7. Quantum algorithms (part 2)

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Differences from factoring algorithm:

- intuitive
- easily visualized
- very little analysis needed

Discussion will be relatively brief.

(ii) Optimality of Grover's algorithm (NC 6.6, KLM 9.3)

Given: $N \in \mathbb{N}$

black box for a function f: $\{1,...,N\} \rightarrow \{0,1\}$

Problem: determine if there is an x s.t. f(x) = 1.

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e.g., 3-SAT. Each instance of size n is a formula of n binary variables and poly(n) clauses:

 $(X_1 \vee \neg X_2) \land (X_2 \vee \neg X_3 \vee X_8) \land \dots =: f(X)$

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Goal: determine if there is an $X = X_1 X_2 \cdots X_N$ with f(x) = 1 (x is called a "satisfying assignment").

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For each x, checking if f(x) = 1 takes poly(n)-time, and is modeled by a query to the blackbox.

Given: $N \in \mathbb{N}$

black box for a function f:{1,...,N} \rightarrow {0,1}

M = # of marked items.

Problem: find an x s.t. f(x) = 1. (a "marked" item)

Variation 3: M is unknown.

Given: $N \in \mathbb{N}$

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Variation 3: M is unknown.

Motivation II: This models database search.

e.g., given a phone book sorted by names and a specific phone number, find whose number it is. Here, M=1, N = # of entries in the phone book.

Focus on variation 2 for now.

Given: $N \in \mathbb{N}$

black box for a function f:{1,...,N} \rightarrow {0,1}

M = # of marked items.

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Classical query complexity: $\Omega(\frac{N}{M})$ (counting argument)

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Classical query complexity: $\int \left(\frac{N}{M}\right)$ (counting argument) Quantum query complexity: $\int \left(\int_{M}^{N}\right)$ (Grover's algorithm)

NB. Quantum is advantageous only when the fraction of marked items is vanishing (needle in a haystack).

Claim: classical query complexity: $\Omega(\frac{N}{M})$

Proof: With M marked items among N, probability not seeing a marked item after t queries

 $= \frac{N-M}{N} \frac{N-M-1}{N-1} \frac{N-M-2}{N-2} \dots \frac{N-M-t+1}{N-t+1}$

Claim: classical query complexity: $\Omega(\frac{N}{M})$

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$$= \frac{N-M}{N} \frac{N-M-I}{N-I} \frac{N-M-2}{N-2} \cdots \frac{N-M-t+I}{N-t+I}$$
$$= \left(I - \frac{M}{N}\right) \left(I - \frac{M}{N-I}\right) \left(I - \frac{M}{N-2}\right) \cdots \left(I - \frac{M}{N-t+I}\right)$$
$$\ge \left(I - \frac{M}{N-t+I}\right)^{t}$$

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$$= \frac{N-M}{N} \frac{N-M-1}{N-1} \frac{N-M-2}{N-2} \cdots \frac{N-M-t+1}{N-t+1}$$

$$= \left(1-\frac{M}{N}\right) \left(1-\frac{M}{N-1}\right) \left(1-\frac{M}{N-2}\right) \cdots \left(1-\frac{M}{N-t+1}\right)$$

$$\geq \left(1-\frac{M}{N-t+1}\right)^{t} \qquad \text{recall } e^{X} = \lim_{k \to \infty} \left(1+\frac{X}{k}\right)^{k}$$

$$= \left(1+\frac{(-1)}{N-t+1}\right)^{\frac{N-t+1}{M} \cdot \frac{M}{N-t+1} t}$$

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 $\approx \left(\frac{1}{\rho}\right)^{\frac{M}{N-t+i}t}$

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 $\approx \left(\frac{1}{e}\right)^{\frac{M}{N-t+i}t} \approx | \text{ unless exponential is far from 0}$ i.e., $Mt = \Omega(N)$

 $t \sim \mathcal{N}\left(\frac{N}{M}\right)$ queries are needed.

