## Quantum algorithms

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## Outline

I. Quantum circuits
II. Elementary quantum algorithms
III. The QFT and phase estimation
IV. Factoring
V. Quantum search
VI. Quantum walk

## Part I

## Quantum circuits

## Quantum circuit model

Quantum circuits are generalizations of Boolean circuits input transformation output (probabilistic)


## Classical bit

Classical bit (bit): $\mathbb{B}:=\{0,1\}$

- Basis state: either 0 or 1
- General state: a probability distribution $p=\left(p_{0}, p_{1}\right)$ on $\mathbb{B}$


## Classical register

Classical register: $\mathbb{B}^{n}:=\underbrace{\mathbb{B} \times \mathbb{B} \times \ldots \times \mathbb{B}}_{n}$

- Basis state: a binary string $x \in \mathbb{B}^{n}$
- General state: a probability distribution $p=\left(p_{x}: x \in \mathbb{B}^{n}\right)$ on $\mathbb{B}^{n}$ (written as a column vector)

Remark: Note that $p$ is a vector with positive entries that is normalized with respect to the $\ell_{1}$-norm (the sum of the absolute values of the entries)

## Classical transformation

Transformations on the classical register $\mathbb{B}$ are described by stochastic matrices

Stochastic matrices preserve the $\ell_{1}$-norm, i.e., probability distributions are mapped on probability distributions

Let $p$ be the state of the register. The state after the transformation $P$ is given by the matrix-vector-product

$$
P p
$$

## Qubit

Quantum bit (qubit): two-dimensional complex Hilbert space $\mathbb{C}^{2}$

- Computational basis states (classical states):

$$
|0\rangle:=\binom{1}{0} \quad \text { and } \quad|1\rangle:=\binom{0}{1}
$$

- General states: superpositions

$$
|\psi\rangle=\binom{\alpha_{0}}{\alpha_{1}}=\alpha_{0}|0\rangle+\alpha_{1}|1\rangle, \quad\left|\alpha_{0}\right|^{2}+\left|\alpha_{1}\right|^{2}=1
$$

the coefficients $\alpha_{0}, \alpha_{1} \in \mathbb{C}$ are called probability amplitudes

## Quantum register

Quantum register: $2^{n}$-dimensional complex Hilbert space $\left(\mathbb{C}^{2}\right)^{\otimes n}$ with tensor product structure

$$
\left(\mathbb{C}^{2}\right)^{\otimes n}:=\underbrace{\mathbb{C}^{2} \otimes \mathbb{C}^{2} \otimes \cdots \otimes \mathbb{C}^{2}}_{n}
$$

- Computational basis states (classical states):

$$
|x\rangle=\left|x_{1}\right\rangle \otimes\left|x_{2}\right\rangle \otimes \cdots \otimes\left|x_{n}\right\rangle, \quad x \in \mathbb{B}^{n}
$$

- General state:

$$
|\psi\rangle=\sum_{x \in \mathbb{B}^{n}} \alpha_{x}|x\rangle, \quad \sum_{x}\left|\alpha_{x}\right|^{2}=1
$$

Remark: Note that $|\psi\rangle$ is a column vector (ket) that is normalized with respect to the $\ell_{2}$-norm (Euclidean norm)

## Quantum transformations

Transformations on the quantum register $\mathcal{H}:=\left(\mathbb{C}^{2}\right)^{\otimes n}$ are described by unitary matrices $U \in \mathcal{U}(\mathcal{H})$

Unitary matrices preserve the $\ell_{2}$-norm
Let $|\psi\rangle \in \mathcal{H}$ be the state of the quantum register; the state after the transformation $U$ is given by the matrix-vector product

$$
U|\psi\rangle
$$

## Quantum circuit

Each transformations $U$ has to be implemented by a quantum circuit, i.e., a sequence of elementary gates

Quantum circuit model $=$ Quantum mechanics + Notion of complexity

## Single qubit gate on two qubits

single-qubit gate $U$ on first qubit


$$
\begin{array}{lll}
|0\rangle \otimes|0\rangle & \mapsto & (U|0\rangle) \otimes|0\rangle \\
|0\rangle \otimes|1\rangle & \mapsto & (U|0\rangle) \otimes|1\rangle \\
|1\rangle \otimes|0\rangle & \mapsto & (U|1\rangle) \otimes|0\rangle \\
|1\rangle \otimes|1\rangle & \mapsto & (U|1\rangle) \otimes|1\rangle
\end{array}
$$

corresponding matrix
$U \otimes I=\left(\begin{array}{c|c}u_{00} \cdot I & u_{01} \cdot I \\ \hline u_{10} \cdot l & u_{11} \cdot I\end{array}\right)=\left(\begin{array}{cccc}u_{00} & 0 & u_{01} & 0 \\ 0 & u_{00} & 0 & u_{01} \\ u_{10} & 0 & u_{11} & 0 \\ 0 & u_{10} & 0 & u_{11}\end{array}\right)$

## Single qubit gate on two qubits

single-qubit gate $U$ on second qubit

$$
-U
$$

$$
\begin{array}{lll}
|0\rangle \otimes|0\rangle & \mapsto & |0\rangle \otimes U|0\rangle \\
|0\rangle \otimes|1\rangle & \mapsto & |0\rangle \otimes U|1\rangle \\
|1\rangle \otimes|0\rangle & \mapsto & |1\rangle \otimes U|0\rangle \\
|1\rangle \otimes|1\rangle & \mapsto & |1\rangle \otimes U|1\rangle
\end{array}
$$

corresponding matrix
$I \otimes U=\left(\begin{array}{c|c}1 \cdot U & 0 \cdot U \\ \hline 0 \cdot U & 1 \cdot U\end{array}\right)=\left(\begin{array}{cccc}u_{00} & u_{01} & 0 & 0 \\ u_{10} & u_{11} & 0 & 0 \\ 0 & 0 & u_{00} & u_{01} \\ 0 & 0 & u_{10} & u_{11}\end{array}\right)$

## Controlled-NOT gate

control: first qubit; target: second qubit

action on basis states of $\mathbb{C}^{2} \otimes \mathbb{C}^{2}$

$$
|c\rangle \otimes|t\rangle \mapsto|c\rangle \otimes|c \oplus t\rangle
$$

corresponding matrix

$$
\left(\begin{array}{ll|ll}
1 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 \\
\hline 0 & 0 & 0 & 1 \\
0 & 0 & 1 & 0
\end{array}\right)=|0\rangle\langle 0| \otimes I_{2}+|1\rangle\langle 1| \otimes X
$$

where $I_{2}=|0\rangle\langle 0|+|1\rangle\langle 1|$ and $X=|0\rangle\langle 1|+|1\rangle\langle 0|$

## Controlled U gate

control: qubit; target: m-qubit register

let $U$ be a unitary acting on the $m$-qubit register
action on basis states of $\mathbb{C}^{2} \otimes\left(\mathbb{C}^{2}\right)^{\otimes m}$

$$
|c\rangle \otimes|t\rangle \mapsto|c\rangle \otimes U^{c}|t\rangle \quad \text { where } b \in \mathbb{B}, t \in \mathbb{B}^{m}
$$

corresponding matrix

$$
\left(\begin{array}{c|c}
I & 0 \\
\hline 0 & U
\end{array}\right)=|0\rangle\langle 0| \otimes I+|1\rangle\langle 1| \otimes U
$$

## Toffoli gate

control: first and second qubits; target: third qubit

action on basis states of $\mathbb{C}^{2} \otimes \mathbb{C}^{2} \otimes \mathbb{C}^{2}$

$$
\left|c_{1}\right\rangle \otimes\left|c_{2}\right\rangle \otimes|t\rangle \mapsto\left|c_{1}\right\rangle \otimes\left|c_{2}\right\rangle \otimes\left|\left(c_{1} \wedge c_{2}\right) \oplus t\right\rangle
$$

corresponding matrix

$$
\left(\begin{array}{ll|ll|ll|ll}
1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
\hline 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\
\hline 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\
0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\
\hline 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0 & 0 & 1 & 0
\end{array}\right)=\left(I_{4}-|11\rangle\langle 11|\right) \otimes I_{2}+|11\rangle\langle 11| \otimes X
$$

## Simulating irreversible gates with Toffoli gate

The classical AND gate is irreversible because if the output is 0 then we cannot determine which of the three possible pairs was the actual input

| $x_{1}$ | $x_{2}$ | $x_{1} \wedge x_{2}$ |
| :---: | :---: | :---: |
| 0 | 0 | 0 |
| 0 | 1 | 0 |
| 1 | 0 | 0 |
| 1 | 1 | 1 |

But it is easy to simulate the AND gate with one Toffoli gate


## Problem of garbage

To simulate irreversible circuits with Toffoli gates, we keep the input and intermediary results to make everything reversible

Consider the function $y=x_{1} \wedge x_{2} \wedge x_{3}$


It is important to not leave any garbage; otherwise, we could not make use of quantum parallelism and constructive interference effects

## Reversible garbage removal

It is always possible to reversibly remove (uncompute) the garbage
In the case $y=x_{1} \wedge x_{2} \wedge x_{3}$, this can be done with the circuit


## Simulating irreversible circuits with Toffoli gates

Let $f:\{0,1\}^{n} \rightarrow\{0,1\}$ be any boolean function
Assume this function can be computed classically using only $t$ classical elementary gates such as AND, OR, NAND

We can implement a unitary $U_{f}$ on $\left(\mathbb{C}^{2}\right)^{\otimes n} \otimes \mathbb{C}^{2} \otimes\left(\mathbb{C}^{2}\right)^{\otimes w}$ such that

$$
U_{f}\left(|x\rangle_{\text {in }} \otimes|y\rangle_{\text {out }} \otimes|0\rangle_{\text {work }}^{\otimes w}\right)=|x\rangle \otimes|y \oplus f(x)\rangle \otimes|0\rangle^{\otimes w}
$$

$U_{f}$ is built from polynomially many in $t$ Toffoli gates and the size $w$ of the workspace register is polynomial in $t$

During the computation the qubits of the workspace register are changed, but at the end they reversibly reset to $|0\rangle^{\otimes \omega}$

## Universal gate set - exact implementation

Each unitary $U \in \mathcal{U}(\mathcal{H})$ can be implemented exactly by quantum circuits using only:

- CNOT gates (acting on adjacent qubits)
- arbitrary single qubit gates


## Gate complexity of unitaries - exact implementation

The gate complexity $\kappa(U)$ of a unitary $U \in \mathcal{U}(\mathcal{H})$ is minimal number of elementary gates needed to implement $U$

For example, quantum Fourier Transform has complexity $O\left(n^{2}\right)$
$\Longrightarrow$ Shor's factorization algorithm

## Universal gate set - approximate implementation

For each $\epsilon \in(0,1)$ and each unitary $U \in \mathcal{U}(\mathcal{H})$, there is a unitary $V$ such that

$$
\|U-V\| \leq \epsilon \quad \text { where } \quad\|U-V\|=\sup _{|\psi\rangle} \|(U-V)|\psi\rangle \|
$$

and $V$ is implemented by quantum circuits using only:

- CNOT gates (acting on adjacent qubits)
- the single qubit gates

$$
H=\frac{1}{\sqrt{2}}\left(\begin{array}{cc}
1 & 1 \\
1 & -1
\end{array}\right) \quad R(\theta)=\left(\begin{array}{cc}
1 & 0 \\
0 & e^{i \theta}
\end{array}\right), \quad \text { with } \theta=\frac{\pi}{4}
$$

There are other universal gate sets

## Gate complexity of unitaries - approximate implementation

The gate complexity $\kappa_{\epsilon}(U)$ of a unitary $U$ is the minimal number of gates (from a universal gate set) need to implement a unitary $V$ with $\|U-V\| \leq \epsilon$

The Solovay-Kitaev theorem implies that

$$
\kappa_{\epsilon}(U)=O\left(\kappa(U) \cdot \log ^{c}(\kappa(U) / \epsilon)\right)
$$

for some small constant $c$

Counting arguments show that most $n$-qubit unitaries have gate complexity exponential in $n$.

