

# Lecture 6:

## Quantum search algorithm

COMP3366

*Quantum algorithms & computing architecture*

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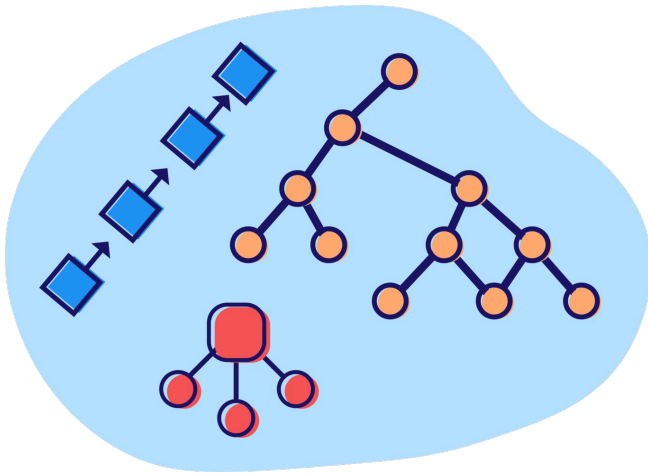
## Objectives:

- **[O1] Concepts:** Quantum search algorithm (Grover's).
- **[O2] Problem solving:** Comprehension of Grover's algorithm, analysis of its performance, the over-cooking issue.
- **[O3] Algorithm design:** Application of phase kickback trick & oracle model in searching, coding your own Grover using Qiskit (Assignment 3).

# **Part I: Search in an unstructured data base**

# Overview

- In some tasks, a quantum computer brings sub-exponential but still very appealing speedups.
- An example: Search an item in an **unstructured** database.



*Structured*



*Unstructured*

- Grover algorithm: quantum runtime  $\approx \sqrt{\text{classical runtime}}$ .

# Oracle model for unstructured search

- Each query to the database is equivalent to one use of the indicator function: (1 = yes and 0 = no):

$$f(x) = \begin{cases} 1 & x = x_0 \\ 0 & x \neq x_0 \end{cases}$$



player

Is the 2<sup>nd</sup> item the  
desired one?

No! ( $f(2) = 0$ )



oracle

$\hat{U} \rightarrow$  apply  $(-1)^{f(x)}$  on  $|x\rangle$

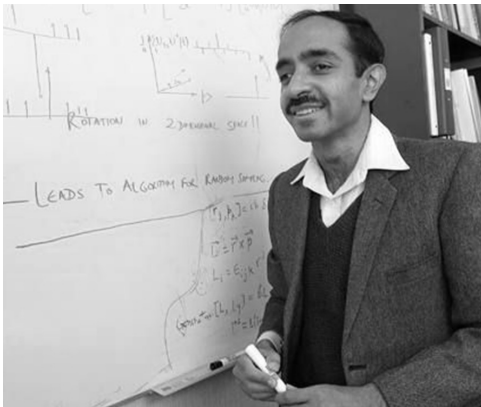
# Classical query complexity

- How many queries to  $f(x)$  does a classical algorithm need?
- As the database is totally unstructured, there is not much it can do.
- Worst-case:  $N - 1$  times!
- Average case: still  $\Omega(N)$ !
- This is a bad (hard) task for classical computers, as there is no clever way to accelerate it.
- Remark: In practice, many database has a good structure, and more efficient search algorithms are possible.
- Can we search faster with a quantum computer?

# **Part II: Grover's algorithm**

# Faster search with a quantum computer

- Search an item in an **unstructured** database of  $N$  items.
- Quantum advantage: quantum runtime  $\approx \sqrt{\text{classical runtime}}$ .



*Lov Grover*

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## Quantum Mechanics Helps in Searching for a Needle in a Haystack

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(Received 4 December 1996)

Quantum mechanics can speed up a range of search applications over unsorted data. For example, imagine a phone directory containing  $N$  names arranged in completely random order. To find someone's phone number with a probability of 50%, any classical algorithm (whether deterministic or probabilistic) will need to access the database a minimum of  $0.5N$  times. Quantum mechanical systems can be in a superposition of states and simultaneously examine multiple names. By properly adjusting the phases of various operations, successful computations reinforce each other while others interfere randomly. As a result, the desired phone number can be obtained in only  $O(\sqrt{N})$  accesses to the database. [S0031-9007(97)03564-3]