Let $|\Psi\rangle = \frac{1}{\sqrt{N}} \sum_{x=1}^{N} |x\rangle$, $V = 2 |\Psi\rangle\langle\Psi| - I$

"reflection" about $|\Psi\rangle$ $\forall |\Psi\rangle = (2|\Psi\rangle\langle\Psi| - I)|\Psi\rangle$ $= 2|\Psi\rangle - |\Psi\rangle = |\Psi\rangle$ $\forall |\Phi\rangle \perp |\Psi\rangle$ $\forall |\Phi\rangle = (2|\Psi\rangle\langle\Psi| - I)|\Phi\rangle = -|\Phi\rangle$

Let $|\Psi\rangle = \frac{1}{\sqrt{N}} \sum_{x=1}^{N} |x\rangle$, $\forall = 2 |\Psi\rangle\langle\Psi| - I$ Blackbox: $U_{f} |x\rangle|\Psi\rangle = |x\rangle|\Psi \oplus f(x)\rangle$ Phase kick back: $U_{f} |x\rangle|-\rangle = (-1)^{f(x)}|x\rangle|-\rangle$

Let
$$|\Psi\rangle = \frac{1}{10} \sum_{x=1}^{N} |x\rangle$$
, $V = 2 |\Psi\rangle\langle\Psi| - I$
Blackbox: $U_{f} |x\rangle| |\Psi\rangle = |x\rangle| |\Psi \oplus f(x)\rangle$
Phase kick back: $U_{f} |x\rangle| - \rangle = (-1)^{f(x)} |x\rangle| - \rangle$

1. Initialize state to $|\psi\rangle|-\rangle$

Let $|\Psi\rangle = \frac{1}{|N|} \sum_{x=1}^{N} |x\rangle$, $V = 2 |\Psi\rangle\langle\Psi| - I$ Blackbox: $U_{f} |x\rangle|y\rangle = |x\rangle|y \oplus f(x)\rangle$ Phase kick back: $U_{f} |x\rangle|-\rangle = (-1)^{f(x)}|x\rangle|-\rangle$

- **1.** Initialize state to $|\Psi\rangle|-\rangle$
- 2. Apply Grover's iteration $G = (\lor \otimes I) U_f$ k times, for k to be determined.

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- 3. Measure 1st register in the computational basis.
- 4. Check if the measurement outcome is a marked item by using $\ U_{f}$.

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2b. Apply $\bigvee = 2 |\Psi\rangle\langle\Psi| - I$ to 1st register : $(\bigvee \otimes I) \bigcup_{f} |\Psi\rangle |-\rangle = (\bigvee \otimes I) (|\Psi\rangle - |3\rangle) |-\rangle$

Let 3 be the marked item $(f(3)=1, f(x)=0 \text{ if } x\neq 3)$. **1. Initial state:** $|\Psi\rangle|-\rangle = \frac{1}{2}(11)+(2)+(3)+(4)(-)$ 2a. Apply $U_f: U_f(\Psi) \to = \pm ((1) + 12) - (3) + (4) = \pm (1) + 12 = \pm (1) + 12 = -13 + 14$ $= (|\Psi\rangle - |3\rangle) |-\rangle$ 2b. Apply $\bigvee = 2 |\psi\rangle\langle\psi| - 1$ to 1st register : $(\bigvee \otimes I) \bigcup_{f} |\Psi\rangle |-\rangle = (\bigvee \otimes I) (|\Psi\rangle - |3\rangle) |-\rangle$ $= \bigvee (|\Psi\rangle - |3\rangle) \otimes |-\rangle$ $= (2|\psi\rangle\langle\psi| - I)(|\psi\rangle - |3\rangle) \otimes |-\rangle$

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Observations from the example:

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- 2. Due to phase-kick back, 2nd register is always in the state $|-\rangle$. Grover's iteration acts on 1st register as $\sqrt{\tilde{u}_{f}}$ where $\tilde{u}_{f}|\times\rangle = (-1)^{f(\times)}|\times\rangle$.
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1. 2nd register stays in the state $|-\rangle$ throughout. 1st register is evolved by $\widetilde{G} = \bigvee \widetilde{U}_{f}$, where $\widetilde{U}_{f} | \times \rangle = (-1)^{f(\times)} | \times \rangle$.