## Quantum measurement

A general measurement is described by a collection $P_{0, \ldots}, P_{m-1}$ of orthogonal projectors such that

$$
\sum_{i=0}^{m-1} P_{i}=I_{\mathcal{H}} \quad \text { where } \mathcal{H} \text { denotes the identity on } \mathcal{H}
$$

Let $|\psi\rangle$ be the state of the quantum register. The probability of obtaining the outcome $i$ is given by

$$
\operatorname{Pr}(i)=\| P_{i}|\psi\rangle \|^{2}
$$

The post-measurement state (collapse of the wavefunction) is

$$
\frac{P_{i}|\psi\rangle}{\| P_{i}|\psi\rangle \|}
$$

## Elementary quantum measurements

A measurement has to be realized by first applying a suitable quantum circuit followed by an elementary measurement

An elementary measurement on the $n$-qubit quantum register $\mathcal{H}$ consists of measuring the first (w.l.o.g.) $m$ qubits ( $m \leq n$ ) with respect to the computational basis

The $2^{m}$ orthogonal projectors $P_{b}$ are labeled by $m$-bit strings $b \in \mathbb{B}^{m}$ and are defined by

$$
P_{b}=\left|b_{1}\right\rangle\left\langle b_{1}\right| \otimes\left|b_{2}\right\rangle\left\langle b_{2}\right| \otimes \cdots \otimes\left|b_{m}\right\rangle\left\langle b_{m}\right| \otimes I_{2^{n-m}}
$$

The probability of obtaining outcome $b$ is given by

$$
\operatorname{Pr}(b)=\| P_{b}|\psi\rangle \|^{2}=\sum_{x_{m+1}, \ldots, x_{n} \in \mathbb{B}}\left|\alpha_{b_{1}, \ldots, b_{m}, x_{m+1}, \ldots, x_{n}}\right|^{2}
$$

## Structure of quantum algorithms

A quantum algorithm consists of

- preparing the initial state $|x\rangle$ with $x \in \mathbb{B}^{n}$,
- applying a quantum circuit of polynomially many in $n$ gates from some universal gate set, and
- performing an elementary measurement

These steps are repeated polynomially many times to collect enough samples and followed by classical post-processing $\Longrightarrow$ solution of the problem

## Hadamard test



The probabilities of obtaining the outcomes 0 and 1 are:

$$
\operatorname{Pr}(0)=\frac{1}{2}(1+\operatorname{Re}\langle\psi| U|\psi\rangle) \quad \operatorname{Pr}(1)=\frac{1}{2}(1-\operatorname{Re}\langle\psi| U|\psi\rangle)
$$

## Hadamard test



$$
\begin{aligned}
& |0\rangle \otimes|\psi\rangle \\
\mapsto & \frac{1}{\sqrt{2}}(|0\rangle+|1\rangle) \otimes|\psi\rangle \\
= & \frac{1}{\sqrt{2}}|0\rangle \otimes|\psi\rangle+\frac{1}{\sqrt{2}}|1\rangle \otimes|\psi\rangle \\
\mapsto & \frac{1}{\sqrt{2}}|0\rangle \otimes|\psi\rangle+\frac{1}{\sqrt{2}}|1\rangle \otimes U|\psi\rangle \\
\mapsto & \frac{1}{2}(|0\rangle+|1\rangle) \otimes|\psi\rangle+\frac{1}{2}(|0\rangle-|1\rangle) \otimes U|\psi\rangle \\
= & \frac{1}{2}|0\rangle \otimes(|\psi\rangle+U|\psi\rangle)+\frac{1}{2}|1\rangle \otimes(|\psi\rangle-U|\psi\rangle) \\
= & |\Phi\rangle
\end{aligned}
$$

## Hadamard test

$$
\begin{aligned}
& \operatorname{Pr}(0) \\
= & \| P|\Phi\rangle \|^{2} \quad \text { with } \quad P=|0\rangle\langle 0| \otimes I \\
= & \|(|0\rangle\langle 0| \otimes I)\left(\frac{1}{2}|0\rangle \otimes(|\psi\rangle+U|\psi\rangle)+\frac{1}{2}|1\rangle \otimes(|\psi\rangle-U|\psi\rangle)\right) \|^{2} \\
= & \| \frac{1}{2}|0\rangle \otimes(|\psi\rangle+U|\psi\rangle) \|^{2} \\
= & \frac{1}{4} \||0\rangle\left\|^{2} \cdot\right\||\psi\rangle+U|\psi\rangle \|^{2} \\
= & \frac{1}{4}\left(\langle\psi|+\langle\psi| U^{\dagger}\right)(|\psi\rangle+U|\psi\rangle) \\
= & \frac{1}{4}\left(\langle\psi \mid \psi\rangle+\langle\psi| U|\psi\rangle+\langle\psi| U^{\dagger}|\psi\rangle+\langle\psi| U^{\dagger} U|\psi\rangle\right) \\
= & \frac{1}{4}(2+\langle\psi| U|\psi\rangle+\overline{\langle\psi| U|\psi\rangle)} \\
= & \frac{1}{2}(1+\operatorname{Re}\langle\psi| U|\psi\rangle)
\end{aligned}
$$

## Hadamard test - Figure it out yourself

How can you estimate the imaginary part of $\langle\psi| U|\psi\rangle$ ?
Hint: Add a simple gate on the control register before the measurement.

## SWAP test - Figure it out yourself

Let $S$ denote the swap gate acting on two qubits

$$
S=|00\rangle\langle 00|+|01\rangle\langle 10|+|10\rangle\langle 01|+|11\rangle\langle 11|
$$

Determine the matrix representation of $S$ with respect to the computational basis

Consider the Hadamard test where the controlled operation is

$$
|0\rangle\langle 0| \otimes I_{4}+|1\rangle\langle 1| \otimes S
$$

and the state of the target register $|\psi\rangle=\left|\psi_{1}\right\rangle \otimes\left|\psi_{2}\right\rangle$
Determine the probability of obtaining 0 and 1 for the cases:

- arbitrary $\left|\psi_{1}\right\rangle$ and $\left|\psi_{2}\right\rangle$,
- $\left\langle\psi_{1} \mid \psi_{2}\right\rangle=0$ (orthogonal), and
- $\left\langle\psi_{1} \mid \psi_{2}\right\rangle=1$ (the same).


## Part II

## Elementary quantum algorithms

## Black box problems

Standard computational problem: determine a property of some input data

- Example: Find the prime factors of $N$

Alternate model: Input is provided by a black box (or oracle)

- Query: On input $x$, black box returns $f(x)$
- Determine a property of $f$ using as few queries as possible
- The minimum number of queries is the query complexity
- Example: Given a black box for $f:\{1,2, \ldots, N\} \rightarrow\{0,1\}$, is there some $x$ such that $f(x)=1$ ?
- Why black boxes?
- Facilitates proving lower bounds
- Can lead to algorithms for standard problems


## Black boxes for reversible/quantum computing

Black box $\quad x-f-f(x)$ is not reversible
Reversible version:


Given a circuit that computes $f$ non-reversibly, we can implement the reversible version with little overhead

Quantum version:

$$
\begin{aligned}
& |x\rangle-\sqrt{|x\rangle-}|x\rangle \\
& |z\rangle-\quad|z \oplus f(x)\rangle
\end{aligned}
$$

A reversible circuit is a quantum circuit

## Deutsch's problem

## Problem

- Given: a black-box function $f:\{0,1\} \rightarrow\{0,1\}$
- Task: determine whether $f$ is constant or balanced


How many queries are needed?

- Classically: 2 queries are necessary and sufficient
- Quantumly: ?


## Toward a quantum algorithm for Deutsch's problem

Quantum black box for $f: \begin{aligned} & |x\rangle-\square-|x\rangle \\ & |z\rangle-f\end{aligned}$

Compute $f$ in superposition:


$$
\begin{aligned}
|0\rangle \otimes|0\rangle & \mapsto \frac{|0\rangle+|1\rangle}{\sqrt{2}} \otimes|0\rangle \\
& \mapsto \frac{1}{\sqrt{2}}(|0\rangle \otimes|f(0)\rangle+|1\rangle \otimes|f(1)\rangle)
\end{aligned}
$$

Can't extract more than one bit of information about $f$

## Phase kickback

$\begin{aligned} \text { Quantum black box for } f: & |x\rangle-\square-|x\rangle \\ & |z\rangle-f-|z \oplus f(x)\rangle\end{aligned}$

Phase kickback:

$$
\begin{array}{r}
|x\rangle \\
\frac{|0\rangle-|1\rangle}{\sqrt{2}}
\end{array}-f-\frac{(-1)^{f(x)}|x\rangle}{\sqrt{2}}
$$

$$
|x\rangle \otimes \frac{1}{\sqrt{2}}(|0\rangle-|1\rangle)=\frac{1}{\sqrt{2}}(|x\rangle \otimes|0\rangle-|x\rangle \otimes|1\rangle)
$$

$$
\mapsto \frac{1}{\sqrt{2}}(|x\rangle \otimes|f(x)\rangle-|x\rangle \otimes|1 \oplus f(x)\rangle)
$$

$$
=|x\rangle \otimes \frac{1}{\sqrt{2}}(|f(x)\rangle-|\overline{f(x)}\rangle)
$$

$$
=\underbrace{(-1)^{f(x)}}_{\text {not necessarily global }}|x\rangle \otimes \frac{1}{\sqrt{2}}(|0\rangle-|1\rangle)
$$

not necessarily global

## Quantum algorithm for Deutsch's problem

$$
\begin{aligned}
|0\rangle & -H=f(0) \oplus f(1) \\
|0\rangle \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} & \\
& \mapsto \frac{(-1)^{f(0)}|0\rangle+(-1)^{f(1)}|1\rangle}{\sqrt{2}} \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} \\
& \mapsto(-1)^{f(0)} \frac{|0\rangle+(-1)^{f(0) \oplus f(1)}|1\rangle}{\sqrt{2}} \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} \\
& \mapsto(-1)^{f(0)}|f(0) \oplus f(1)\rangle \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}}
\end{aligned}
$$

1 quantum query vs. 2 classical queries!

