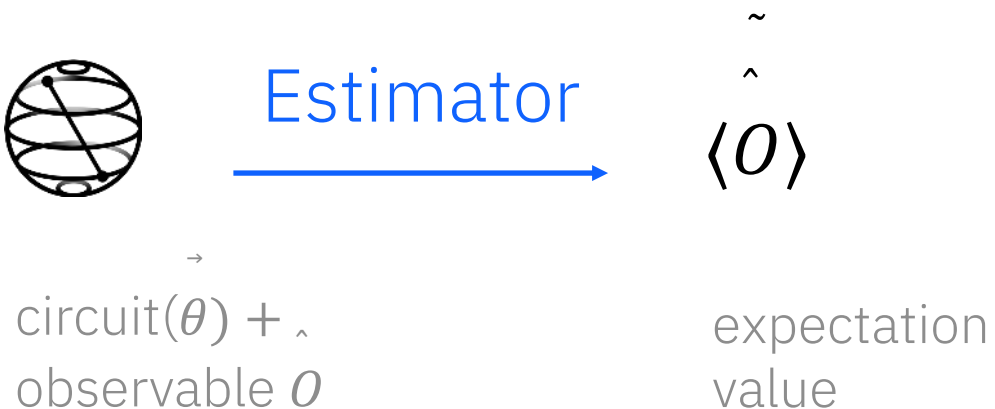
Optimize problem for quantum execution.

```
PassManager(UnitarySynthesis(),
            BasisTranslator(),
            EnlargeWithAncilla(),
            AISwap(),
            Collect1qRuns(),
            Optimize1qGates(),
            Collect2qBlocks(),
            ConsolidateBlocks())
```



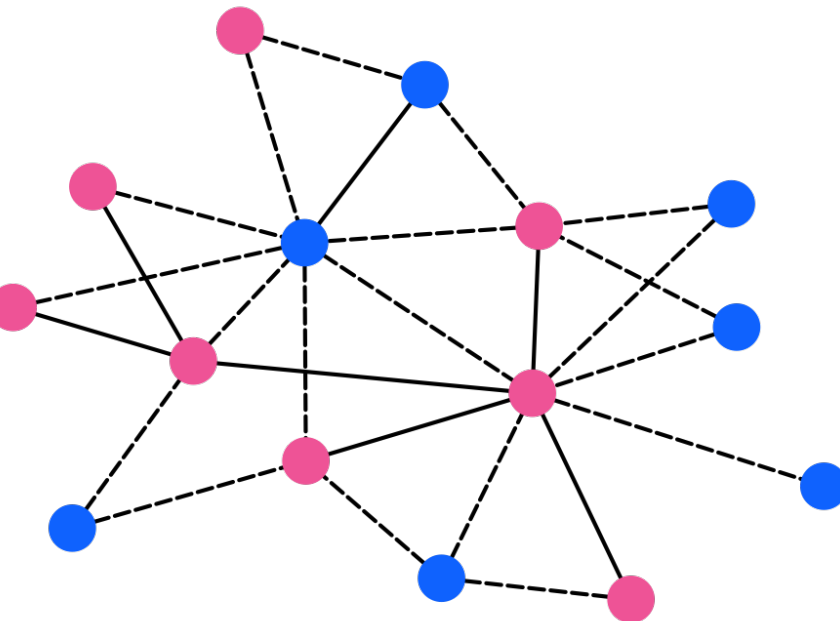
Step 3

Execute using Qiskit Runtime Primitives.



Step 4

Post-process, return result in classical format.



The full promise of
quantum computing at
scale delivered via error
correction

A 100,000-qubit quantum centric super computer lets us move toward practical fault-tolerance.

Low-overhead fault tolerant quantum computing using long-range connectivity

arXiv:2308.07915v1 [quant-ph] 15 Aug 2023

High-threshold and low-overhead fault-tolerant quantum memory

Sergey Bravyi¹, Andrew W. Cross¹, Jay M. Gambetta¹, Dmitri Maslov¹, Patrick Rall², and Theodore J. Yoder¹

¹IBM Quantum, IBM T.J. Watson Research Center, Yorktown Heights, NY 10598 (USA)

²IBM Quantum, MIT-IBM Watson AI Lab, Cambridge, MA 02142 (USA)

August 16, 2023

Abstract

Quantum error correction becomes a practical possibility only if the physical error rate is below a threshold value that depends on a particular quantum code, syndrome measurement circuit, and a decoding algorithm. Here we present an end-to-end quantum error correction protocol that implements fault-tolerant memory based on a family of LDPC codes with a high encoding rate that achieves an error threshold of 0.8% for the standard circuit-based noise model. This is on par with the surface code which has remained an uncontested leader in terms of its high error threshold for nearly 20 years. The full syndrome measurement cycle for a length- n code in our family requires n ancillary qubits and a depth-7 circuit composed of nearest-neighbor CNOT gates. The required qubit connectivity is a degree-6 graph that consists of two edge-disjoint planar subgraphs. As a concrete example, we show that 12 logical qubits can be preserved for ten million syndrome cycles using 288 physical qubits in total, assuming the physical error rate of 0.1%. We argue that achieving the same level of error suppression on 12 logical qubits with the surface code would require more than 4000 physical qubits. Our findings bring demonstrations of a low-overhead fault-tolerant quantum memory within the reach of near-term quantum processors.

1 Introduction

Quantum computing attracted attention due to its ability to offer asymptotically faster solutions to a set of computational problems compared to the best known classical algorithms [1]. It is believed that a scalable functioning quantum computer may help solve computational problems in such areas as scientific discovery, materials research, chemistry, and drug design, to name a few [2, 3, 4, 5].