2. 1st register stays in the span of:

$$|z\rangle = \frac{1}{\sqrt{M}} \sum_{x:f(x)=1} |x\rangle$$
$$|\beta\rangle = \frac{1}{\sqrt{N-M}} \sum_{x:f(x)=0} |x\rangle$$

 $|\downarrow\rangle\rangle$: equal superposition of all marked items $|\beta\rangle\rangle$: equal superposition of all unmarked items

2. 1st register stays in the span of:

$$|\beta\rangle = \frac{1}{\sqrt{N-M}} \sum_{x:f(x)=0} |x\rangle \quad (|\beta\rangle = \frac{1}{\sqrt{3}} (|1\rangle + |2\rangle + |4\rangle)$$

in example)

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in example)

 $|\downarrow\rangle\rangle$: equal superposition of all marked items $|\beta\rangle\rangle$: equal superposition of all unmarked items

Intuition: symmetry among all marked items, and symmetry among all unmarked items.

2. 1st register stays in the span of:

$$|z\rangle = \frac{1}{100} \sum_{x:f(x)=1} |x\rangle, |\beta\rangle = \frac{1}{1000} \sum_{x:f(x)=0} |x\rangle$$

Initial state =
$$|\Psi\rangle = \frac{1}{\sqrt{N}} \sum_{x} \langle x \rangle$$

2. 1st register stays in the span of:

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= $\frac{1}{\sqrt{N}} \left(\sum_{X:f(X)=1} |X\rangle + \sum_{X:f(X)=0} |X\rangle \right)$

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Initial state =
$$|\Psi\rangle = \frac{1}{\sqrt{2}} \sum_{x} |x\rangle$$

$$=\frac{1}{\int N} \left(\sum_{\mathbf{x}: f(\mathbf{x})=1} |\mathbf{x}\rangle + \sum_{\mathbf{x}: f(\mathbf{x})=0} |\mathbf{x}\rangle \right)$$

$$=\frac{\sqrt{M}}{\sqrt{N}} \left(2 \right) + \frac{\sqrt{N-M}}{\sqrt{N}} \left(2 \right)$$

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$$=\frac{\sqrt{M}}{\sqrt{N}} \frac{1}{\sqrt{N}} + \frac{\sqrt{N-M}}{\sqrt{N}} \frac{1}{\sqrt{N}}$$

NB.
$$|z\rangle \perp |\beta\rangle, \langle \Psi|z\rangle = \int_{N}^{M}, \langle \Psi|\beta\rangle = \int_{N}^{N-M}$$

Initial state =
$$|\Psi\rangle = \frac{\sqrt{M}}{\sqrt{N}} |\omega\rangle + \frac{\sqrt{N-M}}{\sqrt{N}} |\beta\rangle$$
.

$$M_{f}: 1d \rightarrow -1d \rightarrow 1\beta \rightarrow 1\beta$$

$$\begin{aligned} \text{Initial state} &= |\Psi\rangle = \frac{\sqrt{M}}{\sqrt{N}} |\lambda\rangle + \frac{\sqrt{N-M}}{\sqrt{N}} |\beta\rangle. \\ \widetilde{U}_{f} &: |\lambda\rangle \rightarrow -|\lambda\rangle, \ |\beta\rangle \rightarrow |\beta\rangle \\ \bigvee &= 2 |\Psi\rangle\langle\Psi| - I, \quad \bigvee |\lambda\rangle = 2 \langle\Psi|\lambda\rangle \ |\Psi\rangle - |\lambda\rangle \end{aligned}$$

Initial state = $|\Psi\rangle = \frac{\sqrt{M}}{\sqrt{N}} |\lambda\rangle + \frac{\sqrt{N-M}}{\sqrt{N}} |\beta\rangle$. $\widetilde{U}_{f} : |\lambda\rangle \rightarrow -|\lambda\rangle, |\beta\rangle \rightarrow |\beta\rangle$ $\bigvee = 2|\Psi\rangle\langle\Psi| - I, \quad \bigvee |\lambda\rangle = 2\langle\Psi|\lambda\rangle |\Psi\rangle - |\lambda\rangle$ $\bigvee |\beta\rangle = 2\langle\Psi|\beta\rangle |\Psi\rangle - |\beta\rangle$

Initial state =
$$|\Psi\rangle = \frac{\sqrt{M}}{\sqrt{N}} |\omega\rangle + \frac{\sqrt{N-M}}{\sqrt{N}} |\beta\rangle$$
.
 $\widetilde{U}_{f} : |\omega\rangle \rightarrow -|\omega\rangle, |\beta\rangle \rightarrow |\beta\rangle$
 $V = 2|\Psi\rangle\langle\Psi| - I, \quad V|\omega\rangle = 2\langle\Psi|\omega\rangle |\Psi\rangle - |\omega\rangle$
 $V|\beta\rangle = 2\langle\Psi|\beta\rangle |\Psi\rangle - |\beta\rangle$
Since $|\Psi\rangle$ is in the span of $|\omega\rangle, |\beta\rangle$.