## The Deutsch-Jozsa problem

## Problem

- Given: a black-box function $f:\{0,1\}^{n} \rightarrow\{0,1\}$
- Promise: $f$ is either constant $\quad(f(x)$ is independent of $x)$ or balanced $(f(x)=0$ for exactly half the values of $x)$
- Task: determine whether $f$ is constant or balanced

How many queries are needed?

- Classically: $2^{n} / 2+1$ queries to answer with certainty
- Quantumly: ?


## Phase kickback for a Boolean function of $n$ bits

Black box function: $\left|x_{1}\right\rangle$

Phase kickback:

$$
\left|x_{1}\right\rangle \otimes \cdots \otimes\left|x_{n}\right\rangle \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} \mapsto(-1)^{f(x)}\left|x_{1}\right\rangle \otimes \cdots \otimes\left|x_{n}\right\rangle \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}}
$$

Quantum algorithm for the Deutsch-Jozsa problem


$$
\begin{aligned}
|0\rangle^{\otimes n} \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} & \mapsto\left(\frac{|0\rangle+|1\rangle}{\sqrt{2}}\right)^{\otimes n} \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} \\
& =\frac{1}{\sqrt{2^{n}}} \sum_{x \in\{0,1\}^{n}}|x\rangle \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}} \\
& \mapsto \frac{1}{\sqrt{2^{n}}} \sum_{x \in\{0,1\}^{n}}(-1)^{f(x)}|x\rangle \otimes \frac{|0\rangle-|1\rangle}{\sqrt{2}}
\end{aligned}
$$

## Hadamard transform

What do the final Hadamard gates do?

$$
\begin{aligned}
H|x\rangle & =\frac{1}{\sqrt{2}}\left(|0\rangle+(-1)^{x}|1\rangle\right) \\
= & \frac{1}{\sqrt{2}} \sum_{y \in\{0,1\}}(-1)^{x y}|y\rangle \\
H^{\otimes n}\left(\left|x_{1}\right\rangle \otimes \cdots \otimes\left|x_{n}\right\rangle\right) & =\bigotimes_{i=1}^{n}\left(\frac{1}{\sqrt{2}} \sum_{y_{i} \in\{0,1\}}(-1)^{x_{i} y_{i}}\left|y_{i}\right\rangle\right) \\
& =\frac{1}{\sqrt{2^{n}}} \sum_{y \in\{0,1\}^{n}}(-1)^{x \cdot y}|y\rangle
\end{aligned}
$$

## Quantum D-J algorithm: Finishing up

$$
\frac{1}{\sqrt{2^{n}}} \sum_{x \in\{0,1\}^{n}}(-1)^{f(x)}|x\rangle \stackrel{H^{\otimes n}}{\mapsto} \frac{1}{2^{n}} \sum_{x, y \in\{0,1\}^{n}}(-1)^{f(x)}(-1)^{x \cdot y}|y\rangle
$$

- If $f$ is constant, the amplitude of $|y\rangle$ is

$$
\pm \frac{1}{2^{n}} \sum_{x \in\{0,1\}^{n}}(-1)^{x \cdot y}= \pm \begin{cases}1 & \text { if } y=0 \ldots 0 \\ 0 & \text { otherwise }\end{cases}
$$

so we definitely measure $0 \ldots 0$

- If $f$ is balanced, the amplitude of $|0 \ldots 0\rangle$ is

$$
\sum_{x \in\{0,1\}^{n}}(-1)^{f(x)}=0
$$

so we measure some nonzero string

## The Deutsch-Jozsa problem: Quantum vs. classical

Above quantum algorithm uses only one query.
Need $2^{n} / 2+1$ classical queries to answer with certainty.
What about randomized algorithms? Success probability arbitrarily close to 1 with a constant number of queries.

Can we get a separation between randomized and quantum computation?

## Simon's problem

## Problem

- Given: a black-box function $f:\{0,1\}^{n} \rightarrow\{0,1\}^{m}$
- Promise: there is some $s \in\{0,1\}^{n}$ such that $f(x)=f(y)$ if and only if $x=y$ or $x=y \oplus s$
- Task: determine $s$

One classical strategy:

- query a random $x$
- repeat until we find $x_{i} \neq x_{j}$ such that $f\left(x_{i}\right)=f\left(x_{j}\right)$
- output $x_{i} \oplus x_{j}$

By the birthday problem, this uses about $\sqrt{2^{n}}$ queries.
It can be shown that this strategy is essentially optimal.

## Quantum algorithm for Simon's problem

Quantum black box: $|x\rangle \otimes|y\rangle \mapsto|x\rangle \otimes|y \oplus f(x)\rangle$

$$
\left(x \in\{0,1\}^{n}, y \in\{0,1\}^{m}\right)
$$



Repeat many times and post-process the measurement outcomes

## Quantum algorithm for Simon's problem: Analysis I

$$
\begin{aligned}
& |0\rangle^{\otimes n} \otimes|0\rangle^{\otimes m} \\
& \mapsto \frac{1}{\sqrt{2^{n}}} \sum_{x \in\{0,1\}^{n}}|x\rangle \otimes|0\rangle^{\otimes m} \\
& \mapsto \frac{1}{\sqrt{2^{n}}} \sum_{x \in\{0,1\}^{n}}|x\rangle \otimes|f(x)\rangle \\
& =\frac{1}{\sqrt{2^{n-1}}} \sum_{x \in R} \frac{|x\rangle+|x \oplus s\rangle}{\sqrt{2}} \otimes|f(x)\rangle
\end{aligned}
$$

for some $R \subset\{0,1\}^{n}$

## Quantum algorithm for Simon's problem: Analysis II

Recall $\boldsymbol{H}^{\otimes n}|x\rangle=\sum_{y \in\{0,1\}^{n}}(-1)^{x \cdot y}|y\rangle$

$$
\begin{aligned}
H^{\otimes n}\left(\frac{|x\rangle+|x \oplus s\rangle}{\sqrt{2}}\right) & =\frac{1}{\sqrt{2^{n+1}}} \sum_{y \in\{0,1\}^{n}}\left[(-1)^{x \cdot y}+(-1)^{(x \oplus s) \cdot y}\right]|y\rangle \\
& =\frac{1}{\sqrt{2^{n+1}}} \sum_{y \in\{0,1\}^{n}}(-1)^{x \cdot y}\left[1+(-1)^{s \cdot y}\right]|y\rangle
\end{aligned}
$$

Two cases:

- if $s \cdot y=0 \bmod 2,1+(-1)^{s \cdot y}=2$
- if $s \cdot y=1 \bmod 2,1+(-1)^{s \cdot y}=0$

Measuring gives a random $y$ orthogonal to $s$ (i.e., $s \cdot y=0$ )

## Quantum algorithm for Simon's problem: Post-processing

Measuring gives a random $y$ orthogonal to $s(s \cdot y=0)$
Repeat $k$ times, giving vectors $y_{1}, \ldots, y_{k} \in\{0,1\}^{n}$; solve a system of $k$ linear equations for $s \in\{0,1\}^{n}$ :

$$
y_{1} \cdot s=0, \quad y_{2} \cdot s=0, \quad \ldots, \quad y_{k} \cdot s=0
$$

How big should $k$ be to give a unique (nonzero) solution?

- Clearly $k \geq n-1$ is necessary
- It can be shown that $k=O(n)$ suffices
$O(n)$ quantum queries, $O\left(n^{3}\right)$ quantum gates
Compare to $\Omega\left(2^{n / 2}\right)$ classical queries (even for bounded error)


## Recap

We have seen several examples of quantum algorithms that outperform classical computation:

- Deutsch's problem: 1 quantum query vs. 2 classical queries
- Deutsch-Jozsa problem: 1 quantum query vs. $2^{\Omega(n)}$ classical queries (deterministic)
- Simon's problem: $O(n)$ quantum queries vs. $2^{\Omega(n)}$ classical queries (randomized)

Quantum algorithms for more interesting problems build on the tools used in these examples.

## Exercise: One-out-of-four search

Let $f:\{0,1\}^{2} \rightarrow\{0,1\}$ be a black-box function taking the value 1 on exactly one input. The goal is to find the unique $\left(x_{1}, x_{2}\right) \in\{0,1\}^{2}$ such that $f\left(x_{1}, x_{2}\right)=1$.

- Write the truth tables of the four possible functions $f$.
- How many classical queries are needed to solve the problem?
- Suppose $f$ is given as a quantum black box $U_{f}$ acting as

$$
\left|x_{1}, x_{2}, y\right\rangle \mapsto\left|x_{1}, x_{2}, y \oplus f\left(x_{1}, x_{2}\right)\right\rangle .
$$

Determine the output of the following quantum circuit for each of the possible black-box functions $f$ :


- Show that the four possible outputs obtained in the previous part are pairwise orthogonal. What can you conclude about the quantum query complexity of one-out-of-four search?


