

Future of Quantum Computing

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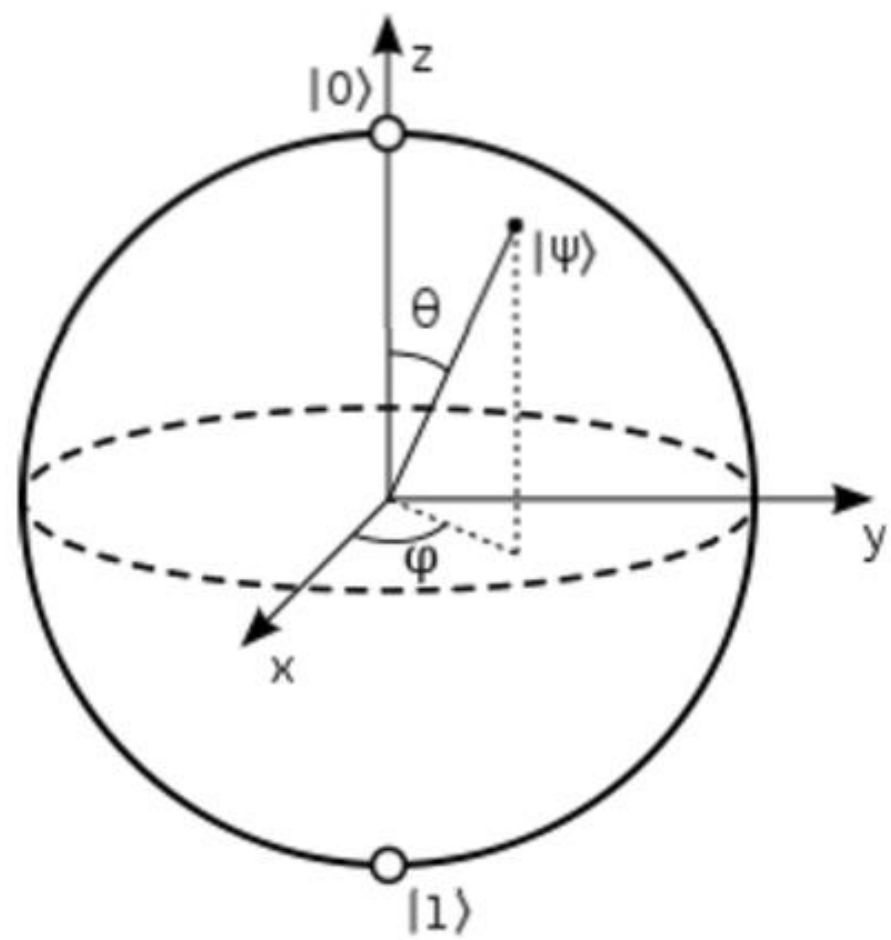


What is a quantum computer?

- a computer that relies on special memory, "quantum bit", to perform massively parallel computing.

What is a quantum bit?

- a basic unit of memory that uses superposition of "quantum" effect (entanglement) to store information.
- a "qubit" stores the probability of information. It represents both "1" and "0" at the same time.



What is the advantage?

- It is very very fast compared to conventional computers
- It has very large memory, example 10-qubit is

10 Qubits

- `>>> print(2**(2**10))`
- 179769313486231590772930519078902473361797697894230657273
430081157732675805500963132708477322407536021120113879871
393357658789768814416622492847430639474124377767893424865
485276302219601246094119453082952085005768838150682342462
881473913110540827237163350510684586298239947245938479716
304835356329624224137216
- `>>>`

How to make a quantum bit?

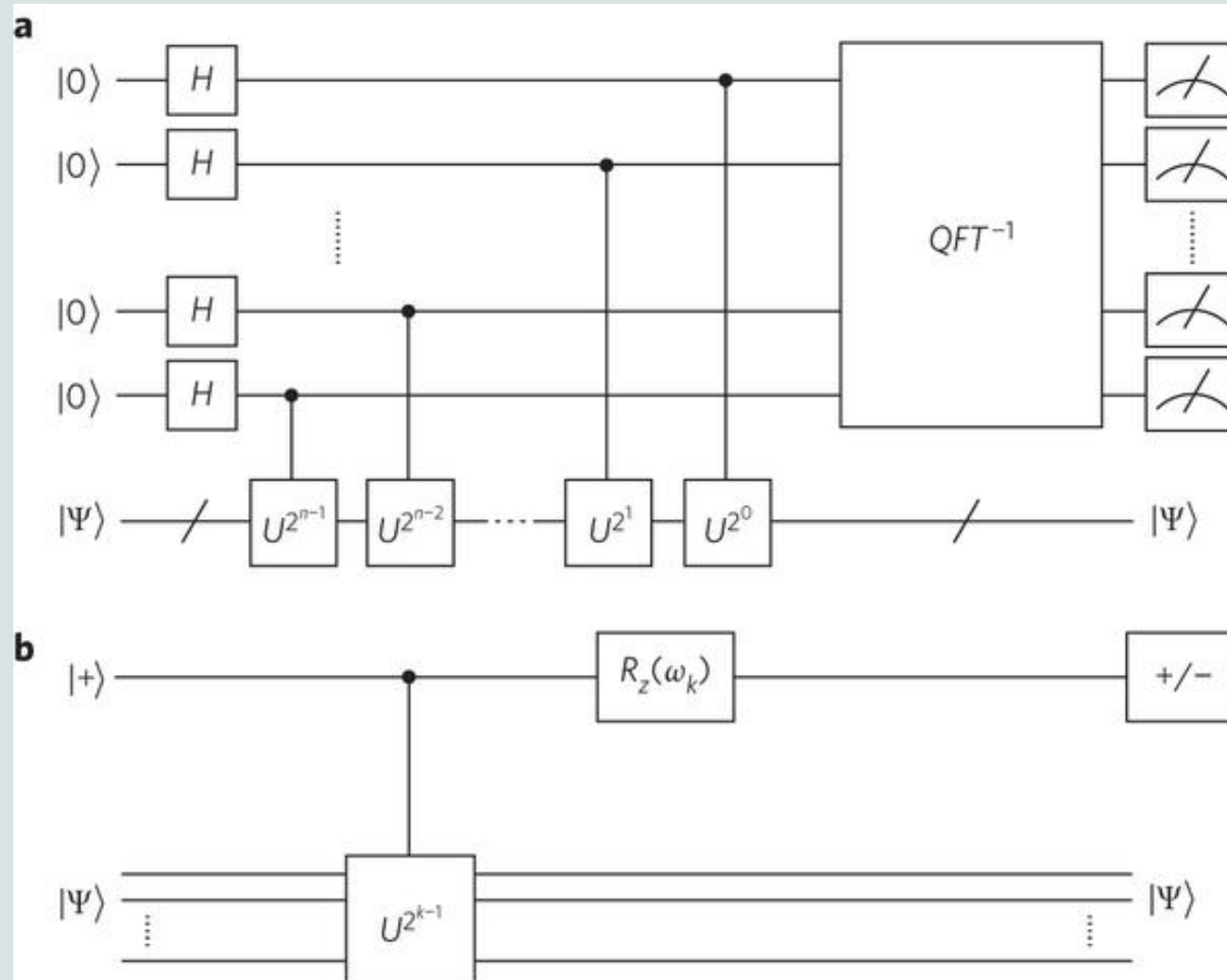
- "quantum effect"
- photon entanglement
- cold atom
- electron spin

Technology to implement Quantum bits

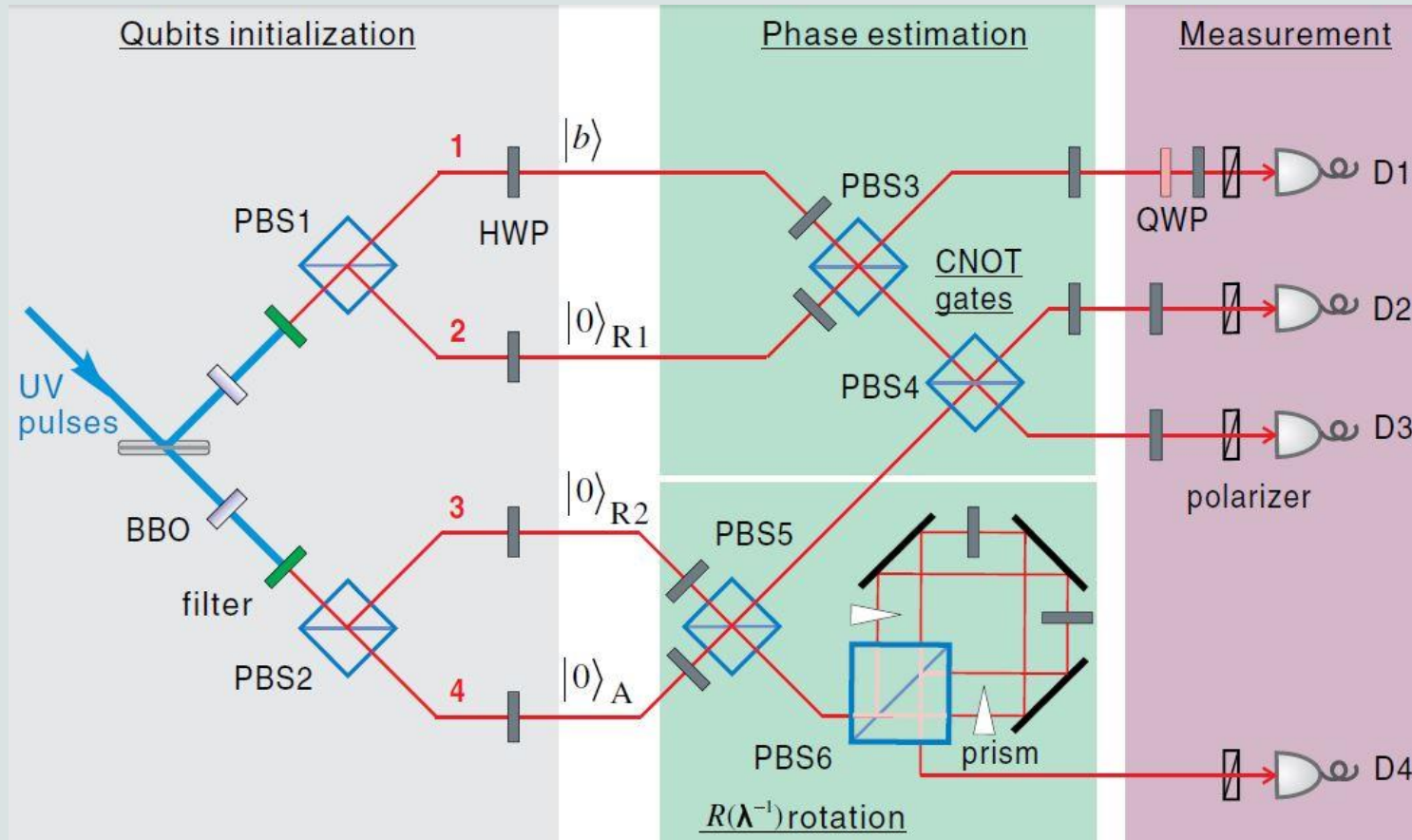
System	τ_Q	τ_{op}	$n_{op} = \lambda^{-1}$
Nuclear spin	$10^{-2} - 10^8$	$10^{-3} - 10^{-6}$	$10^5 - 10^{14}$
Electron spin	10^{-3}	10^{-7}	10^4
Ion trap (In^+)	10^{-1}	10^{-14}	10^{13}
Electron – Au	10^{-8}	10^{-14}	10^6
Electron – GaAs	10^{-10}	10^{-13}	10^3
Quantum dot	10^{-6}	10^{-9}	10^3
Optical cavity	10^{-5}	10^{-14}	10^9
Microwave cavity	10^0	10^{-4}	10^4