# The quantum oracle for search

- Recall: A quantum oracle for  $f$  is a unitary acting on the system  $|x\rangle$  and a qubit ancilla  $|q\rangle$

$$|x\rangle|q\rangle \mapsto |x\rangle|q \oplus f(x)\rangle$$

- For data base search, we have the indicator function

$$f(x) = \begin{cases} 1 & x = x_0 \\ 0 & x \neq x_0 \end{cases}$$

- Recall the “phase kickback” trick for the oracle: Prepare the ancilla in  $|-\rangle$  :
  - If  $f(x) = 0 \Rightarrow |x\rangle|-\rangle \mapsto |x\rangle|-\rangle$
  - If  $f(x) = 1 \Rightarrow |x\rangle|-\rangle \mapsto |x\rangle(-|-\rangle)$

The phase  $\pm 1$  can be kicked onto the system as:

$$O = \sum_x (-1)^{f(x)} |x\rangle\langle x|$$

# Quantum search (Grover's) algorithm

- **Input:**

1. Oracle  $O: |x\rangle|q\rangle \mapsto |x\rangle|q \oplus f(x)\rangle$ . Here  $f(x_0) = 1$  and  $f(x) = 0$  for  $x \neq x_0$ .

- **Output:**

1. An estimate  $\widehat{x}_0$  of  $x_0$ .

- **Query complexity:**  $O(\sqrt{N})$

$n := \lceil \log N \rceil$  is the number of qubits.

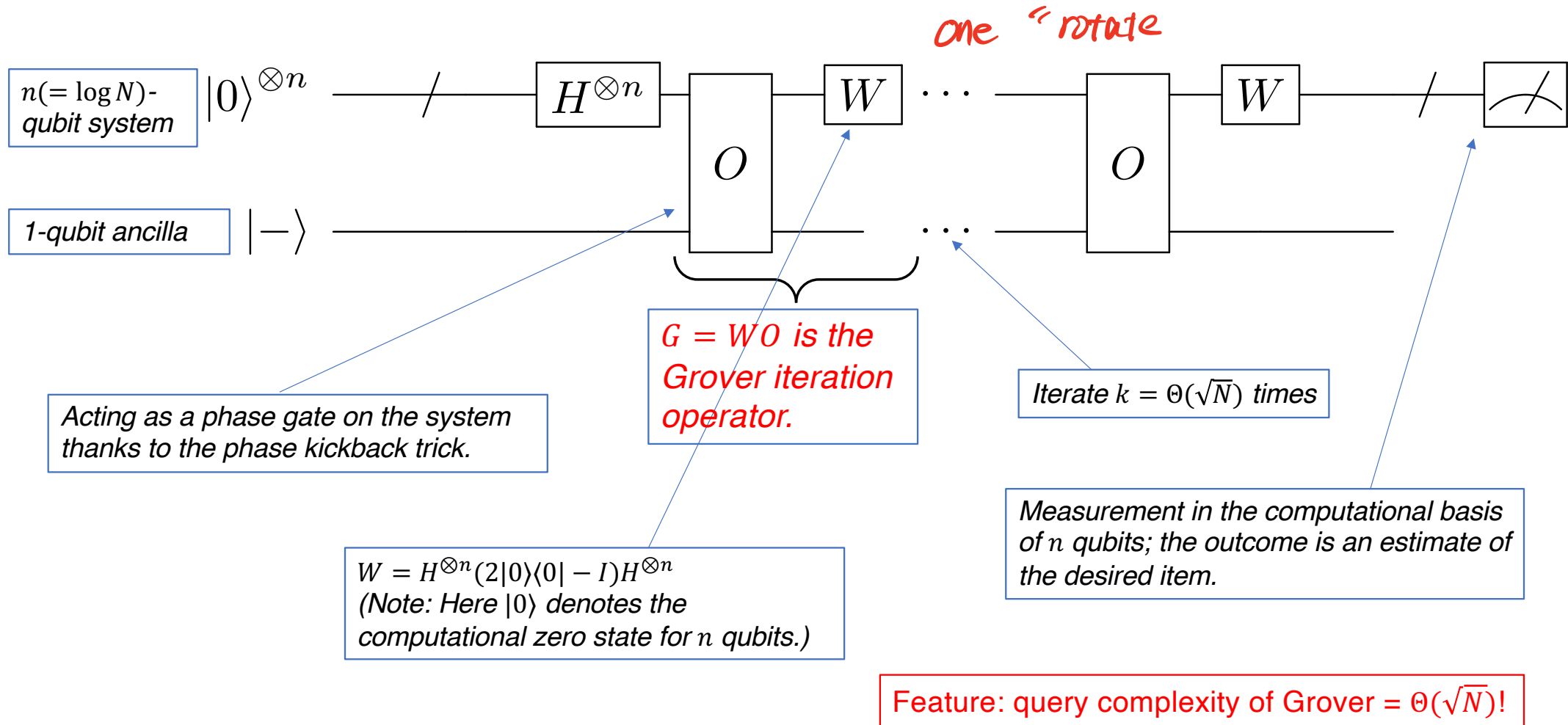
- **Circuit depth:**  $O(n \cdot \sqrt{N})$