## Part III

The QFT and phase estimation

## Quantum phase estimation

## Problem

We are given a unitary $U$ and an eigenvector $|\psi\rangle$ of $U$ with unknown eigenvalue

We seek to determine its eigenphase $\varphi \in[0,1)$ such that

$$
U|\psi\rangle=e^{2 \pi i \varphi}|\psi\rangle
$$

More precisely, we want to obtain an estimate $\hat{\varphi}$ such that

$$
\operatorname{Pr}\left(|\hat{\varphi}-\varphi| \leq \frac{1}{2^{n}}\right) \geq \frac{3}{4}
$$

The deviation $|\hat{\varphi}-\varphi|$ is computed modulo 1

## Phase kick back

$$
\begin{aligned}
\frac{|0\rangle+|1\rangle}{\sqrt{2}} \otimes|\psi\rangle & =\frac{|0\rangle}{\sqrt{2}} \otimes|\psi\rangle+\frac{|1\rangle}{\sqrt{2}} \otimes|\psi\rangle \\
& \mapsto \frac{|0\rangle}{\sqrt{2}} \otimes|\psi\rangle+\frac{|1\rangle}{\sqrt{2}} \otimes U|\psi\rangle \\
& =\frac{|0\rangle}{\sqrt{2}} \otimes|\psi\rangle+\frac{|1\rangle}{\sqrt{2}} \otimes e^{2 \pi i \varphi}|\psi\rangle \\
& =\frac{|0\rangle}{\sqrt{2}} \otimes|\psi\rangle+\frac{e^{2 \pi i \varphi}|1\rangle}{\sqrt{2}} \otimes|\psi\rangle \\
& =\frac{|0\rangle+e^{2 \pi i \varphi}|1\rangle}{\sqrt{2}} \otimes|\psi\rangle
\end{aligned}
$$

## Phase kick back

The eigenstate $|\psi\rangle$ in the target register emerges unchanged
$\Rightarrow$ It suffices to focus on the control register
The state $|0\rangle+|1\rangle$ of the control qubit is changed to $|0\rangle+e^{2 \pi i \varphi}|1\rangle$ by phase kick back

## Hadamard test + phase kick back



$$
\begin{aligned}
& \frac{|0\rangle+e^{2 \pi i \varphi}|1\rangle}{\sqrt{2}} \\
\mapsto & \frac{1}{2}\left((|0\rangle+|1\rangle)+e^{2 \pi i \varphi}(|0\rangle-|1\rangle)\right) \\
\mapsto & \left.\frac{1}{2}\left(\left(1+e^{2 \pi i \varphi}\right)|0\rangle+\left(1-e^{2 \pi i \varphi}\right)|1\rangle\right)\right):=|\varphi\rangle
\end{aligned}
$$

## Hadamard test + phase kick back

$$
\left.|\varphi\rangle=\frac{1}{2}\left(\left(1+e^{2 \pi i \varphi}\right)|0\rangle+\left(1-e^{2 \pi i \varphi}\right)|1\rangle\right)\right)
$$

The probability of obtaining 0 is

$$
\begin{aligned}
\operatorname{Pr}(0) & =\||0\rangle\langle 0||\varphi\rangle \|^{2} \\
& =\left|\frac{1}{2}\left(1+e^{2 \pi i \varphi}\right)\right|^{2} \\
& \left.\left.=\frac{1}{4} \right\rvert\, e^{\pi i \varphi}+e^{-\pi i \varphi}\right)\left.\right|^{2} \\
& =\frac{1}{4}|2 \cos (\pi \varphi)|^{2} \\
& =\cos ^{2}(\pi \varphi)=\frac{1}{2}(1+\cos (2 \pi \varphi))
\end{aligned}
$$

## Phase kick back due to higher powers of $U$

For arbitrary $k$, we obtain

since

$$
U^{2^{k}}|\psi\rangle=e^{2 \pi i 2^{k} \varphi}|\psi\rangle
$$

## Phase kick back part of phase estimation



We set

$$
|\varphi\rangle:=\frac{|0\rangle+e^{2 \pi i 2^{n-1} \varphi}|1\rangle}{\sqrt{2}} \otimes \frac{|0\rangle+e^{2 \pi i 2^{n-2} \varphi}|1\rangle}{\sqrt{2}} \otimes \cdots \otimes \frac{|0\rangle+e^{2 \pi i 2^{0} \varphi}|1\rangle}{\sqrt{2}}
$$

## Binary fractions

Assume that the eigenphase $\varphi$ is an exact $n$-bit binary fraction, i.e.,

$$
\varphi=0 . x_{1} x_{2} \ldots x_{n}=\sum_{i=1}^{n} \frac{x_{i}}{2^{i}}
$$

For arbitrary $k \in\{0, \ldots, n-1\}$, we have

$$
\begin{aligned}
2^{k} \varphi & =x_{1} x_{2} \ldots x_{k} \cdot x_{k+1} \ldots x_{n} \\
e^{2 \pi i 2^{k} \varphi} & =e^{2 \pi i\left(x_{1} x_{2} \ldots x_{k} \cdot x_{k+1} \ldots x_{n}\right)} \\
& =e^{2 \pi i\left(x_{1} x_{2} \ldots x_{k}+0 . x_{k+1} \ldots x_{n}\right)} \\
& =e^{2 \pi i\left(x_{1} x_{2} \ldots x_{k}\right)} \cdot e^{2 \pi i\left(0 . x_{k+1} \ldots x_{n}\right)} \\
& =e^{2 \pi i\left(0 . x_{k+1} \ldots x_{n}\right)}
\end{aligned}
$$

## Phase kick back part of phase estimation



$$
\begin{aligned}
& \frac{|0\rangle+e^{2 \pi i 0 . x_{n}}|1\rangle}{\sqrt{2}} \\
& \frac{|0\rangle+e^{2 \pi i 0 . x_{n-1} x_{n}}|1\rangle}{\sqrt{2}} \\
& \frac{|0\rangle+e^{2 \pi i 0 . x_{1} \ldots x_{n-1} x_{n}}|1\rangle}{\sqrt{2}} \\
& |\psi\rangle
\end{aligned}
$$

## Quantum Fourier transform

The quantum Fourier transform $F$ is defined by

$$
\begin{aligned}
& F\left(\left|x_{n}\right\rangle \otimes\left|x_{n-1}\right\rangle \otimes \cdots \otimes\left|x_{1}\right\rangle\right) \\
= & \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{n}}|1\rangle}{\sqrt{2}} \otimes \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{n-1} x_{n}}|1\rangle}{\sqrt{2}} \otimes \cdots \otimes \frac{|0\rangle+e^{2 \pi i 0 . x_{1} x_{2} \ldots x_{n}}|1\rangle}{\sqrt{2}}
\end{aligned}
$$

We use inverse quantum Fourier transform $F^{\dagger}$ to obtain the bits of the eigenphase

Note: QFT is defined by $F\left|x_{1}\right\rangle \otimes\left|x_{2}\right\rangle \otimes \cdots \otimes\left|x_{n}\right\rangle=\left|0 . x_{1} x_{2} \ldots x_{n}\right\rangle$ in the literature; we use the above definition for the sake of notational simplicity (otherwise, we would have to include the so-called bit-reversal)

## Quantum circuit for phase estimation



## Inverse quantum Fourier transform for 3 bits



The phase shift $R_{k}$ is defined by

$$
R_{k}:=\left[\begin{array}{cc}
1 & 0 \\
0 & e^{2 \pi i / 2^{k}}
\end{array}\right]
$$

## QPE: least significant bit - top qubit

$$
\begin{aligned}
& \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{3}}|1\rangle}{\sqrt{2}} \stackrel{\ulcorner H}{\llcorner\cdot} \\
& \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{3}}|1\rangle}{\sqrt{2}}=\frac{|0\rangle+(-1)^{x_{3}}|1\rangle}{\sqrt{2}} \\
& \xrightarrow{H}\left|x_{3}\right\rangle
\end{aligned}
$$

## QPE: second bit - middle qubit

$$
\begin{aligned}
& \left|x_{3}\right\rangle \otimes \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{2} x_{3}}|1\rangle}{\sqrt{2}} \\
& \xrightarrow{\operatorname{ctr} I R_{2}^{\dagger}} \quad\left|x_{3}\right\rangle \otimes \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{2} 0}|1\rangle}{\sqrt{2}} \\
& \xrightarrow{I \otimes H} \quad\left|x_{3}\right\rangle \otimes\left|x_{2}\right\rangle
\end{aligned}
$$

## QPE: most significant bit - bottom qubit



$$
\begin{array}{ll} 
& \left|x_{3}\right\rangle \otimes\left|x_{2}\right\rangle \otimes \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{1} x_{2} x_{3}}|1\rangle}{\sqrt{2}} \\
\xrightarrow{\text { ctrl } R_{3}^{\dagger}} & \left|x_{3}\right\rangle \otimes\left|x_{2}\right\rangle \otimes \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{1} x_{2} 0}|1\rangle}{\sqrt{2}} \\
\xrightarrow{\text { ctr } R_{2}^{\dagger}} & \left|x_{3}\right\rangle \otimes\left|x_{2}\right\rangle \otimes \frac{|0\rangle+e^{2 \pi i 0 \cdot x_{1} 00}|1\rangle}{\sqrt{2}} \\
\xrightarrow{\|\| H} & \left|x_{3}\right\rangle \otimes\left|x_{2}\right\rangle \otimes\left|x_{1}\right\rangle
\end{array}
$$

## Summary of phase estimation circuit

We use phase kick back due to the controlled $U^{2^{k}}$ gate to prepare the state

$$
\frac{|0\rangle+e^{2 \pi i 0 . x_{k+1} x_{k+2} \ldots x_{n}}|1\rangle}{\sqrt{2}}
$$

Using the previously determined bits $x_{k+2}, \ldots, x_{n}$, we change this state to

$$
\frac{|0\rangle+e^{2 \pi i 0 . x_{k+1} 0 \ldots 0}|1\rangle}{\sqrt{2}}=\frac{|0\rangle+(-1)^{x_{k}}|1\rangle}{\sqrt{2}}
$$

We apply the Hadamard gate to obtain

$$
\left|x_{k+1}\right\rangle
$$

The controlled phase shifts enable us to reduce the problem of determining each bit to distinguishing between $|+\rangle$ and $|-\rangle$ (deterministic Hadamard test)

## Special case: exact $n$-bit binary fraction

Assume that $\varphi$ is an exact $n$-bit binary fraction, i.e., $\varphi=0 . x_{1} \ldots x_{n-1} x_{n}$

$\Rightarrow$ The measurment of the qubits yields the bits $x_{n}, x_{n-1}, \ldots, x_{1}$ deterministically

## General case: arbitrary eigenphases

Let $\varphi$ be arbitrary
Unless $\varphi$ is an exact $n$-bit fraction, the application of the inverse quantum Fourier transform

$$
F^{\dagger}|\varphi\rangle
$$

produces a superposition of $n$-bit strings

## Geometric summation

Lemma
We have

$$
\begin{aligned}
& \sum_{y=0}^{N-1} e^{2 \pi i \theta y}=N \text { for } \theta=0 \\
& \sum_{y=0}^{N-1} e^{2 \pi i \theta y}=\frac{1-e^{2 \pi i N \theta}}{1-e^{2 \pi i \theta}} \text { for } \theta \in(0,1)
\end{aligned}
$$

Assume that $\theta=\frac{x}{N}$ for some $x \in[0, N-1]$
$\Rightarrow$ We have

$$
\sum_{y=0}^{N-1} e^{2 \pi i \frac{x}{N} y}=N \delta_{x, 0}
$$

## Probability of obtaining a certain estimate

Lemma
Let $x=\sum_{k=1}^{n} x_{i} 2^{n-i}$ and $\varphi_{x}:=0 . x_{1} x_{2} \ldots x_{n}=\frac{x}{2^{n}}$ be the corresponding $n$-bit fraction