Initial state =
$$|\Psi\rangle = \frac{\sqrt{m}}{\sqrt{n}} |\lambda\rangle + \frac{\sqrt{n-m}}{\sqrt{n}} |\beta\rangle$$
.
 $\widetilde{U}_{f} : |\lambda\rangle \rightarrow -|\lambda\rangle, |\beta\rangle \rightarrow |\beta\rangle$
 $V = 2|\Psi\rangle\langle\Psi| - I, \quad V|\lambda\rangle = 2\langle\Psi|\lambda\rangle |\Psi\rangle - |\lambda\rangle$
 $V|\beta\rangle = 2\langle\Psi|\beta\rangle |\Psi\rangle - |\beta\rangle$
Since $|\Psi\rangle$ is in the span of
 $|\lambda\rangle, |\beta\rangle$, so are $V|\lambda\rangle, V|\beta\rangle$.
So, each of V and \widetilde{U}_{f} preserves the span of $|\lambda\rangle, |\beta\rangle$.

Initial state =
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Since $|\Psi\rangle$ is in the span of
 $|\lambda\rangle, |\beta\rangle$, so are $V|\lambda\rangle, V|\beta\rangle$.
So, each of V and \widetilde{U}_{f} preserves the span of $|\lambda\rangle, |\beta\rangle$.
Algorithm starts with $|\Psi\rangle$ (in the span of $|\lambda\rangle, |\beta\rangle$)

Algorithm starts with $|\Psi\rangle$ (in the span of $|\lambda\rangle, (\beta)$) and applies $\forall \widetilde{M_{+}}$ k times. So, 1st register is always in the span of $|\lambda\rangle, (\beta)$.

<u>Analysis of Grover's algorithm</u>:

We can restrict the analysis to the span of $\{\lambda\}, [\beta]$. Initial state = $|\Psi\rangle = \frac{\sqrt{M}}{\sqrt{N}} |\lambda\rangle + \frac{\sqrt{N-M}}{\sqrt{N}} |\beta\rangle$.

What does Grover's iteration $\bigvee \widetilde{\mathfrak{U}}_{\mathfrak{f}}$ do?

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What does Grover's iteration $\bigvee \widetilde{\mathfrak{U}}_{\mathfrak{f}}$ do?

By linearity, suffices to check its action on a spanning set: $|\beta\rangle$, $|\Psi\rangle$.



Each marked angle is $\theta/2$ where $\sin \frac{\theta}{2} = \int_{N}^{M}$.

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Each marked angle is 9/2where $\sin \frac{9}{2} = \int_{N}^{M}$.

For the initial state $|\beta\rangle$

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For the initial state *β* rotation of angle θ

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For the initial state \β) rotation of angle θ

Each marked $\left(\mathcal{L}\right)$ angle is 9/2 where $\sin \frac{\Theta}{2} = \int \frac{M}{N}$. 14) $|\beta\rangle$ For the initial state $|\Psi\rangle$ rotation of angle θ

By linearity, suffices to check its action on a spanning set: $|\beta\rangle$, $|\Psi\rangle$.



 $\langle , \ \rangle \widetilde{\mathcal{U}}_{f}$ is a <u>rotation</u> of angle θ in the $(\beta), (z)$ plane. 2 reflections (anti-clockwise) make a rotation!



 $\chi', \forall \widetilde{\mathcal{U}}_{f}$ is a rotation of angle θ in the (ه), (ع) plane. (anti-clockwise)

Optimal # of Grover's iteration

Goal: rotate $|\Psi\rangle$ to as close to $|\lambda\rangle$ as possible



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(meas [J]) in comp basis gives an outcome that is a marked item.) superposition of marked states Each marked (2) angle is 9/2 where $\sin \frac{\Theta}{2} = \prod_{N}^{M}$. initial state

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Optimal # of Grover's iteration

Goal: rotate $|\Psi\rangle$ to as close to $|\lambda\rangle$ as possible

(meas () in comp basis gives an outcome that is a marked item.)

