



Quantum Searching

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Overview

- Quantum Searching
- Quantum Counting
- Searching when you don't know the number of elements

QUANTUM SEARCHING

Searching problem

Consider $f : \{0,1\}^n \rightarrow \{0,1\}$

Given $U_f : |x\rangle \mapsto (-1)^{f(x)} |x\rangle$

find an x satisfying $f(x) = 1$.

Application

Consider a 3-SAT formula

$$\ddot{O} = C_1 \wedge C_2 \wedge \dots \wedge C_M$$

$$C_j = (y_{j,1} \vee y_{j,2} \vee y_{j,3})$$

$$y_{j,k} \in \{x_1, x_2, \dots, x_n, \bar{x}_1, \bar{x}_2, \dots, \bar{x}_n\}$$

For a given assignment $\mathbf{x} = x_1 x_2 \dots x_n$

$$f_{\ddot{O}}(\mathbf{x}) = \begin{cases} 1 & \text{if } \mathbf{x} \text{ satisfies } \ddot{O} \\ 0 & \text{otherwise} \end{cases}$$

Some ideas

For simplicity, let's start by assuming that $f(x) = 1$ has exactly one solution, $x = w$.

IDEA: Prepare

$$\sum_x \frac{1}{\sqrt{2^n}} |x\rangle = \frac{1}{\sqrt{2^n}} |w\rangle + \sum_{x \neq w} \frac{1}{\sqrt{2^n}} |x\rangle$$

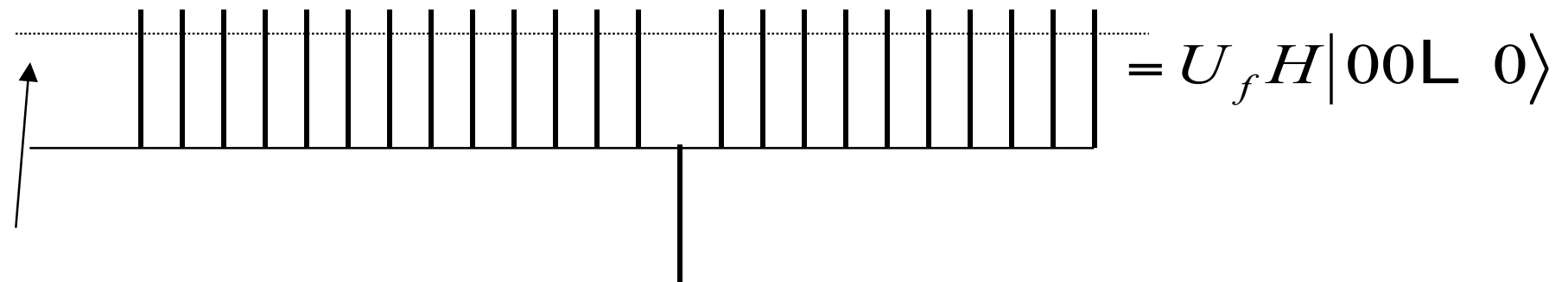
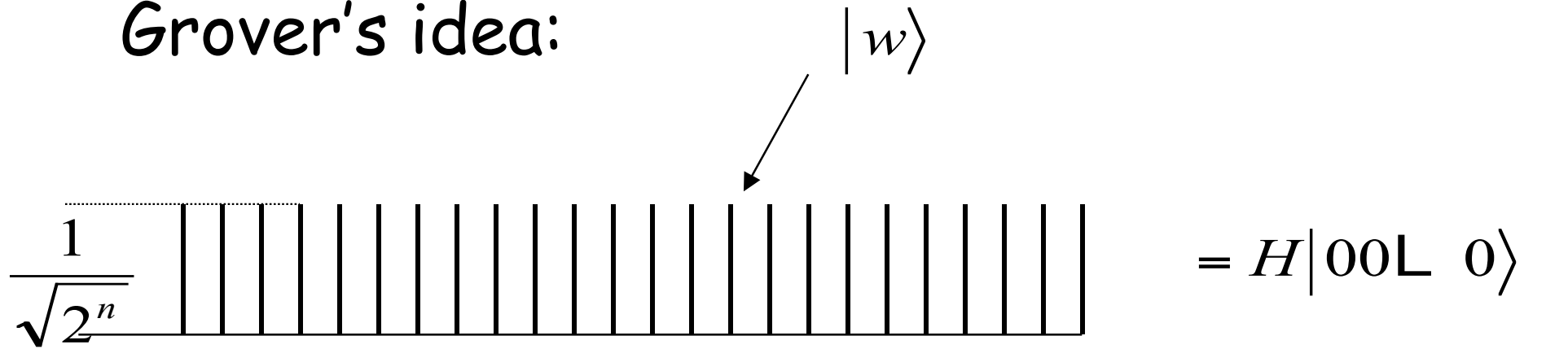
Keep this