- **Accuracy:**  $\widehat{x}_0 = x_0$  with probability  $\approx 1 - N^{-1}$

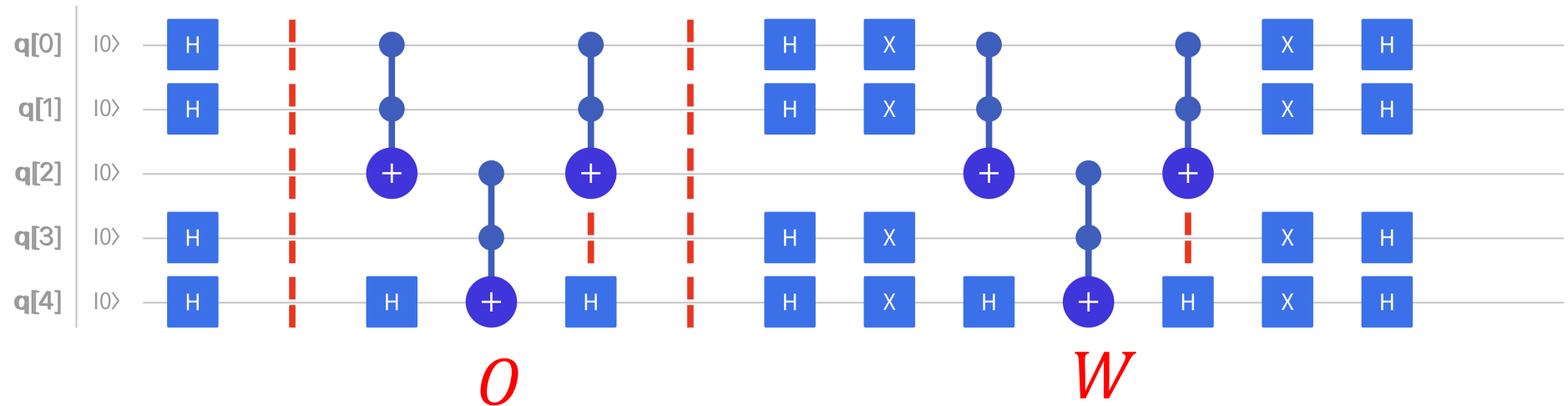
- **Procedure:**

1. Initialize the  $n$ -qubit main register and a qubit ancilla in  $|0\rangle_S |-\rangle_A$
2. Perform Hadamard transform  $H^{\otimes n}$  on  $S$ .
3. **Repeat** the below procedure for  $k \approx \left\lceil \frac{\pi\sqrt{N}}{4} \right\rceil$  times:
  - 1) Apply the oracle  $O$ .
  - 2) Apply  $W = H^{\otimes n}(2|0\rangle\langle 0| - I)H^{\otimes n}$  on  $S$ .
4. Measure  $S$  in the computational basis and **output the outcome as an estimate  $\widehat{x}_0$ .**

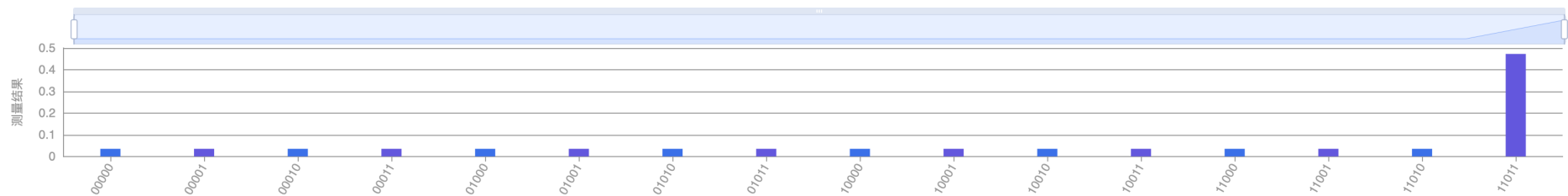
# The circuit of Grover



# Run Grover on OriginQC ( $N = 2^4$ )



Verify the functionality.  
Why 5 qubits? What is the target item?