After k iterations, state is $(k + \frac{1}{2}) \Theta$ from the $|\beta\rangle$ axis. Want $(k + \frac{1}{2}) \Theta \approx \frac{\pi}{2}$. superposition of marked states Each marked $\left(\mathcal{A}\right)$ angle is 9/2 where $\sin \frac{\Theta}{2} = \frac{M}{N}$. (Uf14) initial state →Iβ)

 $\chi', \forall \widetilde{\mathcal{U}}_{f}$ is a rotation of angle θ in the (م), (2) plane. (anti-clockwise)

Want
$$\left(K + \frac{1}{2}\right) \otimes \approx \frac{\pi}{2}$$
.
Recall $S_{in} \frac{\Theta}{2} = \int_{N}^{M}$ is very small, so, $\frac{\Theta}{2} \approx \int_{N}^{M}$.

Want
$$\left(K + \frac{1}{2} \right) \otimes \approx \frac{\pi}{2}$$
.
Recall $S_{1n} \frac{\Theta}{2} = \int \frac{M}{N}$ is very small, so, $\frac{\Theta}{2} \approx \int \frac{M}{N}$.
Solving $\left(K + \frac{1}{2} \right) 2 \int \frac{M}{N} \approx \frac{\pi}{2}$.
 $K \approx \frac{\pi}{4} \int \frac{N}{M} - \frac{1}{2}$.

We take k to be the integer closest to $\frac{\pi}{4}\int_{M}^{N} - \frac{1}{2}$.

Grover's algorithm:

Let $|\Psi\rangle = \frac{1}{\sqrt{N}} \sum_{x=1}^{N} |x\rangle$, $\forall = 2 |\Psi\rangle\langle\Psi| - I$ Blackbox: $U_{f} |x\rangle|_{Y} = |x\rangle|_{Y} \oplus f(x)$ Phase kick back: $U_{f} |x\rangle|_{-} = (-1)^{f(x)} |x\rangle|_{-}$

- **1.** Initialize state to $|\Psi\rangle|-\rangle$
- 2. Apply Grover's iteration $G = (\lor \otimes I) U_f$ k times, for k to be determined.
- 3. Measure 1st register in the computational basis.
- 4. Check if the measurement outcome is a marked item by using U_{f} . Yes with prob close to 1.

Repeat t = O(1) times, prob failure $\sim \exp(-t)$.

<u>Summary:</u>

We proved that quantum query complexity of the unstructured search problem (variation 2) is $O(\overline{\mathbb{N}})$.

Optimality: part (ii) of topic07-2

Further question:

What is the circuit complexity of the algorithm?

Circuit complexity of Grover's algorithm

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For simplicity, N = 2^n.
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State initialization: $(n+1) \ 0$ states, apply X to the last qubit, and then apply Hadamard gates to all.

Circuit complexity of Grover's algorithm

For simplicity, $N = 2^n$.

State initialization: $(n+1) \ b$ states, apply X to the last qubit, and then apply Hadamard gates to all.

Computational basis measurement: n "individual-qubit" measurements along \0>, \1>

Remains to implement $V \simeq 2 | \forall \times \forall I - I$.

Implementing V

Lemma:
$$V = 2 [\forall X \forall | -I = H^{\otimes n} (2 | \circ X \circ | -I) H^{\otimes n}.$$

Proof:
Lemma:
$$V = 2 [YXY| - I = H^{\otimes n} (2 | 0 \times 0 | - I) H^{\otimes n}$$
.
Proof:

Since
$$H^{\otimes n} | o \rangle = [\Psi \rangle$$
,
 $H^{\otimes n} | o \rangle \langle o | H^{\otimes n} = [\Psi \rangle \Psi \Psi$.