The probability of obtaining the estimate $\varphi_{x}$ is

$$
\operatorname{Pr}(x)=\frac{1}{2^{2 n}} \frac{\sin ^{2}\left(2^{n} \pi\left(\varphi-\varphi_{x}\right)\right)}{\sin ^{2}\left(\pi\left(\varphi-\varphi_{x}\right)\right)}
$$

## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Examples of probability distributions for different $\varphi$



## Probability of obtaining a certain estimate

## Proof.

The probability of obtaining the estimate $\varphi_{x}$ is

$$
\begin{aligned}
& \operatorname{Pr}(x) \\
= & \left.\left|\langle x| F^{\dagger}\right| \varphi\right\rangle\left.\right|^{2} \\
= & \left|\left\langle\varphi_{x} \mid \varphi\right\rangle\right|^{2} \\
= & \frac{1}{2^{2 n}}\left|\sum_{y=0}^{2^{n}-1} e^{2 \pi i\left(\varphi-\varphi_{x}\right) y}\right|^{2} \quad \text { geometric summation } \\
= & \frac{1}{2^{2 n}}\left|\frac{1-e^{2 \pi i\left(2^{n}\left(\varphi-\varphi_{x}\right)\right)}}{1-e^{2 \pi i\left(\varphi-\varphi_{x}\right)}}\right|^{2} \quad\left|1-e^{i 2 \theta}\right|=\left|e^{-i \theta}-e^{i \theta}\right|=2|\sin \theta| \\
= & \frac{1}{2^{2 n}} \frac{\sin ^{2}\left(2^{n} \pi\left(\varphi-\varphi_{x}\right)\right)}{\sin ^{2}\left(\pi\left(\varphi-\varphi_{x}\right)\right)}
\end{aligned}
$$

## Lower bound on success probability

Theorem
Let $x$ be such that $\frac{x}{2^{n}} \leq \varphi<\frac{x+1}{2^{n}}$
The probability of returning one of the two closest n-bit fractions $\varphi_{x}$ and $\varphi_{x+1}$ is at least $\frac{8}{\pi^{2}}$

## Proof of lower bound on success probability

$\operatorname{Pr}($ success $):=\operatorname{Pr}(x)+\operatorname{Pr}(x+1)$

$$
=\frac{1}{2^{2 n}}\left(\left|\sum_{y=0}^{2^{n}-1} e^{2 \pi i\left(\varphi-\varphi_{x}\right) y}\right|^{2}+\left|\sum_{y=0}^{2^{n}-1} e^{2 \pi i\left(\varphi-\varphi_{x}\right) y}\right|^{2}\right)
$$

This function attains its minimum at $\varphi=\frac{1}{2}\left(\varphi_{x}+\varphi_{x+1}\right) \quad \Rightarrow$

$$
\begin{aligned}
\operatorname{Pr}(\text { success }) & \geq \frac{2}{2^{2 n}}\left(\left\lvert\, \sum_{y=0}^{2^{n}-1} e^{\left.\left.2 \pi i \frac{y}{2^{n+1}}\right|^{2}\right)}\right.\right. \\
& \geq \frac{2}{2^{2 n}} \frac{4}{4 \sin ^{2}\left(\frac{\pi}{2^{n+1}}\right)} \\
& \geq \frac{8}{\pi^{2}}
\end{aligned}
$$

The last inequality follows from $\frac{1}{|\sin \theta|^{2}} \geq \frac{1}{|\theta|^{2}}$

## Summary of phase estimation

We are given a unitary $U$ and an eigenvector $|\psi\rangle$ of $U$ with unknown eigenphase $\varphi$

We obtain an estimate $\hat{\varphi}$ such that

$$
\operatorname{Pr}\left(|\hat{\varphi}-\varphi| \leq \frac{1}{2^{n}}\right) \geq \frac{8}{\pi^{2}}
$$

To do this, we need invoke each of the controlled $U, U^{2}, \ldots, U^{2^{n-1}}$ gates once

We can boost the success probability to $1-\epsilon$ by repeating the above algorithm $O(\log (1 / \epsilon))$ times and outputting the median of the outcomes

## Phase estimation applied to superpositions of eigenstates

We are given a unitary $U$ with eigenvectors $\left|\psi_{i}\right\rangle$ and corresponding eigenphases $\varphi_{i}$

Let

$$
|\psi\rangle=\sum_{i} \alpha_{i}\left|\psi_{i}\right\rangle
$$

What happens if we apply phase estimation to $|0\rangle^{\otimes n} \otimes|\psi\rangle$ ?
After the $n$ phase kick-backs due to $U^{2^{0}}, U^{2^{1}}, \ldots U^{2^{n-1}}$, we obtain

$$
\sum_{i} \alpha_{i}\left|\varphi_{i}\right\rangle \otimes\left|\psi_{i}\right\rangle
$$

After applying the inverse quantum Fourier transform, we obtain

$$
\sum_{i} \alpha_{i}\left|\tilde{x}_{i}\right\rangle \otimes\left|\psi_{i}\right\rangle
$$

where $\left|\tilde{x}_{i}\right\rangle$ denotes a superpositions of $n$-bit estimates of $\varphi_{i}$

## Part IV

Factoring

## The fundamental theorem of arithmetic

Theorem
Every positive integer larger than 1 can be factored as a product of prime numbers, and this factorization is unique (up to the order of the factors).

$$
N=2^{n_{2}} \times 3^{n_{3}} \times 5^{n_{5}} \times 7^{n_{7}} \times \cdots
$$

## Examples

$$
15=3 \times 5
$$

$239815173914273=15485863 \times 15486071$

3107418240490043721350750 0358885679300373460228427 2754572016194882320644051 8081504556346829671723286 7824379162728380334154710 7310850191954852900733772 4822783525742386454014691 736602477652346609

16347336458092538484 43133883865090859841 78367003309231218111 08523893331001045081 51212118167511579

19008712816648221131 26851573935413975471 89678996851549366663 85390880271038021044 98957191261465571
"The problem of distinguishing prime numbers from composite numbers and of resolving the latter into their prime factors is known to be one of the most important and useful in arithmetic. It has engaged the industry and wisdom of ancient and modern geometers to such an extent that it would be superfluous to discuss the problem at length... Further, the dignity of the science itself seems to require that every possible means be explored for the solution of a problem so elegant and so celebrated."

- Carl Friedrich Gauss, Disquisitiones Arithmeticæ (1801)


Alice
M
message
primes $p, q$
$n \quad n=p q$
$e \quad \begin{gathered}e \in \mathbb{Z}_{(p-1)(q-1)}^{\times} \\ \text {encryption key }\end{gathered}$
$d:=e^{-1} \bmod (p-1)(q-1)$
$C:=M^{e} \bmod n$ ciphertext

C $\quad C^{d}=M^{e d} \bmod n=M$

## Order finding

## Definition

Given $a, N \in \mathbb{Z}$ with $\operatorname{gcd}(a, N)=1$, the order of a modulo $N$ is the smallest positive integer $r$ such that $a^{r} \equiv 1(\bmod N)$.

## Problem

- Given: $a, N \in \mathbb{Z}$ with $\operatorname{gcd}(a, N)=1$
- Task: find the order of a modulo $N$


## Spectrum of a cyclic shift

Let $P$ be a cyclic shift modulo $r: P|x\rangle=|x+1 \bmod r\rangle$
Claim. For any $k \in \mathbb{Z}$, the state $\left|u_{k}\right\rangle:=\frac{1}{\sqrt{r}} \sum_{x=0}^{r-1} e^{-2 \pi i k x / r}|x\rangle$ is an eigenstate of $P$.

$$
\text { Proof. } \begin{align*}
U\left|u_{k}\right\rangle & =\frac{1}{\sqrt{r}} \sum_{x=0}^{r-1} e^{-2 \pi \mathrm{i} k x / r}|x+1 \bmod r\rangle \\
& =\frac{1}{\sqrt{r}} \sum_{x=0}^{r-1} e^{2 \pi \mathrm{i} k / r} e^{-2 \pi \mathrm{i} k(x+1) / r}|x+1 \bmod r\rangle \\
& =e^{2 \pi i k / r} \frac{1}{\sqrt{r}} \sum_{x=1}^{r} e^{-2 \pi \mathrm{i} x / / r}|x \bmod r\rangle \\
& =e^{2 \pi i k / r}\left|u_{k}\right\rangle
\end{align*}
$$

## The multiplication-by-a map

Define $U$ by $U|x\rangle=|a x\rangle$ for $x \in \mathbb{Z}_{N}$.

Computing $U$ :

$$
\begin{aligned}
|x, 0\rangle & \mapsto|x, a x\rangle & & \text { (reversible multiplication by } a) \\
& \mapsto|a x, x\rangle & & \text { (swap) } \\
& \mapsto|a x, 0\rangle & & \text { (uncompute reversible division by } a)
\end{aligned}
$$

High powers of $U$ can be implemented efficiently using repeated squaring

## Spectrum of the multiplication-by-a map

Define $U$ by $U|x\rangle=|a x\rangle$ for $x \in \mathbb{Z}_{N}$.
Claim. Let $r$ be the order of a modulo $N$. For any $k \in \mathbb{Z}$, the state

$$
\left|u_{k}\right\rangle:=\frac{1}{\sqrt{r}} \sum_{x=0}^{r-1} e^{-2 \pi i k x / r}\left|a^{x} \bmod N\right\rangle
$$

is an eigenstate of $U$ with eigenvalue $e^{2 \pi i k / r}$.

Proof.
Same as for the cyclic shift, due to the isomorphism

$$
x \bmod r \quad \leftrightarrow \quad a^{x} \bmod N
$$

## Order finding and phase estimation

$U\left|u_{k}\right\rangle=e^{2 \pi i k / r}\left|u_{k}\right\rangle$
Phase estimation of $U$ on $\left|u_{k}\right\rangle$ can be used to approximate $k / r$.

Problems:

- We don't know $r$, so we can't prepare $\left|u_{k}\right\rangle$.
- We only get an approximation of $k / r$.
- Even if we knew $k / r$ exactly, $k$ and $r$ could have common factors.