# **Part III: Working principle of Grover's algorithm**

## Preliminary: Reducing to a 2-D plane

- Initially, the space is  $N$  – dimensional:  
There are  $N$  items  $1, \dots, N$ , each assigned a state  $|1\rangle, \dots, |N\rangle$
- However, there are actually only 2 types of states:  $|x_0\rangle$  and others.
- Observing this symmetry, we can define

$$|x_0^\perp\rangle := \frac{1}{\sqrt{N-1}} \sum_{x \neq x_0} |x\rangle$$

and Grover's algorithm works within the 2-D space spanned by  $\{|x_0\rangle, |x_0^\perp\rangle\}$ .

- The initial state of Grover  $|e_0\rangle = \frac{1}{\sqrt{N}} (|x_0\rangle + \sqrt{N-1}|x_0^\perp\rangle)$ .  
The Grover's oracle has action  $O = (-1)|x_0\rangle\langle x_0| + |x_0^\perp\rangle\langle x_0^\perp|$  in this space.

# Step 1: Grover iteration operator as a rotation

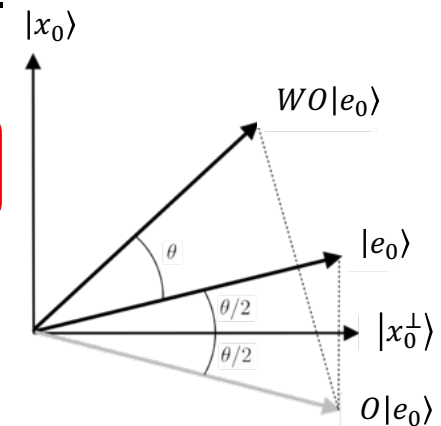
- The spirit of Grover: To rotate  $|e_0\rangle$  towards  $|x_0\rangle$  in the 2-D plane.
- **Problem:** How? We don't know  $x_0$ !
- A bit of geometry: Within a 2D plane **2 reflections = 1 rotation!**
- The oracle is the **1<sup>st</sup> reflection**:

$$O = \sum_x (-1)^{f(x)} |x\rangle\langle x| = -2|x_0\rangle\langle x_0| + I$$

This is a reflection about  $x_0^\perp$ :  $O|x_0^\perp\rangle = |x_0^\perp\rangle$ ,  $O|x_0\rangle = -|x_0\rangle$ .

- $W = H^{\otimes n}(2|0\rangle\langle 0| - I)H^{\otimes n}$  is the **2<sup>nd</sup> reflection**.

(Since  $H, |0\rangle$  are fixed and do not depend on  $x_0$ , we can construct  $W$  without the oracle.)



Reflections in the plane spanned by  $|x_0\rangle, |e_0\rangle$

Exercise:  
Prove that  $W$  is a reflection about  $|e_0\rangle$ .

$$|\psi_0\rangle = |0\rangle^{\otimes n}, \quad n = \lceil \log_2 N \rceil \quad \leftarrow \text{考虑多目标情况.}$$

$$|\psi_1\rangle = H^{\otimes n} |\psi_0\rangle = |+\rangle^{\otimes n} = \frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} |j\rangle \quad \rightarrow \text{为简化, 我们设 } N=2^n$$

$$|x_0\rangle = \sum_{x \in x_0} \frac{1}{\sqrt{m}} |x\rangle, \quad |x_0^\perp\rangle = \frac{1}{\sqrt{N-m}} \sum_{x \notin x_0} |x\rangle$$

$$\text{i.e. } |\psi_1\rangle = \frac{\sqrt{m}}{\sqrt{N}} |x_0\rangle + \frac{\sqrt{N-m}}{\sqrt{N}} |x_0^\perp\rangle = \sin\theta |x_0\rangle + \cos\theta |x_0^\perp\rangle$$

我们考虑 算子  $\{ |x_0\rangle, |x_0^\perp\rangle \}$  !

