



@zapatacomputing

ZAPATA COMPUTING

Challenges and opportunities in practical near-term quantum computers

June 5, 2019

ITU Workshop on Quantum Information Technology

Christopher Savoie, PhD, JD

Founder and CEO



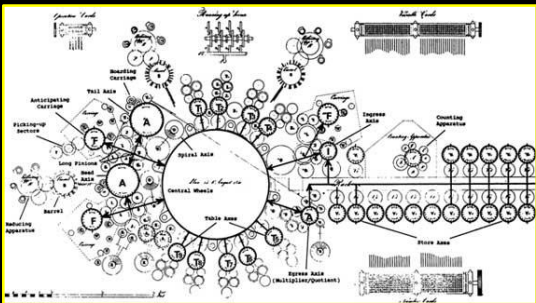
@cjsavoie

Abacus

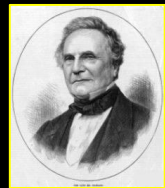


2700 BC

Thermodynamics



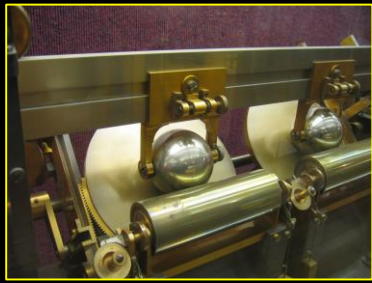
Analytical engine
Application: polynomial function evaluation
Failed due to limited funding and Babbage's own longevity



Charles Babbage



Ada Lovelace



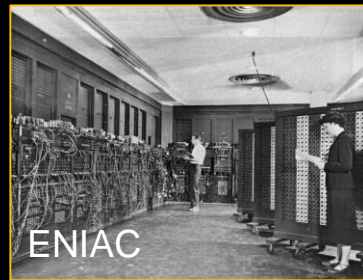
Ball-and-disk integrator
Special-purpose computer for industrial measurements e.g. areas and volumes

1886



Alan Turing

Theory of computation



ENIAC

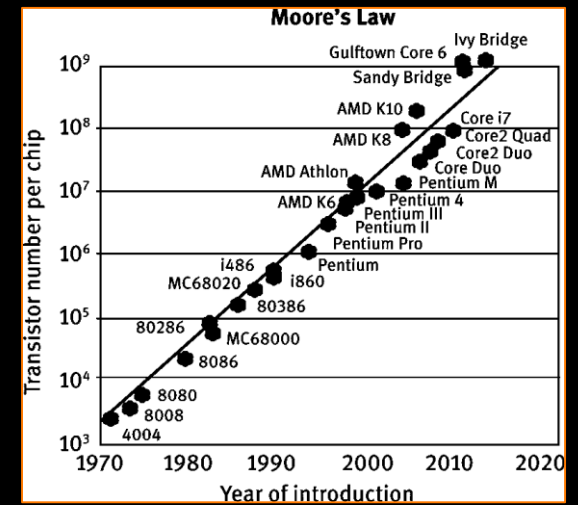
General-purpose computer that **significantly accelerated** Monte Carlo calculation over manual computation for Manhattan project

1945



John von Neumann

Electronic computers I (vacuum tube)
Electronic computers II (transistors)



Mechanical computers

Electro-Mechanical computers

Electronic computers I (vacuum tube)

1837

Electromagnetism

1890

Quantum physics



Tabulation machine
Application: Data tabulation
Special-purpose computer that demonstrated **clear advantage** over manual process for 1890 US census

1957



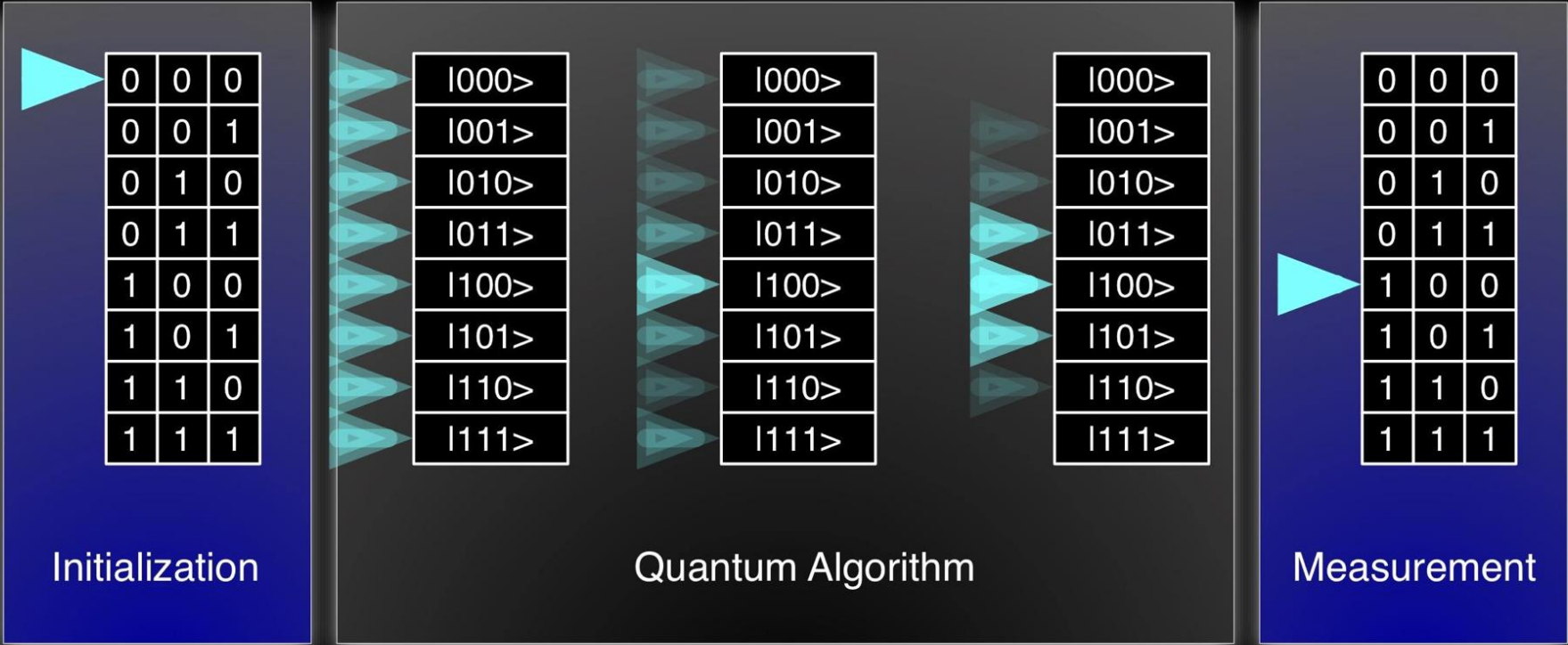
IBM 608
First commercial computer which is fully based on transistors

“...50 percent reduction in physical size and a 90 percent reduction in power requirements over comparable vacuum tube models...”