The main obstacle to building a quantum computer is the fragility of quantum information, owing to various sources of noise affecting it. Since isolating a quantum computer from external effects and controlling it to induce a desired computation are in conflict with each other, noise appears to be inevitable. The sources of noise include imperfections in qubits, materials used, controlling apparatus, State Preparation and Measurement (SPAM) errors, and a variety of external factors ranging from local man-made, such as stray electromagnetic fields, to those inherent to the Universe, such as cosmic rays. See Ref. [6] for a summary. While some sources of noise can be eliminated with better control [7], materials [8], and shielding [9, 10, 11], a number of other sources appear to be difficult if at all possible to remove. The latter kind can include spontaneous and stimulated emission in trapped ions [12, 13], and the interaction with the bath (Purcell Effect) [14] in superconducting circuits—covering both leading quantum technologies. Thus, error correction becomes a key requirement for building a functioning scalable quantum computer.

The road is
clear to
extending
quantum utility

Development Roadmap

	2019	2020	2021	2022	2023	2024	2025	2026+
	Run quantum circuits on the IBM cloud	Demonstrate and prototype quantum algorithms and applications	Run quantum programs 100x faster with Qiskit Runtime	Bring dynamic circuits to Qiskit Runtime to unlock more computations	Enhancing applications with elastic computing and parallelization of Qiskit Runtime	Improve accuracy of Qiskit Runtime with scalable error mitigation	Scale quantum applications with circuit knitting toolbox controlling Qiskit Runtime	Increase accuracy and speed of quantum workflows with integration of error correction into Qiskit Runtime

Data Scientist					Prototype quantum software functions	Quantum software applications		
						Machine learning Natural science Optimization		

Researchers		Quantum algorithm & application modules			Middleware for Quantum			
		Machine learning Natural science Optimization			Quantum Serverless	Intelligent orchestration	Circuit Knitting Toolbox	Circuit libraries

Quantum Physicist	Circuits		Qiskit Runtime				
		QSAM3	Dynamic circuits	Execution Modes	Error suppression and mitigation		Error correction

Falcon 27 qubits	Hummingbird 65 qubits	Eagle 127 qubits	Osprey 433 qubits	Condor 1,121 qubits	Flamingo 1,386+ qubits	Kookaburra 4,158+ qubits	Scaling to 10K-100K qubits with classical and quantum communication
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Heron 133 qubits x p	Crossbill 408 qubits
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Executed by IBM
 On target

Development Roadmap

	2016–2019	2020	2021	2022	2023	2024	2025	2026	2027	2028	2029	2033+
	Run quantum circuits on the IBM Quantum Platform	Release multi-dimensional roadmap publicly with initial aim focused on scaling	Enhancing quantum execution speed by 100x with Qiskit Runtime	Bring dynamic circuits to unlock more computations	Enhancing quantum execution speed by 5x with quantum serverless and Execution modes	Improving quantum circuit quality and speed to allow 5K gates with parametric circuits	Enhancing quantum execution speed and parallelization with partitioning and quantum modularity	Improving quantum circuit quality to allow 7.5K gates	Improving quantum circuit quality to allow 10K gates	Improving quantum circuit quality to allow 15K gates	Improving quantum circuit quality to allow 100M gates	Beyond 2033, quantum-centric supercomputers will include 1000's of logical qubits unlocking the full power of quantum computing
Data Scientist						Platform						
						Code assistant	Functions	Mapping Collection	Specific Libraries			General purpose QC libraries
Researchers					Middleware							
					Quantum Serverless	Transpiler Service	Resource Management	Circuit Knitting x P	Intelligent Orchestration			Circuit libraries
Quantum Physicist			Qiskit Runtime									
	IBM Quantum Experience		QASM3	Dynamic circuits	Execution Modes	Heron (5K)	Flamingo (5K)	Flamingo (7.5K)	Flamingo (10K)	Flamingo (15K)	Starling (100M)	Blue Jay (1B)
	Early	Falcon		Eagle		Error Mitigation	Error Mitigation	Error Mitigation	Error Mitigation	Error Mitigation	Error correction	Error correction
	Canary 5 qubits, Albatross 16 qubits, Penguin 20 qubits, Prototype 53 qubits	Benchmarking 27 qubits		Benchmarking 127 qubits		5k gates, 133 qubits, Classical modular, 133x3 = 399 qubits	5k gates, 156 qubits, Quantum modular, 156x7 = 1092 qubits	7.5k gates, 156 qubits, Quantum modular, 156x7 = 1092 qubits	10k gates, 156 qubits, Quantum modular, 156x7 = 1092 qubits	15k gates, 156 qubits, Quantum modular, 156x7 = 1092 qubits	100M gates, 200 qubits, Error corrected modularity	1B gates, 2000 qubits, Error corrected modularity