$$O|x_0\rangle = -|x_0\rangle, \quad O|x_0^\perp\rangle = |x_0^\perp\rangle$$

i.e.  $O$  is “ $-Z$ ” on  $\{ |x_0\rangle, |x_0^\perp\rangle \}$

$$\text{i.e. } O \Rightarrow \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} \text{ on } \{ |x_0\rangle, |x_0^\perp\rangle \}$$

$$W = H^{\otimes n} (2|x_0\rangle\langle x_0| - I) H^{\otimes n} = 2|\psi_1\rangle\langle\psi_1| - I \Rightarrow 2 \begin{pmatrix} \sin^2\theta & \sin\theta\cos\theta \\ \sin\theta\cos\theta & \cos^2\theta \end{pmatrix} - I$$

$$= \begin{pmatrix} -\cos 2\theta & \sin 2\theta \\ \sin 2\theta & \cos 2\theta \end{pmatrix}$$

$\begin{pmatrix} \sin\theta \\ \cos\theta \end{pmatrix}$  on  $\{ |x_0\rangle, |x_0^\perp\rangle \}$ !



$$\therefore \text{Grover 算子 } G = WU = \begin{pmatrix} \cos 2\theta & \sin 2\theta \\ -\sin 2\theta & \cos 2\theta \end{pmatrix} = R(-2\theta)$$

→ 旋转矩阵

$$\text{i.e. } G|u_1\rangle = R(-2\theta) \cdot (\sin\theta|x_0\rangle + \cos\theta|x_0^\perp\rangle) = \begin{pmatrix} \sin 3\theta \\ \cos 3\theta \end{pmatrix}$$

经历  $k$  次  $G$ , 得到:

$$\sin((k+1)\theta)|x_0\rangle + \cos((k+1)\theta)|x_0^\perp\rangle$$

此时测量, 得到  $|x_0\rangle$  中计算基的概率:

$$\langle x_0 | \sim = \sin((k+1)\theta) \rightarrow 1$$

$$\text{when } (2k+1)\theta \rightarrow \frac{\pi}{2}.$$

# The Grover iteration operator

- $G := WO$  is a rotation. What do we know about it?

- Restricting to the plane spanned by  $\{|x_0^\perp\rangle, |x_0\rangle\}$ ,

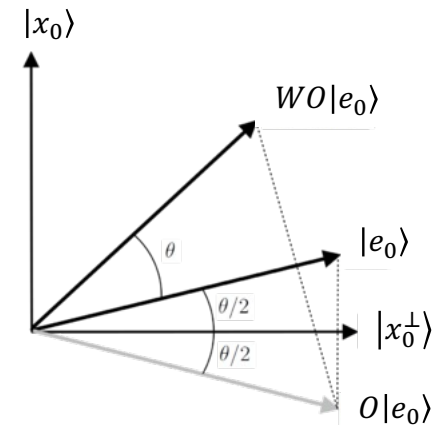
with  $|x_0^\perp\rangle := \sqrt{\frac{N-1}{N}} \sum_{x \neq x_0} |x\rangle$ , we can rewrite  $W, O$  as

$$W = \begin{pmatrix} -\cos \theta & \sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}, O = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

with  $\theta := 2 \arcsin \sqrt{1/N}$ .

- The Grover iteration is **a rotation from  $|x_0^\perp\rangle$  to  $|x_0\rangle$  by  $\theta$ :**

$$G = WO = - \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$$



Exercise:  
Verify the above  
matrix forms of  $W, O$ .

## Step 2: the optimal number of iterations

- **Question:** How many times should we apply  $G$ ?
- The initial state is  $|e_0\rangle = \begin{pmatrix} \cos \theta/2 \\ \sin \theta/2 \end{pmatrix}$ .  $G$  is a rotation by  $\theta = 2 \arcsin \sqrt{1/N}$ .
- $\Rightarrow$  After  $k$  iterations the state becomes  $\begin{pmatrix} \cos(\frac{1}{2}+k)\theta \\ \sin(\frac{1}{2}+k)\theta \end{pmatrix}$ .
- The target is  $|x_0\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$ .  $\Rightarrow$  The minimum  $k$  satisfies

$$\left(\frac{1}{2} + k\right) \theta \approx \frac{\pi}{2}$$

- **Optimal iteration number for Grover:**

We should choose  $k$  to be the closest integer  $k^*$  to

$$\frac{\pi}{4 \arcsin \sqrt{1/N}} - \frac{1}{2}$$

Exercise:

See what happens if we over-rotate (e.g.,  $k$  being twice the optimal value).