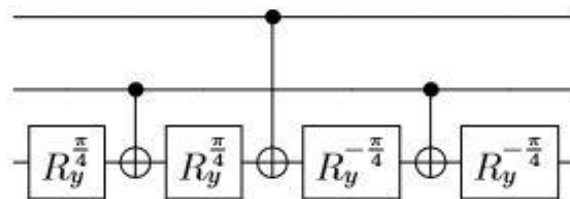
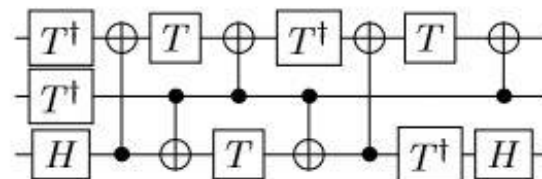
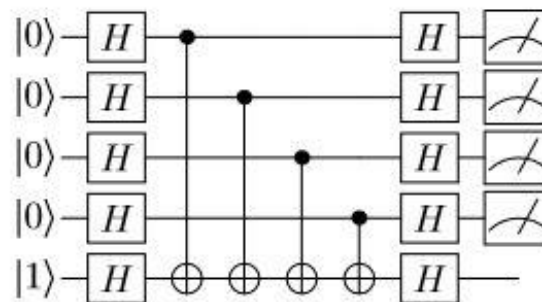
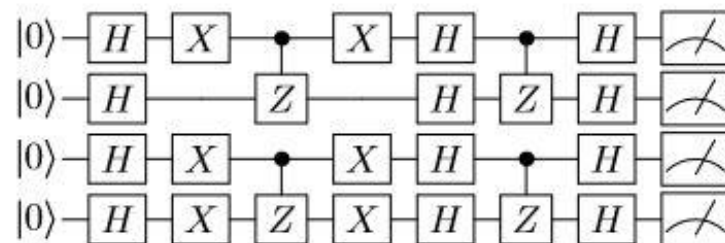
Figure 7.1. Crude estimates for decoherence times τ_Q (seconds), operation times τ_{op} (seconds), and maximum number of operations $n_{op} = \lambda^{-1} = \tau_Q/\tau_{op}$ for various candidate physical realizations of interacting systems of quantum bits. Despite the number of entries in this table, only three fundamentally different qubit representations are given: spin, charge, and photon. The ion trap utilizes either fine or hyperfine transitions of a trapped atom (Section 7.6), which correspond to electron and nuclear spin flips. The estimates for electrons in gold and GaAs, and in quantum dots are given for a charge representation, with an electrode or some confined area either containing an electron or not. In optical and microwave cavities, photons (of frequencies from gigahertz to hundreds of terahertz) populating different modes of the cavities represent the qubit. Take these estimates with a grain of salt: they are only meant to give some perspective on the wide range of possibilities.

Quantum circuits



Quantum circuits



a**b****c****d**

Quantum algorithms

- computer programs that work on quantum computers

Famous algorithms

- Shor's integer factorization
- Given an integer N , find its prime factors

Quantum Algorithms

1994 Peter Shor

a quantum algorithm for
integer factorization
formulated .



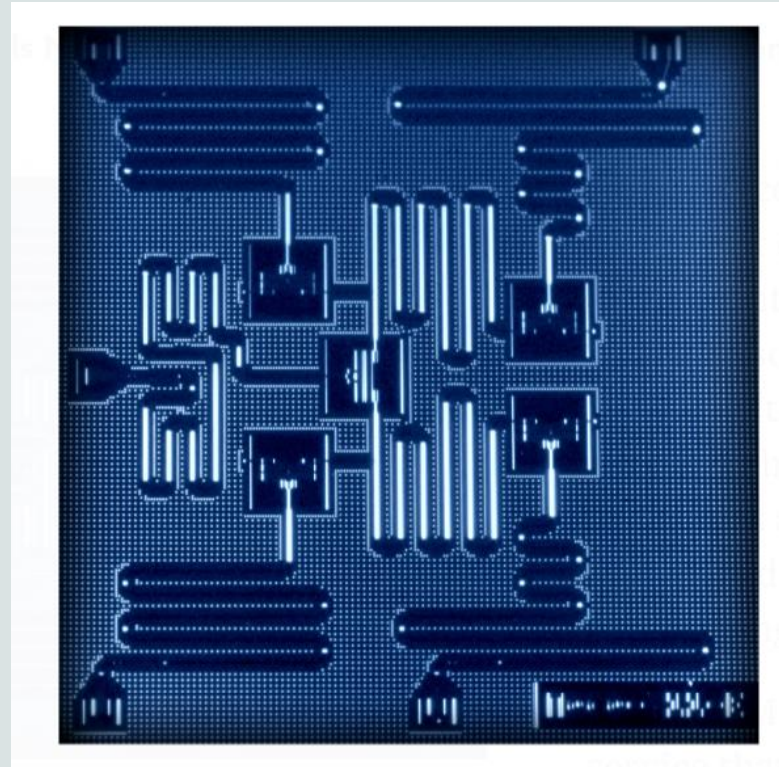
Shor's algorithm

The factorization also needs huge amount of quantum gates. It increases with N as $(\log N)^3$. Thus factoring of a 4096-bit number requires 4,947,802,324,992 quantum gates.