## Estimating $k / r$ in superposition

A useful identity:

$$
\sum_{k=0}^{r-1} e^{2 \pi i k x / r}= \begin{cases}r & \text { if } x=0 \\ 0 & \text { otherwise }\end{cases}
$$

Consider

$$
\begin{aligned}
\frac{1}{\sqrt{r}} \sum_{k=0}^{r-1}\left|u_{k}\right\rangle & =\frac{1}{r} \sum_{k, x=0}^{r-1} e^{-2 \pi i k x / r}\left|a^{x} \bmod N\right\rangle \\
& =\left|a^{0} \bmod N\right\rangle=|1\rangle
\end{aligned}
$$

Phase estimation:

$$
|0\rangle \otimes|1\rangle=\frac{1}{\sqrt{r}} \sum_{k=0}^{r-1}|0\rangle \otimes\left|u_{k}\right\rangle \mapsto \frac{1}{\sqrt{r}} \sum_{k=0}^{r-1}|\widetilde{k / r}\rangle \otimes\left|u_{k}\right\rangle
$$

Measurement gives an approximation of $k / r$ for a random $k$

## Continued fractions

## Problem

Given samples $x$ of the form $\left\lfloor k \frac{2^{n}}{r}\right\rfloor,\left\lceil k \frac{2^{n}}{r}\right\rceil(k \in\{0,1, \ldots, r-1\})$, determine $r$.

Continued fraction expansion:

$$
\frac{x}{2^{n}}=\frac{1}{a_{1}+\frac{1}{a_{2}+\frac{1}{a_{3}+\cdots}}}
$$

Gives an efficiently computable sequence of rational approximations
Theorem
If $2^{n} \geq N^{2}$, then $k / r$ is the closest convergent of the CFE to $x / 2^{n}$ among those with denominator smaller than $N$.

Since $r<N$, it suffices to take $n=2 \log _{2} N$

## Common factors

If $\operatorname{gcd}(k, r)=1$, then the denominator of $k / r$ is $r$

## Fact

The probability that $\operatorname{gcd}(k, r)=1$ for a random $k \in\{0,1, \ldots, r-1\}$ is

$$
\frac{\phi(r)}{r}=\Omega\left(\frac{1}{\log \log r}\right)
$$



Thus $\Omega(\log \log N)$ repetitions suffice to give $r$ with constant probability

Alternatively, find two (or more) denominators and take their least common multiple; then $O(1)$ repetitions suffice

## Factoring $\rightarrow$ finding a nontrivial factor

Suppose we want to factor the positive integer $N$.
Since primality can be tested efficiently, it suffices to give a procedure for finding a nontrivial factor of $N$ with constant probability.

```
function factor(N)
if N is prime
    output N
else
    repeat
        x=find_nontrivial_factor(N)
    until success
    factor(x)
    factor(N/x)
end if
```

We can assume $N$ is odd, since it is easy to find the factor 2 .
We can also assume that $N$ contains at least two distinct prime powers, since it is easy to check if it is a power of some integer.

## Reduction of factoring to order finding

Factoring $N$ reduces to order finding in $\mathbb{Z}_{N}^{\times}$[Miller 1976].
Choose $a \in\{2,3, \ldots, N-1\}$ uniformly at random.
If $\operatorname{gcd}(a, N) \neq 1$, then it is a nontrivial factor of $N$.
If $\operatorname{gcd}(a, N)=1$, let $r$ denote the order of $a$ modulo $N$.
Suppose $r$ is even. Then

$$
\begin{gathered}
a^{r}=1 \bmod N \\
\left(a^{r / 2}\right)^{2}-1 \stackrel{\Downarrow}{\Uparrow}=0 \bmod N \\
\left(a^{r / 2}-1\right)\left(a^{r / 2}+1\right)=0 \bmod N
\end{gathered}
$$

so we might hope that $\operatorname{gcd}\left(a^{r / 2}-1, N\right)$ is a nontrivial factor of $N$.

## Miller's reduction

## Question

Given $\left(a^{r / 2}-1\right)\left(a^{r / 2}+1\right)=0 \bmod N$, when does $\operatorname{gcd}\left(a^{r / 2}-1, N\right)$ give a nontrivial factor of $N$ ?

Note that $a^{r / 2}-1 \neq 0 \bmod N$ (otherwise the order of a would be $r / 2$, or smaller).

So it suffices to ensure that $a^{r / 2}+1 \neq 0 \bmod N$.

## Lemma

Suppose $a \in \mathbb{Z}_{N}^{\times}$is chosen uniformly at random, where $N$ is an odd integer with at least two distinct prime factors. Then with probability at least $1 / 2$, the order $r$ of $a$ is even and $a^{r / 2} \neq-1 \bmod N$.

## Proof (part 1 of 2 )

Let $N=p_{1}^{m_{1}} \times \cdots \times p_{k}^{m_{k}} \quad\left(p_{i}\right.$ distinct odd primes, $\left.k \geq 2\right)$

$$
\begin{aligned}
a & =a_{i} \bmod p_{i}^{m_{i}} \\
r_{i} & =\text { order of } a_{i} \bmod p_{i}^{m_{i}} \\
2^{c_{i}} & =\text { largest power of } 2 \text { that divides } r_{i}
\end{aligned}
$$

Claim 1. If $r$ is odd or $a^{r / 2}+1=0 \bmod N$, then $c_{1}=\cdots=c_{k}$.
Since $r=\operatorname{lcm}\left(r_{1}, \ldots, r_{k}\right), r$ is odd iff $c_{1}=\ldots=c_{k}=0$.
If $r$ is even and $a^{r / 2}=-1 \bmod N$, then $a^{r / 2}=-1 \bmod p_{i}^{m_{i}}$ for each $i$, so $r_{i}$ does not divide $r / 2$; but notice that $r_{i}$ does divide $r$. Hence $r / r_{i}$ is an odd integer for each $i$, and every $r_{i}$ must contain the same number of powers of 2 as $r$.

## Proof (part 2 of 2 )

Claim 2. $\operatorname{Pr}\left(c_{i}=\right.$ any particular value $) \leq 1 / 2$
(Then the lemma follows, since in particular $\operatorname{Pr}\left(c_{1}=c_{2}\right) \leq 1 / 2$.)


Since $\mathbb{Z}_{p_{i}}^{\times}$is cyclic and of even order, exactly half its elements have the maximal value of $c_{i}$, so in particular the probability of any particular $c_{i}$ is at most $1 / 2$.

## Shor's algorithm

Input: Integer $N$
Output: A nontrivial factor of $N$

1. Choose a random $a \in\{2,3, \ldots, N-1\}$
2. Compute $\operatorname{gcd}(a, N)$; if it is not 1 then it is a nontrivial factor, and otherwise we continue
3. Perform phase estimation with the multiplication-by-a operator $U$ on the state $|1\rangle$ using $n=2 \log _{2} N$ bits of precision
4. Compute the continued fraction expansion of the estimated phase, and find the best approximation with denominator less than $N$; call the result $r$
5. Compute $\operatorname{gcd}\left(a^{r / 2}-1, N\right)$. If it is a nontrivial factor of $N$, we are done; if not, go back to step 1

## Quantum vs. classical factoring algorithms

Best known classical algorithm for factoring $N$

- Proven running time: $2^{O\left((\log N)^{1 / 2}(\log \log N)^{1 / 2}\right)}$
- With plausible heuristic assumptions: $2^{O\left((\log N)^{1 / 3}(\log \log N)^{1 / 3}\right)}$

Shor's quantum algorithm

- QFT modulo $2^{n}$ with $n=O(\log N)$ : takes $O\left(n^{2}\right)$ steps
- Modular exponentiation: compute $a^{x}$ for $x<2^{n}$. With repeated squaring, takes $O\left(n^{3}\right)$ steps
- Running time of Shor's algorithm: $O\left(\log ^{3} N\right)$


## Beyond factoring

There are many fast quantum algorithms based on related ideas

- Computing discrete logarithms
- Decomposing abelian/solvable groups
- Estimating Gauss sums
- Counting points on algebraic curves
- Computations in number fields (Pell's equation, etc.)
- Abelian hidden subgroup problem
- Non-abelian hidden subgroup problem?

Part V
Quantum search

## Unstructured search

Quantum computers can quadratically outperform classical computers at a very basic computational task, called unstructured search.

There is a set $X$ containing $N$ items, some of which are marked
We are given a Boolean black box $f: X \rightarrow\{0,1\}$ that indicates whether a given item is marked

The problem is to decide if any item is marked, or alternatively, to find a marked item given that one exists

## Applications of unstructured search

Unstructured search can be thought of as a model for solving problems in NP by brute force search

If a problem is in NP, then we can efficiently recognize a solution, so one way to find a solution is to solve unstructured search

Of course, this may not be the best way to find a solution in general, even if the problem is NP-hard

We don't know if NP-hard problems are really "unstructured"

## Unstructured search

It is obvious that even a randomized classical algorithm needs $\Omega(N)$ queries to decide if any item is marked

On the other hand, a quantum algorithm can do much better!

## Phase oracle

We assume that we a unitary operator $U$ satisfying

$$
U|x\rangle=(-1)^{f(x)}|x\rangle=\left\{\begin{aligned}
|x\rangle & x \text { is not marked } \\
-|x\rangle & x \text { is marked }
\end{aligned}\right.
$$

## Target state

We consider the case where there is exactly one $x \in X$ element that is marked; call this element $m$

Our goal is to prepare the state $|m\rangle$

## Initial state

We have no information about which item might be marked
$\Rightarrow$ We take

$$
|\psi\rangle:=\frac{1}{\sqrt{N}} \sum_{x=1}^{N}|x\rangle
$$

as the initial state

## Rough idea behind Grover search

We start with the initial state $|\psi\rangle$
We prepare the target state $|m\rangle$ by implementing a rotation that moves $|\psi\rangle$ toward $|m\rangle$

We realize the rotation with the help of two reflections

Visualization of a reflection in $\mathbb{R}^{2}$


Visualization of a reflection in $\mathbb{R}^{2}$


## Visualization of a reflection in $\mathbb{R}^{2}$



## Reflections

$U=I-2|m\rangle\langle m|$ is a reflection about the target state $|m\rangle$
$V=I-2|\psi\rangle\langle\psi|$ is the reflection around about the initial state $|\psi\rangle$ :

$$
\begin{aligned}
V|\psi\rangle & =-|\psi\rangle \\
V\left|\psi^{\perp}\right\rangle & =\left|\psi^{\perp}\right\rangle
\end{aligned}
$$

for any state $\left|\psi^{\perp}\right\rangle$ orthogonal to $|\psi\rangle$

## Structure of Grover

The algorithm is as follows:

- start in $|\psi\rangle$,
- apply the Grover iteration $G:=V U$ some number of times,
- make a measurement, and hope that the outcome is $m$


