

## GROVER'S ALGORITHM

Solves the problem of "unstructured search" in the "black-box query model"

### UNSTRUCTURED SEARCH PROBLEM :

Given database  $X = \{x_1, x_2, \dots, x_N\}$

and  $f: X \rightarrow \{0, 1\}$

find  $x \in X$  satisfying  $f(x) = 1$ .

### BLACK-BOX QUERY MODEL :

Algorithm only accesses  $f$  via input/output queries

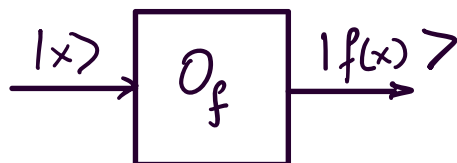
Doesn't rely on any details of circuit implementing  $f$ .

Q: How many queries to  $f$  will solve unstructured search?

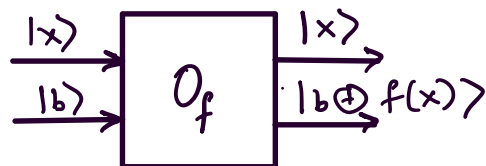
A: Classically, need  $N$  queries worst case.

Quantumly, only need  $O(\sqrt{N})$  queries!

### SETTING :



not reversible



reversible  
implementation

## "PHASE KICKBACK"

Given access to  $O_f$ , implement  $|x\rangle \rightarrow (-1)^{f(x)} |x\rangle$

$$\begin{aligned} \text{Query } O_f \text{ on } |x\rangle \otimes |-\rangle \\ = |x0\rangle - |x1\rangle \end{aligned}$$

$$\begin{aligned} O_f(|x0\rangle - |x1\rangle) &= |x f(x)\rangle - |x \overline{f(x)}\rangle \\ &= |x0\rangle - |x1\rangle = |x\rangle \otimes |-\rangle \quad \text{if } f(x)=0 \\ &= |x1\rangle - |x0\rangle = -|x\rangle \otimes |-\rangle \quad \text{if } f(x)=1 \end{aligned}$$

$$O_f(|x\rangle \otimes |-\rangle) \rightarrow (-1)^{f(x)} |x\rangle \otimes |-\rangle$$

Assume gate  $O_{f\pm} : |x\rangle \rightarrow (-1)^{f(x)} |x\rangle$

## GROVER SEARCH :

Simplifying assumption :  $\exists$  unique  $x^*$  s.t.  $f(x)=0$  if  $x=x^*$   
 $= 1$  otherwise

and  $N = 2^n$

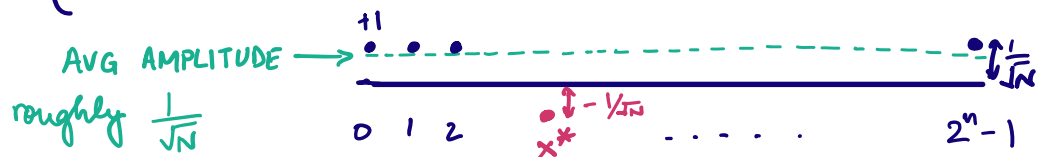
General strategy to build Quantum algorithms

① Prepare a uniform superposition of inputs

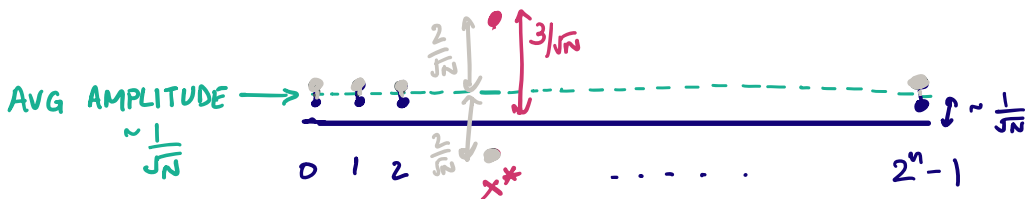
Apply  $H^{\otimes n} |0^n\rangle \rightarrow \sum_{x \in \{0,1\}^n} |x\rangle$

② Query the function in superposition

$$O_{f^{\pm}} \left( \sum_{x \in \{0,1\}^n} |x\rangle \right) \rightarrow \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle$$