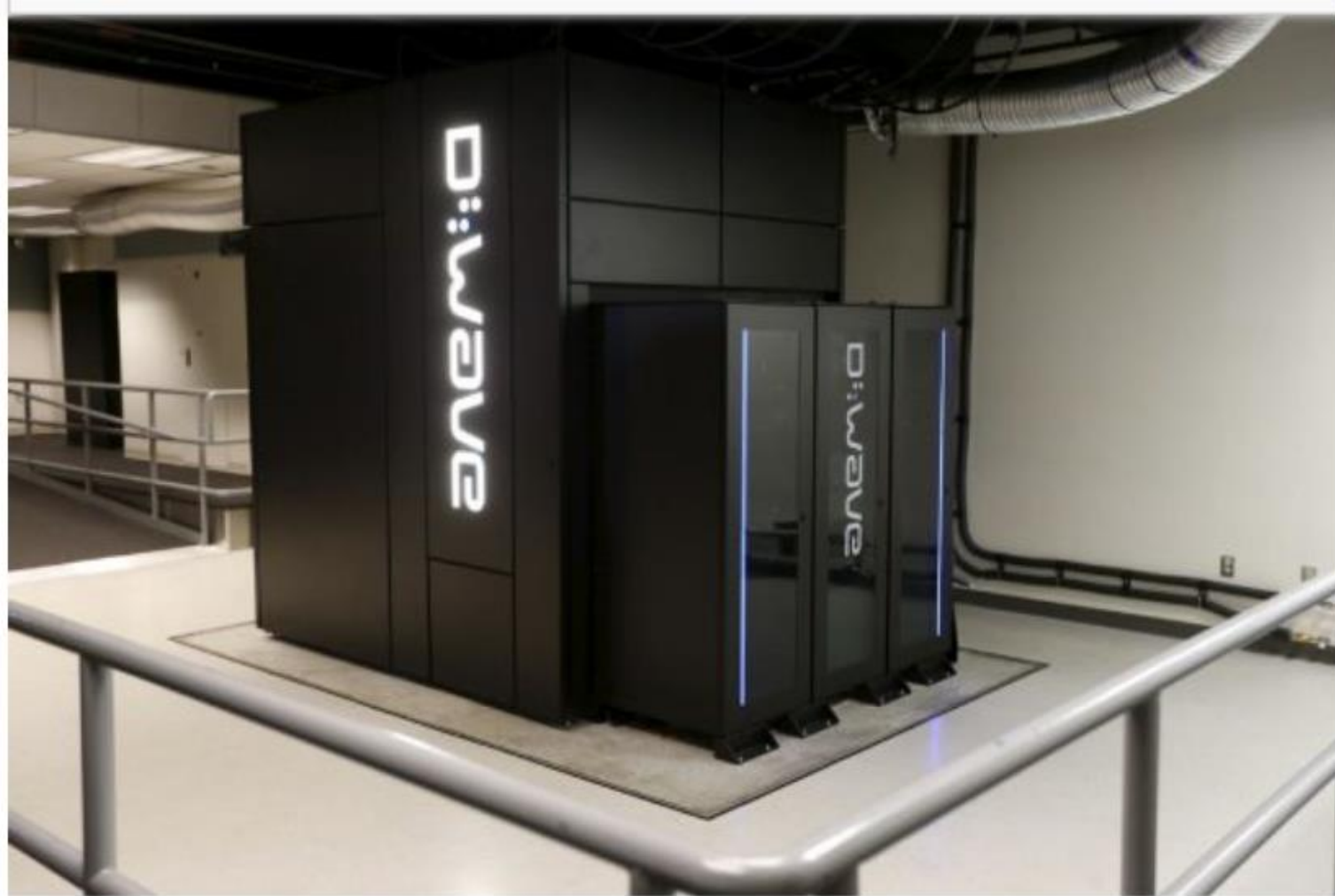
Example of quantum computers

- ibm 5 qubits
- D-wave two, quantum annealing

IBM 5 qubits processor



Google Nasa, D-Wave 2x machine



STEPHEN LAM / REUTERS

Claim Quantum Computer is fast

- google quantum lab's paper
- claim of 100,000,000x speed up

What is the Computational Value of Finite Range Tunneling?

Vasil S. Denchev,¹ Sergio Boixo,¹ Sergei V. Isakov,¹ Nan Ding,¹ Ryan Babbush,¹ Vadim Smelyanskiy,¹ John Martinis,² and Hartmut Neven¹

¹*Google Inc., Venice, CA 90291, USA*

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(Dated: December 31, 2015)

Quantum annealing (QA) has been proposed as a quantum enhanced optimization heuristic exploiting tunneling. Here, we demonstrate how finite range tunneling can provide considerable computational advantage. For a crafted problem designed to have tall and narrow energy barriers separating local minima, the D-Wave 2X quantum annealer achieves significant runtime advantages relative to Simulated Annealing (SA). For instances with 945 variables, this results in a time-to-99%-success-probability that is $\sim 10^8$ times faster than SA running on a single processor core. We also compared physical QA with Quantum Monte Carlo (QMC), an algorithm that emulates quantum tunneling on classical processors. We observe a substantial constant overhead against physical QA: D-Wave 2X again runs up to $\sim 10^8$ times faster than an optimized implementation of QMC on a single core. We note that there exist heuristic classical algorithms that can solve most instances of Chimera structured problems in a timescale comparable to the D-Wave 2X. However, we believe that such solvers will become ineffective for the next generation of annealers currently being designed. To investigate whether finite range tunneling will also confer an advantage for problems of practical interest, we conduct numerical studies on binary optimization problems that cannot yet be represented on quantum hardware. For random instances of the number partitioning problem, we find numerically that QMC, as well as other algorithms designed to simulate QA, scale better than SA and better than the best known classical algorithms for this problem. We discuss the implications of these findings for the design of next generation quantum annealers.

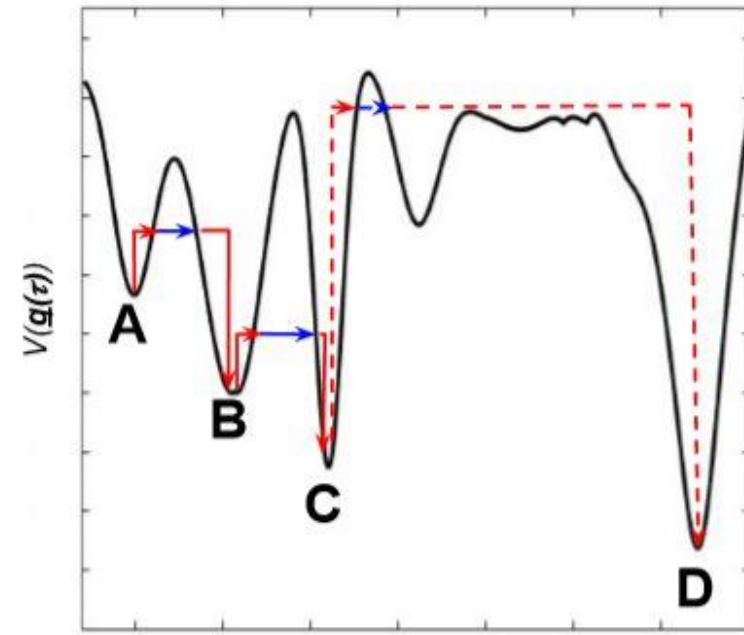
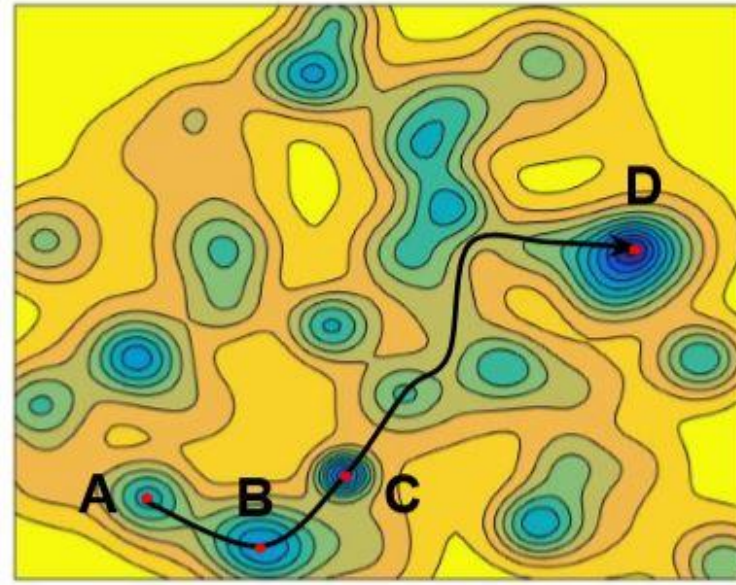
I. INTRODUCTION

Simulated annealing (SA) [1] is perhaps the most widely used algorithm for global optimization of pseudo-Boolean functions with little known structure. The objective function for this general class of problems is

standard time-dependent Hamiltonian used for QA is

$$H(t) = -A(t) \sum_{j=1}^N \sigma_j^x + B(t) H_P, \quad (2)$$

where H_P is written as in Eq. (1) but with the spin variables replaced with ± 1 Pauli matrices acting on qubits



My own example of quantum computation

- compact genetic algorithm by quantum computers
- exponential speedup

An Implementation of Compact Genetic Algorithm on a Quantum Computer

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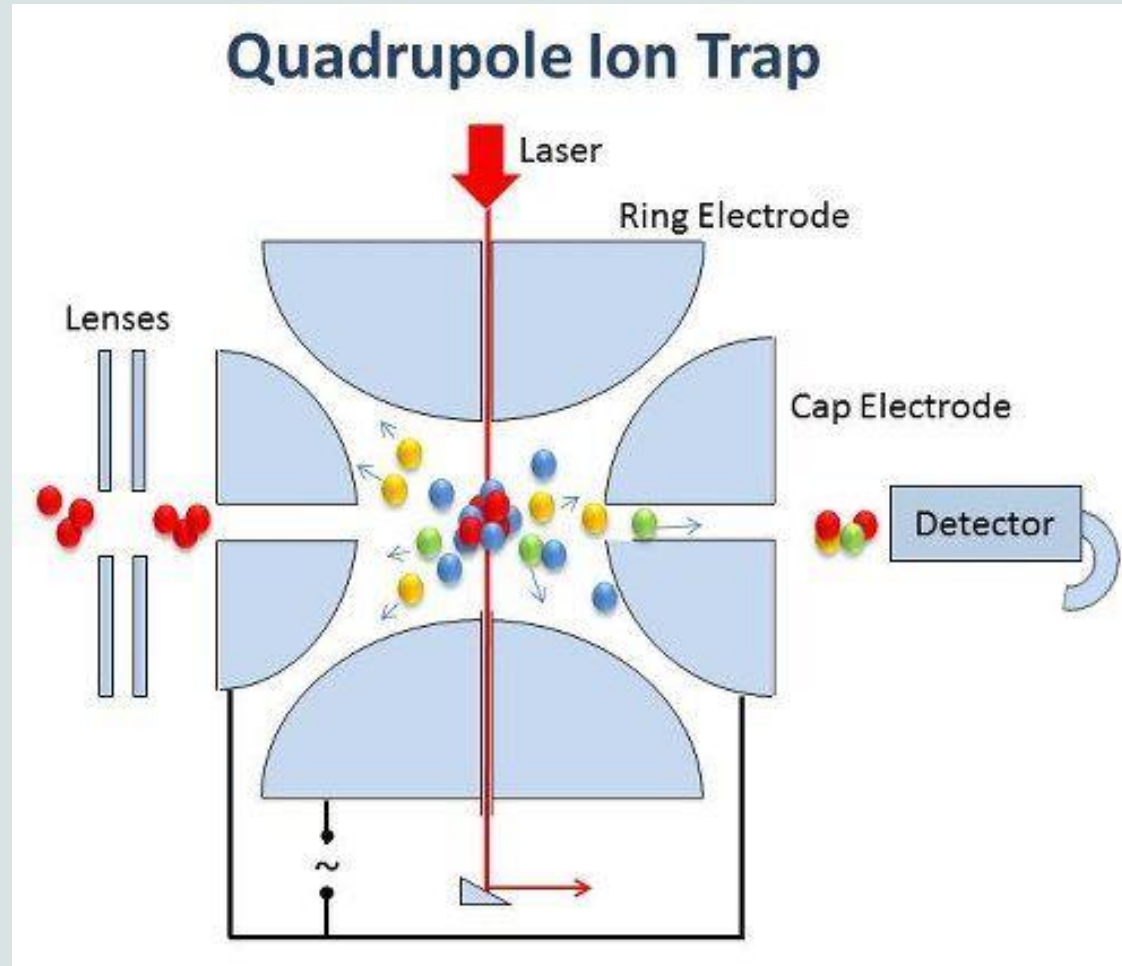
Abstract— Programming a quantum computer posts a challenge. It is not straight forward to transfer the current programming skill on a classical computer to a quantum computer. This work presents an example of programming a quantum computer. The compact genetic algorithm is used as a target as it is powerful and popular method in evolutionary computation. A quantum bit (qubit) concept was introduced as a basis for storing information. The representation of quantum register has benefits over

