# Quantum algorithms: an overview 

Ashley Montanaro<br>School of Mathematics, University of Bristol

14 November 2019
erc


Pic: Google

## Quantum computers

Quantum computers are designed to do things that classical computers cannot. But to achieve a quantum speedup requires a quantum algorithm.

## Quantum computers

Quantum computers are designed to do things that classical computers cannot. But to achieve a quantum speedup requires a quantum algorithm.

Most quantum algorithms can be divided into 5 categories:

| 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 |

## Quantum computers

Quantum computers are designed to do things that classical computers cannot. But to achieve a quantum speedup requires a quantum algorithm.

Most quantum algorithms can be divided into 5 categories:

| 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 |

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]

## 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]

Different variants of this task include:

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


## Analogue simulation

## Problem

Given a Hamiltonian $H$ describing a physical system, find a Hamiltonian $H^{\prime}$ that encodes $H$, and allows physically meaningful (static or dynamic) information about $H$ to be determined.

## Analogue simulation

## Problem

Given a Hamiltonian $H$ describing a physical system, find a Hamiltonian $H^{\prime}$ that encodes $H$, and allows physically meaningful (static or dynamic) information about $H$ to be determined.

- $H^{\prime}$ should be "easier" to prepare in the lab than $H$.


## Analogue simulation

## Problem

Given a Hamiltonian $H$ describing a physical system, find a Hamiltonian $H^{\prime}$ that encodes $H$, and allows physically meaningful (static or dynamic) information about $H$ to be determined.

- $H^{\prime}$ should be "easier" to prepare in the lab than $H$.
- Even very simple quantum systems can be universal analogue quantum simulators [Cubitt, AM, Piddock, 1701.05182]


## Analogue simulation

## Problem

Given a Hamiltonian $H$ describing a physical system, find a Hamiltonian $H^{\prime}$ that encodes $H$, and allows physically meaningful (static or dynamic) information about $H$ to be determined.

- $H^{\prime}$ should be "easier" to prepare in the lab than $H$.
- 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])


## Digital simulation

## Dynamics simulation

Given a Hamiltonian $H$ describing a physical system, and an initial state $\left|\psi_{0}\right\rangle$ of that system, produce the state

$$
\left|\psi_{t}\right\rangle=e^{-i H t}\left|\psi_{0}\right\rangle .
$$

Given such an output state, measurements can be performed to determine quantities of interest about the state.

## Digital simulation

## Dynamics simulation

Given a Hamiltonian $H$ describing a physical system, and an initial state $\left|\psi_{0}\right\rangle$ of that system, produce the state

$$
\left|\psi_{t}\right\rangle=e^{-i H t}\left|\psi_{0}\right\rangle .
$$

Given such an output state, measurements can be performed to determine quantities of interest about the state.

- 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])


## Digital simulation

## Static simulation (e.g.)

Given a Hamiltonian $H$ describing a physical system, produce the ground (lowest energy) state of $H$.

## Digital simulation

## Static simulation (e.g.)

Given a Hamiltonian $H$ describing a physical system, produce the ground (lowest energy) state of $H$.

- Given such a state, measurements can be performed to determine quantities of interest about the state.


## Digital simulation

## Static simulation (e.g.)

Given a Hamiltonian $H$ describing a physical system, produce the ground (lowest energy) state of $H$.

- Given such a state, measurements can be performed to determine quantities of interest about the state.
- 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.


## Digital simulation

## Static simulation (e.g.)

Given a Hamiltonian $H$ describing a physical system, produce the ground (lowest energy) state of $H$.

- Given such a state, measurements can be performed to determine quantities of interest about the state.
- 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].


## Integer factorisation

## Problem

Given an $n$-digit integer $N=p \times q$ for primes $p$ and $q$, determine $p$ and $q$.

## Integer factorisation

## Problem

Given an $n$-digit integer $N=p \times q$ for primes $p$ and $q$, determine $p$ and $q$.

- The best (classical!) algorithm we have for factorisation (the number field sieve) runs in time

$$
\exp \left(O\left(n^{1 / 3}(\log n)^{2 / 3}\right)\right)
$$

## Integer factorisation

## Problem

Given an $n$-digit integer $N=p \times q$ for primes $p$ and $q$, determine $p$ and $q$.

- The best (classical!) algorithm we have for factorisation (the number field sieve) runs in time

$$
\exp \left(O\left(n^{1 / 3}(\log n)^{2 / 3}\right)\right)
$$

- The RSA cryptosystem that underlies Internet security is based around the hardness of this task.
- That is, if we can factorise large integers efficiently, we can break RSA.