Lemma:
$$V = 2 | \forall X \forall | -I = H^{\otimes n} (2 | \circ X \circ | -I) H^{\otimes n}$$
.
Proof:

Since $H^{\otimes n} | o \rangle = | \Psi \rangle$, $H^{\otimes n} | o \rangle \langle o | H^{\otimes n} = | \Psi \times \Psi |$.

$$\sum_{n=2}^{\infty} |V| = 2 |Y| \times |V| - I$$

= 2 H^{@n} |0 × 0| H^{@n} - I

Lemma:
$$V = 2 | \forall X \forall | -I = H^{\otimes n} (2 | \circ X \circ | -I) H^{\otimes n}$$
.
Proof:

Since $H^{\otimes n} | o \rangle = | \Psi \rangle$, $H^{\otimes n} | o \rangle \langle o | H^{\otimes n} = | \Psi \times \Psi |$. $\downarrow, \forall = 2 | \Psi \times \Psi | - I$ $= 2 H^{\otimes n} | o \rangle \langle o | H^{\otimes n} - I$ $= H^{\otimes n} (2 | o \rangle \langle o | - I) H^{\otimes n}$ $\downarrow, (H^{\otimes n})^{2} = I$

Lemma:
$$V = 2|\forall X \forall | -I = H^{\otimes n} (2|0 \times 0| -I) H^{\otimes n}$$
.
Proof:

Since $H^{\otimes n} | o \rangle = | \Psi \rangle$, $H^{\otimes n} | o \rangle \langle o | H^{\otimes n} = | \Psi \rangle \Psi |$. $\vdots, \forall = 2 | \Psi \rangle \Psi | - I$ $= 2 H^{\otimes n} | o \rangle \langle o | H^{\otimes n} - I$ $= H^{\otimes n} (2 | o \rangle \langle o | - I) H^{\otimes n}$ $\vdots (H^{\otimes n})^2 = I$

So we can implement V by applying n Hadamard gates, then $2 \log |-1$, and n Hadamard gates again.

This gate takes |0> to |0>, and |x> to |x> on all other computational basis states.

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Same as negating all n bits, then mapping |1..1> to -|1...1> & keeping all other |x>'s the same (this is a control-control-...-control-Z), and finally negating all n bits again.

<u>Implementing $2 \mid 0 \mid 0 \mid - \underline{T}$ up to a "-" sign:</u>



<u>Implementing 2\oXol - I up to a "-" sign:</u>



Figure 4.10. Network implementing the $C^{n}(U)$ operation, for the case n = 5.

from NC

<u>Implementing 2\oXol - I up to a "-" sign:</u>



Figure 4.10. Network implementing the $C^n(U)$ operation, for the case n = 5.

Figure 4.9. Implementation of the Toffoli gate using Hadamard, phase, controlled-NOT and $\pi/8$ gates.

from NC

<u>Implementing $V = 2|\Psi \times |-1|$ up to a "-" sign:</u>



Figure 4.10. Network implementing the $C^n(U)$ operation, for the case n = 5.

Figure 4.9. Implementation of the Toffoli gate using Hadamard, phase, controlled-NOT and $\pi/8$ gates.

<u>Grover algorithm summary:</u>



<u>What if M (the # marked items) is unknown?</u> (vars 1,3)

Use a quantum algorithm to estimate M using $()(\mathbb{N})$ queries, with accuracy $()(\mathbb{N})$.

Such algorithms can be Grover like or based on phase estimation. (Reading exercise)

Alternative: trying M=1,2,4,8,... etc works too. You see the marked item in the final check if and only if M is about right.

7. Quantum algorithms (part 2)

/ (i) Grover's search algorithm (NC 6.1, KLM 8.1-8.2, M 4)

Differences from factoring algorithm:

- intuitive
- easily visualized
- very little analysis needed

Discussion will be relatively brief.

 \rightarrow (ii) Optimality of Grover's algorithm (NC 6.6, KLM 9.3)

Optimality of Grover's algorithm

<u>Theorem</u> Given there is either no marked item or a unique marked item, $\Omega(\sqrt{N})$ queries are required to determine which case holds.

<u>Corollary</u> $\Omega(\sqrt{N})$ queries are required to determine if there is a marked item, or to find one.

Proof: Most general algorithm with T queries



The "no marked item case" also corresponds to Uf = I.

Let $| \Upsilon_t \rangle$ be the state before the (t+1)-st query, in the absence of marked items.