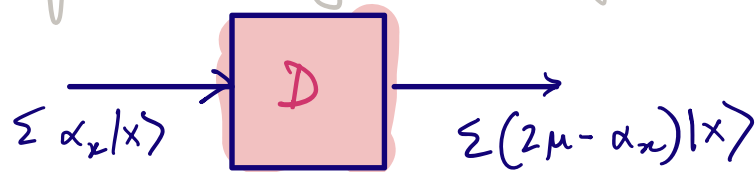


③ Flip amplitudes about the average



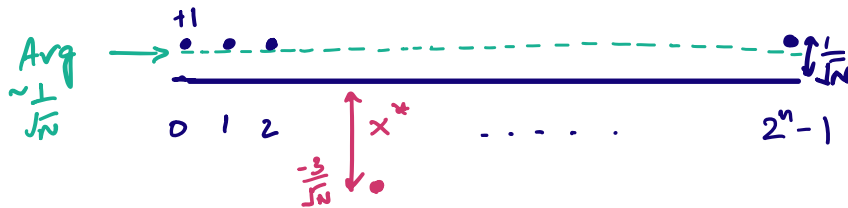
The state is roughly:  $|\psi_1\rangle = \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} (2f(x)+1) |x\rangle$

[Our goal is to bring it to roughly  $|x^*\rangle$ ]

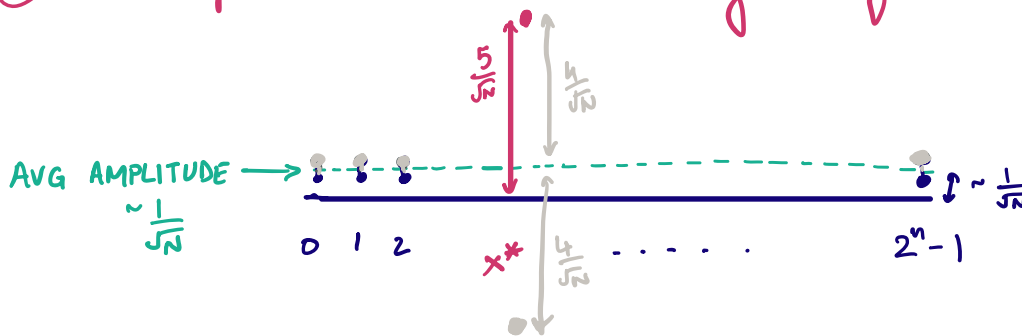


(4) Query the function again, on  $|\psi_1\rangle$

$$O_{f \pm} \left( \sum_{x \in \{0,1\}^n} (2f(x)+1) |x\rangle \right) \rightarrow \sum_{x \in \{0,1\}^n} (-1)^{f(x)} (2f(x)+1) |x\rangle$$



(5) Flip around the average again



Repeat  $O(\sqrt{N})$  times : a) Query  $O_{f \pm}$   
 b) Flip around the average

Roughly, each time amplitude on  $x^*$   $\uparrow$  by  $> \frac{1}{\sqrt{N}}$   
 After sufficiently many tries, amplitude on  $x^* \sim 1$ .

## IMPLEMENTING $\mathcal{D}$ :



$$\text{Convert } \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \rightarrow \sum_{x \in \{0,1\}^n} (2\mu - \alpha_x) |x\rangle$$

Is there a unitary that achieves this?

$$-\mathbb{I} \left( \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \right) \rightarrow - \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle$$

$$\langle +^n | = \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} \langle x |$$

$$\langle +^n | \left( \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \right) \rightarrow \frac{\sum_{x \in \{0,1\}^n} \alpha_x}{\sqrt{N}} = \frac{\mu N}{\sqrt{N}} = \mu \sqrt{N}$$

$$2 | +^n \rangle \left( \langle +^n | \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \right) \rightarrow 2 \left( \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} |x\rangle \right) \mu \sqrt{N}$$

$$= 2\mu \sum_{x \in \{0,1\}^n} |x\rangle$$

$$\left( 2 | +^n \rangle \langle +^n | - \mathbb{I} \right) \left( \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle \right) \rightarrow \sum_{x \in \{0,1\}^n} 2\mu - \alpha_x |x\rangle$$

$\mathcal{D}$ : efficient unitary implementation?

$$\text{Let } Z_0 = 2 |0^n\rangle\langle 0^n| - I$$

What does this do?

$$Z_0 |x\rangle = 2 |0^n\rangle\langle 0^n| x\rangle - |x\rangle$$

When  $|x\rangle = |0^n\rangle$ , then  $Z_0 |x\rangle = |0^n\rangle = |x\rangle$

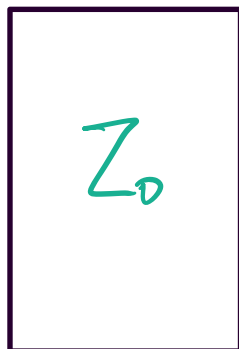
When  $|x\rangle \neq |0^n\rangle$ , then  $Z_0 |x\rangle = -|x\rangle$

So  $Z_0$  does nothing when  $x = 0^n$   
and flips the phase whenever  $x \neq 0^n$ .