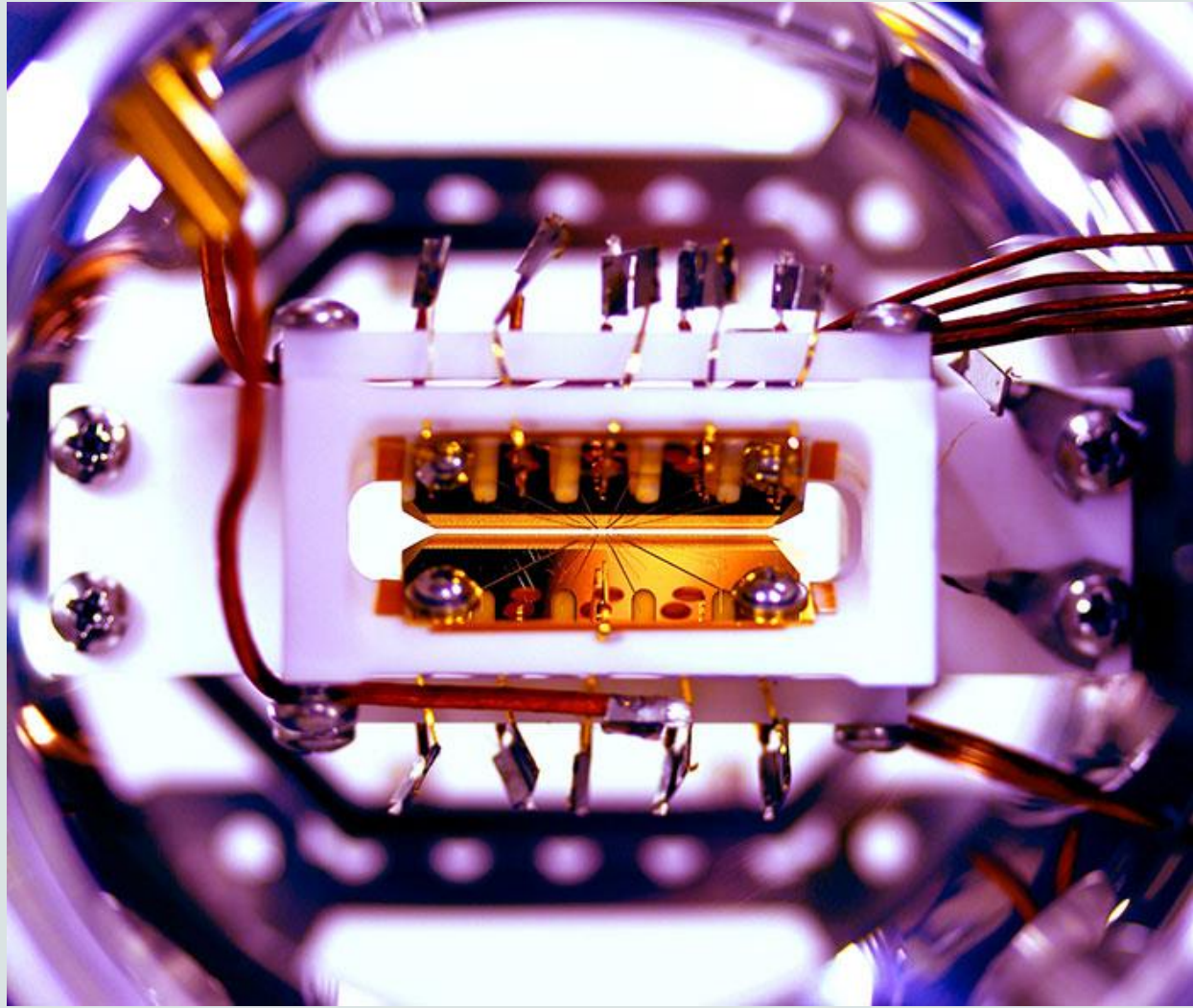
Computation Language) [5]—[7] as an emulator of quantum computer.

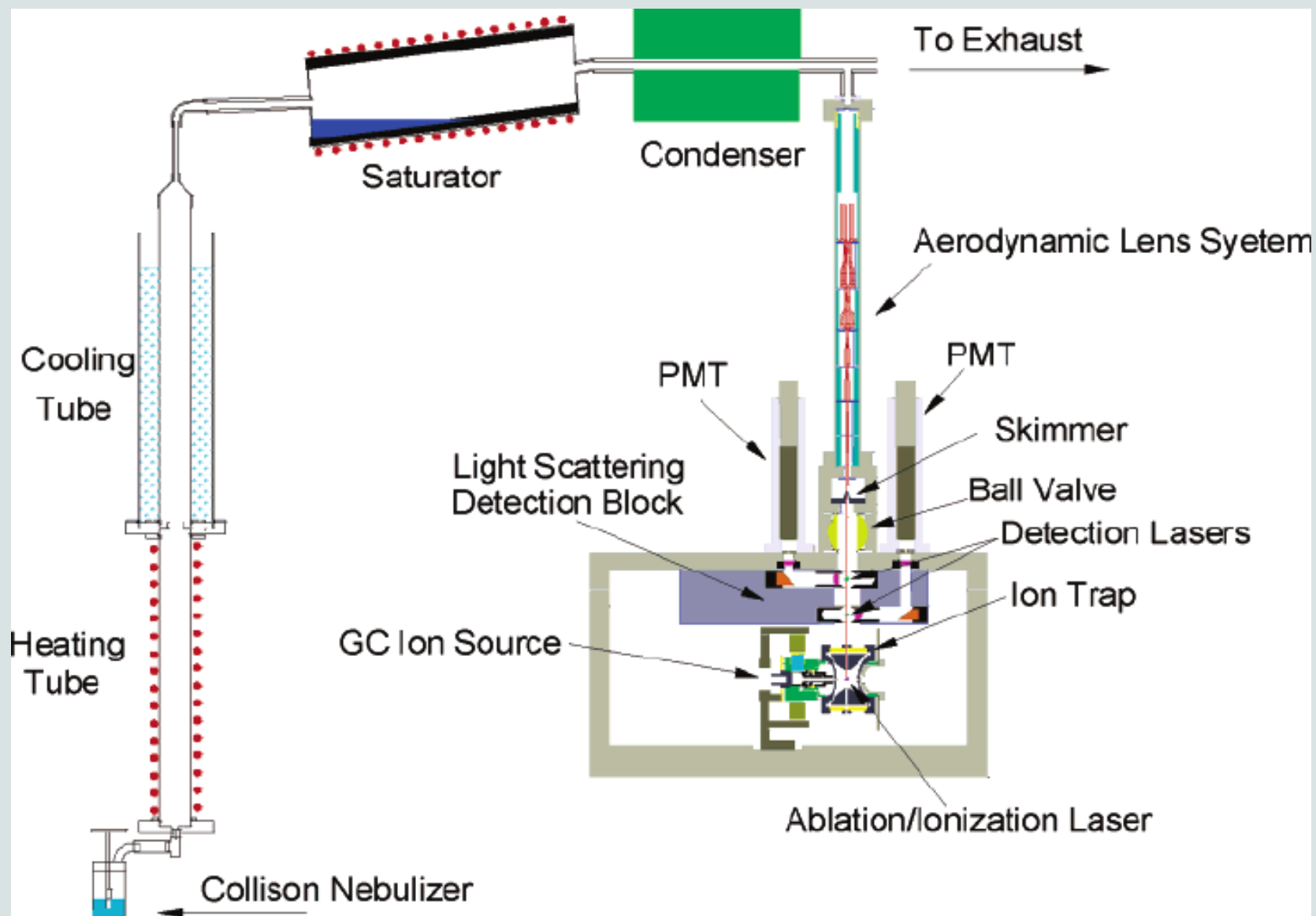
II. COMPACT GENETIC ALGORITHMS

Genetic algorithms are adaptive search algorithms based on the idea of biological evolution such as natural selection, cross over and mutation. Compact genetic algorithm represents the