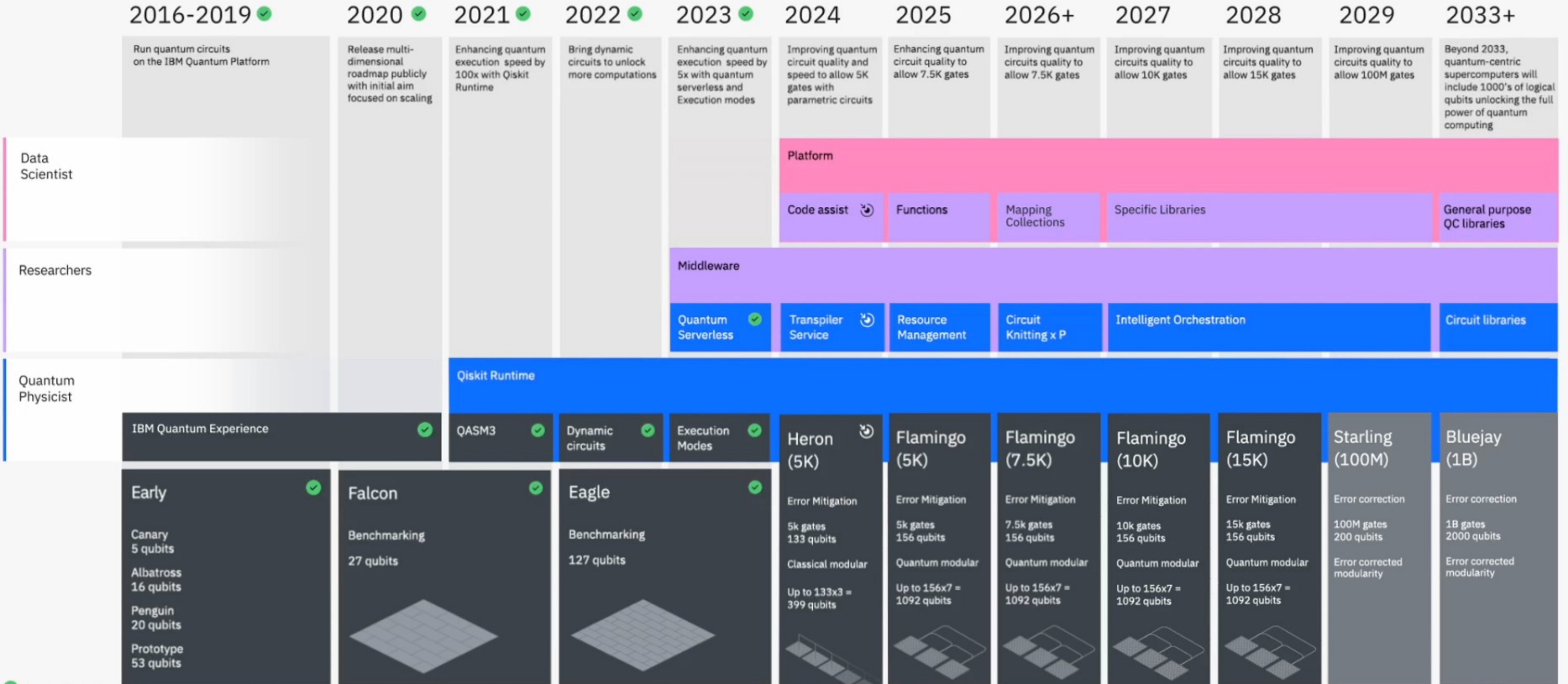
Innovation Roadmap

Software Innovation	IBM Quantum Experience	Qiskit Circuit and operator API with compilation to multiple targets	Application modules Modules for domain specific application and algorithm workflows	Qiskit Runtime Performance and abstract through Primitives	Serverless Demonstrate concepts of quantum centric-supercomputing	AI enhanced quantum Prototype demonstrations of AI enhanced circuit transpilation	Resource management System partitioning to enable parallel execution	Scalable circuit knitting Circuit partitioning with classical reconstruction at HPC scale	Error correction decoder Demonstration of a quantum system with real-time error correction decoder				
Hardware Innovation	Early Canary 5 qubits, Penguin 20 qubits, Albatross 16 qubits, Prototype 53 qubits	Falcon Demonstrate scaling with I/O routing with Bump bonds	Hummingbird Demonstrate scaling with multiplexing readout	Eagle Demonstrate scaling with MLW and TSV	Osprey Enabling scaling with high density signal delivery	Condor Single system scaling and fridge capacity	Flamingo Demonstrate scaling with modular connectors	Kookaburra Demonstrate scaling with nonlocal c-coupler	Cockatoo Demonstrate path to improved quality with logical communication	Starling Demonstrate path to improved quality with logical gates			
						Heron Architecture based on tunable-couplers	Crossbill m- coupler						