## Invariant subspace

Observe that span $\{|m\rangle,|\psi\rangle\}$ is a $U$ - and $V$-invariant subspace, and both the inital and target states belong to this subspace
$\Rightarrow$ It suffices to understand the restriction of $V U$ to this subspace
Consider an orthonormal basis $\{|m\rangle,|\phi\rangle\}$ for $\operatorname{span}\{|m\rangle,|\psi\rangle\}$
The Gram-Schmidt process yields

$$
|\phi\rangle=\frac{|\psi\rangle-\alpha|m\rangle}{\sqrt{1-\alpha^{2}}}
$$

where $\alpha:=\langle m \mid \psi\rangle=1 / \sqrt{N}$

## Invariant subspace

Now in the basis $\{|m\rangle,|\phi\rangle\}$, we have

$$
\begin{aligned}
&|\psi\rangle=\sin \theta|m\rangle+\cos \theta|\phi\rangle \text { where } \sin \theta=\langle m \mid \psi\rangle=1 / \sqrt{N} \\
& U=\left(\begin{array}{cc}
-1 & 0 \\
0 & 1
\end{array}\right) \\
& V=I-2|\psi\rangle\langle\psi| \\
&=\left(\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right)-2\binom{\sin \theta}{\cos \theta}(\sin \theta \\
&\cos \theta) \\
&=\left(\begin{array}{cc}
1-2 \sin ^{2} \theta & -2 \sin \theta \cos \theta \\
-2 \sin \theta \cos \theta & 1-2 \cos ^{2} \theta
\end{array}\right) \\
&=-\left(\begin{array}{cc}
-\cos 2 \theta & \sin 2 \theta \\
\sin 2 \theta & \cos 2 \theta
\end{array}\right)
\end{aligned}
$$

## Grover iteration within the invariant subspace

$\Rightarrow$ We find

$$
\begin{aligned}
V U & =-\left(\begin{array}{cc}
-\cos 2 \theta & \sin 2 \theta \\
\sin 2 \theta & \cos 2 \theta
\end{array}\right)\left(\begin{array}{cc}
-1 & 0 \\
0 & 1
\end{array}\right) \\
& =-\left(\begin{array}{cc}
\cos 2 \theta & \sin 2 \theta \\
-\sin 2 \theta & \cos 2 \theta
\end{array}\right)
\end{aligned}
$$

This is a rotation up to a minus sign

## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Visualization of first Grover iteration



## Grover search

Geometrically, $U$ is a reflection around the $|m\rangle$ axis and $V$ is a reflection around the $|\psi\rangle$ axis, which is almost but not quite orthogonal to the $|m\rangle$ axis

The product of these two reflections is a clockwise rotation by an angle $2 \theta$, up to an overall minus sign

From this geometric picture, or by explicit calculation using trig identities, it is easy to verify that

$$
(V U)^{k}=(-1)^{k}\left(\begin{array}{cc}
\cos 2 k \theta & \sin 2 k \theta \\
-\sin 2 k \theta & \cos 2 k \theta
\end{array}\right)
$$

## Grover search

Recall that our initial state is $|\psi\rangle=\sin \theta|m\rangle+\cos \theta|\phi\rangle$
How large should $k$ be before $(V U)^{k}|\psi\rangle$ is close to $|m\rangle$ ?
We start an angle $\theta$ from the $|\phi\rangle$ axis and rotate toward $|m\rangle$ by an angle $2 \theta$ per iteration
$\Rightarrow$ To rotate by $\pi / 2$, we need

$$
\begin{aligned}
\theta+2 k \theta & =\pi / 2 \\
k & \approx \frac{\pi}{4} \theta^{-1} \approx \frac{\pi}{4} \sqrt{N}
\end{aligned}
$$

## Grover search

It is easy to calculate that

$$
\left.\left|\langle m|(V U)^{k}\right| \psi\right\rangle\left.\right|^{2}=\sin ^{2}((2 k+1) \theta)
$$

This is the probability that, after $k$ steps of the algorithm, a measurement reveals the marked state

We are solving a completely unstructured search problem with $N$ possible solutions, yet we can find a unique solution in only $O(\sqrt{N})$ queries!

While this is only a polynomial separation, it is very generic, and it is surprising that we can obtain a speedup for a search in which we have so little information to go on

## Grover search

It can also be shown that this quantum algorithm is optimal
Any quantum algorithm needs at least $\Omega(\sqrt{N})$ queries to find a marked item (or even to decide if some item is marked)

## Multiple solutions

Assume that there are $t$ marked items
$\Rightarrow$ There is a two-dimensional invariant subspace spanned by $\operatorname{span}\{|\mu\rangle,|\psi\rangle\}$ where

$$
|\mu\rangle=\frac{1}{\sqrt{t}} \sum_{x \text { marked }}|x\rangle
$$

is the uniform superposition of all solutions
The Gram-Schmidt process yields the ONB $\{|\mu\rangle,|\phi\rangle\}$ where

$$
|\phi\rangle=\frac{1}{\sqrt{N-t}} \sum_{x \text { unmarked }}|x\rangle
$$

is the uniform superposition of all non-solutions

## Invariant subspace

Now in the basis $\{|\mu\rangle,|\phi\rangle\}$, we have

$$
\begin{aligned}
|\psi\rangle & =\sin \theta|\mu\rangle+\cos \theta|\phi\rangle \text { where } \sin \theta=\langle\mu \mid \psi\rangle=\sqrt{\frac{t}{N}} \\
V U & =-\left(\begin{array}{cc}
\cos 2 \theta & \sin 2 \theta \\
-\sin 2 \theta & \cos 2 \theta
\end{array}\right)
\end{aligned}
$$

## Overshooting

The success probability is given by

$$
\sin ((2 k+1) \theta) \text { where } \sin \theta=\sqrt{\frac{t}{N}}
$$

$\Rightarrow$ We need to apply $V U$

$$
k \approx \frac{\pi}{4} \sqrt{\frac{N}{t}}
$$

times
Due to the oscillatory behaviour of the success probability it is important not to overshoot, i.e., to choose a number of iterations that it too large, so that the probability starts decreasing

## Quantum counting

The eigenvalues of

$$
-V U=\left(\begin{array}{cc}
\cos 2 \theta & \sin 2 \theta \\
-\sin 2 \theta & \cos 2 \theta
\end{array}\right)
$$

are $e^{i 2 \theta}$ and $e^{-i 2 \theta}$
The initial state $|\psi\rangle$ is a superposition of the two eigenvectors corresponding to the above two eigenvalues
$\Rightarrow$ Using phase estimation, we can obtain an estimate $\tilde{\theta}$ such that

$$
|\theta-\tilde{\theta}| \leq \epsilon
$$

by invoking the controlled version of $-V U$

$$
O(1 / \epsilon) \text { times }
$$

## Quantum counting

Using the estimate $\tilde{\theta}$, we obtain an estimate $\tilde{t}$ satifying

$$
|t-\tilde{t}| \leq(2 \sqrt{t N}+\epsilon) \epsilon
$$

## Quantum counting

We use the following two inequalities:

$$
\begin{aligned}
& |\sin \theta+\sin \tilde{\theta}| \leq 2 \sin \theta+|\theta-\tilde{\theta}| \leq 2 \sqrt{\frac{t}{N}}+\epsilon \\
& |\sin \theta-\sin \tilde{\theta}| \leq|\theta-\tilde{\theta}| \leq \epsilon
\end{aligned}
$$

We have

$$
\begin{aligned}
\left|\frac{t}{N}-\frac{\tilde{t}}{N}\right| & =\left|\sin ^{2} \theta-\sin ^{2} \tilde{\theta}\right| \\
& =|\sin \theta+\sin \tilde{\theta}||\sin \theta-\sin \tilde{\theta}| \\
& \leq\left(2 \sqrt{\frac{t}{N}}+\epsilon\right) \epsilon
\end{aligned}
$$

## Amplitude amplification

Assume that there is a classical (randomized) algorithm that produces a solution to some problem with probability $p$

Assume that we can recognized if the output produced by the algorithm is a valid solution or not
$\Rightarrow$ We repeat the algorithm until we obtain a solution
The expected number of times we have to repeat is $O(1 / p)$ (geometric random variable)

Quantum amplitude amplification makes it possible to reduce the complexity to $O(1 / \sqrt{p})$

## Part VI

## Quantum walk

## Randomized algorithms

Randomness is an important tool in computer science
Black-box problems

- Huge speedups are possible (Deutsch-Jozsa: $2^{\Omega(n)}$ vs. $O(1)$ )
- Polynomial speedup for some total functions (game trees: $\Omega(n)$ vs. $\left.O\left(n^{0.754}\right)\right)$

Natural problems

- Majority view is that derandomization should be possible ( $\mathrm{P}=\mathrm{BPP}$ )
- Randomness may give polynomial speedups (Schöning algorithm for $k$-SAT)
- Can be useful for algorithm design


## Random walk

Graph $G=(V, E)$


Two kinds of walks:

- Discrete time
- Continuous time


## Random walk algorithms

Undirected $s-t$ connectivity in log space

- Problem: given an undirected graph $G=(V, E)$ and $s, t \in V$, is there a path from $s$ to $t$ ?
- A random walk from $s$ eventually reaches $t$ iff there is a path
- Taking a random walk only requires log space
- Can be derandomized (Reingold 2004), but this is nontrivial

Markov chain Monte Carlo

- Problem: sample from some probability distribution (uniform distribution over some set of combinatorial objects, thermal equilibrium state of a physical system, etc.)
- Create a Markov chain whose stationary distribution is the desired one
- Run the chain until it converges


## Continuous-time quantum walk



$$
\begin{array}{cc}
A=\left(\begin{array}{ccccc}
0 & 1 & 1 & 0 & 0 \\
1 & 0 & 0 & 1 & 1 \\
1 & 0 & 0 & 1 & 0 \\
0 & 1 & 1 & 0 & 1 \\
0 & 1 & 0 & 1 & 0
\end{array}\right) & L=\left(\begin{array}{ccccc}
-2 & 1 & 1 & 0 & 0 \\
1 & -3 & 0 & 1 & 1 \\
1 & 0 & -2 & 1 & 0 \\
0 & 1 & 1 & -3 & 1 \\
0 & 1 & 0 & 1 & -2
\end{array}\right) \\
\text { adjacency matrix } & \text { Laplacian }
\end{array}
$$

Random walk on $G$

- State: probability $p_{v}(t)$ of being at vertex $v$ at time $t$
- Dynamics: $\frac{\mathrm{d}}{\mathrm{d} t} \vec{p}(t)=-L \vec{p}(t)$

Quantum walk on $G$

- State: amplitude $q_{v}(t)$ to be at vertex $v$ at time $t$ (i.e., $|\psi(t)\rangle=\sum_{v \in V} q_{v}(t)|v\rangle$ )
- Dynamics: $\mathrm{i} \frac{\mathrm{d}}{\mathrm{d} t} \vec{q}(t)=-L \vec{q}(t)$