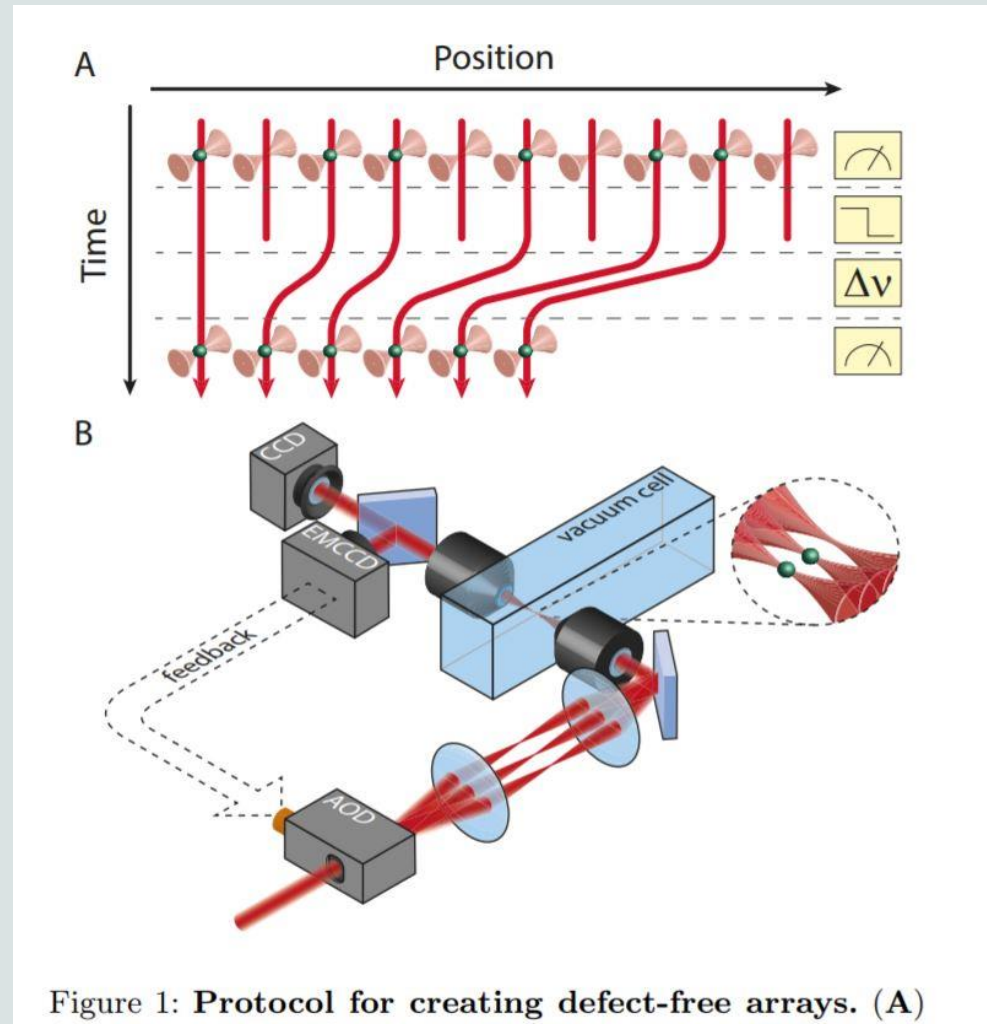
Quantum computers

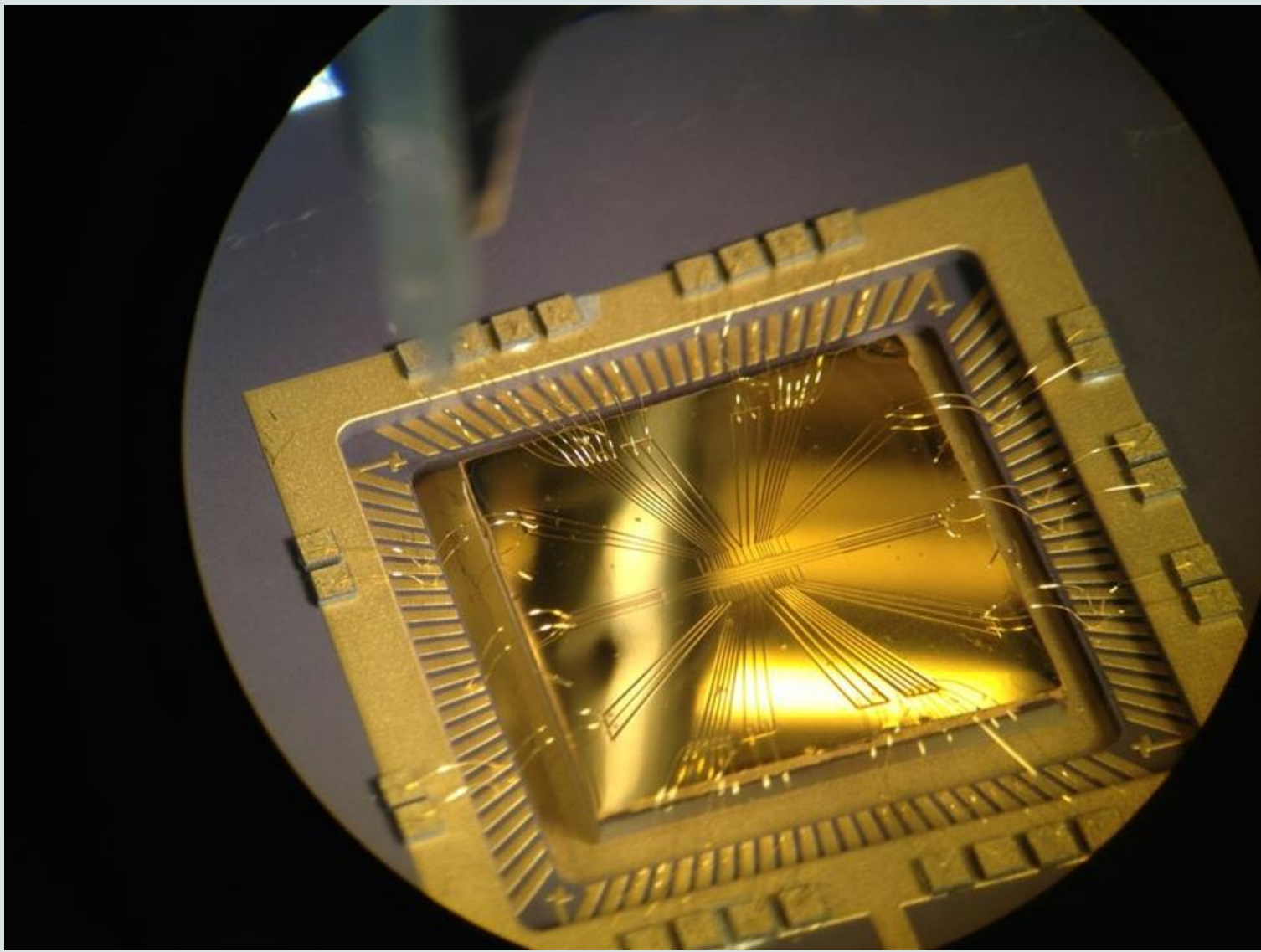






Large array of ion-trap system







Review

Quantum computational supremacy

Aram W. Harrow & Ashley Montanaro *Nature* **549**, 203–209 (14 September 2017)

doi:10.1038/nature23458

[Download Citation](#)[Computer science](#) [Quantum information](#)

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Abstract

The field of quantum algorithms aims to find ways to speed up the solution of computational problems by using a quantum computer. A key milestone in this field will be when a universal quantum computer performs a computational task that is beyond the capability of any classical computer, an event known as quantum supremacy. This would



Quantum Physics

A blueprint for demonstrating quantum supremacy with superconducting qubits

C. Neill, P. Roushan, K. Kechedzhi, S. Boixo, S. V. Isakov, V. Smelyanskiy, R. Barends, B. Burkett, Y. Chen, Z. Chen, B. Chiaro, A. Dunsworth, A. Fowler, B. Foxen, R. Graff, E. Jeffrey, J. Kelly, E. Lucero, A. Megrant, J. Mutus, M. Neeley, C. Quintana, D. Sank, A. Vainsencher, J. Wenner, T. C. White, H. Neven, J. M. Martinis

(Submitted on 19 Sep 2017)

Fundamental questions in chemistry and physics may never be answered due to the exponential complexity of the underlying quantum phenomena. A desire to overcome this challenge has sparked a new industry of quantum technologies with the promise that engineered quantum systems can address these hard problems. A key step towards demonstrating such a system will be performing a computation beyond the capabilities of any classical computer, achieving so-called quantum supremacy. Here, using 9 superconducting qubits, we demonstrate an immediate path towards quantum supremacy. By individually tuning the qubit parameters, we are able to generate thousands of unique Hamiltonian evolutions and probe the output probabilities. The measured probabilities obey a universal distribution, consistent with uniformly sampling the full Hilbert-space. As the number of qubits in the algorithm is varied, the system continues to explore the exponentially growing number of states. Combining these large datasets with techniques from machine learning allows us to construct a model which accurately predicts the measured probabilities. We demonstrate an application of these algorithms by systematically increasing the disorder and observing a transition from delocalized states to localized states. By extending these results to a system of 50 qubits, we hope to address scientific questions that are beyond the capabilities of any classical computer.