## Integer factorisation

## Problem

Given an $n$-digit integer $N=p \times q$ for primes $p$ and $q$, determine $p$ and $q$.

- The best (classical!) algorithm we have for factorisation (the number field sieve) runs in time

$$
\exp \left(O\left(n^{1 / 3}(\log n)^{2 / 3}\right)\right)
$$

- The RSA cryptosystem that underlies Internet security is based around the hardness of this task.
- That is, if we can factorise large integers efficiently, we can break RSA.


## Theorem [Shor quant-ph/9508027]

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

## Shor's algorithm: complexity comparison

Very roughly (ignoring constant factors!):

| Number of digits | Timesteps (quantum) | Timesteps (classical) |
| :---: | :---: | :---: |
| 100 | $10^{6}$ | $\sim 4 \times 10^{5}$ |
| 1,000 | $10^{9}$ | $\sim 5 \times 10^{15}$ |
| 10,000 | $10^{12}$ | $\sim 1 \times 10^{41}$ |

## Shor's algorithm: complexity comparison

Very roughly (ignoring constant factors!):

| Number of digits | Timesteps (quantum) | Timesteps (classical) |
| :---: | :---: | :---: |
| 100 | $10^{6}$ | $\sim 4 \times 10^{5}$ |
| 1,000 | $10^{9}$ | $\sim 5 \times 10^{15}$ |
| 10,000 | $10^{12}$ | $\sim 1 \times 10^{41}$ |

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

- A 1 MHz clock speed quantum computer in 11 days.


## Shor's algorithm: complexity comparison

Very roughly (ignoring constant factors!):

| Number of digits | Timesteps (quantum) | Timesteps (classical) |
| :---: | :---: | :---: |
| 100 | $10^{6}$ | $\sim 4 \times 10^{5}$ |
| 1,000 | $10^{9}$ | $\sim 5 \times 10^{15}$ |
| 10,000 | $10^{12}$ | $\sim 1 \times 10^{41}$ |

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

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


## Grover's algorithm

One of the most basic problems in computer science is unstructured search.

## Grover's algorithm

One of the most basic problems in computer science is unstructured search.

- Imagine we have access to a function $f:\{0,1\}^{n} \rightarrow\{0,1\}$ which we treat as a black box.


## Grover's algorithm

One of the most basic problems in computer science is unstructured search.

- Imagine we have access to a function $f:\{0,1\}^{n} \rightarrow\{0,1\}$ which we treat as a black box.
- We want to find an $x$ such that $f(x)=1$.



## Grover's algorithm

One of the most basic problems in computer science is unstructured search.

- Imagine we have access to a function $f:\{0,1\}^{n} \rightarrow\{0,1\}$ which we treat as a black box.
- We want to find an $x$ such that $f(x)=1$.

- 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 $O\left(\sqrt{2^{n}}\right)$ queries to $f$ (and bounded failure probability).


## Applications of Grover's algorithm

Grover's algorithm gives a speedup over naïve algorithms for any decision problem in the complexity class NP, i.e. where we can verify the solution efficiently.

## Applications of Grover's algorithm

Grover's algorithm gives a speedup over naïve algorithms for any decision problem in the complexity class NP, i.e. where we can verify the solution efficiently.

- For example, in the Circuit SAT problem we would like to find an input to a circuit on $n$ bits such that the output is 1 :



## Applications of Grover's algorithm

Grover's algorithm gives a speedup over naïve algorithms for any decision problem in the complexity class NP, i.e. where we can verify the solution efficiently.

- For example, in the Circuit SAT problem we would like to find an input to a circuit on $n$ bits such that the output is 1 :

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


## Applications of Grover's algorithm

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 $\mathcal{A}$ such that

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

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.

## Applications of Grover's algorithm

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

- Finding the minimum of $n$ numbers in $O(\sqrt{n})$ time [Dürr+Høyer quant-ph/9607014]


## Applications of Grover's algorithm

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

- Finding the minimum of $n$ numbers in $O(\sqrt{n})$ time [Dürr+Høyer quant-ph/9607014]
- Determining connectivity of an $n$-vertex graph in $O\left(n^{3 / 2}\right)$ time [Dürr et al quant-ph/0401091]


## Applications of Grover's algorithm

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

- Finding the minimum of $n$ numbers in $O(\sqrt{n})$ time [Dürr+Høyer quant-ph/9607014]
- Determining connectivity of an $n$-vertex graph in $O\left(n^{3 / 2}\right)$ time [Dürr et al quant-ph/0401091]
- Finding a collision in a 2-1 function $f:[n] \rightarrow[n]$ in $O\left(n^{1 / 3}\right)$ time [Brassard et al quant-ph/9705002]
- ...