"Re-scramble" this

Repeat roughly $\sqrt{2^n}$ times.

Must do this with legal quantum operations

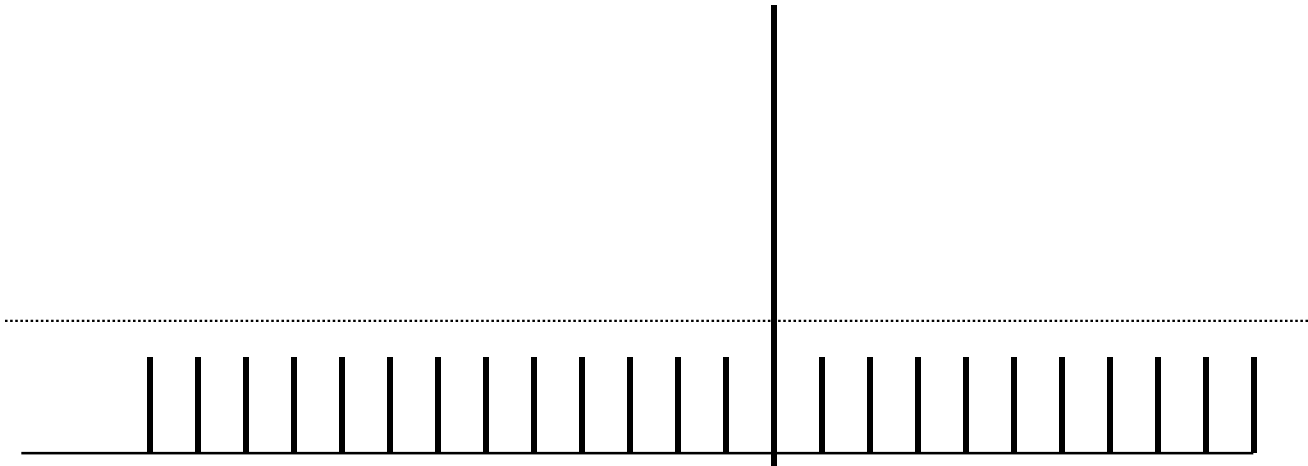
Grover's idea:



"mean value"