Subjects: **Quantum Physics (quant-ph)**

Cite as: **arXiv:1709.06678 [quant-ph]**

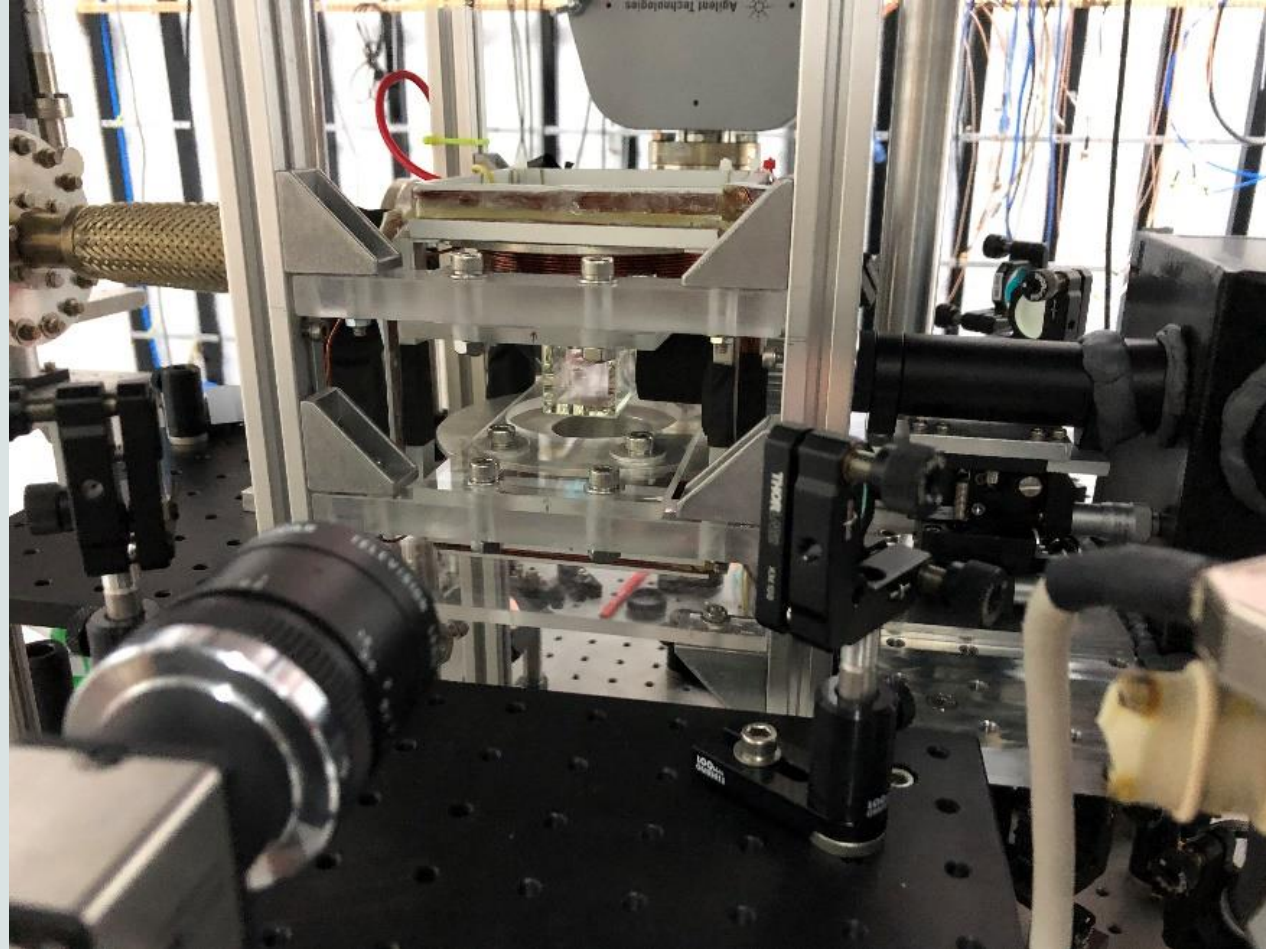
(or **arXiv:1709.06678v1 [quant-ph]** for this version)

IBM 50 qubits quantum computer





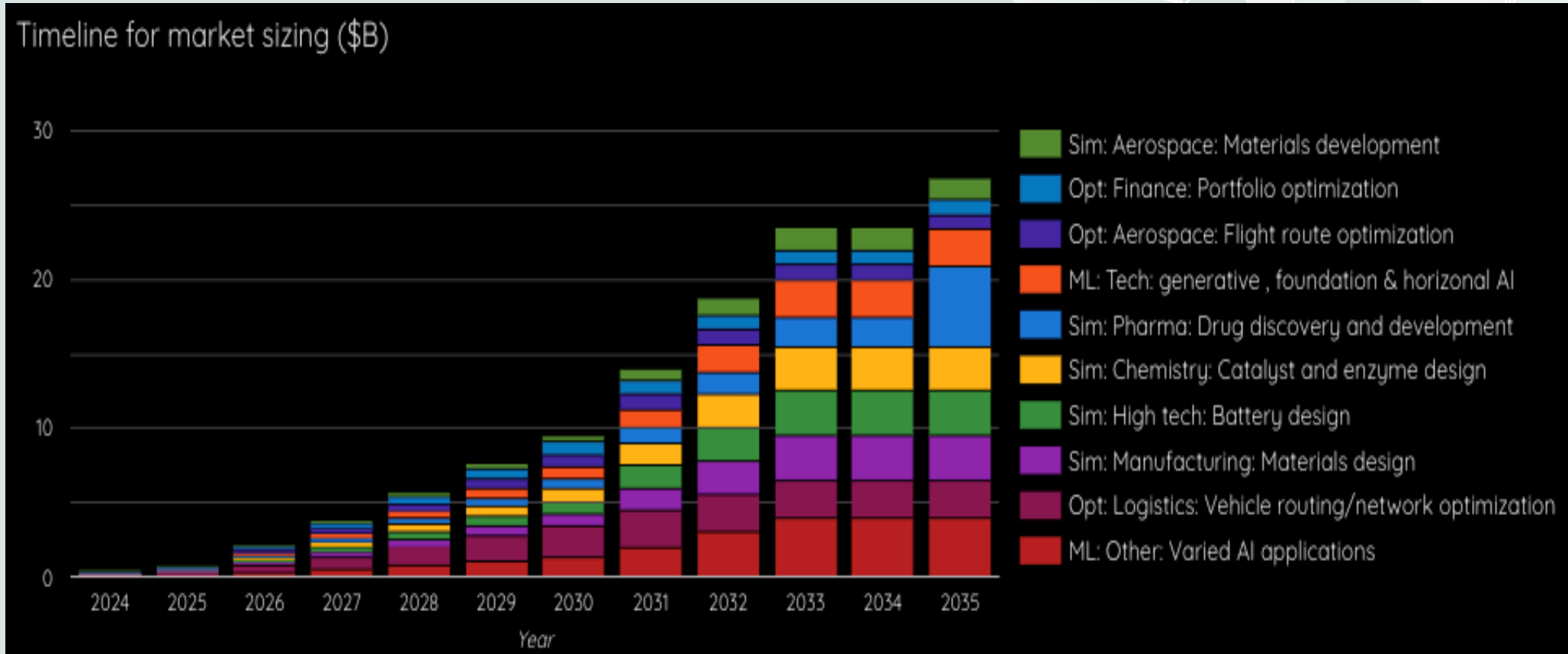
Cold Atom Trap experiment at Physics, Chiangmai University, 2020 (5 Rubidium atoms)