## Applications of Grover's algorithm

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

- Finding the minimum of $n$ numbers in $O(\sqrt{n})$ time [Dürr+Høyer quant-ph/9607014]
- Determining connectivity of an $n$-vertex graph in $O\left(n^{3 / 2}\right)$ time [Dürr et al quant-ph/0401091]
- Finding a collision in a 2-1 function $f:[n] \rightarrow[n]$ in $O\left(n^{1 / 3}\right)$ time [Brassard et al quant-ph/9705002]
. . .

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 $O\left(\sigma^{2} / \epsilon^{2}\right)$.


## Quantum speedup of backtracking algorithms

Backtracking is a general approach to solve constraint satisfaction problems (CSPs).

## Quantum speedup of backtracking algorithms

Backtracking is a general approach to solve constraint satisfaction problems (CSPs).

- An instance of a CSP on $n$ variables $x_{1}, \ldots, x_{n}$ is specified by a sequence of constraints, all of which must be satisfied by the variables.


## Quantum speedup of backtracking algorithms

Backtracking is a general approach to solve constraint satisfaction problems (CSPs).

- An instance of a CSP on $n$ variables $x_{1}, \ldots, x_{n}$ is specified by a sequence of constraints, all of which must be satisfied by the variables.
- We might want to find one assignment to $x_{1}, \ldots, x_{n}$ that satisfies all the constraints, or list all such assignments.


## Quantum speedup of backtracking algorithms

Backtracking is a general approach to solve constraint satisfaction problems (CSPs).

- An instance of a CSP on $n$ variables $x_{1}, \ldots, x_{n}$ is specified by a sequence of constraints, all of which must be satisfied by the variables.
- We might want to find one assignment to $x_{1}, \ldots, x_{n}$ that satisfies all the constraints, or list all such assignments.
- A simple example: graph 3-colouring.



## Quantum speedup of backtracking algorithms

Backtracking is a general approach to solve constraint satisfaction problems (CSPs).

- An instance of a CSP on $n$ variables $x_{1}, \ldots, x_{n}$ is specified by a sequence of constraints, all of which must be satisfied by the variables.
- We might want to find one assignment to $x_{1}, \ldots, x_{n}$ that satisfies all the constraints, or list all such assignments.
- A simple example: graph 3-colouring.


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

## Quantum speedup of backtracking algorithms

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 $O(\sqrt{T}$ poly $(n))$.

## Quantum speedup of backtracking algorithms

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 $O(\sqrt{T}$ poly $(n))$.

This is a near-quadratic speedup, assuming that $T \gg \operatorname{poly}(n)$.

## Quantum speedup of backtracking algorithms

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 $O(\sqrt{T}$ poly $(n))$.

This is a near-quadratic speedup, assuming that $T \gg \operatorname{poly}(n)$. Backtracking is one of the most useful classical algorithmic techniques known in practice.

## Quantum speedup of backtracking algorithms

## 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 $O(\sqrt{T}$ poly $(n))$.

This is a near-quadratic speedup, assuming that $T \gg \operatorname{poly}(n)$. Backtracking is one of the most useful classical algorithmic techniques known in practice.
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:

## Solving linear equations

Given access to a $d$-sparse $N \times N$ matrix $A$, and $b \in \mathbb{R}^{N}$, output $x$ such that $A x=b$.

## "Solving" linear equations

A basic task in mathematics and engineering:

## Solving linear equations

Given access to a $d$-sparse $N \times N$ matrix $A$, and $b \in \mathbb{R}^{N}$, output $x$ such that $A x=b$.

One "quantum" way of thinking about the problem:

## "Solving" linear equations

Given the ability to produce the quantum state $|b\rangle=\sum_{i=1}^{N} b_{i}|i\rangle$, and access to $A$ as above, produce the state $|x\rangle=\sum_{i=1}^{N} x_{i}|i\rangle$.

## "Solving" linear equations

A basic task in mathematics and engineering:

## Solving linear equations

Given access to a $d$-sparse $N \times N$ matrix $A$, and $b \in \mathbb{R}^{N}$, output $x$ such that $A x=b$.