## Step 3: success probability

- Query complexity of Grover =  $\Theta(\sqrt{N})$ , as  $k^* = \frac{\pi}{4 \arcsin \sqrt{1/N}} - \frac{1}{2} \approx \frac{\pi}{4} \sqrt{N}$ .
- **Question**: What is the probability of success (error)?
- When query complexity =  $k$ , the state becomes  $\begin{pmatrix} \cos\left(\frac{1}{2}+k\right)\theta \\ \sin\left(\frac{1}{2}+k\right)\theta \end{pmatrix}$ .
- $\Rightarrow$  the probability of correctly outputting  $x_0$  is  $P = \left(\sin\left(\frac{1}{2}+k\right)\theta\right)^2$ .
- Since  $k$  is the closest integer (gap  $\leq 1/2$ ) to  $\frac{\pi}{2\theta} - \frac{1}{2}$  with  $\theta = 2 \arcsin \sqrt{1/N}$ , we have

$$P \geq 1 - \left(\frac{\pi}{2} - \left(\frac{1}{2} + k\right)\theta\right)^2 \geq 1 - \frac{\theta^2}{4} \approx 1 - \frac{1}{N}$$

# Optimality of Grover's scaling



Can we do better than  $O(\sqrt{N})$ ?

Maybe another exponential speedup like Shor?

- No!  
Searching an item in an unstructured database of size  $N$  requires  $\Omega(\sqrt{N})$  queries to the oracle.
- In another word, **Grover's algorithm achieves the optimal scaling  $\Theta(\sqrt{N})$ !**
- Proof: See the bonus material.

# **Part III:**

# **Improvements on Grover**

## Multi-item search

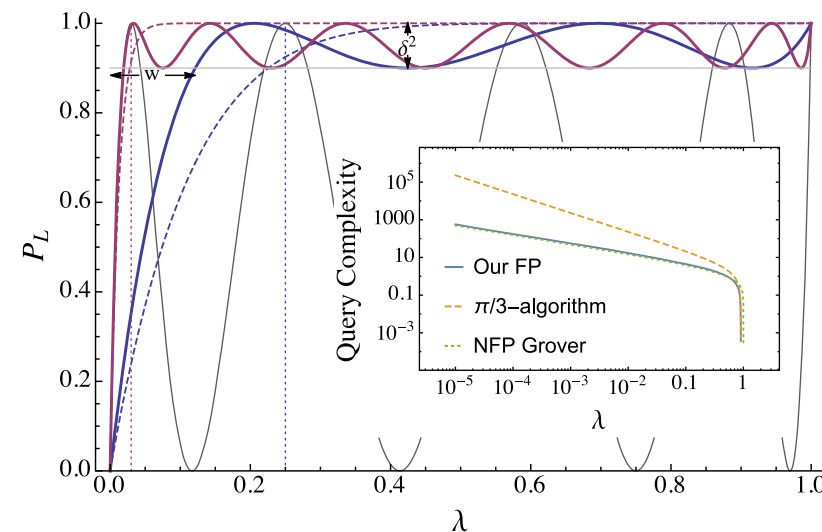
- What if there are  $M \geq 1$  items, and the task is to find any one of them amount  $N$  items?
- This can be easily remedied, by replacing every  $N$  in the original algorithm by  $N/M$ !

Exercise:  
Verify the above statement.

- The (modified) Grover algorithm yields a random but desired item within  $O(\sqrt{N/M})$  queries to the oracle, with success probability  $\approx 1 - M/N$ !

# Remedy for overcooking

- Grover algorithm has a caveat that it requires knowledge of the ratio  $N/M$ .
- Otherwise, we have the overcooking problem:
- Remedies:
  - Estimate  $N/M$  by sampling before applying Grover.
  - Fixed-point quantum search algorithms (Grover'05, Yoder-Low-Chuang'14)
- Using fixed-point quantum search, the success probability will not vanish under overcooking.
- **Question:** Can you design a quantum algorithm for estimating  $N/M$ ?



Success prob. vs  $\lambda$  ( $\sim \frac{M}{N}$ ) in [Yoder-Low-Chuang'14]



## **Discussion: Grover vs Shor, HHL**

- We have so far learned two categories of quantum algorithms.
- What are their differences and similarities?



# Summary

- Grover: a quantum search algorithm in an unstructured database.
- Optimality of Grover<sup>\*</sup>.
- Improvements on the quantum search algorithm.