Experiment setup



Global View

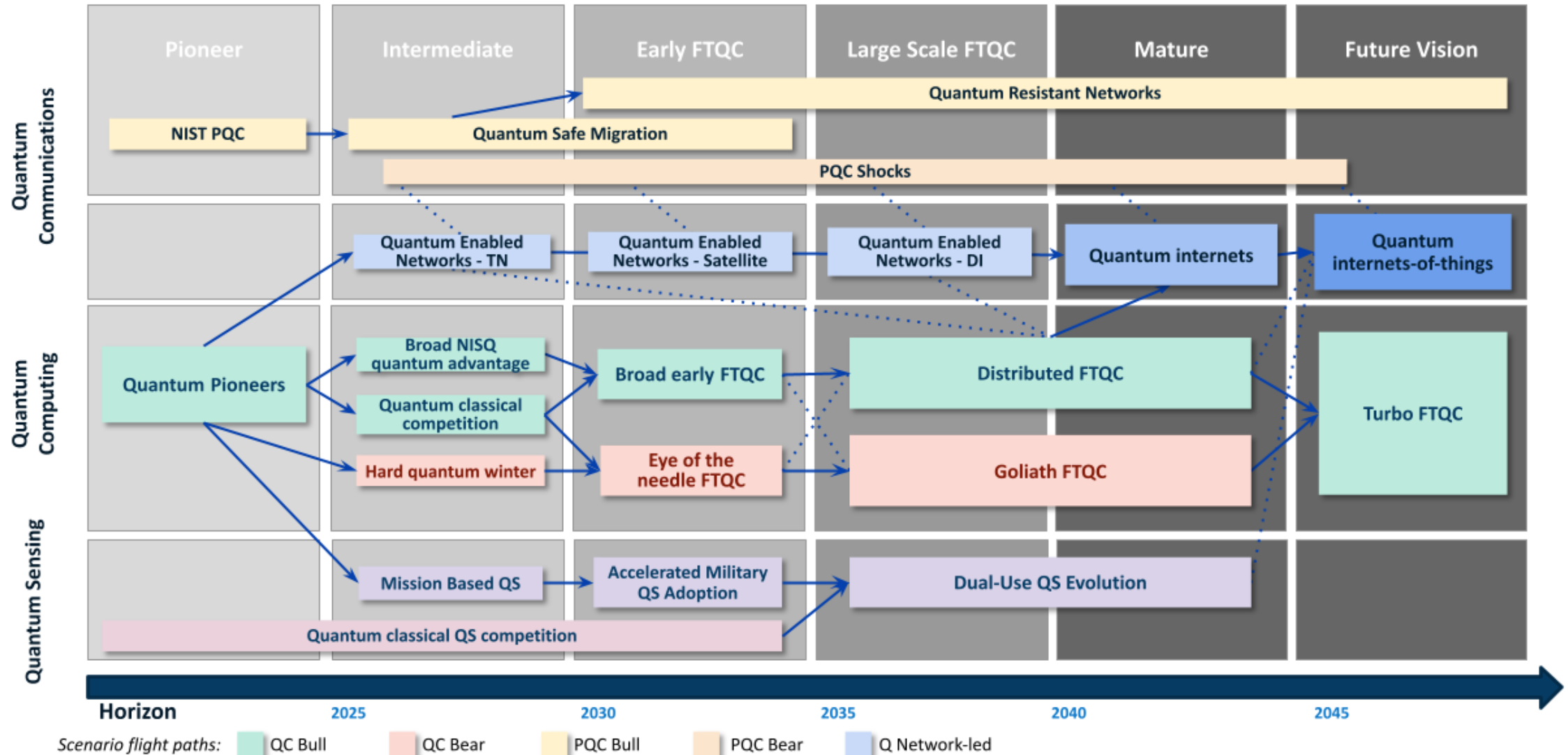


Quantum computing addressable market 2024-2035

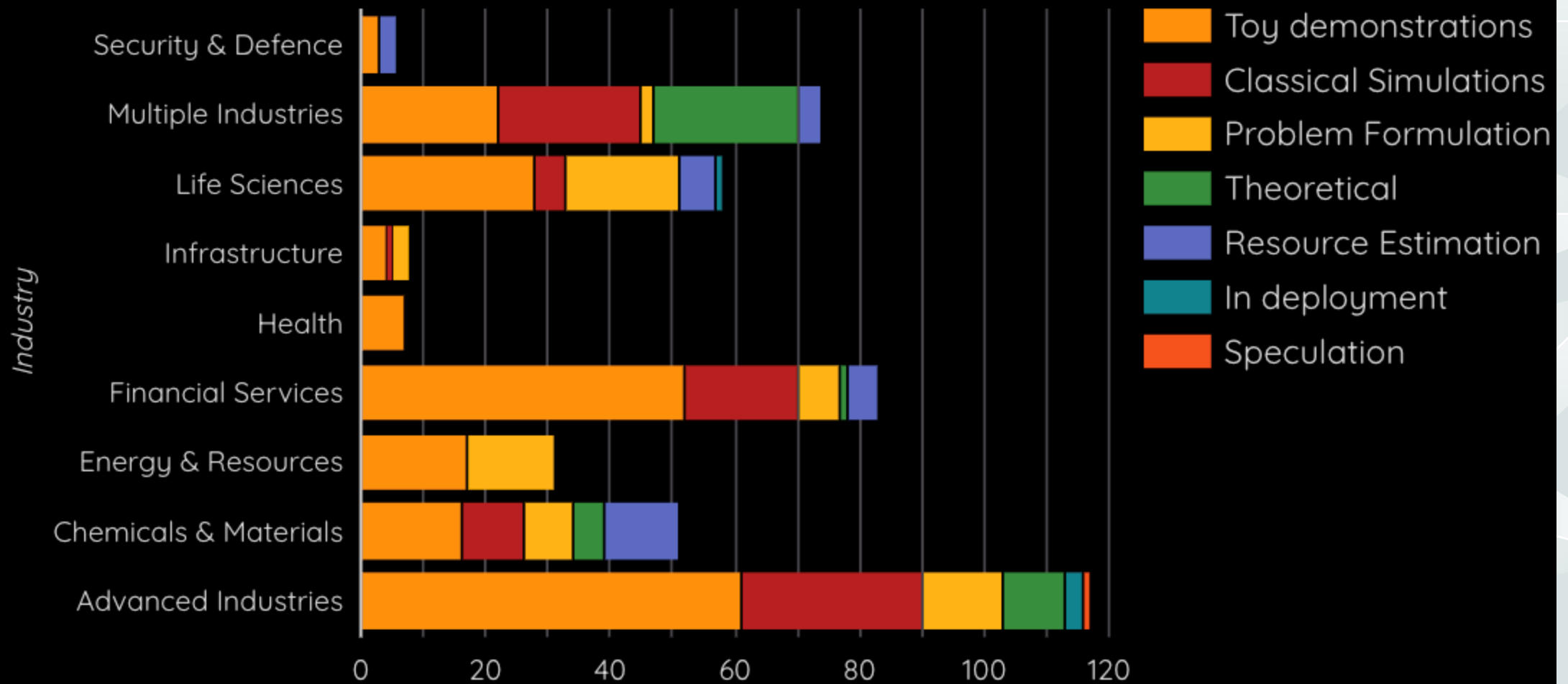
Source: Quantum computing report 2024

Quantum market scenarios across quantum eras

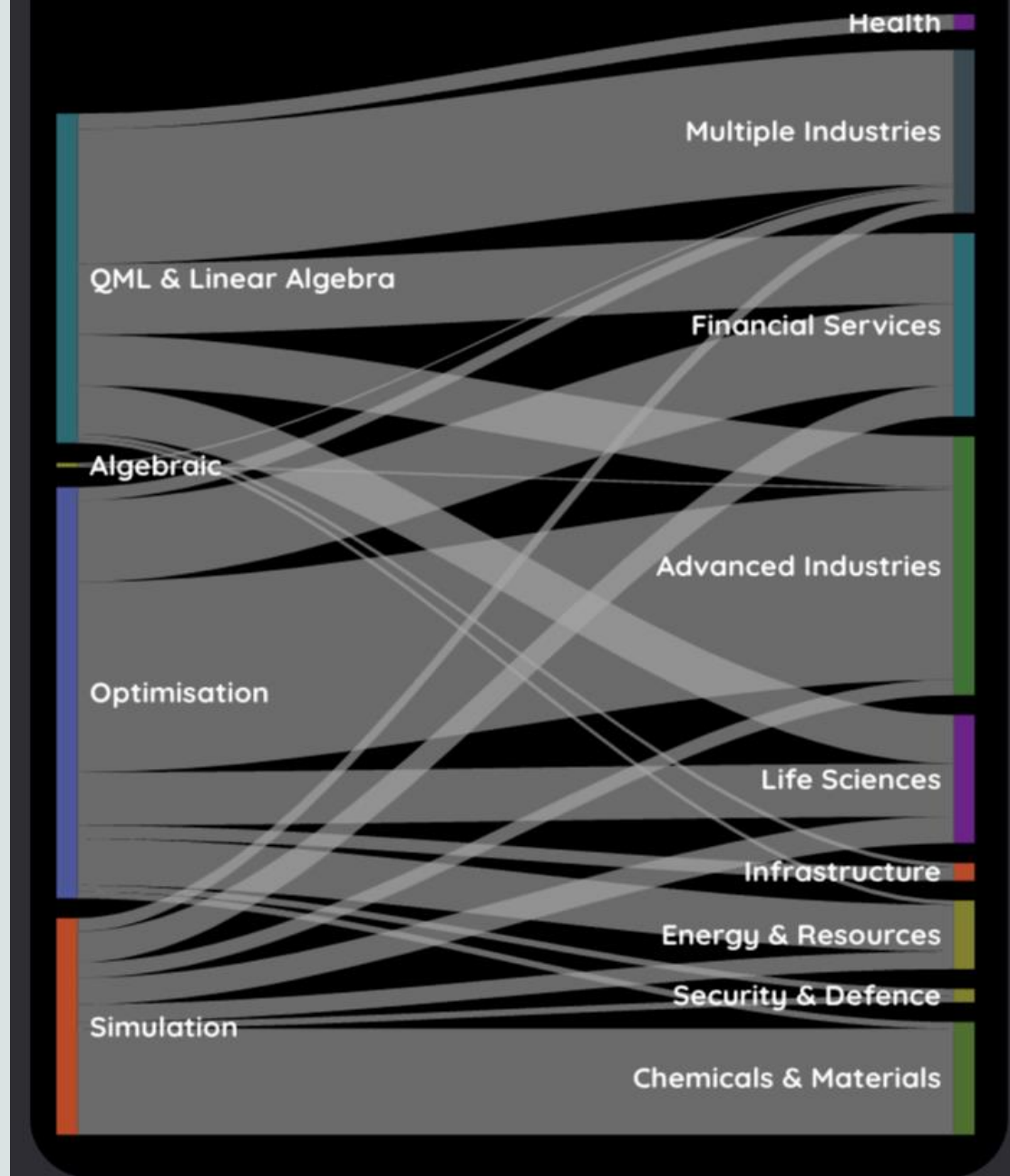
GQI QAM Interactive Model



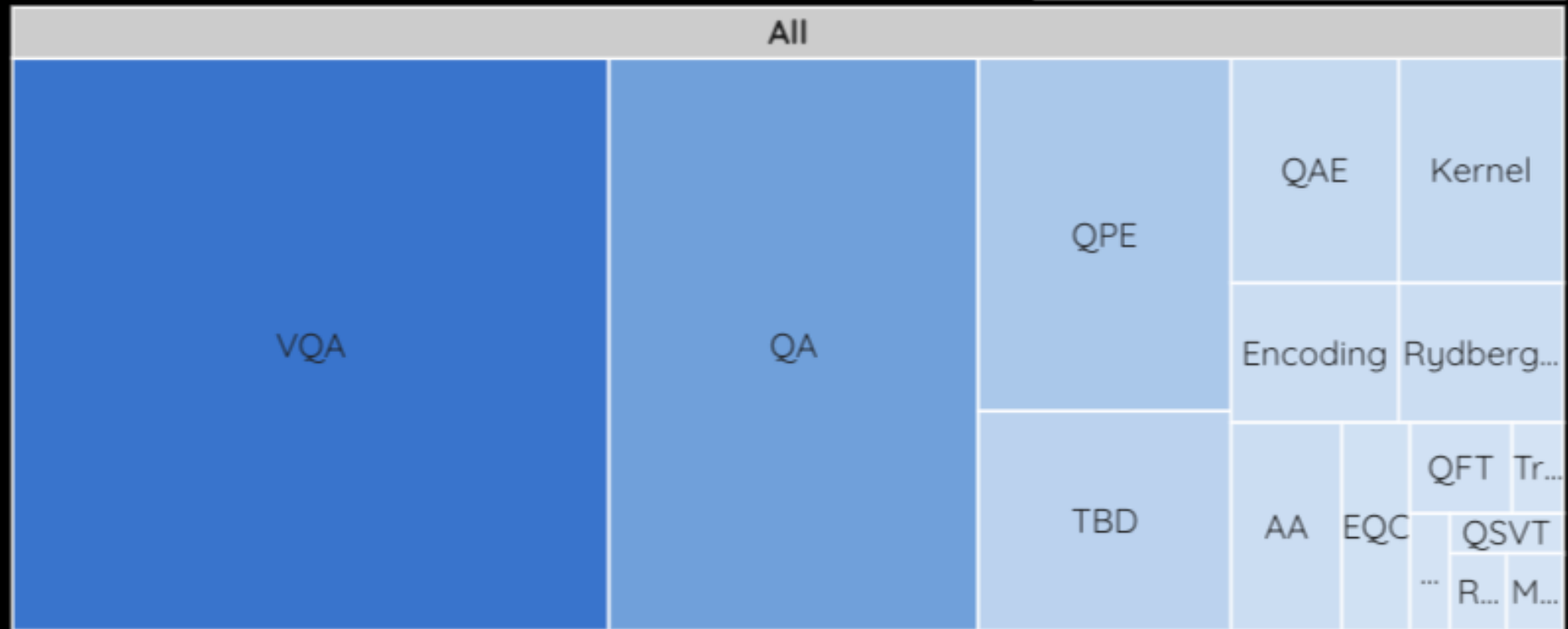
Use cases distribution by industry and implementation status