Executed by IBM

On target



Development Roadmap



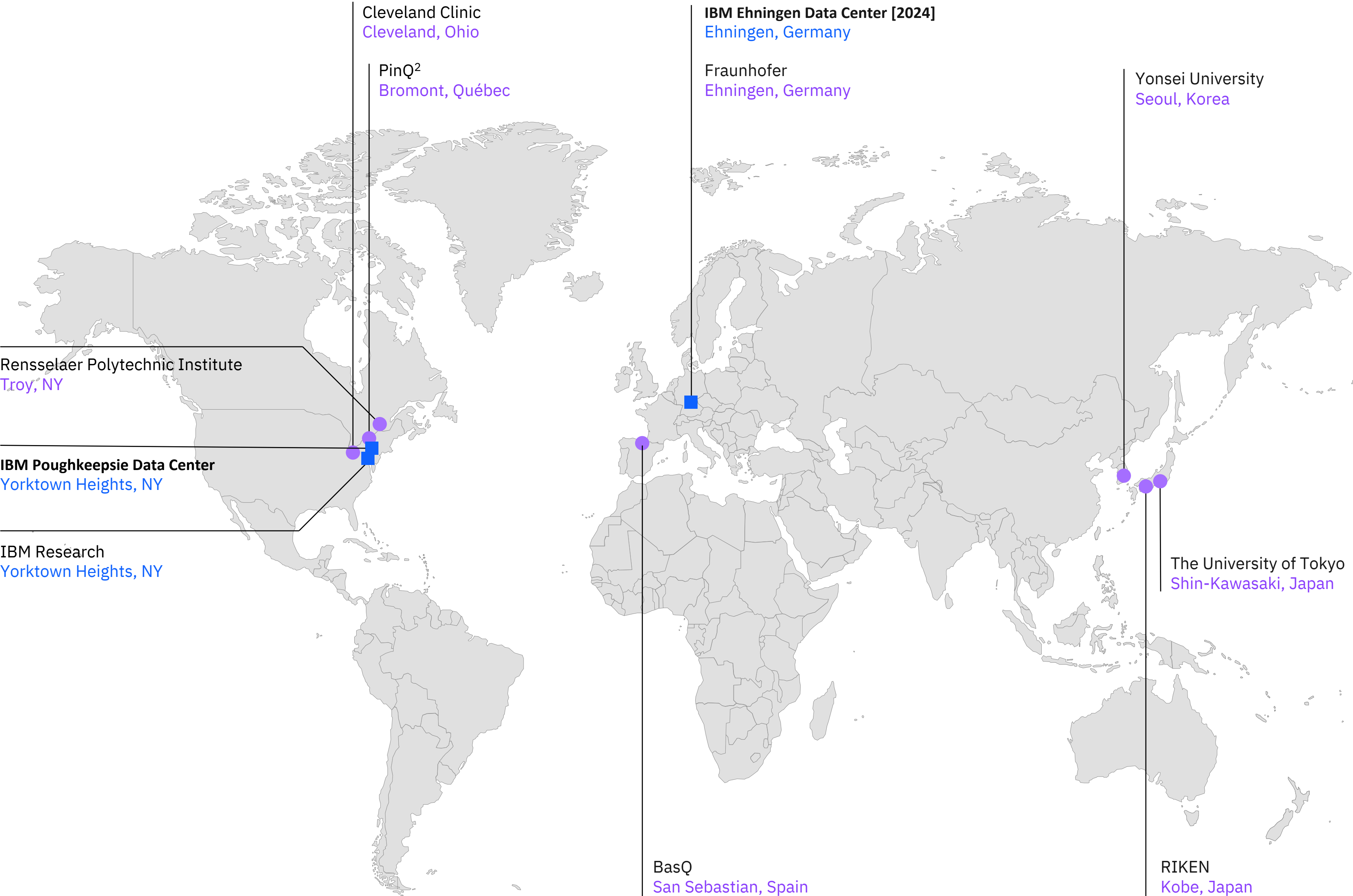
Executed by IBM
 On target

Innovation Roadmap



 Executed by IBM
 On target

Scaling collaboration through Quantum Computation Centers



Technical Working Groups

Bringing together experts from **classical + quantum + industry** to identify the most pressing scientific challenges in the area today

Biological Sciences & Drug Discovery

Image processing, Biomarkers – Omics technologies, Radiotherapy planning, Clinical trial optimization and design, Disease mechanism...



High Energy Physics

*Anomaly detection
Lattice Gauge Theory
Sensing...*



Materials

*Materials properties
Dynamics of chemical systems
Ground and excited states...*



Optimization

*Risk analysis
Portfolio Optimization
Transaction Settlement...*



Tackling challenges for HCLS with quantum computing

Goals

Establish that now is the time to focus on HCLS for which quantum computing (through quantum chemistry and QML algorithms) will be of paramount importance

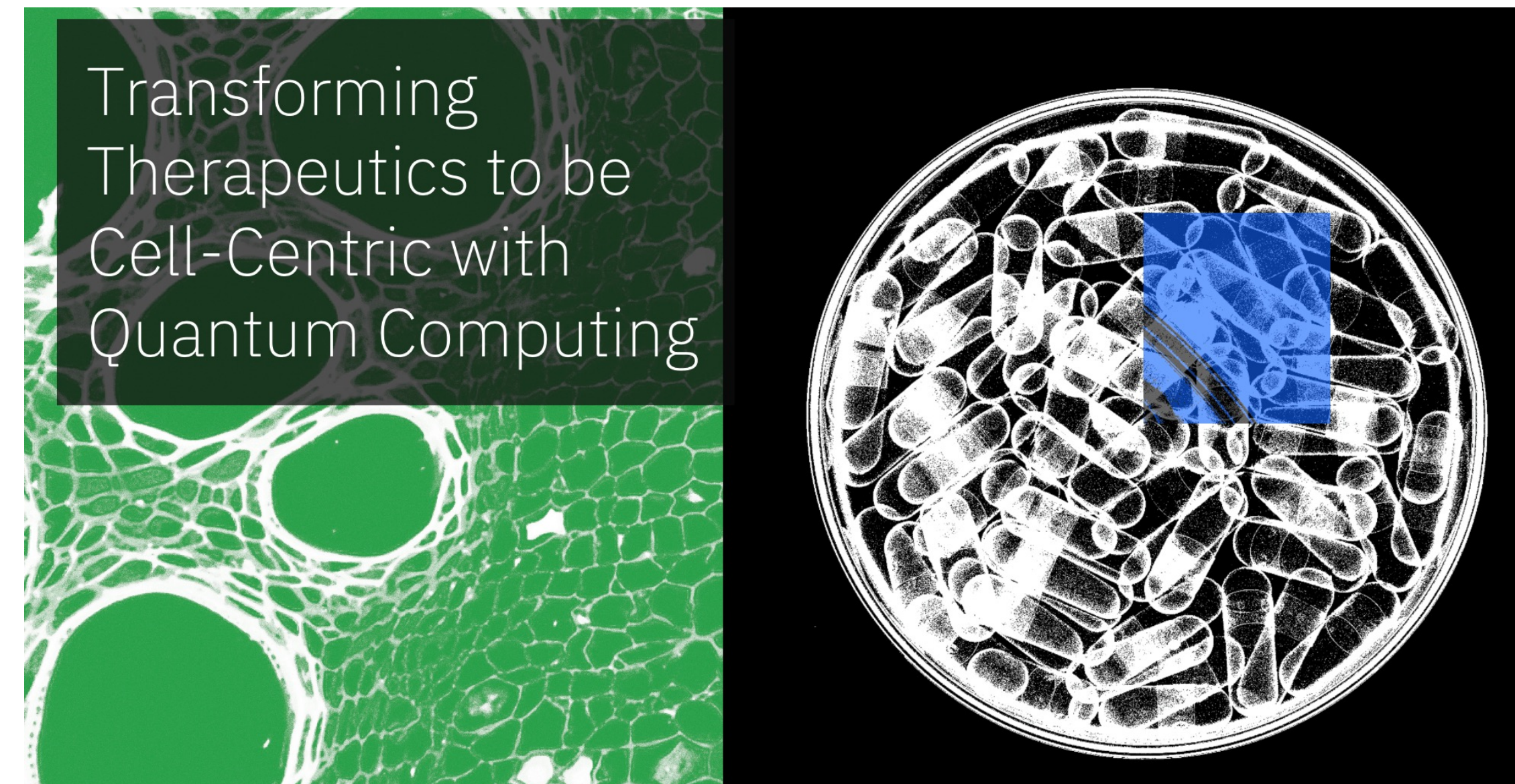
Gain consensus on key scientific questions for quantum computing and HCLS.