There is a classical circuit for this:

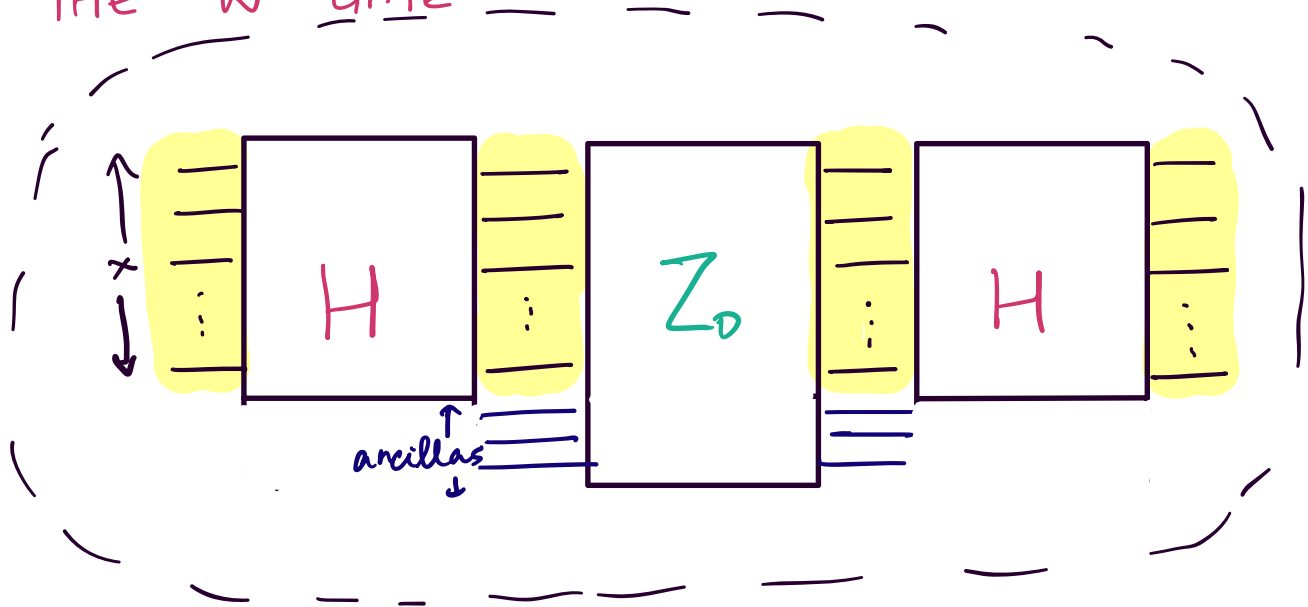
Output  $x$  if  $\text{OR}(x_1, x_2, \dots, x_n) = 0$   
- $x$  otherwise

You already know to convert this to reversible quantum circuit



$A$  is same as  $Z_0$ , except in Hadamard basis,  
 i.e. do nothing when  $x = |+\rangle^n$   
 and flip phase on all other Had. basis states

THE  $A$  GATE



Let us check

$$\begin{aligned}
 & H^{\otimes n} Z_0 H^{\otimes n} \\
 &= H^{\otimes n} (2|0^n\rangle\langle 0^n| - \mathbb{I}) H^{\otimes n} \\
 &= (2 H^{\otimes n} |0^n\rangle\langle 0^n| - H^{\otimes n}) H^{\otimes n} \\
 &= 2|+\rangle\langle +| - \mathbb{I}
 \end{aligned}$$

## HOW MANY ITERATIONS ?

In each step, apply  $D_{f\pm}$  followed by  $\mathcal{D}$ .