Problem domain to industry mapping



Distribution of used computational approaches

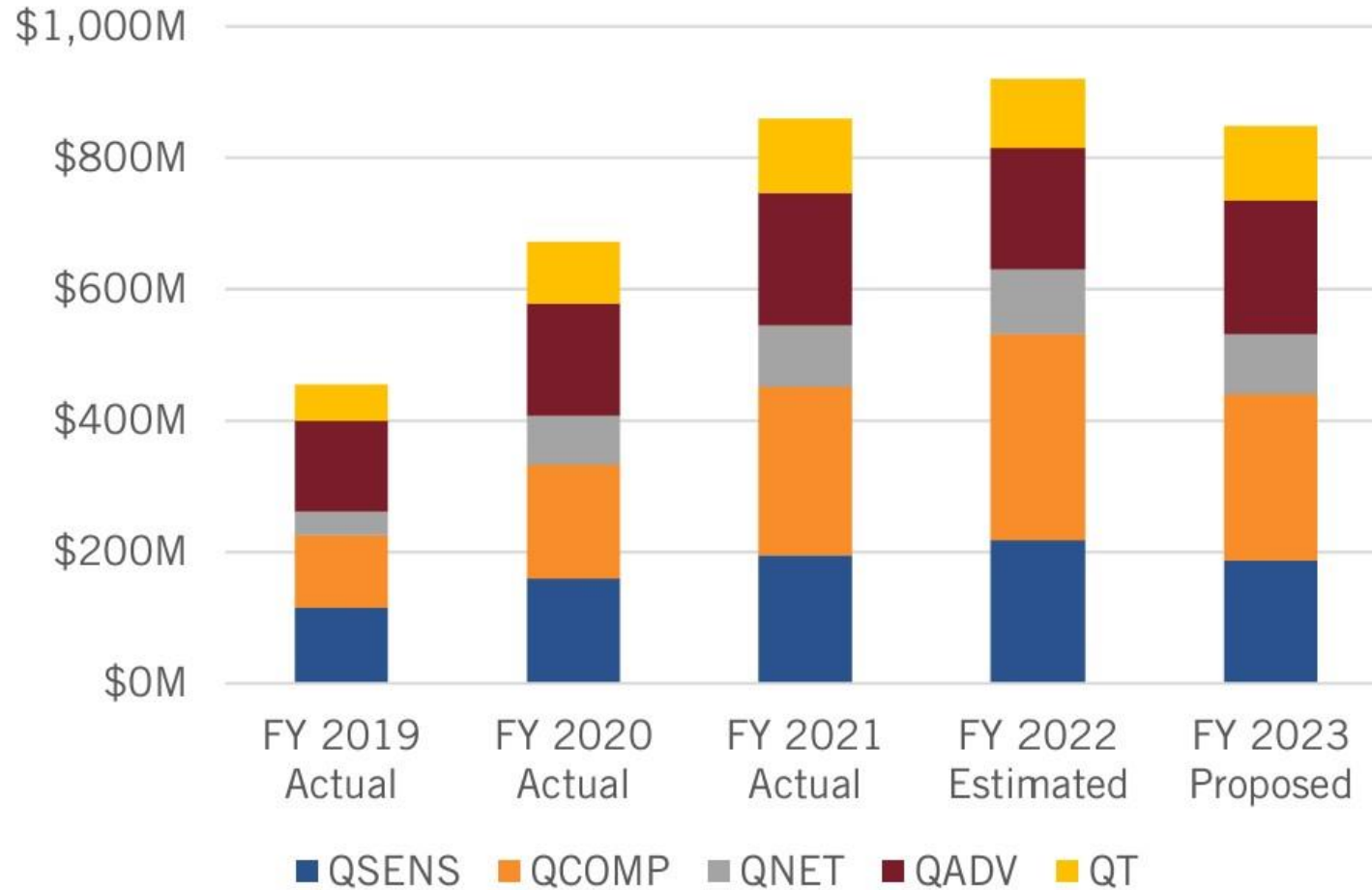


US Quantum Policy

Quantum Information Science (QIS)
encompass:

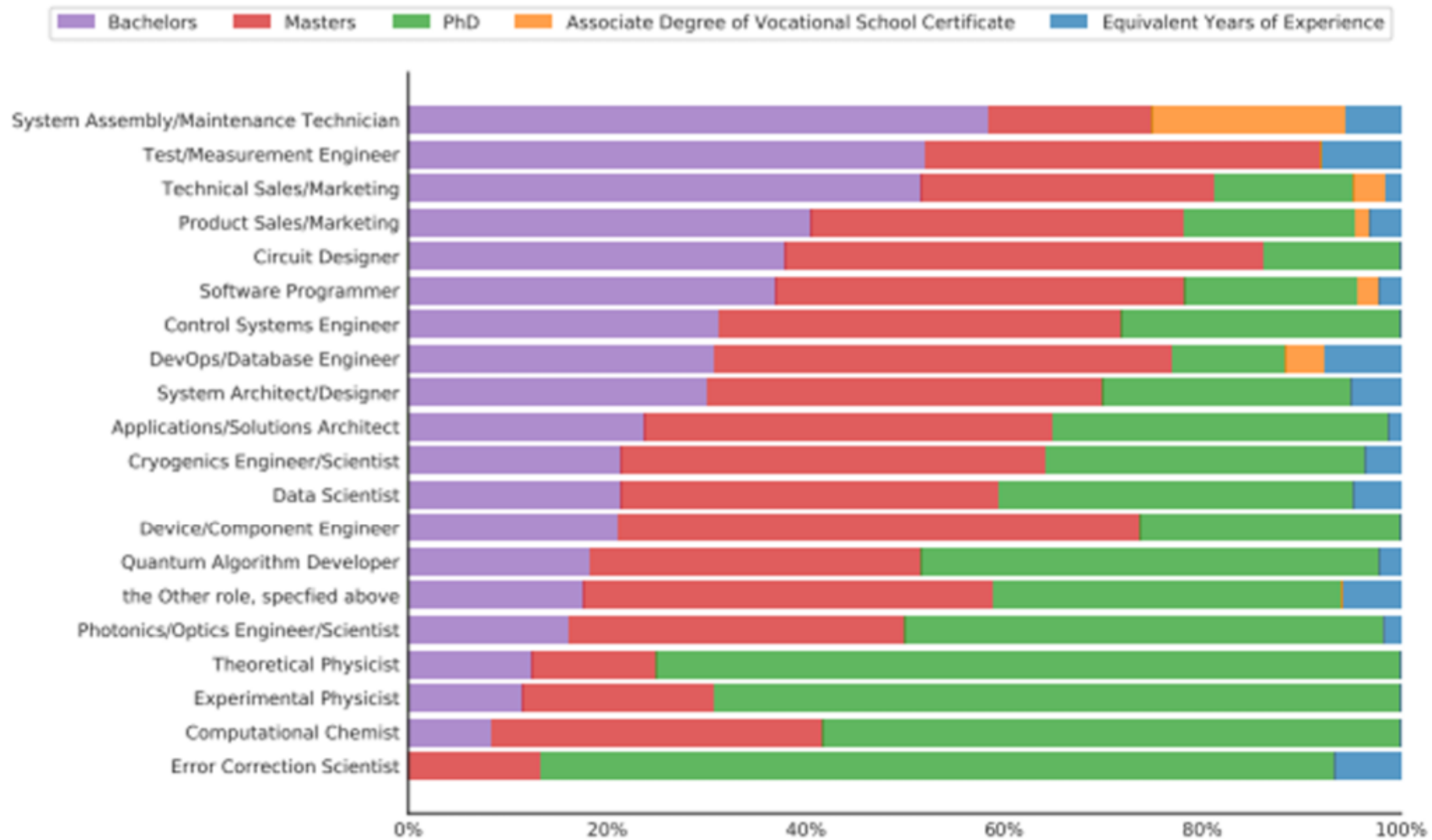
- Quantum sensing and metrology
- Quantum computing
- Quantum networking
- QIS for advancing fundamental science
- Quantum technology

Figure 2: U.S. Quantum Information Science R&D by program component area¹²



Source: The U.S. Approach to Quantum Policy By Hodan Omaar, October 10, 2023

Figure 5: The distribution of degrees needed for different job roles in the quantum industry⁵⁰



Source: The U.S. Approach to Quantum Policy By Hodan Omaar, October 10, 2023

Looking into the future

- Studying quantum computer give rise to new ideas
- Quantum “thinking” promises a new kind of method to solve really difficult problems
- Quantum Computers will change the face of computing forever

Contributors (student team 2019)





Thank you