Quantum computation and quantum information

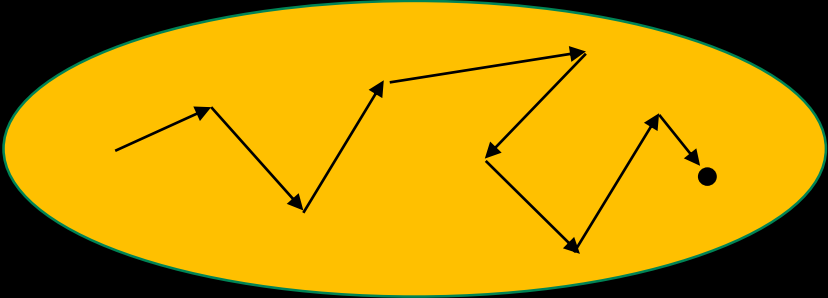
Quantum computers

A quantum computer is a collection of qubits that can be coherently controlled to process quantum information

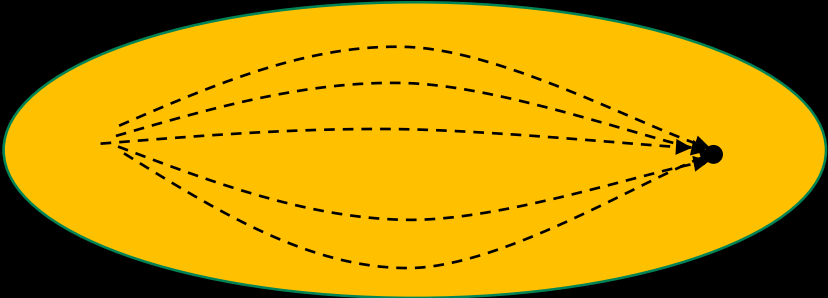


8 states for 3 qubits
16 states for 4 qubits
32 states for 5 qubits
⋮
 2^{300} states for 300 qubits

Number of Hydrogen atoms in the Universe: 2^{285}



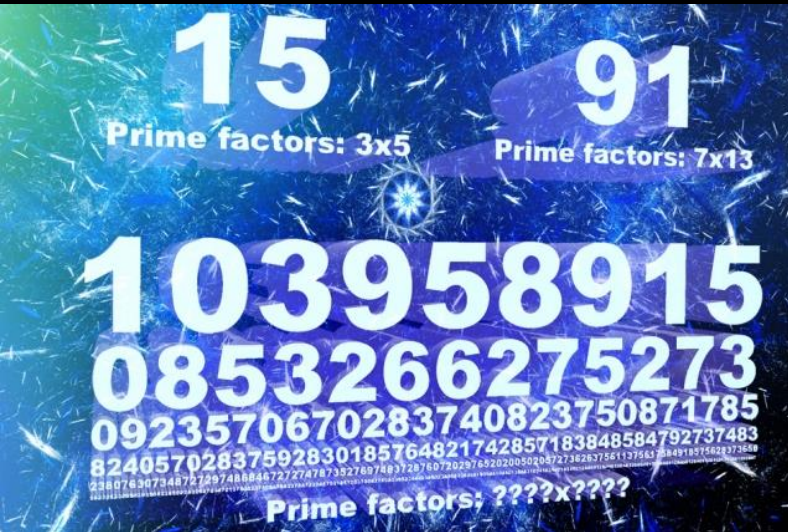
Classical computer



Quantum computer

Integer factorization

Shor's algorithm (1997)



Information security

Potential threat to encryption protocols such as RSA, ECC

Unstructured search

Grover's algorithm (1996)



Database

Brute-force search among a large number of items

Quantum simulation

Aspuru-Guzik et al. (2005)



Chemistry

Accurate calculation of molecular properties of drugs and materials

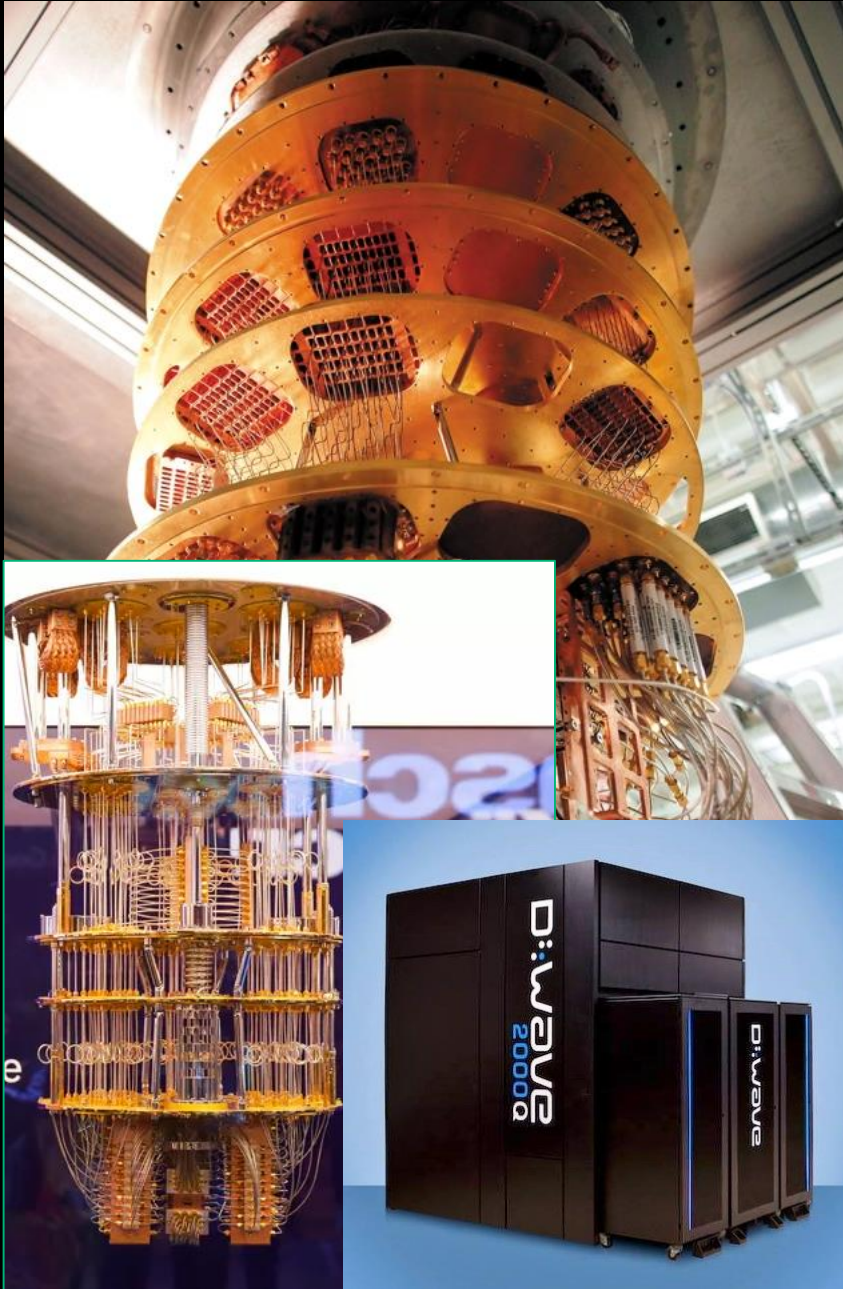
Shor, Peter W. "Polynomial-Time Algorithms for Prime Factorization and Discrete Logarithms on a Quantum Computer", *SIAM J. Comput.*, 26 (5): 1484–1509, 1997

Grover, L.K. "A fast quantum mechanical algorithm for database search", *Proceedings, 28th Annual ACM Symposium on the Theory of Computing*, p. 212, 1996

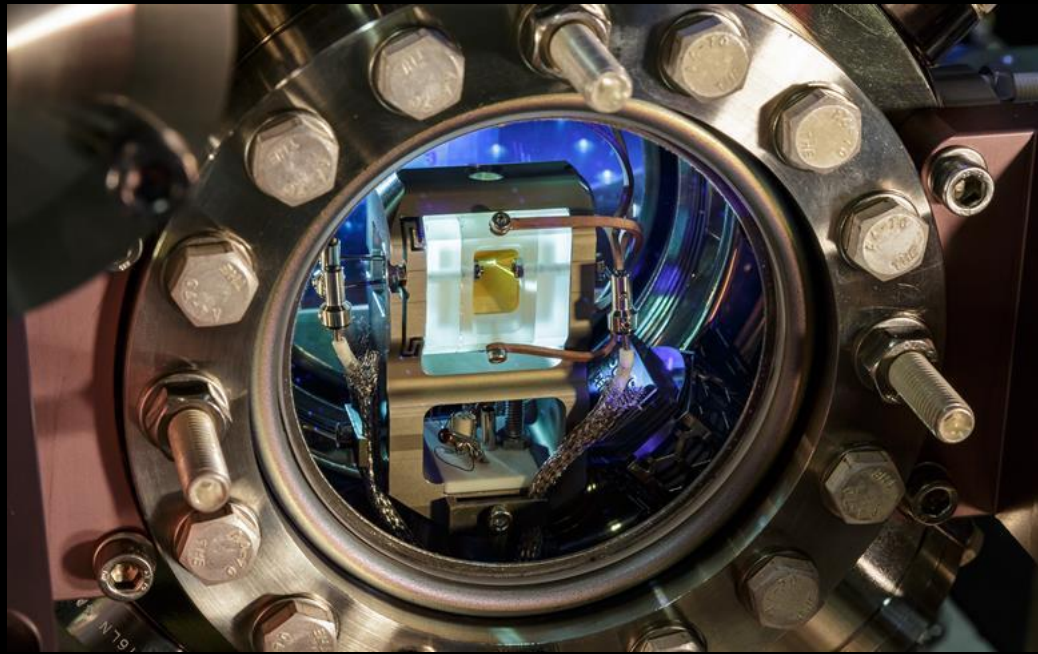
Aspuru-Guzik, A., A. Dutoi, P. J. Love, M. Head-Gordon. "Simulated quantum computation of molecular energies", *Science*, Vol. 309, Issue 5741, pp. 1704-1707, 2005