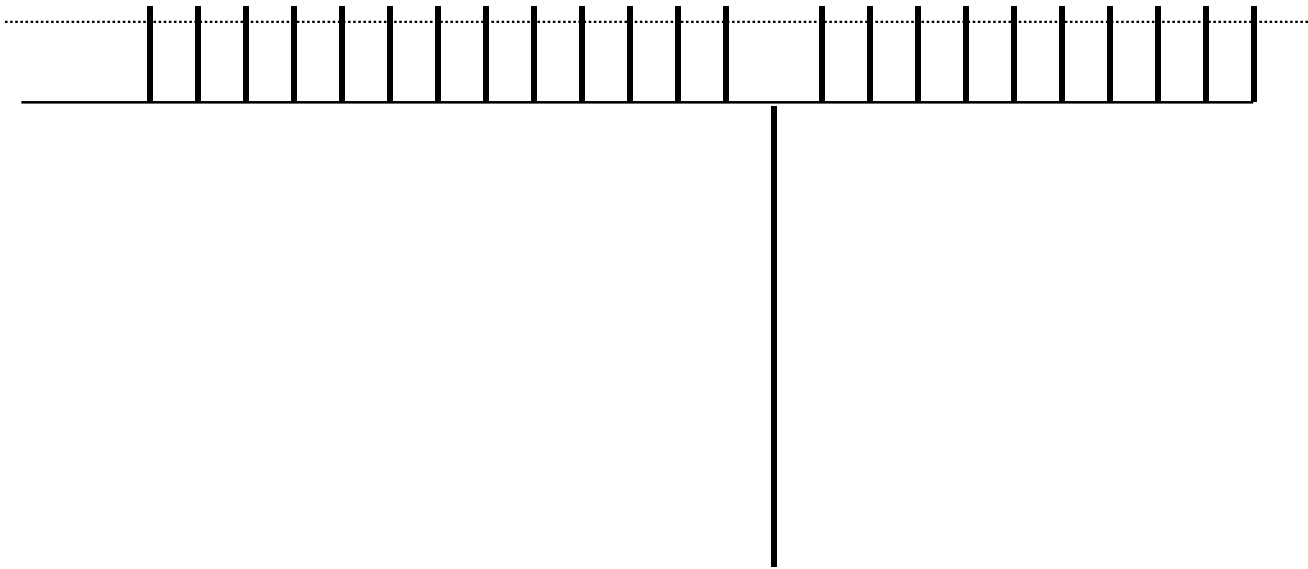
Must do this with legal quantum operations

"invert about the mean"



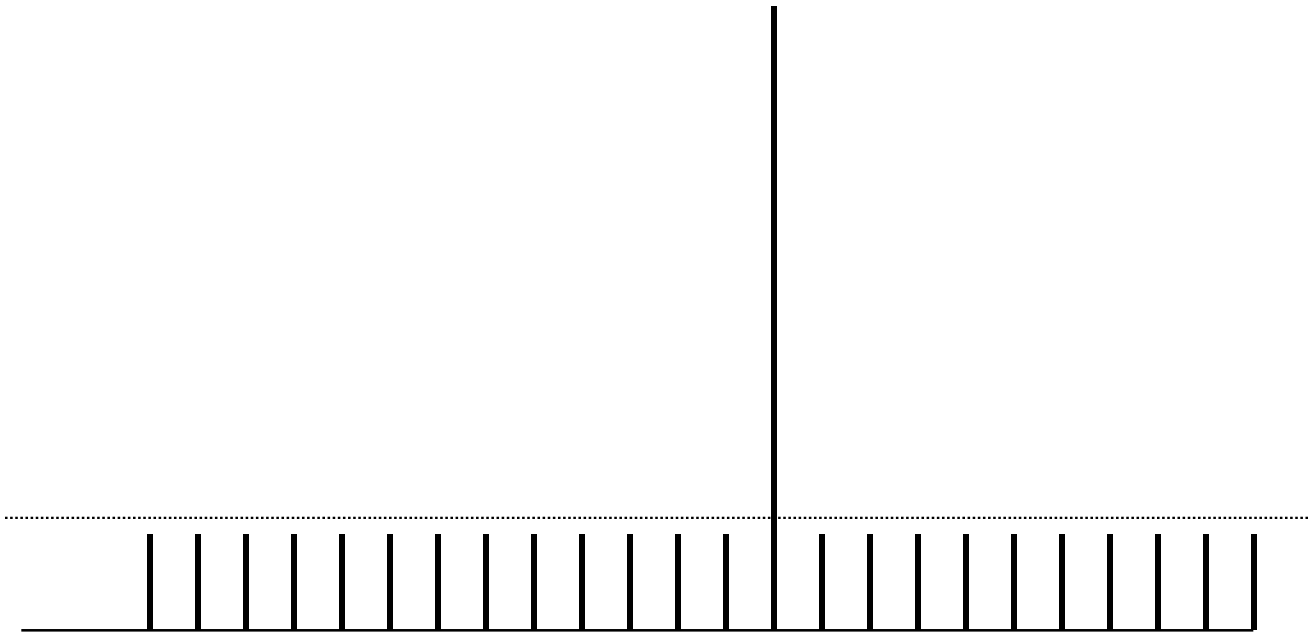
$$\begin{aligned} &= \left(\sum H U_0 H \right) U_f H |00L \ 0\rangle \\ &= \sum U_{H|0\rangle} U_f H |00L \ 0\rangle \end{aligned}$$