Establish a roadmap/perspective on key challenges over next 3-6 years

Towards quantum-enabled cell-centric therapeutics

Saugata Basu¹, Jannis Born², Aritra Bose³, Sara Capponi^{4,5}, Dimitra Chalkia⁶, Timothy A Chan^{7,8}, Hakan Doga⁹, Maark Goldsmith¹⁰, Tanvi Gujarati⁹, Aldo Guzmán-Sáenz³, Dimitrios Iliopoulos⁶, Gavin Jones⁹, Stefan Knecht¹⁰, Dhiraj Madan¹¹, Sabrina Maniscalco¹⁰, Nicola Mariella¹², Joseph Morrone³, Pushpak Pati², Daniel Platt³, Maria Anna Rapsomaniki², Anupama Ray¹¹, Kahn Rhrissorrakrai³, Omar Shehab¹⁴, Ivano Tavernelli¹³, Meltem Tolunay⁹, Filippo Utro³, Stefan Woerner¹³, Sergiy Zhuk¹², Jeannette Garcia^{†9}, and Laxmi Parida^{†3}

<https://arxiv.org/abs/2307.05734>



Approaches to studying high energy physics problems with quantum

Goals

Deep dive look at the state of High Energy Physics challenges (from both experiment and theory)

Understand how quantum computing can boost the solution of open challenges in HEP

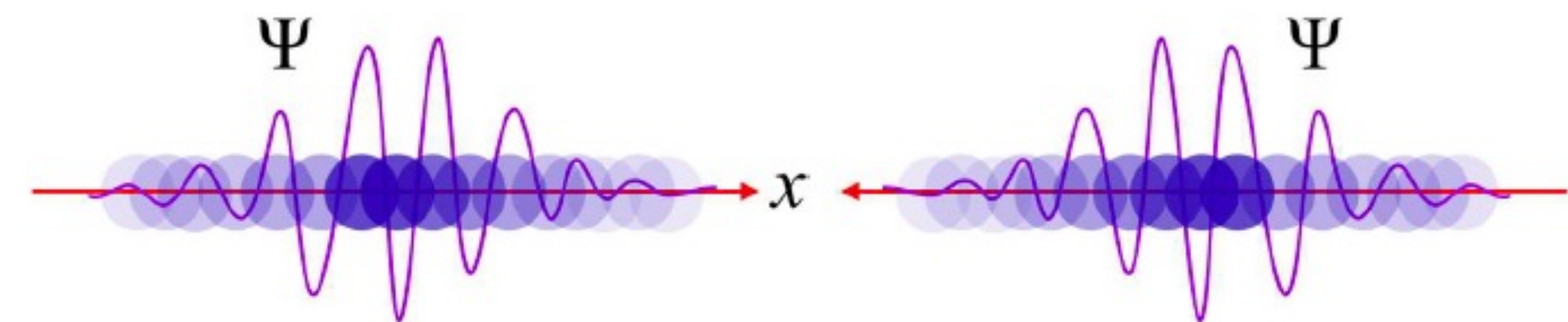
Demonstrate quantum advantage with near-term devices on relevant HEP problems

Catalyze new developments within HEP

Quantum Computing for High-Energy Physics State of the Art and Challenges Summary of the QC4HEP Working Group

Alberto Di Meglio,^{1,*} Karl Jansen,^{2,3,†} Ivano Tavernelli,^{4,‡} Constantia Alexandrou,^{5,3} Srinivasan Arunachalam,⁶ Christian W. Bauer,⁷ Kerstin Borras,^{8,9} Stefano Carrazza,^{10,1} Arianna Crippa,^{2,11} Vincent Croft,¹² Roland de Putter,⁶ Andrea Delgado,¹³ Vedran Dunjko,¹² Daniel J. Egger,⁴ Elias Fernández-Combarro,¹⁴ Elina Fuchs,^{1,15,16} Lena Funcke,¹⁷ Daniel González-Cuadra,^{18,19} Michele Grossi,¹ Jad C. Halimeh,^{20,21} Zoë Holmes,²² Stefan Kühn,² Denis Lacroix,²³ Randy Lewis,²⁴ Donatella Lucchesi,^{25,26,1} Miriam Lucio Martinez,^{27,28} Federico Meloni,⁸ Antonio Mezzacapo,⁶ Simone Montangero,^{25,26} Lento Nagano,²⁹ Voica Radescu,³⁰ Enrique Rico Ortega,^{31,32,33,34} Alessandro Roggero,^{35,36} Julian Schuhmacher,⁴ Joao Seixas,^{37,38,39} Pietro Silvi,^{25,26} Panagiotis Spentzouris,⁴⁰ Francesco Tacchino,⁴ Kristan Temme,⁶ Koji Terashi,²⁹ Jordi Tura,^{12,41} Cenk Tüysüz,^{2,11} Sofia Vallecorsa,¹ Uwe-Jens Wiese,⁴² Shinjae Yoo,⁴³ and Jinglei Zhang^{44,45}