Cao, Y., J. Romero, J. Olson, M. Degroot, P. Johnson, et al. "Quantum chemistry in the age of quantum computing", *Chemical Review*. 2018. To appear.

Superconducting quantum computer

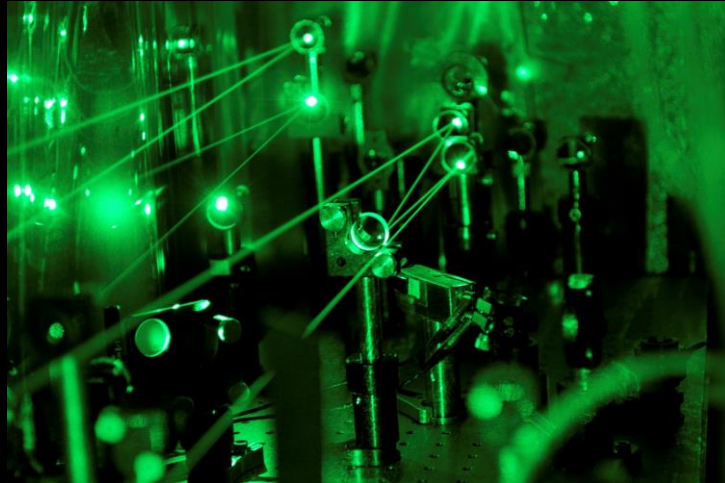


Trapped ion quantum computer



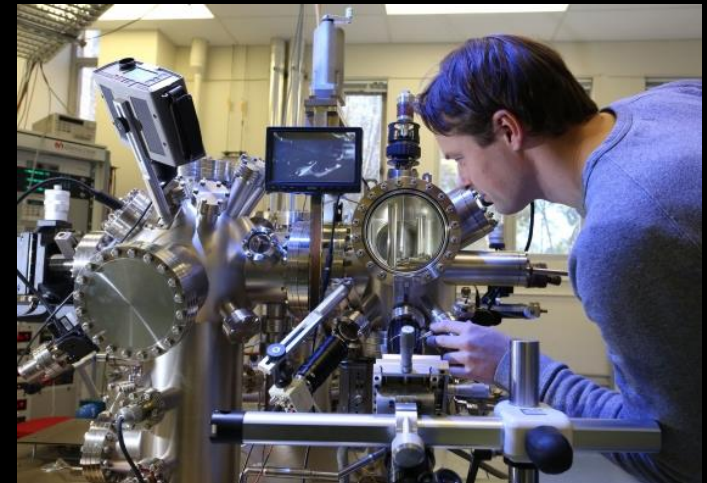
Source: © IQOQI / M. R. Knabl
The quantum computer uses trapped calcium ions as qubits

Photonic quantum computer

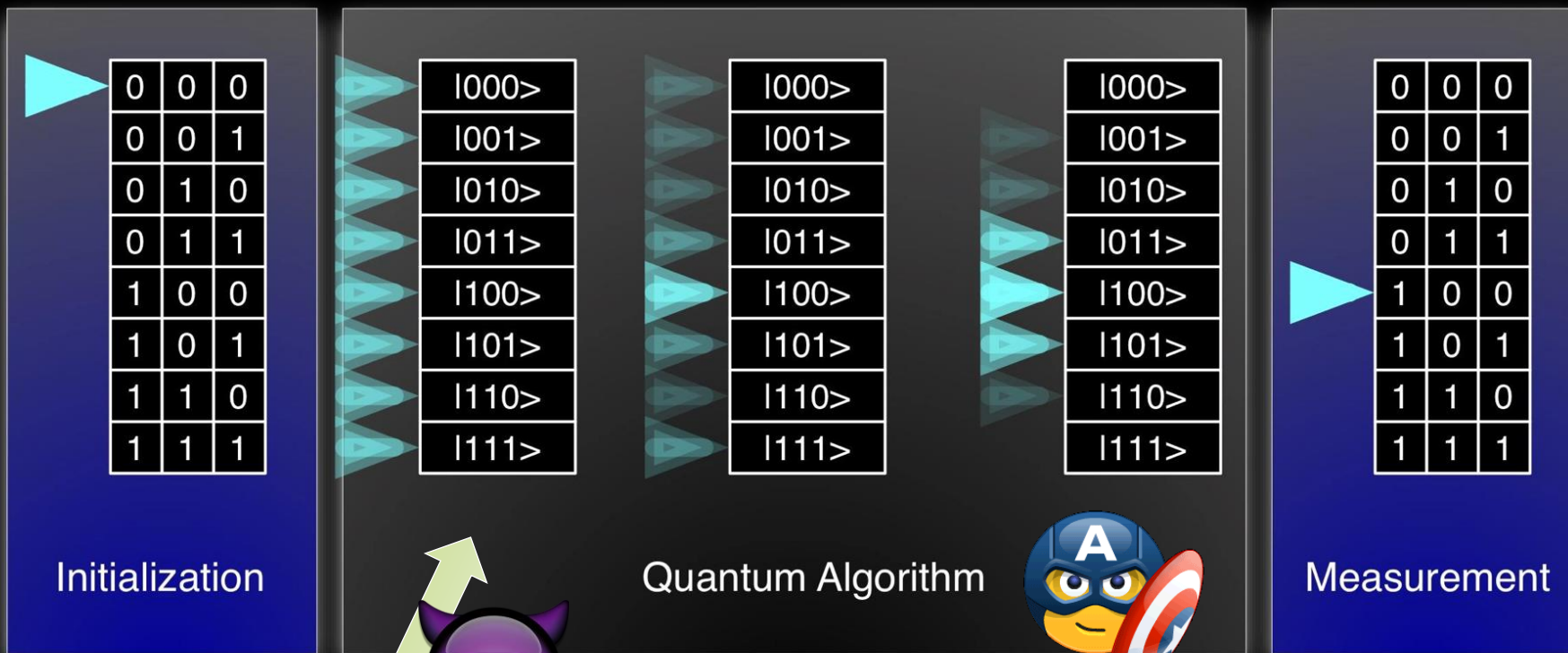


Credit: Zhejiang Daily

Silicon-based quantum computer



Source: UNSW Sydney



State-of-the-art QEC techniques require **~10000** physical qubits for each logical qubit

If the physical qubits are not too noisy, QEC can in principle allow for quantum computation of arbitrary duration, a regime known as **fault tolerance**

Environment

Decoherence
The physics of QC becomes **classical**

Quantum Error Correction (QEC)
Each **logical qubit** is encoded into **multiple physical qubits**

Resource estimate for quantum simulation problems of practical interest:
~1 million physical qubits, hours of QC



Current hardware: **<100 qubits, ~0.001s**

What useful applications can be found *before* scalable, fault-tolerant quantum computers are built?