# Homework

- **Review** the lecture slides; you may find the review questions in the next slides helpful.

Try the exercises in the slides and discuss with your classmates.

- Arithmetic details like continued fractions are not important.  
Instead, you should understand how the quantum subroutine works and **master the tricks of algorithm design** (phase-kick, partial measurement ...)
- **Attempt Q5 in Assignment 2 and Q1 in Assignment 3.**
- Optional: Read p248-255 of *Quantum Computation and Quantum Information* by Nielsen and Chuang.

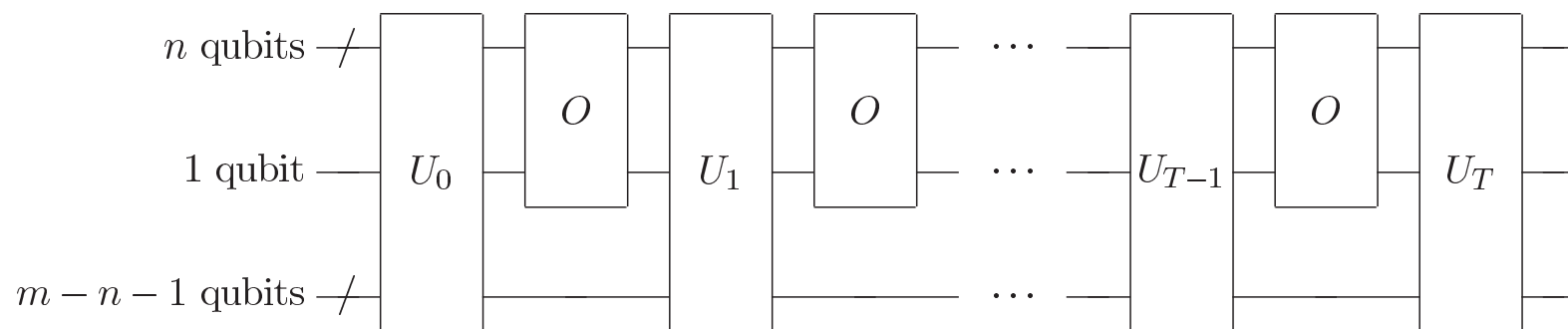
# Review questions

- Review the phase kickback trick, which is very useful in quantum algorithm design. Summarize how is it used in Deutsch-Jozsa, QPE (Shor), and Grover.
- In the original version of Grover, find the probability of getting the correct answer if we stop after  $m$  interactions. What is the probability when  $N = m$ ?
- What happens to Grover if we don't know  $N$ ? Are there any solutions?
- Compared to best (existing) classical algorithms, Shor is exponentially faster but Grover is only quadratically faster. What is the cause of this difference?

# **Bonus content\*: Optimality of Grover's algorithm**

# A general quantum algorithm for searching

- Suppose an arbitrary algorithm makes  $K$  queries to the oracle  $O_x = -2|x\rangle\langle x| + I$ .
- The algorithm does not know  $x$ , so it **must** be of the following general structure:



where  $U_1, \dots, U_K$  are generic unitary operations. They may consist of one or many elementary gates.

- Our goal is to show that  $K = \Omega(\sqrt{N})$ , and thus no algorithm can do better than Grover in terms of query complexity!

# A general quantum algorithm for searching

- Consider the two families of states: for  $k = 0, 1, \dots, K$  we have

$$|\psi_k^x\rangle = U_k O_x U_{k-1} O_x \cdots U_1 O_x |\psi_0\rangle$$

$$|\psi_k\rangle = U_k U_{k-1} \cdots U_1 |\psi_0\rangle$$

- Remark:  $|\psi_k^x\rangle$  is the state after, and  $|\psi_k\rangle$  is a reference state obtained by removing the oracles from the algorithm.
- We require that the algorithm finds the correct answer up to a (small) error  $\epsilon$  when it terminates, for every  $x = 1, 2, \dots, N$ .

This can be characterized by the following requirement:

- Faithfulness:**  $\| |x\rangle - |\psi_K^x\rangle \|^2 \leq \epsilon \quad \forall x \in \{1, \dots, N\}$ .  $\| |a\rangle \| := \sqrt{\langle a|a \rangle}$

## Proof strategy

- Consider the overall impact that the oracle has on the system state:

$$D_K := \sum_{x=1}^N |||\psi_K\rangle - |\psi_K^x\rangle||^2$$

- We shall establish the bound on  $K$  (in terms of  $N$ ), by constructing both a lower bound and an upper bound on  $D_K$ .