<https://arxiv.org/abs/2307.03236>



Optimization

Identifying the next steps for optimization

Goals

Engage with quantum & classical optimization communities

Identify and work on key questions to scale quantum optimization towards quantum advantage

Establish a set of benchmarks to track progress towards quantum advantage (in optimization and/or finance)

The screenshot shows a GitHub repository page for 'Quantum Algorithms Benchmarks'. At the top, it displays 'main' branch, '9 branches', and '0 tags'. A 'Go to file' button and a 'Code' button are visible. Below this is a table of recent commits:

Commit	Author	Message	Time
828b4d7	eggerdj	Merge pull request #26 from da66/...	2 weeks ago
121 commits			
6 months ago		Update README.md	6 months ago
7 months ago		Added ChebPE benchmarks	7 months ago
2 weeks ago		Added version table	2 weeks ago
last month		Added knapsack to solution_summary	last month
6 months ago		Revert "Remove IBM from License"	6 months ago
6 months ago		Update README.md	6 months ago

The main content area shows the README.md file with the following text:

Quantum Algorithms Benchmarks (work in progress)

Welcome to this quantum algorithms benchmark repository! The goal of which is the collection of benchmarks that enable comparability between existing and future quantum algorithms and/or methods-directly provided by the quantum community.

This repository is made of three sub-modules each reflecting an area where quantum computing has the potential to improve a computational task:

- [Machine Learning](#),
- [Optimization](#),
- [Monte Carlo Simulation](#).

Vision

The repository tracks the progress of quantum computing algorithms and applications. It is made of libraries that include community-proposed benchmarking problems and corresponding solutions to these problems. Solutions to the benchmark problems may be run on any type of quantum hardware or simulator, e.g., superconducting qubits, annealers, or trapped ions. The progress of quantum computing towards larger problems is tracked by performance metrics defined in each sub-module. Everybody is welcome to propose benchmark problems that they believe are relevant as well as new solutions to existing benchmarks.

How to contribute

The proposed solutions to a benchmark problem may either be executed on simulators or hardware. The platform on which the solution was obtained and the corresponding settings must be clearly indicated. Refer to the READMEs of the sub-modules for detailed instructions on how to contribute: [Machine Learning](#), [Optimization](#), and [Monte Carlo Simulation](#).

The right sidebar contains repository metadata:

- About:** No description, website, or topics provided.
- Readme:** Readme
- License:** Apache-2.0 license
- Code of conduct:** Code of conduct
- Activity:** Activity
- Stars:** 10 stars
- Watching:** 5 watching
- Forks:** 2 forks
- Report repository:** Report repository
- Releases:** No releases published
- Packages:** No packages published
- Contributors:** 7 contributors
- Languages:** Jupyter Notebook 99.1%, Python 0.9%

<https://github.com/qiskit-community/quantum-algorithms-benchmarks>

Identifying the next steps for optimization

Goals

Engage with quantum & classical optimization communities

Identify and work on key questions to scale quantum optimization towards quantum advantage

Establish a set of benchmarks to track progress towards quantum advantage (in optimization and/or finance)

Quantum Optimization: Potential, Challenges, and the Path Forward*

Amira Abbas,¹ Andris Ambainis,² Brandon Augustino,³ Andreas Bärttschi,⁴ Harry Buhrman,¹ Carleton Coffrin,⁴ Giorgio Cortiana,⁵ Vedran Dunjko,⁶ Daniel J. Egger,⁷ Bruce G. Elmegreen,⁸ Nicola Franco,⁹ Filippo Fratini,¹⁰ Bryce Fuller,¹¹ Julien Gacon,^{7,12} Constantin Gonciulea,¹³ Sander Gribling,¹⁴ Swati Gupta,³ Stuart Hadfield,^{15,16} Raoul Heese,¹⁷ Gerhard Kircher,¹⁰ Thomas Kleinert,¹⁸ Thorsten Koch,^{19,20} Georgios Korpas,^{21,22} Steve Lenk,²³ Jakub Marecek,²² Vanio Markov,¹³ Guglielmo Mazzola,²⁴ Stefano Mensa,²⁵ Naeimeh Mohseni,⁵ Giacomo Nannicini,²⁶ Corey O’Meara,⁵ Elena Peña Tapia,⁷ Sebastian Pokutta,^{19,20} Manuel Proissl,⁷ Patrick Rebstrost,²⁷ Emre Sahin,²⁵ Benjamin C. B. Symons,²⁵ Sabine Törnøw,²⁸ Víctor Valls,²⁹ Stefan Woerner,⁷ Mira L. Wolf-Bauwens,⁷ Jon Yard,³⁰ Sheir Yarkoni,³¹ Dirk Zechiel,¹⁸ Sergiy Zhuk,²⁹ and Christa Zoufal⁷

<https://arxiv.org/abs/2312.02279>

Table IV. An overview of state-of-the-art experimental realizations of optimization algorithms on gate-based quantum computers with more than 15 variables. In cases where data was not made available in the corresponding publication or the accompanying data repository, we denote this in the respective field with *N/A*. *AR* denotes the approximation ratio, given based on the *mean* and the *best* sample value of the experiment, *n.n. grid* stands for nearest neighbor grid. Furthermore, JSP, FVQE and GQAOA abbreviate job shop scheduling problem, filtering variational quantum eigensolver, and greedy QAOA, respectively.