NISQ era

- Noisy Intermediate Scale Quantum (NISQ) devices
- Limited quantum computational resource

- Qubit fidelity
- Gate fidelity
- Qubit connectivity
- SPAM error

Builds on decades of quantum system and electrical engineering



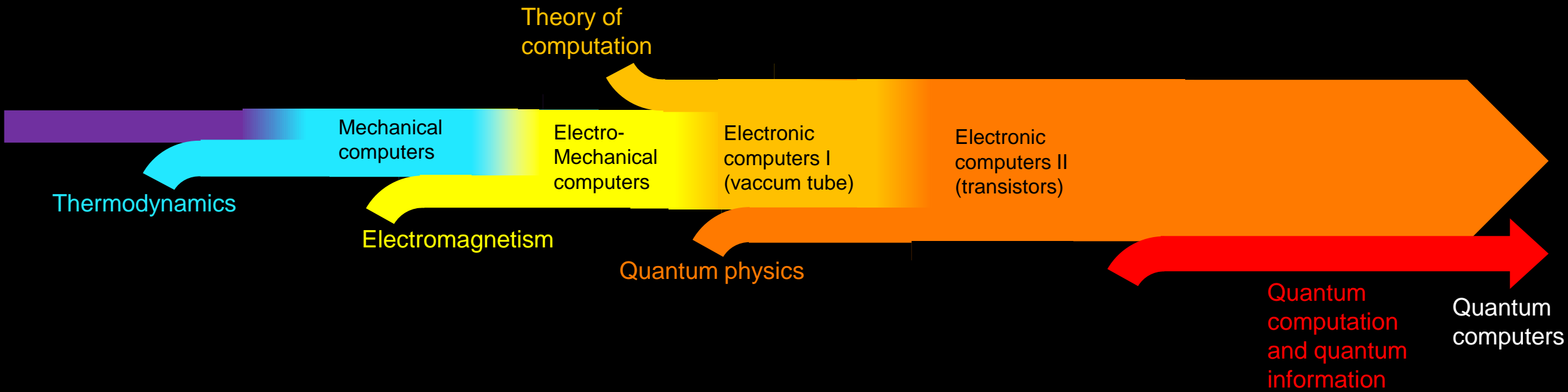
- Need to optimize *all* layers of abstraction

- Quantum algorithm
- Quantum circuit
- Compilation
- Hardware instruction

Builds on decades of computer science and engineering



(Useful near-term quantum applications)



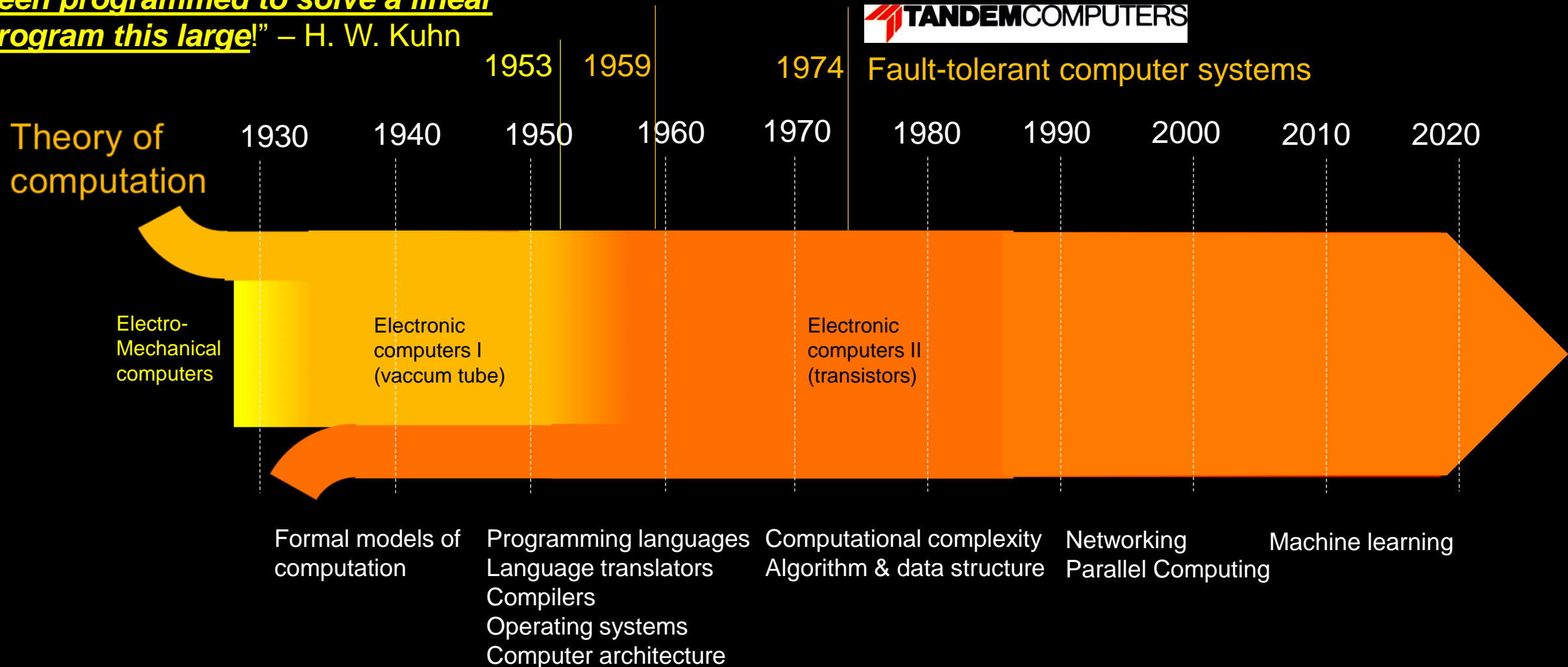
Electro-
Mechanical
computers

Theory of
computation

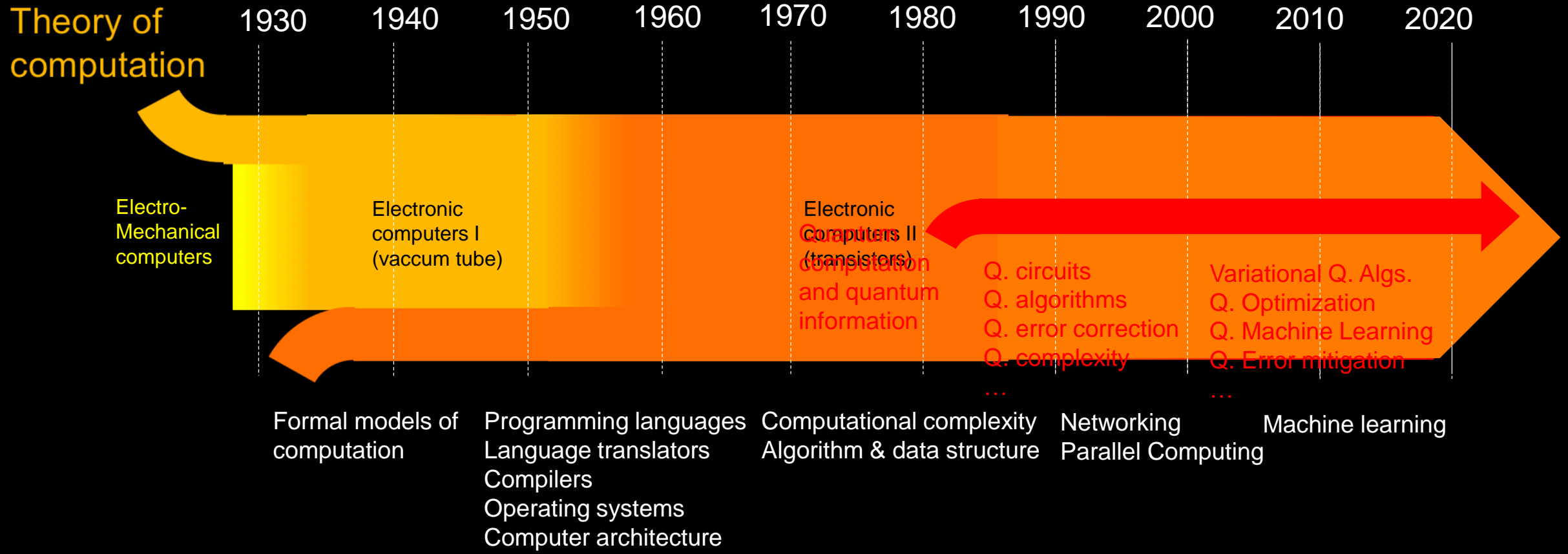


Solving a 10×10 assignment problem with the simplex method on the SEAC, consisted of 256 Williamson cathode ray tubes, took 20 minutes. “In 1953, there was no machine in the world that had been programmed to solve a linear program this large!” – H. W. Kuhn

