### Quantum algorithms: an overview

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Algorithm	Speedup	Example
Simulation of quantum systems	Exponential	Lloyd
Breaking cryptographic codes	Exponential	Shor
Optimisation / combinatorial search	Square-root	Grover
High-dimensional linear algebra	Exponential?	HHL
Quantum heuristics	Unknown	QAOA

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The Quantum Algorithm Zoo currently lists 404 papers on quantum algorithms.

### **Quantum simulation**

The most important early application of quantum computers is likely to be quantum simulation: modelling a quantum-mechanical system on a quantum computer.

Applications include quantum chemistry, superconductivity, metamaterials, high-energy physics, ... [Georgescu et al 1308.6253]

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Different variants of this task include:

- Analogue vs. digital simulation
- Static vs. dynamics simulation

#### Problem

Given a Hamiltonian H describing a physical system, find a Hamiltonian H' that encodes H, and allows physically meaningful (static or dynamic) information about H to be determined.

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- Even very simple quantum systems can be universal analogue quantum simulators [Cubitt, AM, Piddock, 1701.05182]
- Analogue quantum simulators with > 50 qubits have been implemented experimentally (e.g. [Zhang et al, 1708.01044])

### **Dynamics simulation**

Given a Hamiltonian *H* describing a physical system, and an initial state  $|\psi_0\rangle$  of that system, produce the state

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- No efficient classical algorithm is known for this task (in full generality), but efficient quantum algorithms exist for many physically reasonable cases.
- A topic of very active research (e.g. [Childs et al 1711.10980])

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- There is good evidence that producing the ground state is hard (QMA-complete) in the worst case, but it may be easy for physical systems of interest.
- One approach: optimise over quantum circuits using a variational algorithm [McClean et al 1509.04279].

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#### Theorem [Shor quant-ph/9508027]

There is a quantum algorithm which finds the prime factors of an *n*-digit integer in time  $O(n^3)$ .

## Shor's algorithm: complexity comparison

Very roughly (ignoring constant factors!):

Number of digits	Timesteps (quantum)	Timesteps (classical)
100	106	$\sim 4  imes 10^5$
1,000	$10^{9}$	$\sim 5  imes 10^{15}$
10,000	10 <sup>12</sup>	$\sim 1  imes 10^{41}$

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Based on these figures, a 10,000-digit number could be factorised by:

- A 1MHz clock speed quantum computer in 11 days.
- The fastest computer on the Top500 supercomputer list (~  $10^{17}$  operations per second) in ~  $3 \times 10^{16}$  years.

(see e.g. [Gidney+Ekerå 1905.09749] for a more detailed analysis, showing that a 2048-digit integer can be factorised in 8 hours with 23 million physical qubits)

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• On a classical computer, this task could require  $2^n$  queries to f in the worst case. But on a quantum computer, Grover's algorithm [Grover quant-ph/9605043] can solve the problem with  $O(\sqrt{2^n})$  queries to f (and bounded failure probability).

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• Grover's algorithm improves the runtime from  $O(2^n)$  to  $O(2^{n/2})$ : applications to design automation, circuit equivalence, model checking, ...

An important generalisation of Grover's algorithm is known as amplitude amplification.

Amplitude amplification [Brassard et al quant-ph/0005055]

Assume we are given access to a "checking" function f, and a probabilistic algorithm A such that

 $\Pr[\mathcal{A} \text{ outputs } w \text{ such that } f(w) = 1] = \epsilon.$ 

Then we can find *w* such that f(w) = 1 with  $O(1/\sqrt{\epsilon})$  uses of *f*.

Gives a quadratic speed-up over classical algorithms which are based on heuristics.

These primitives can be used to obtain many speedups over classical algorithms, e.g.:

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• ...

They can also speed up Monte Carlo methods [AM 1504.06987, Hamoudi+Magniez 1807.06456]:

• The mean of a random variable with variance  $\sigma^2$  can be approximated up to  $\epsilon$  in time roughly  $O(\sigma/\epsilon)$ , as opposed to the classical  $O(\sigma^2/\epsilon^2)$ .

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Backtracking algorithms solve CSPs by "trial and error": exploring a tree of partial solutions.

#### Theorem [AM 1509.02374] (informal)

If there is a classical backtracking algorithm which solves a CSP by exploring a tree of partial solutions of size *T*, there is a quantum algorithm that solves the CSP in time  $O(\sqrt{T} \operatorname{poly}(n))$ .

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Some applications:

- Quantum speedup of the Travelling Salesman Problem on bounded-degree graphs [Moylett, Linden and AM 1612.06203]
- Finding shortest vectors in lattices for cryptographic applications [Alkim et al. '15, del Pino et al. '16]
- Accelerating classical branch-and-bound algorithms for optimisation problems [AM 1906.10375]

# "Solving" linear equations

A basic task in mathematics and engineering:

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Theorem: If *A* has condition number  $\kappa$  (=  $||A^{-1}|| ||A||$ ),  $|x\rangle$  can be approximately produced in time poly(log *N*, *d*,  $\kappa$ ) [Harrow et al 0811.3171]

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- Some applications of this algorithm include:
  - Electromagnetic scattering cross-sections using the finite element method [Clader et al 1301.2340] [AM+Pallister 1512.05903]

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- Recommendation systems and other problems in machine learning (e.g. [Kerenidis+Prakash 1603.08675]) but note "quantum-inspired" competition [Tang 1807.04271]!

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Examples:

- The adiabatic algorithm / quantum annealing [Farhi et al quant-ph/0001106]
- The Quantum Approximate Optimisation Algorithm (QAOA) [Hogg+Portnov quant-ph/0006090, Farhi et al 1411.4028]

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Evidence that they outperform classical algorithms is mixed, but we at least know they are probably hard to simulate classically [Farhi+Harrow 1602.07674].

## Analysing real quantum algorithm complexity

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Some fully worked-out applications with large speedups (for quantum runtime  $\sim$  1 day) include:

- Nitrogen fixation [Reiher et al 1605.03590]
- Many-body localisation [Childs et al 1711.10980]
- Other problems in quantum chemistry and condensed-matter physics, e.g. [Babbush et al 1805.03662]
- Integer factorisation [Kutin quant-ph/0609001] [Gidney and Ekerå 1905.09749]

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In constraint satisfaction the speedups are smaller and quantum hardware requirements larger...

• Graph colouring / boolean satisfiability: speedup factor of  $\sim 10^5$  (ignoring cost of fault-tolerance processing) but  $\sim 10^{12}$  physical qubits required [Campbell et al 1810.05582]

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Further reading:

Quantum algorithms: an overview [AM, 1511.04206]

### Thanks!