Repeat



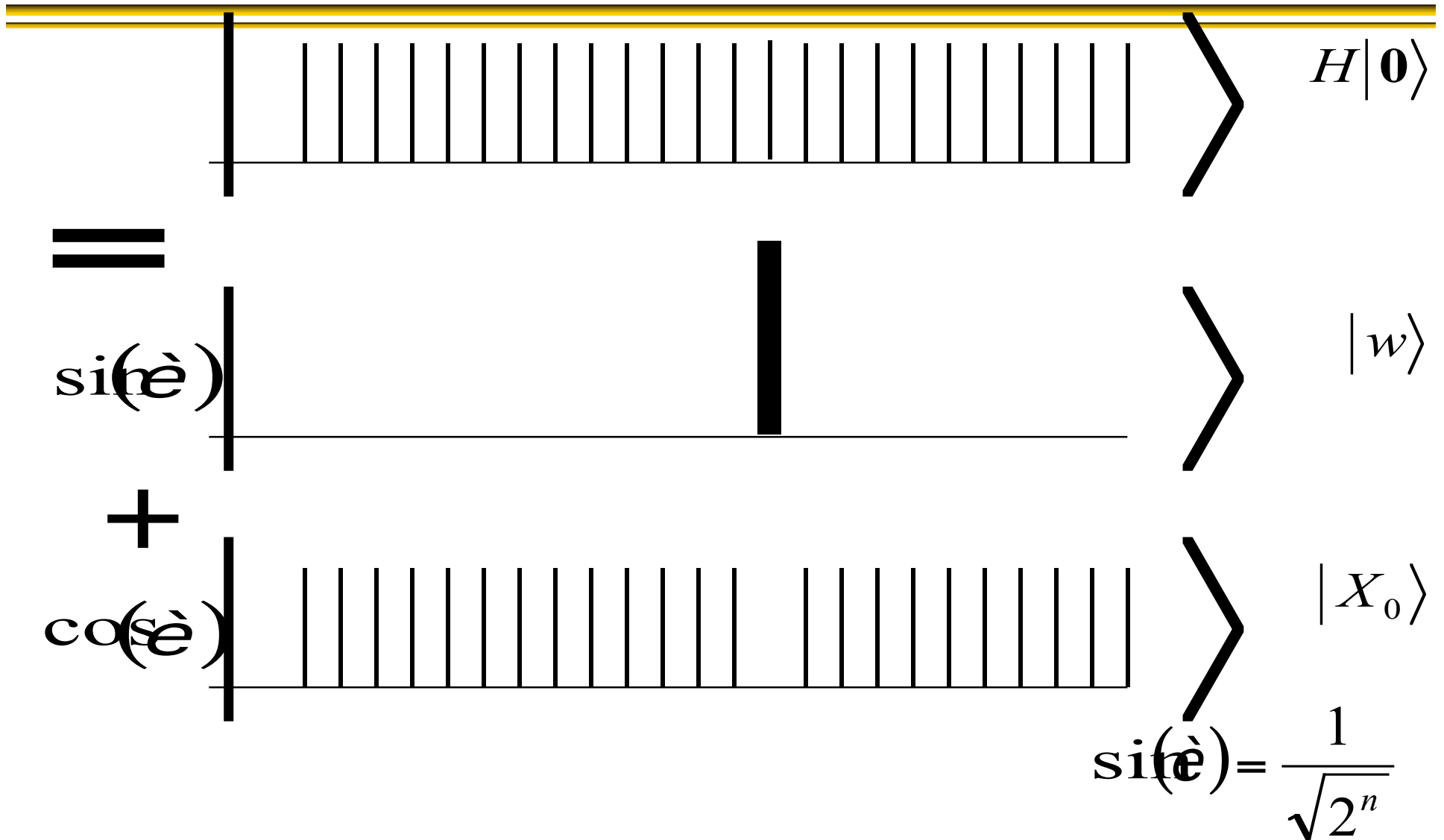
$$= U_f (\square H U_0 H) U_f H |00L \ 0\rangle$$