“My Ph.D. thesis on computational quantum mechanics was done on the IBM 650, a 2000-word drum memory machine. I believe that the need to be very economical on this computer may have led to my early interest in optimal algorithms and computational complexity.” –J. F. Traub



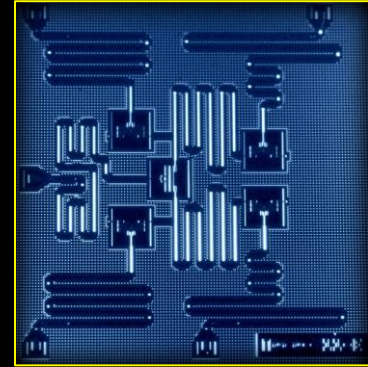
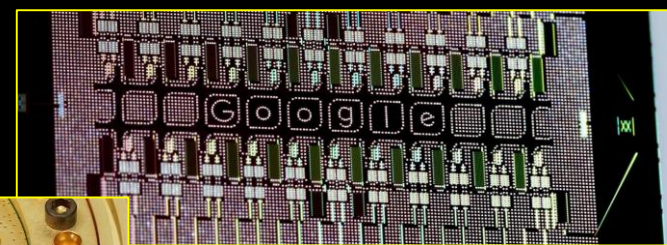
Accelerated progress in quantum algorithm developments vs. classical computation



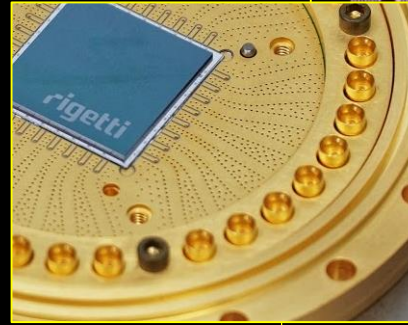


NISQ era

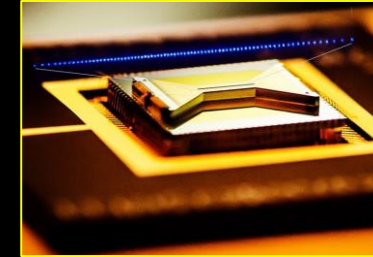
Google 22-qubit Foxtail Xmon



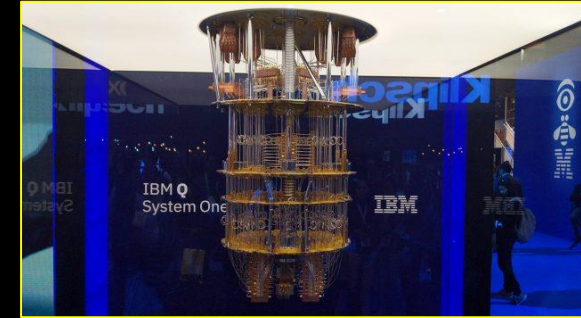
IBM 5-qubit processor (Q Experience)



Rigetti 19Q device



IonQ 11-qubit device



IBM System One (20 qubits)