- Lower bound in terms of  $N$ :

By faithfulness,  $|\psi_K^x\rangle \approx |x\rangle$ , whereas  $|\psi_K\rangle$  is independent of  $|x\rangle$  and cannot be close to every  $|x\rangle$  at the same time.

- Upper bound in terms of  $K$ :

The two states  $|\psi_K\rangle, |\psi_K^x\rangle$  differ only by the oracle action ( $K$  times), which is subject to a “speed limit” on how much the oracle can change the state in each iteration.



## An upper bound on $D_K$

- Consider the quantity:

$$\begin{aligned}
 D_{k+1} &= \sum_x \|U_k |\psi_k\rangle - O_x U_k |\psi_k^x\rangle\|^2 \\
 &= \sum_x \| |\psi_k\rangle - O_x |\psi_k^x\rangle \|^2 = \sum_x \| (O_x - I) |\psi_k^x\rangle + (|\psi_k^x\rangle - |\psi_k\rangle) \|^2 \\
 &\leq \sum_x (\| (O_x - I) |\psi_k^x\rangle \| + \| |\psi_k^x\rangle - |\psi_k\rangle \|^2) \qquad O_x = -2|x\rangle\langle x| + I \\
 &= \sum_x 4|\langle x | \psi_k^x \rangle|^2 + 4|\langle x | \psi_k^x \rangle| \cdot \| |\psi_k^x\rangle - |\psi_k\rangle \| + \| |\psi_k^x\rangle - |\psi_k\rangle \|^2 \\
 &\leq 4 + 4\sqrt{D_k} + D_k = (\sqrt{D_k} + 2)^2
 \end{aligned}$$

## An upper bound on $D_K$

- All together, we get the inductive bound:

$$\sqrt{D_{k+1}} \leq \sqrt{D_k} + 2.$$

- Since  $D_0 = 0$ , we have the upper bound as

$$D_K \leq 4K^2.$$

## A lower bound on $D_K$

- Faithfulness:  $\| |x\rangle - |\psi_k^x\rangle \| \leq \epsilon$  for every  $x = 1, \dots, N$ .
- Consider the quantity:

$$D_k := \sum_{x=1}^N \| |\psi_k\rangle - |\psi_k^x\rangle \|^2$$

- We can lower bound it as

$$\begin{aligned} D_k &\geq \sum_{x=1}^N (\| |\psi_k\rangle - |x\rangle \| - \| |x\rangle - |\psi_k^x\rangle \|)^2 \\ &= \sum_{x=1}^N \| |\psi_k\rangle - |x\rangle \|^2 - 2 \| |x\rangle - |\psi_k^x\rangle \| \cdot \| |\psi_k\rangle - |x\rangle \| + \| |x\rangle - |\psi_k^x\rangle \|^2 \end{aligned}$$

## A lower bound on $D_K$

- We can lower bound  $D_K$  as

$$D_K \geq \sum_{x=1}^N \left( \|\psi_K\rangle - |x\rangle\|^2 - 2\|\psi_K\rangle - |x\rangle\| \cdot \|\psi_K^x\rangle - |x\rangle\| + \|\psi_K^x\rangle - |x\rangle\|^2 \right)$$

$$T_1 = \sum_x (2 - 2|\langle\psi_K|x\rangle|) \geq 2N - 2\sqrt{N}$$

$$T_2 \leq 2\sqrt{T_3 T_1}$$

$$T_3 = \sum_x \|\psi_K^x\rangle - |x\rangle\|^2 \leq N \cdot \epsilon$$

Faithfulness:  $\|\psi_K^x\rangle - |x\rangle\|^2 \leq \epsilon$

- Summarizing, we get the lower bound as:

$$D_K \geq \left( \sqrt{T_1} - \sqrt{T_3} \right)^2 \geq (2 - 2\sqrt{2\epsilon})N - O(\sqrt{N})$$

# Optimality of Grover's scaling

- Combining both bounds, we get:

$$4K^2 \geq D_K \geq (2 - 2\sqrt{2\epsilon})N - O(\sqrt{N})$$

- Together it implies that  $K = \Omega(\sqrt{N})$ !
- Conclusion:  
The  $\sqrt{N}$ -scaling of search in an unstructured database is optimal!
- Grover's algorithm achieves the optimal scaling  $O(\sqrt{N})$ !