Let $| \mathcal{V}_{t}^{\times} \rangle$ be the state before the (t+1)-st query, if the marked item is x.

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Recall Holevo-Helstrom Theorem (topic04) that two non-orthogonal states (a, b) are hard to distinguish if $||a\rangle - |b\rangle||$ is too small.

If $|a\rangle, |b\rangle$ are somewhat distinguishable, it is necessary that $||a\rangle-|b\rangle|| \ge c$ (some constant).

Let $|\Upsilon_t\rangle$ be the state before the (t+1)-st query, in the absence of marked items.

Let $| \mathcal{V}_{t}^{x} \rangle$ be the state before the (t+1)-st query, if the marked item is x.

If t queries enable us to determine if there is a marked item, it holds that: $\| | \psi_t \rangle - | \psi_t^{\times} \rangle \| \ge C$.

Let $| \Upsilon_t \rangle$ be the state before the (t+1)-st query, in the absence of marked items.

Let $| \mathcal{V}_{t}^{\times} \rangle$ be the state before the (t+1)-st query, if the marked item is x.

If t queries enable us to determine if there is a marked item, it holds that: $\| || \psi_t \rangle - || \psi_t^* \rangle \| \ge C$.

When we say "the algorithm works", it works for any marked item x, so:

$$\mathcal{D}_{t} := \sum_{\chi=1}^{N} \| | | \psi_{t} \rangle - | \psi_{t}^{\chi} \rangle \| \ge C N .$$
(Similar if alg works for average case input.)

How does \mathcal{P}_{t} change with each query?

- 1

For one x:

$\| | | \Psi_{t+1} \rangle - | \Psi_{t+1}^{sc} \rangle \|$ $= \| V_t | \Psi_t \rangle - V_t U_t | \Psi_t^{sc} \rangle \|$

For one x:

$$\begin{cases} \|W|a\rangle - W|b\rangle\|^{2} \\ = (\langle a|W^{t} - \langle b|W^{t}\rangle (W|a) - W|b\rangle) \\ = (\langle a| - \langle b|\rangle (|a\rangle - |b\rangle) \\ = \||a\rangle - |b\rangle\|^{2} \end{cases}$$

For one x:

 $\||| || \langle | | | \rangle - || \langle | | \rangle \|$ $\begin{cases} \|W^{(a)} - W^{(b)}\|^{2} \\ = (\langle a|W^{t} - \langle b|W^{t}\rangle (W^{(a)} - W^{(b)}) \\ = (\langle a| - \langle b|\rangle (|a\rangle - |b\rangle) \\ = \||a\rangle - |b\rangle\|^{2} \end{cases}$ $= \| V_t | \Psi_t \rangle - V_t U_t | \Psi_t^{\infty} \rangle \|$ $= \| | \Psi_t \rangle - | \Psi_t^{sc} \rangle + | \Psi_t^{sc} \rangle - | \Psi_t^{sc} \rangle \|$ $= \| | \Psi_{t} \rangle - | \Psi_{t}^{x} \rangle \| + \| | \Psi_{t}^{x} \rangle - \| \Psi_{t}^{x} | \Psi_{t}^{x} \rangle \|$ by ∆ inequality bound this diff induced by recursively 1 use of \mathbb{V}_{f} on \mathbb{V}_{t}^{∞}

Let
$$|\Psi_{t}^{x}\rangle = \sum_{y=1}^{N} \langle y_{y}, t | y \rangle | \phi_{y}^{t}\rangle$$
. Use method 2 to express
bipartite state, topic03-02, p7.

Let
$$|\Psi_{t}^{\prime}\rangle = \sum_{y=1}^{N} \langle y_{y,t} | y \rangle | \phi_{y}^{t}\rangle$$
. Use method 2 to express bipartite state, topic03-02, p7.

computational basis on input to Uf register not acted on by Uf

$$\begin{split} \mathcal{U}_{f} | \mathcal{Y}_{t}^{x} \rangle - | \mathcal{Y}_{t}^{x} \rangle &= (-2Ix \times I + I) | \mathcal{Y}_{t}^{x} \rangle - | \mathcal{Y}_{t}^{x} \rangle \\ &= -2Ix \times I | \mathcal{Y}_{t}^{x} \rangle \\ &= -2Ix \times I | \mathcal{Y}_{t}^{x} \rangle \end{split}$$

WLOG, Uf used with phase kick-back (use blackboard).