2000

2010

2016

2020



2007

2011

2013

2015

2017

2018

2019



D-Wave 16-qubit Orion quantum annealer



D-Wave One, 128-qubits

D-Wave Two, 512-qubits

D-Wave 2X, 1152-qubits

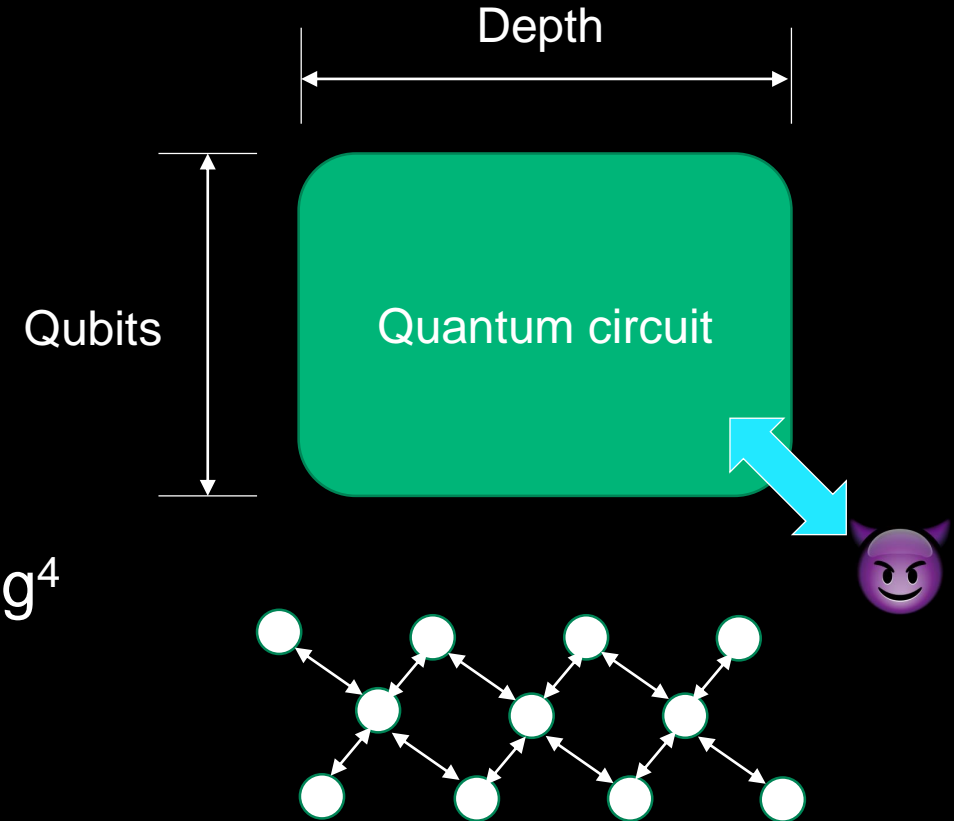
D-Wave 2000Q, 2048-qubits



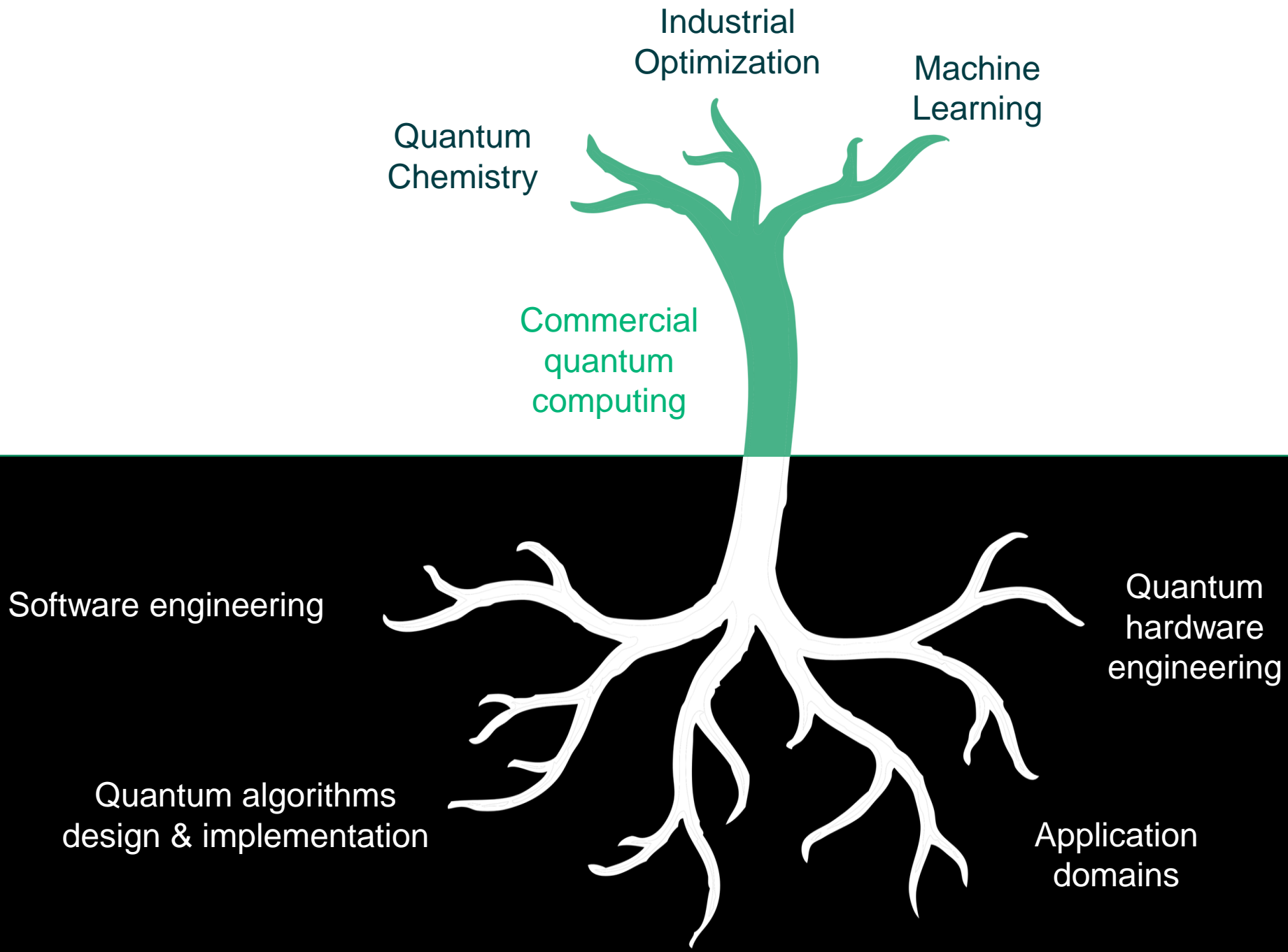
QC capability and performance metrics

It's not just about qubit counts.

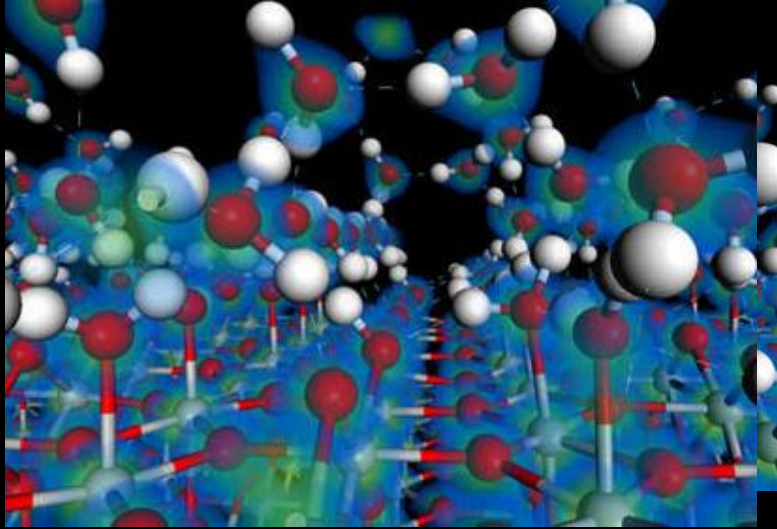
- **Qubit fidelity, gate fidelity**
 - Quantum volume¹
- **Qubit connectivity**
 - Ansatz expressive power in quantum chemistry² and supervised learning³
 - Embedding overhead in quantum annealing⁴
- **Cost to reproduce classically**
 - "Quantum supremacy" comparison⁵



[1] Andrew W. Cross, Lev S. Bishop, Sarah Sheldon, Paul D. Nation, Jay M. Gambetta. 2018. arXiv:1811.12926 [quant-ph]
[2] A. Kandala, A. Mezzacapo, K. Temme, M. Takita, M. Brink, J. M. Chow, J. M. Gambetta. Nature 549, 242. 2017. arXiv:1704.05018 [quant-ph]
[3] D. Zhu, N. M. Linke, M. Benedetti, K. A. Landsman, N. H. Nguyen, C. H. Alderete, A. Perdomo-Ortiz, N. Korda, A. Garfoot, C. Brecque, L. Egan, O. Perdomo, C. Monroe. 2018. arXiv:1812.08862 [quant-ph]
[4] Z. Li, N. S. Dattani, X. Chen, X. Liu, H. Wang, R. Tanburn, H. Chen, X. Peng, and J. Du, (2017), arXiv:1706.08061 [quant-ph]
[5] Igor L. Markov, Aneeqa Fatima, Sergei V. Isakov, Sergio Boixo. 2018. arXiv:1807.10749 [quant-ph]

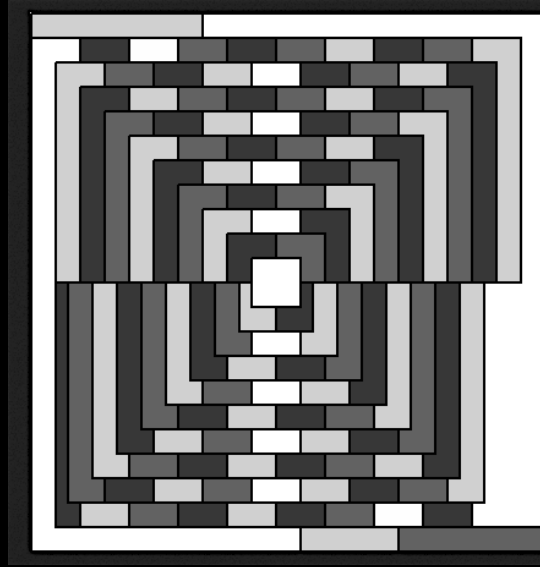


Quantum simulation



Variational quantum eigensolver
Hybrid quantum-classical simulator
Adiabatic quantum computing

Combinatorial optimization



Q. Approx. Optimization Algorithm
Q. Alternating Operator Ansatz
Q. Annealing / Optical Ising machines

Machine learning



Variational circuit classifier
Kernel PCA
Quantum Boltzmann machine
Quantum autoencoder
Quantum GAN

	Supervised	Unsupervised
NISQ	Variational quantum circuit classifier [122]–[128] Kernel-based quantum-classical classifier [124], [125] Quantum Boltzmann machine [133], [134] Quantum training of classical Boltzmann machine [136]	Quantum autoencoder [129]–[131] Hybrid quantum-classical variational autoencoder [132] Hybrid quantum-classical Helmholtz machine [135] Quantum circuit-based generative modeling [137] Learning probabilistic graphical models [138] Quantum generative adversarial networks [117] Hybrid quantum-classical clustering [139] Quantum Boltzmann machine [133], [134] Quantum training of classical Boltzmann machine [136]
FTQC	Quantum-enhanced classical Boltzmann machine [140] Quantum nearest-neighbor classification [141] Quantum least-squares regression [143] Quantum support vector machine [145] Quantum perceptron models [147], [148] Quantum Bayesian inference [149] Quantum-enhanced Bayesian deep learning [150]	Quantum k -means clustering [141] Quantum principal component analysis [142] Quantum generative adversarial networks [144] Quantum Hopfield network [146] Quantum-enhanced classical Boltzmann machine [140]