Repeat

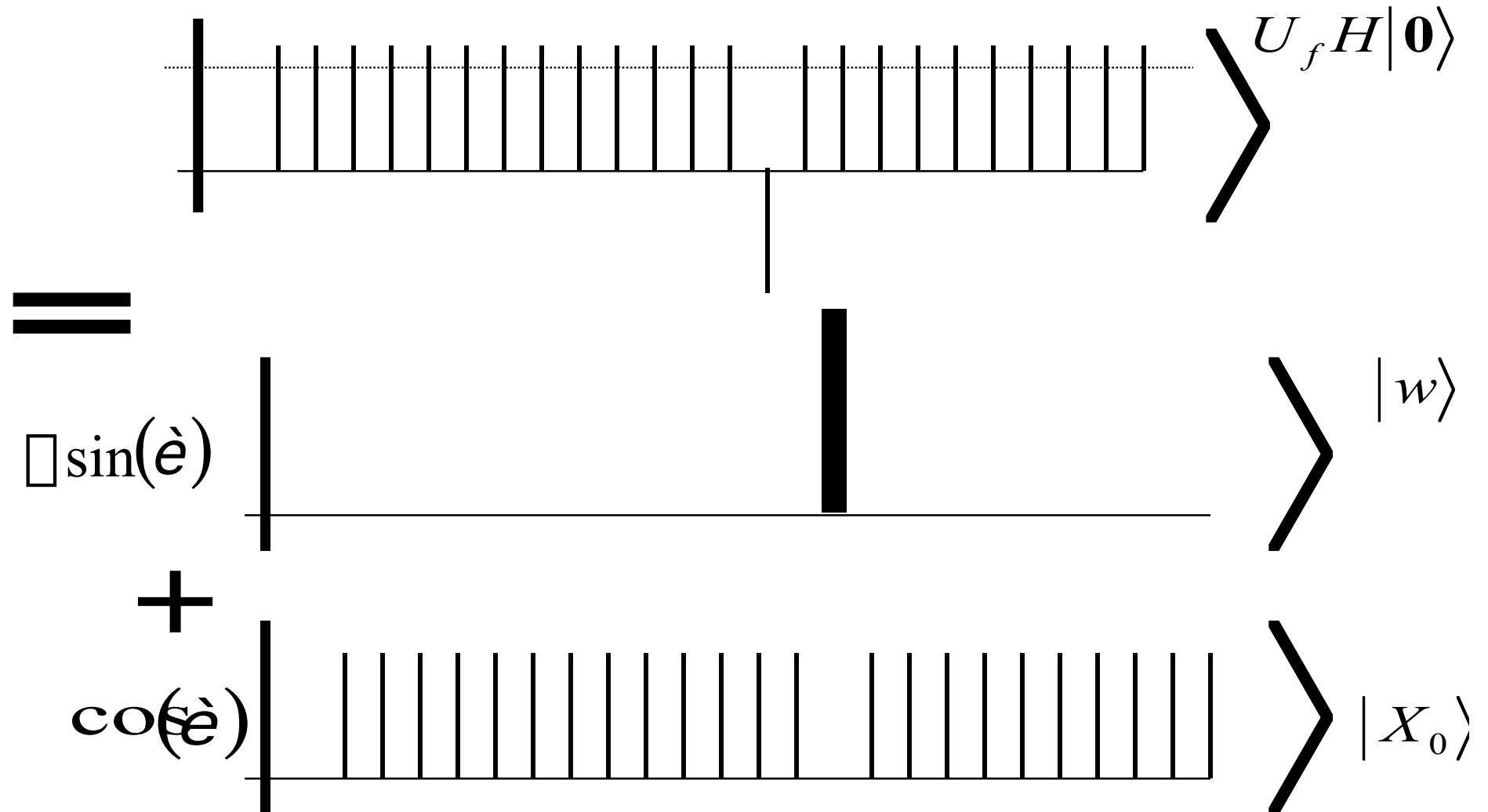


$$= (\square HU_0 H) U_f (\square HU_0 H) U_f H |00L \ 0\rangle$$