Problem	Algorithm	Qubits	Density	AR		Depth	Year	Ref.
				mean	best			
Sherrington–Kirkpatrick	QAOA	17	100%	0.61	N/A	$1 \leq p \leq 3$	2021	[360]
MAXCUT (R3R)	QAOA	20	16%	0.64	1	$p = 2$	2023	[377]
MAXCUT (R3R)	QAOA	20	16%	0.94	1	$p \leq 10$	2023	[513]
MAXCUT (R3R)	QAOA	22	14%	0.67	N/A	$1 \leq p \leq 3$	2021	[360]
MAXCUT (n.n. grid)	QAOA	23	13%	0.72	N/A	$1 \leq p \leq 5$	2021	[360]
QUBO (JSP)	FVQE	23	N/A	0.88	N/A	$1 \leq p \leq 2$	2022	[514]
MAXCUT (heavy-hex.)	QAOA	27	8%	N/A	1	$p = 2$	2022	[376]
MAXCUT (R3R)	QAOA	30	10%	0.59	0.83	$p = 2$	2023	[377]
MAXCUT (R3R)	QAOA	32	10%	0.88	1	$p \leq 10$	2023	[513]
MAXCUT (R3R)	QAOA	32	10%	N/A	1	$p = 2$	2023	[205]
MAXCUT (R3R)	QAOA	40	8%	0.58	0.78	$p = 2$	2023	[377]
Sherrington–Kirkpatrick	GQAOA	72	100%	0.92	N/A	$p = 1$	2023	[208]
QUBO (heavy-hex.)	QAOA	127	2%	0.67	0.85	$1 \leq p \leq 2$	2023	[411]
PUBO (heavy-hex.)	QAOA	127	2%	0.65	0.84	$1 \leq p \leq 2$	2023	[411]
PUBO (heavy-hex.)	QAOA	127	2%	0.73	0.89	$1 \leq p \leq 5$	2023	[515]
QUBO (heavy-hex.)	QAOA	414	0.6%	0.57	0.69	$p = 1$	2023	[515]
PUBO (heavy-hex.)	QAOA	414	0.6%	0.56	0.68	$p = 1$	2023	[515]

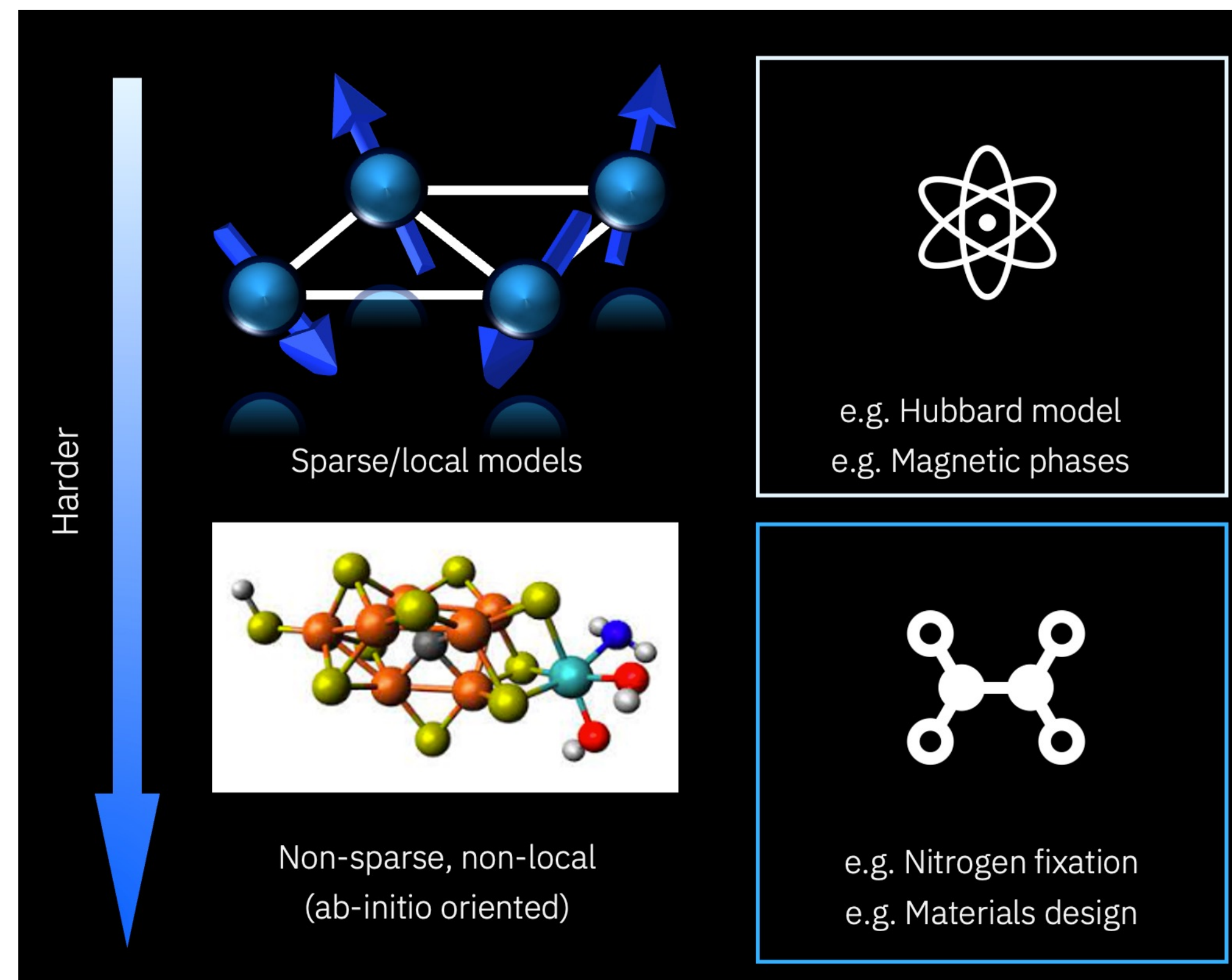
Workflows for materials simulations

Goals

Identify key algorithms that make optimal use of quantum computing and HPC

Identify use cases in materials science for quantum/HPC algorithms

How best to model materials with quantum computers?