High-tech

- Machine learning and artificial intelligence, such as neural networks
- Search
- Bidding strategies for advertisements
- Cybersecurity
- Online and product marketing
- Software verification and validation

IBM

Telstra

Alibaba

Baidu

Google

Samsung

Microsoft



Industrial goods

- Logistics: scheduling, planning, product distribution, routing
- Automotive: traffic simulation, e-charging station and parking search, autonomous driving
- Semiconductors: manufacturing, such as chip layout optimization
- Aerospace: R&D and manufacturing, such as fault-analysis, stronger polymers for airplanes
- Material science: effective catalytic converters for cars, battery cell research, more-efficient materials for solar cells, and property engineering uses such as OLEDs

Airbus

BMW

NASA

Volkswagen

Northrop
Grumman

Lockheed
Martin

Daimler

Honeywell

Raytheon

Bosch



Chemistry and Pharma

- Catalyst and enzyme design, such as nitrogenase
- Pharmaceuticals R&D, such as faster drug discovery
- Bioinformatics, such as genomics
- Patient diagnostics for health care, such as improved diagnostic capability for MRI

BASF

JSR

Biogen

DuPont

Dow
Chemical

Amgen



Finance

- Trading strategies
- Portfolio optimization
- Asset pricing
- Risk analysis
- Fraud detection
- Market simulation

J.P. Morgan

Barclays

Commonwealth
Bank

Goldman
Sachs



Energy

- Network design
- Energy distribution
- Oil well optimization

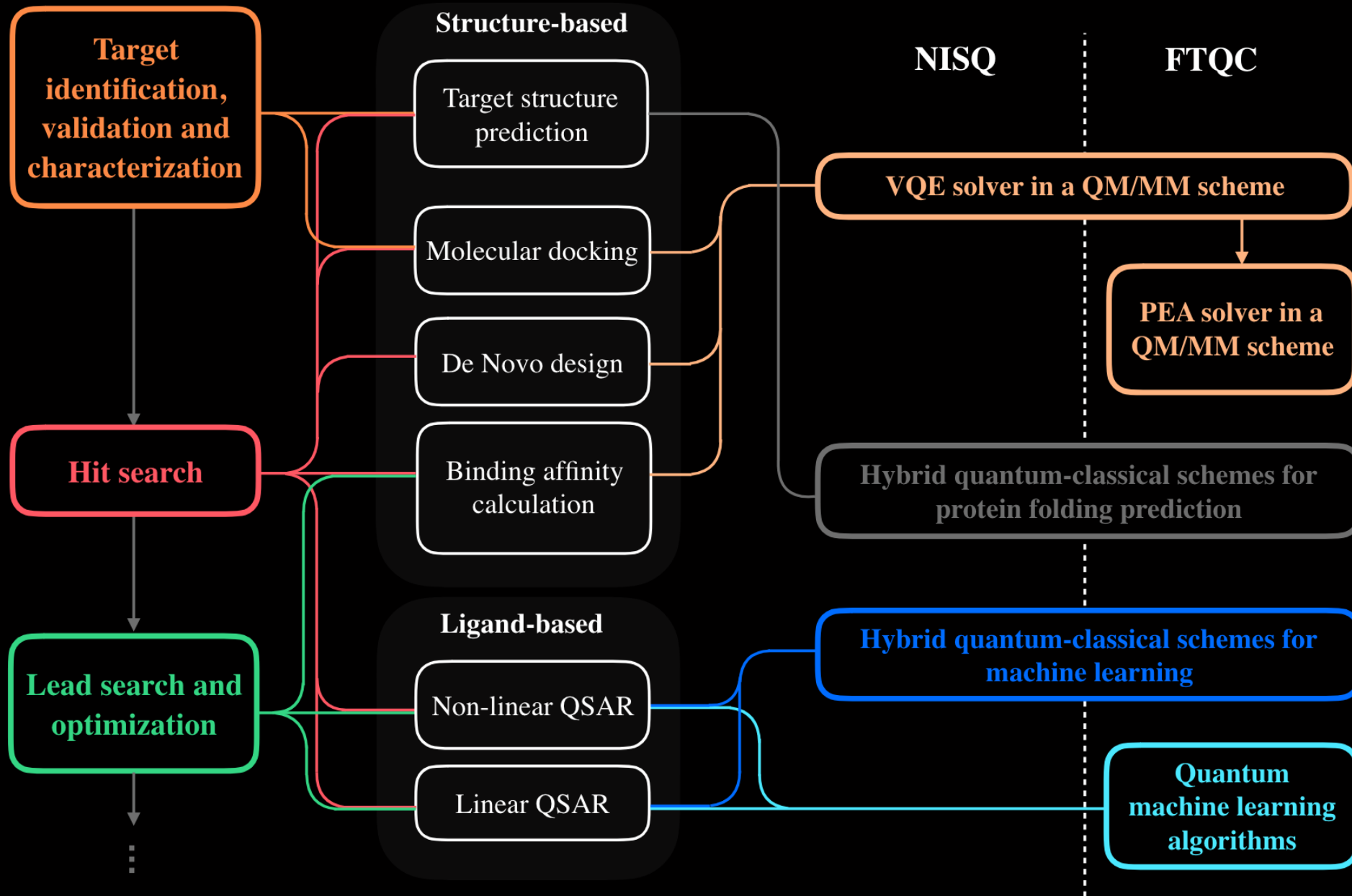
Dubai
Electricity &
Water Authority

BP

Drug discovery pipeline

CADD

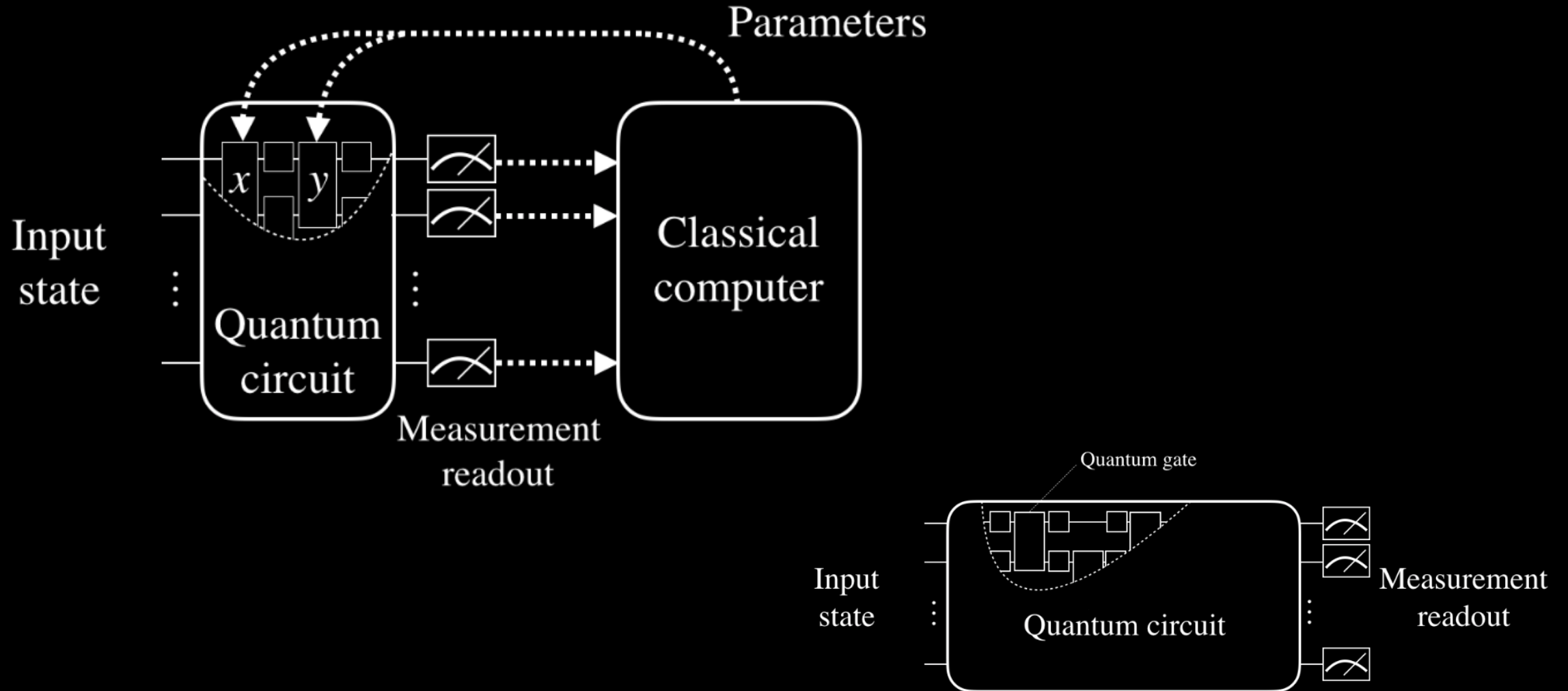
Quantum techniques



Y. Cao, J. Romero, A. Aspuru-Guzik. *Potential of quantum computing for drug discovery*. IBM Journal of Research and Development. Vol. 62, Issue 6. 2018