Let  $\alpha^{(t)}$  = Amplitude on  $x^*$  after  $t^{\text{th}}$  step

$\beta^{(t)}$  = Amplitude on  $x \neq x^*$  after  $t^{\text{th}}$  step

$\mu^{(t)}$  = Avg amplitude after  $D_{f\pm}$  in  $t^{\text{th}}$  step  
 $= -\alpha^{(t)} + \frac{N-1}{N} \beta^{(t)}$

$$\alpha^{(0)} = \beta^{(0)} = \frac{1}{\sqrt{N}}$$

After  $D_{f\pm}$ ,  $x^*$  has amplitude  $\left(-\frac{1}{\sqrt{N}}\right)$

$$\text{Avg. amplitude } \mu^{(1)} = \frac{1}{N} \left( \frac{N-1}{\sqrt{N}} + \frac{(-1)}{\sqrt{N}} \right) = \frac{N-2}{N\sqrt{N}}$$

After  $\mathcal{D}$ ,  $x^*$  has amplitude  $2\mu^{(1)} - \left(-\frac{1}{\sqrt{N}}\right)$

$$\alpha^{(1)} = 2\mu^{(1)} + \alpha^{(0)}$$



**CLAIM 1:**

For any  $t$ ,  $a^{(t+1)} \leq \alpha^{(t)} + \frac{2}{\sqrt{N}}$

Proof

Since sum of squares of amplitudes = 1

$$\Rightarrow (N-1)(\beta^{(t)})^2 \leq 1 \Rightarrow \beta^{(t)} \leq \frac{1}{\sqrt{N-1}}$$

$$\mu^{(t)} = \alpha^{(t)} + \frac{N-1}{N} \beta^{(t)} \leq \frac{N-1}{N} \beta^{(t)} = \frac{\sqrt{N-1}}{N}$$

$$\begin{aligned} \alpha^{(t+1)} &= 2\mu^{(t)} + \alpha^{(t)} \\ &= \frac{2\sqrt{N-1}}{N} + \alpha^{(t)} \\ &\leq \frac{2}{\sqrt{N}} + \alpha^{(t)} \end{aligned}$$

## CLAIM 2:

Suppose  $\alpha^{(t)} \geq \frac{1}{2}$ ,  $N \geq 4$ .

Then,  $\alpha^{(t+1)} \geq \alpha^{(t)} + \frac{1}{\sqrt{N}}$

Proof. Since sum of squares of amplitudes is 1,  
 $(\alpha^{(t)})^2 + (N-1)(\beta^{(t)})^2 = 1$

$$(\beta^{(t)})^2 = \frac{1 - (\alpha^{(t)})^2}{N-1} \geq \frac{3}{4(N-1)}$$

$$\mu^{(t)} = \frac{-\alpha^{(t)}}{N} + \frac{(N-1)}{N} \beta^{(t)}$$

$$\geq -\frac{1}{2N} + \left(\frac{N-1}{N}\right) \sqrt{\frac{3}{4(N-1)}}$$

$$\geq -\frac{1}{2N} + \frac{\sqrt{N-1}}{2N} \cdot \sqrt{3} \geq \frac{1}{2\sqrt{N}} \text{ for } N \geq 4$$

$$\alpha^{(t+1)} = 2\mu^{(t)} + \alpha^{(t)} \geq \frac{1}{\sqrt{N}} + \alpha^{(t)}$$

Claims 1 and 2 together tell us  
that for  $\alpha^{(t)} \leq \frac{1}{2}$ ,  $N > 4$ ,

$$\frac{1}{\sqrt{N}} + \alpha^{(t)} \leq \alpha^{(t+1)} \leq \frac{2}{\sqrt{N}} + \alpha^{(t)}$$

$\alpha^{(0)} = \frac{1}{\sqrt{N}}$ . For  $t \leq \frac{\sqrt{N}}{8}$ , by

$$\begin{aligned} \alpha^{(t+1)} &\leq \frac{2}{\sqrt{N}} \cdot \frac{\sqrt{N}}{8} + \alpha_0 \\ &\leq \frac{1}{\sqrt{N}} + \frac{1}{4} < \frac{1}{2} \end{aligned}$$

So,  $\alpha^{(t+1)} \geq \alpha^{(t)} + \frac{1}{\sqrt{N}}$  for  $t \leq \frac{\sqrt{N}}{8}$

$$\Rightarrow \alpha^{(\sqrt{N}/8)} \geq \frac{1}{\sqrt{N}} \cdot \frac{\sqrt{N}}{8} > \frac{1}{8}$$

So, measuring in computational basis after  $\frac{\sqrt{N}}{8}$  steps gives 0.125 prob. of finding  $x^*$ .  
Repeat 100 times!