Let
$$|\Psi_{t}^{x}\rangle = \sum_{y=1}^{N} a_{y,t} |y\rangle| \phi_{y}^{t}\rangle$$
. Use method 2 to express
bipartite state, topic03-02, p7.
computational basis on input to Uf register not acted on by Uf
 $U_{t} |\Psi_{t}^{x}\rangle - |\Psi_{t}^{x}\rangle = (-2ixxi1+I) |\Psi_{t}^{x}\rangle - |\Psi_{t}^{x}\rangle$ WLOG, Uf used with phase
kick-back (use blackboard).
 $= -2ixxi1 |\Psi_{t}^{x}\rangle$
 $= -2ix\rangle dx_{it} |\phi_{x}^{t}\rangle$
 $||U_{t} |\Psi_{t}^{x}\rangle - |\Psi_{t}^{x}\rangle|| = 2 |dx_{it}|$
 $\int_{1}^{1} ||\Psi_{t+1}\rangle - |\Psi_{t+1}^{x}\rangle|| \leq ||\Psi_{t}\rangle - |\Psi_{t}^{x}\rangle|| + |||\Psi_{t}^{x}\rangle - U_{t} |\Psi_{t}^{x}\rangle||$
 $= ||\Psi_{t}\rangle - |\Psi_{t}^{x}\rangle|| + 2 |dx_{it}|$
 $\leq 2 \sum_{j=0}^{t} |dx_{ij}|| (re cursive argument)$

$$CN \leq \sum_{j=1}^{N} || |\Psi_{T}\rangle - |\Psi_{T}^{*}\rangle||$$

$$\sum_{j=1}^{N} \sum_{j=0}^{T-1} |J_{x,j}||$$

from last page

$$CN \leq \sum_{j=0}^{N} || |Y_{T}\rangle - |Y_{T}^{*}\rangle||$$

$$\sum_{j=0}^{N} \sum_{j=0}^{T-1} |J_{x,j}||$$

$$= 2 \sum_{j=0}^{T-1} |N_{x,j}||$$

from last page

*.*1).

$$CN \leq \sum_{j=0}^{N} || |\Psi_{T}\rangle - |\Psi_{T}^{*}\rangle ||$$

$$\leq \sum_{j=0}^{N} 2 \sum_{j=0}^{T-1} || x_{ij}| \qquad \text{from last page}$$

$$= 2 \sum_{j=0}^{T-1} \sum_{j=0}^{N} || x_{ij}| \qquad (auchy-Schwarzineq)$$

$$= 2 \sum_{j=0}^{T-1} \sum_{j=0}^{N} || x_{ij}|^{2} \qquad (auchy-Schwarzineq)$$

$$= 2 \sum_{j=0}^{T-1} \sum_{j=0}^{N} || x_{ij}|^{2} \qquad Here b = (1,1,...,1).$$

$$= 2 \sum_{j=0}^{T-1} \sum_{j=0}^{N} || x_{ij}|^{2} \qquad 1, by def of || Y_{ij}^{*}\rangle$$

$$CN \leq \sum_{j=0}^{N} || |\Psi_{T}\rangle - |\Psi_{T}^{*}\rangle ||$$

$$\leq \sum_{j=0}^{N} 2 \sum_{j=0}^{T-1} || x_{j}| \qquad \text{from last page}$$

$$= 2 \sum_{j=0}^{T-1} \sum_{j=0}^{N} || x_{j}| \qquad (auchy-Schwarz'ineq)$$

$$= 2 \sum_{j=0}^{T-1} \sqrt{N} \cdot \sqrt{\sum_{\lambda \in I}^{N} || x_{j}|^{2}} \qquad (auchy-Schwarz'ineq)$$

$$= 2 \sum_{j=0}^{T-1} \sqrt{N} \cdot \sqrt{\sum_{\lambda \in I}^{N} || x_{j}|^{2}} \qquad Here b = (1,1,...,1).$$

$$= 2 \sum_{j=0}^{T-1} \sqrt{N} = 2 T \sqrt{N}.$$

 $(T \ge \Omega(TN))$