Near-term QC paradigm

Hybrid model that maximizes the use of both quantum and classical processors



What can you do with existing devices?

Some examples of demo experiments run on hardware

	Algorithm	# qubits	Circuit depth	Reference
ML	Variational circuit classifier	2	3	Partner app - Rigetti / Zapata ¹
ML / Chem	Quantum autoencoder	2, 3	<8	Partner app - Rigetti / Zapata ¹
Chem	VQE for H ₂ ground state	2-4	<7	SC qubits (Google ² , IBM ³), trapped ion ⁴
Optimization	Factorization	3	<10	IBM Hackathon - VQF ⁵ (Zapata)
QI	Mermin inequality violation	3-5	<15	IBM Q Experience ⁶
QI	Quantum info. scrambling	7	<10	Trapped ion QC ⁷
ML / Opt.	Clustering	19	5	Rigetti 19Q ⁸

[1] S. Sim, Y. Cao, J. Romero, P. D. Johnson, A. Aspuru-Guzik. 2018. arXiv:1810.10576 [quant-ph]

[2] P. J. J. O'Malley et al. *Phys. Rev. X* 6, 031007 (2016). arXiv:1512.06860 [quant-ph]

[3] A. Kandala, A. Mezzacapo, K. Temme, M. Takita, M. Brink, J. M. Chow, J. M. Gambetta. *Nature* 549, 242 (2017). arXiv:1704.05018 [quant-ph]

[4] C. Hempel et al. *Phys. Rev. X* 8, 031022 (2018). arXiv:1803.10238 [quant-ph]

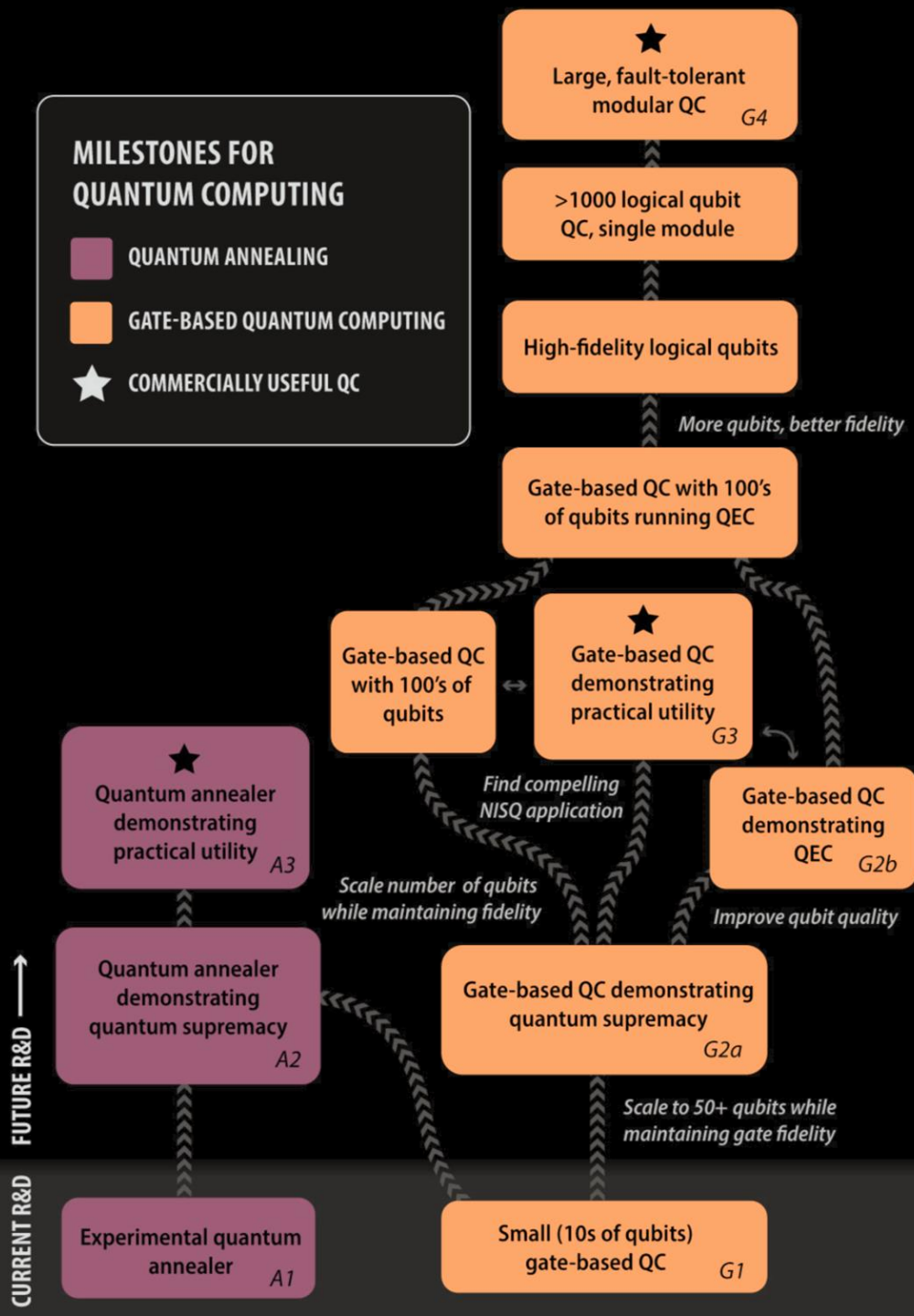
[5] E. R. Anschuetz, J. P. Olson, A. Aspuru-Guzik, Y. Cao. *Lecture Notes in Computer Science*, Vol. 11413, Ch. 7. 2018. arXiv:1808.08927 [quant-ph]

[6] Daniel Alsina, José Ignacio Latorre. *Phys. Rev. A* 94, 012314 (2016). arXiv:1605.04220 [quant-ph]

[7] K. A. Landsman, C. Figgatt, T. Schuster, N. M. Linke, B. Yoshida, N. Y. Yao, C. Monroe. *Nature* 567 61-65 (2019). arXiv:1806.02807 [quant-ph]

[8] J. S. Otterbach et al. 2017. arXiv:1712.05771 [quant-ph]

QC milestones



National Academies of Sciences, Engineering, and Medicine. 2018. *Quantum Computing: Progress and Prospects*. Washington, DC: The National Academies Press. <https://doi.org/10.17226/25196>.

Summary

Before general-purpose fault-tolerant devices mature, special-purpose noisy devices are already commercially relevant

Hardware and algorithm / theory co-evolve in moving the goal posts closer to each other

Government funding is crucial for sustaining the field in the early stage towards more commercial maturity

For quantum computing, we are building on decades of accumulation in science, technology and enterprise know-hows in classical computation. Timeline of progress will likely be accelerated significantly as a result.

More algorithmic innovation is needed to maximize the use of NISQ devices