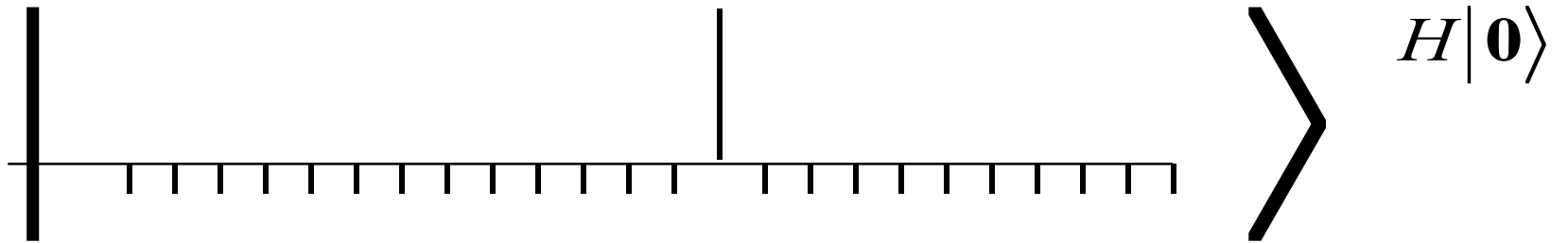
A nice way to analyze this



A nice way to analyze this