Paper coming soon!!!

Bringing together experts from **classical + quantum + industry** to identify the most pressing scientific challenges in the area today

Coming in 2024: Sustainability

Biological Sciences & Drug Discovery

Image processing, Biomarkers – Omics technologies, Radiotherapy planning, Clinical trial optimization and design, Disease mechanism...



High Energy Physics Quantum

*Anomaly detection
Lattice Gauge Theory
Sensing...*



Materials

*Materials properties
Dynamics of chemical systems
Ground and excited states...*



Optimization

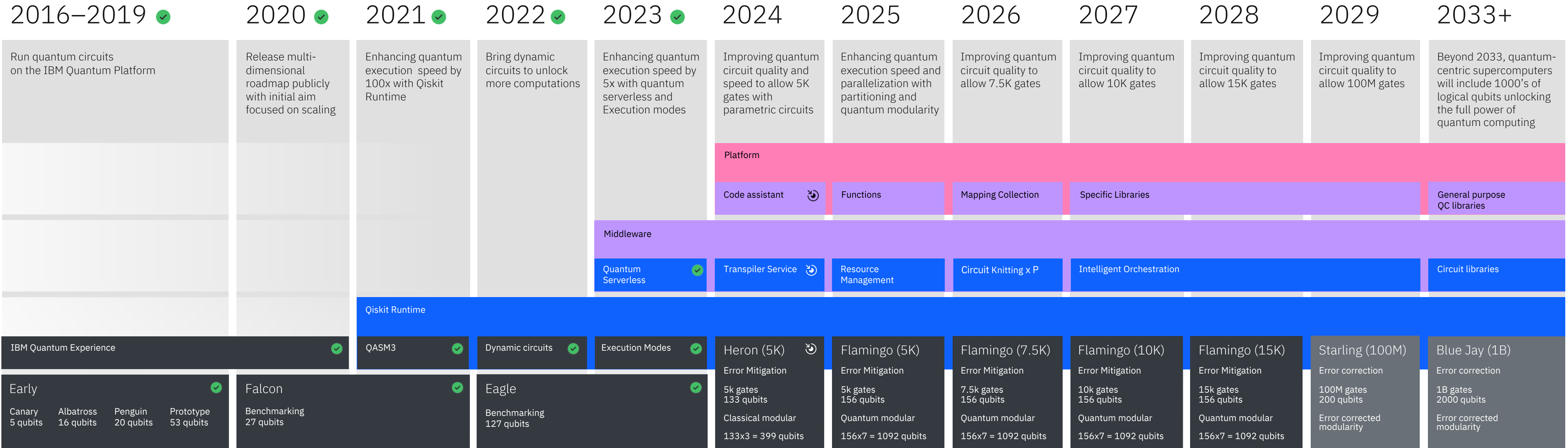
*Risk analysis
Portfolio Optimization
Transaction Settlement...*



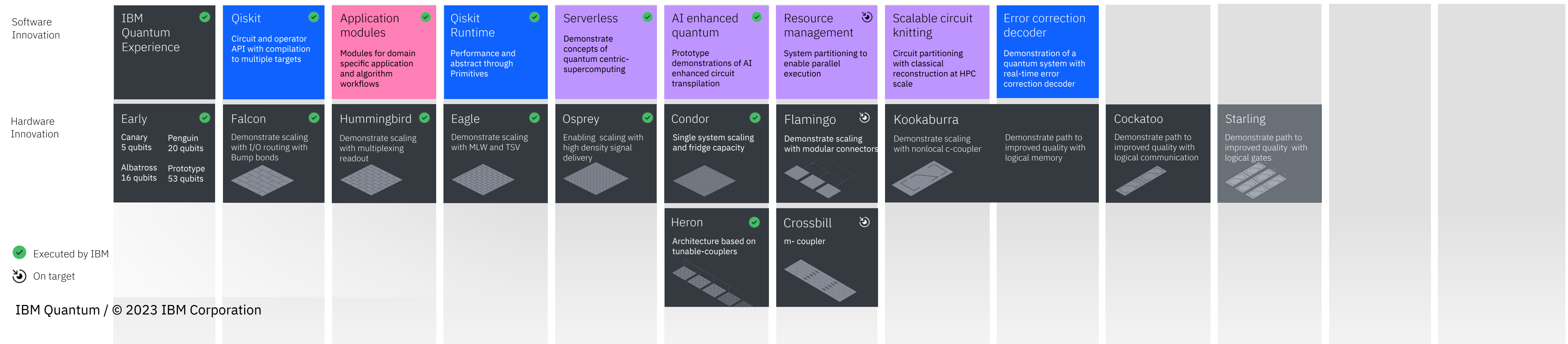
Working group interest form:

<https://airtable.com/appL9KALjibTDaueh/shrmeubXExazSH9vv>

Development Roadmap



Innovation Roadmap



Executed by IBM

On target