One "quantum" way of thinking about the problem:

## "Solving" linear equations

Given the ability to produce the quantum state $|b\rangle=\sum_{i=1}^{N} b_{i}|i\rangle$, and access to $A$ as above, produce the state $|x\rangle=\sum_{i=1}^{N} x_{i}|i\rangle$.

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

## Notes on this algorithm

The algorithm (approximately) produces a state $|x\rangle$ such that we can extract some information from $|x\rangle$. Is this useful?

## Notes on this algorithm

The algorithm (approximately) produces a state $|x\rangle$ such that we can extract some information from $|x\rangle$. Is this useful?

- We could use this to e.g. determine whether two sets of linear equations have (approximately) the same solution not clear how to do this classically.


## Notes on this algorithm

The algorithm (approximately) produces a state $|x\rangle$ such that we can extract some information from $|x\rangle$. Is this useful?

- We could use this to e.g. determine whether two sets of linear equations have (approximately) the same solution not clear how to do this classically.
- Achieving a similar runtime classically would imply that all quantum computations could be simulated!


## Notes on this algorithm

The algorithm (approximately) produces a state $|x\rangle$ such that we can extract some information from $|x\rangle$. Is this useful?

- We could use this to e.g. determine whether two sets of linear equations have (approximately) the same solution not clear how to do this classically.
- Achieving a similar runtime classically would imply that all quantum computations could be simulated!

Some applications of this algorithm include:

- Electromagnetic scattering cross-sections using the finite element method [Clader et al 1301.2340] [AM+Pallister 1512.05903]


## Notes on this algorithm

The algorithm (approximately) produces a state $|x\rangle$ such that we can extract some information from $|x\rangle$. Is this useful?

- We could use this to e.g. determine whether two sets of linear equations have (approximately) the same solution not clear how to do this classically.
- Achieving a similar runtime classically would imply that all quantum computations could be simulated!

Some applications of this algorithm include:

- Electromagnetic scattering cross-sections using the finite element method [Clader et al 1301.2340] [AM+Pallister 1512.05903]
- "Solving" differential equations [Leyton+Osborne 0812.4423] [Berry 1010.2745]


## Notes on this algorithm

The algorithm (approximately) produces a state $|x\rangle$ such that we can extract some information from $|x\rangle$. Is this useful?

- We could use this to e.g. determine whether two sets of linear equations have (approximately) the same solution not clear how to do this classically.
- Achieving a similar runtime classically would imply that all quantum computations could be simulated!

Some applications of this algorithm include:

- Electromagnetic scattering cross-sections using the finite element method [Clader et al 1301.2340] [AM+Pallister 1512.05903]
- "Solving" differential equations [Leyton+Osborne 0812.4423] [Berry 1010.2745]
- Recommendation systems and other problems in machine learning (e.g. [Kerenidis+Prakash 1603.08675]) - but note "quantum-inspired" competition [Tang 1807.04271]!


## Quantum heuristics

Some quantum optimisation algorithms might be more efficient than our best classical algorithms, but we can't prove this rigorously...

## Quantum heuristics

Some quantum optimisation algorithms might be more efficient than our best classical algorithms, but we can't prove this rigorously...

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]


## Quantum heuristics

Some quantum optimisation algorithms might be more efficient than our best classical algorithms, but we can't prove this rigorously...

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]

These algorithms try to find good solutions to hard combinatorial optimisation problems (e.g. MAX-CUT).

## Quantum heuristics

Some quantum optimisation algorithms might be more efficient than our best classical algorithms, but we can't prove this rigorously...

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]

These algorithms try to find good solutions to hard combinatorial optimisation problems (e.g. MAX-CUT).

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

## Analysing real quantum algorithm complexity

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]


## Analysing real quantum algorithm complexity

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]

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]


## Conclusions

There are many quantum algorithms, solving many different problems, some of which achieve substantial speedups over their classical counterparts.

## Conclusions

There are many quantum algorithms, solving many different problems, some of which achieve substantial speedups over their classical counterparts.

Important future research directions include:

- Finding more practical applications for these algorithms;
- Analysing their complexity in detail;
- New ideas for quantum algorithm design.


## Conclusions

There are many quantum algorithms, solving many different problems, some of which achieve substantial speedups over their classical counterparts.

Important future research directions include:

- Finding more practical applications for these algorithms;
- Analysing their complexity in detail;
- New ideas for quantum algorithm design.


## Further reading:

Quantum algorithms: an overview [AM, 1511.04206]

Thanks!