## Random vs. quantum walk on the line



Classical:


Quantum:


## Random vs. quantum walk on the hypercube



Classical random walk: reaching $11 \ldots 1$ from $00 \ldots 0$ is exponentially unlikely

Quantum walk: with $A=\sum_{j=1}^{n} X_{j}$,

$$
e^{-\mathrm{i} A t}=\prod_{j=1}^{n} e^{-\mathrm{i} x_{j} t}=\bigotimes_{j=1}^{n}\left(\begin{array}{cc}
\cos t & -\mathrm{i} \sin t \\
-\mathrm{i} \sin t & \cos t
\end{array}\right)
$$

## Glued trees problem



Black-box description of a graph

- Vertices have arbitrary labels
- Label of 'in' vertex is known
- Given a vertex label, black box returns labels of its neighbors
- Restricts algorithms to explore the graph locally


## Glued trees problem: Classical query complexity



Let $n$ denote the height of one of the binary trees
Classical random walk from 'in': probability of reaching 'out' is $2^{-\Omega(n)}$ at all times

In fact, the classical query complexity is $2^{\Omega(n)}$

## Glued trees problem: Exponential speedup

Column subspace


$$
\begin{aligned}
& |\operatorname{col} j\rangle:=\frac{1}{\sqrt{N_{j}}} \sum_{v \in \text { column } j}|v\rangle \\
& N_{j}:= \begin{cases}2^{j} & \text { if } j \in[0, n] \\
2^{2 n+1-j} & \text { if } j \in[n+1,2 n+1]\end{cases}
\end{aligned}
$$

Reduced adjacency matrix

$$
\begin{aligned}
& \langle\operatorname{col} j| A|\operatorname{col} j+1\rangle \\
& \quad= \begin{cases}\sqrt{2} & \text { if } j \in[0, n-1] \\
\sqrt{2} & \text { if } j \in[n+1,2 n] \\
2 & \text { if } j=n\end{cases}
\end{aligned}
$$

## Discrete-time quantum walk: Need for a coin

Quantum analog of discrete-time random walk?
Unitary matrix $U \in \mathbb{C}^{|V| \times|V|}$ with $U_{v w} \neq 0$ iff $(v, w) \in E$
Consider the line:


Define walk by $|x\rangle \mapsto \frac{1}{\sqrt{2}}(|x-1\rangle+|x+1\rangle)$ ?
But then $|x+2\rangle \mapsto \frac{1}{\sqrt{2}}(|x+1\rangle+|x+3\rangle)$, so this is not unitary!
In general, we must enlarge the state space.

## Discrete-time quantum walk on a line



Add a "coin": state space $\operatorname{span}\{|x\rangle \otimes|\leftarrow\rangle,|x\rangle \otimes|\rightarrow\rangle: x \in \mathbb{Z}\}$
Coin flip: $C:=I \otimes H$
Shift: $\quad \begin{aligned} & S|x\rangle \otimes|\leftarrow\rangle=|x-1\rangle \otimes|\leftarrow\rangle \\ & S|x\rangle \otimes|\rightarrow\rangle=|x+1\rangle \otimes|\rightarrow\rangle\end{aligned}$
Walk step: SC


## The Szegedy walk

State space: $\operatorname{span}\{|v\rangle \otimes|w\rangle,|w\rangle \otimes|v\rangle:(v, w) \in E\}$
Let $W$ be a stochastic matrix (a discrete-time random walk)

$$
\begin{aligned}
\text { Define }\left|\psi_{v}\right\rangle & :=|v\rangle \otimes \sum_{w \in V} \sqrt{W_{w v}}|w\rangle \quad\left(\text { note }\left\langle\psi_{v} \mid \psi_{w}\right\rangle=\delta_{v, w}\right) \\
R & :=2 \sum_{v \in V}\left|\psi_{v}\right\rangle\left\langle\psi_{v}\right|-I \\
S(|v\rangle \otimes|w\rangle) & :=|w\rangle \otimes|v\rangle
\end{aligned}
$$

Then a step of the walk is the unitary operator $U:=S R$

## Spectrum of the walk

$$
\text { Let } T:=\sum_{v \in V}\left|\psi_{v}\right\rangle\langle v| \text {, so } R=2 T T^{\dagger}-I
$$

Theorem (Szegedy)
Let $W$ be a stochastic matrix. Suppose the matrix

$$
\sum_{v, w} \sqrt{W_{v w} W_{w v}}|w\rangle\langle v|
$$

has an eigenvector $|\lambda\rangle$ with eigenvalue $\lambda$. Then

$$
\frac{I-e^{ \pm i \arccos \lambda} S}{\sqrt{2\left(1-\lambda^{2}\right)}} T|\lambda\rangle
$$

are eigenvectors of $U=S R$ with eigenvalues

$$
e^{ \pm i \arccos \lambda}
$$

## Proof of Szegedy's spectral theorem

Proof sketch.
Straightforward calculations give

$$
\begin{gathered}
T T^{\dagger}=\sum_{v \in V}\left|\psi_{v}\right\rangle\left\langle\psi_{v}\right| \quad T^{\dagger} T=I \\
T^{\dagger} S T=\sum_{v, w \in V} \sqrt{W_{v w} W_{w v}}|w\rangle\langle v|=\sum_{\lambda}|\lambda\rangle\langle\lambda|
\end{gathered}
$$

which can be used to show

$$
U(T|\lambda\rangle)=S T|\lambda\rangle \quad U(S T|\lambda\rangle)=2 \lambda S T|\lambda\rangle-T|\lambda\rangle
$$

Diagonalizing within the subspace $\operatorname{span}\{T|\lambda\rangle, S T|\lambda\rangle\}$ gives the desired result.

Exercise. Fill in the details

## Random walk search algorithm

Given $G=(V, E)$, let $M \subset V$ be a set of marked vertices
Start at a random unmarked vertex
Walk until we reach a marked vertex:

$$
\begin{aligned}
W_{v w}^{\prime} & := \begin{cases}1 & w \in M \text { and } v=w \\
0 & w \in M \text { and } v \neq w \\
W_{v w} & w \notin M .\end{cases} \\
& =\left(\begin{array}{cc}
W_{M} & 0 \\
V & 1
\end{array}\right) \quad\left(W_{M}: \text { delete marked rows and columns of } W\right)
\end{aligned}
$$

Question. How long does it take to reach a marked vertex?

## Classical hitting time

Take $t$ steps of the walk:

$$
\begin{aligned}
\left(W^{\prime}\right)^{t} & =\left(\begin{array}{cc}
W_{M}^{t} & 0 \\
V\left(I+W_{M}+\cdots+W_{M}^{t-1}\right) & I
\end{array}\right) \\
& =\left(\begin{array}{cc}
W_{M}^{t} & 0 \\
V \frac{I-W_{M}^{t}}{I-W_{M}} & I
\end{array}\right)
\end{aligned}
$$

Convergence time depends on how close $\left\|W_{M}\right\|$ is to 1 , which depends on the spectrum of $W$

## Lemma

Let $W=W^{T}$ be a symmetric Markov chain. Let the second largest eigenvalue of $W$ be $1-\delta$, and let $\epsilon=|M| /|V|$ (the fraction of marked items). Then the probability of reaching a marked vertex is $\Omega(1)$ after $t=O(1 / \delta \epsilon)$ steps of the walk.

## Quantum walk search algorithm

Start from the state $\frac{1}{\sqrt{N-|M|}} \sum_{v \notin M}\left|\psi_{v}\right\rangle$
Consider the walk $U$ corresponding to $W^{\prime}$ :

$$
\sum_{v, w \in V} \sqrt{W_{v, w}^{\prime} W_{w, v}^{\prime}}|w\rangle\langle v|=\left(\begin{array}{cc}
W_{M} & 0 \\
0 & 1
\end{array}\right)
$$

Eigenvalues of $U$ are $e^{ \pm i \arccos \lambda}$ where the $\lambda$ are eigenvalues of $W_{M}$
Perform phase estimation on $U$ with precision $O(\sqrt{\delta \epsilon})$

- no marked items $\Longrightarrow$ estimated phase is 0
- $\epsilon$ fraction of marked items $\Longrightarrow$ nonzero phase with probability $\Omega(1)$

Further refinements give algorithms for finding a marked item

## Grover's algorithm revisited

## Problem

Given a black box $f: X \rightarrow\{0,1\}$, is there an $x$ with $f(x)=1$ ?
Markov chain on $N=|X|$ vertices:

$$
W:=\frac{1}{N}\left(\begin{array}{ccc}
1 & \cdots & 1 \\
\vdots & \ddots & \vdots \\
1 & \cdots & 1
\end{array}\right)=|\psi\rangle\langle\psi|, \quad|\psi\rangle:=\frac{1}{\sqrt{N}} \sum_{x \in X}|x\rangle
$$

Eigenvalues of $W$ are $0,1 \Longrightarrow \delta=1$
Hard case: one marked vertex, $\epsilon=1 / N$
Hitting times

- Classical: $O(1 / \delta \epsilon)=O(N)$
- Quantum: $O(1 / \sqrt{\delta \epsilon})=O(\sqrt{N})$


## Element distinctness

## Problem

Given a black box $f: X \rightarrow Y$, are there distinct $x, x^{\prime}$ with $f(x)=f\left(x^{\prime}\right)$ ?

Let $N=|X|$; classical query complexity is $\Omega(N)$
Consider a quantum walk on the Hamming graph $H(N, M)$

- Vertices: $\left\{\left(x_{1}, \ldots, x_{M}\right): x_{i} \in X\right\}$
- Store the values $\left(f\left(x_{1}\right), \ldots, f\left(x_{M}\right)\right)$ at vertex $\left(x_{1}, \ldots, x_{M}\right)$
- Edges between vertices that differ in exactly one coordinate


## Element distinctness: Analysis

Spectral gap: $\delta=O(1 / M)$
Fraction of marked vertices: $\epsilon \geq\binom{ N-2}{M-2} /\binom{N}{M}=\Theta\left(M^{2} / N^{2}\right)$
Quantum hitting time: $O(1 / \sqrt{\delta \epsilon})=O(N / \sqrt{M})$
Quantum query complexity:

- $M$ queries to prepare the initial state
- 2 queries for each step of the walk (compute $f$, uncompute $f$ )
- Overall: $M+O(N / \sqrt{M})$

Choose $M=N^{2 / 3}$ : query complexity is $O\left(N^{2 / 3}\right) \quad$ (optimal!)

## Quantum walk algorithms

Quantum walk search algorithms

- Spatial search
- Finding a triangle in a graph
- Checking matrix multiplication
- Testing if a black-box group is abelian

Evaluating Boolean formulas
Exponential speedup for a natural problem?