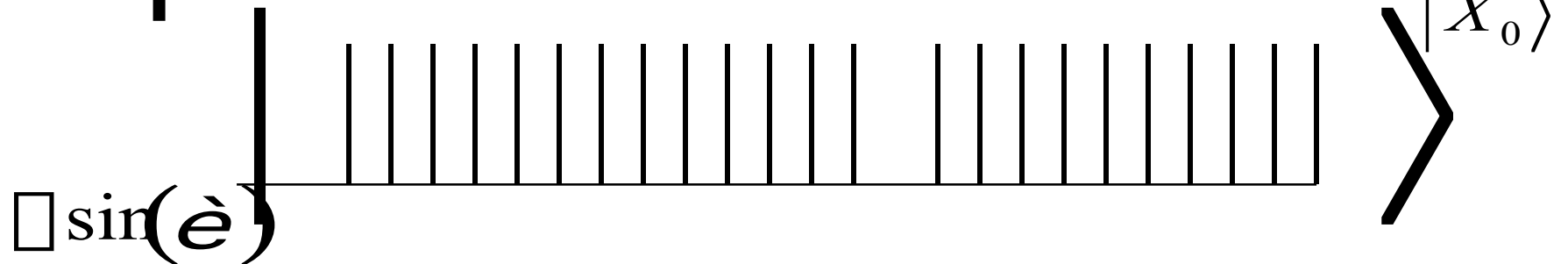
Definition



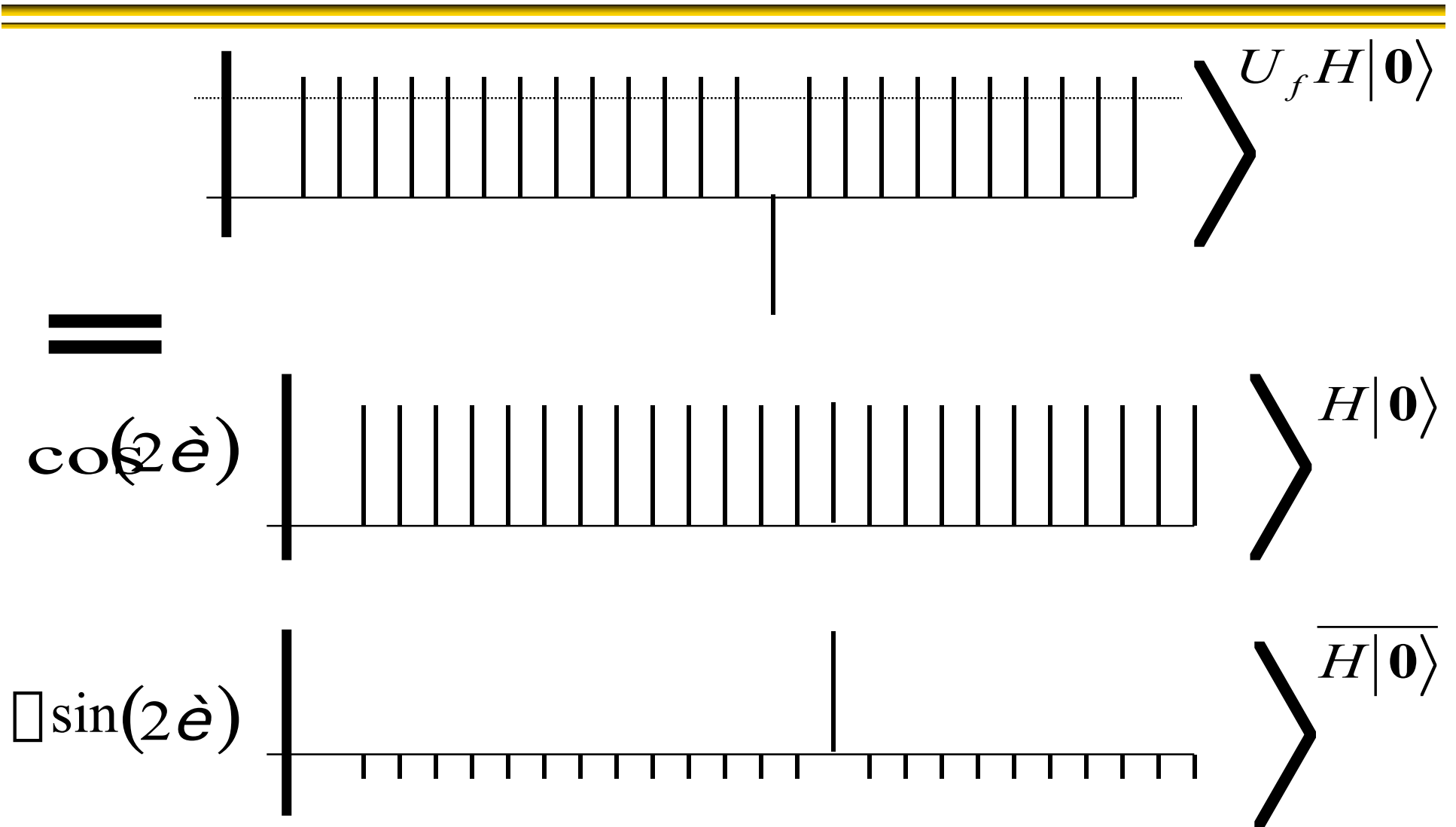
=



+



Note that



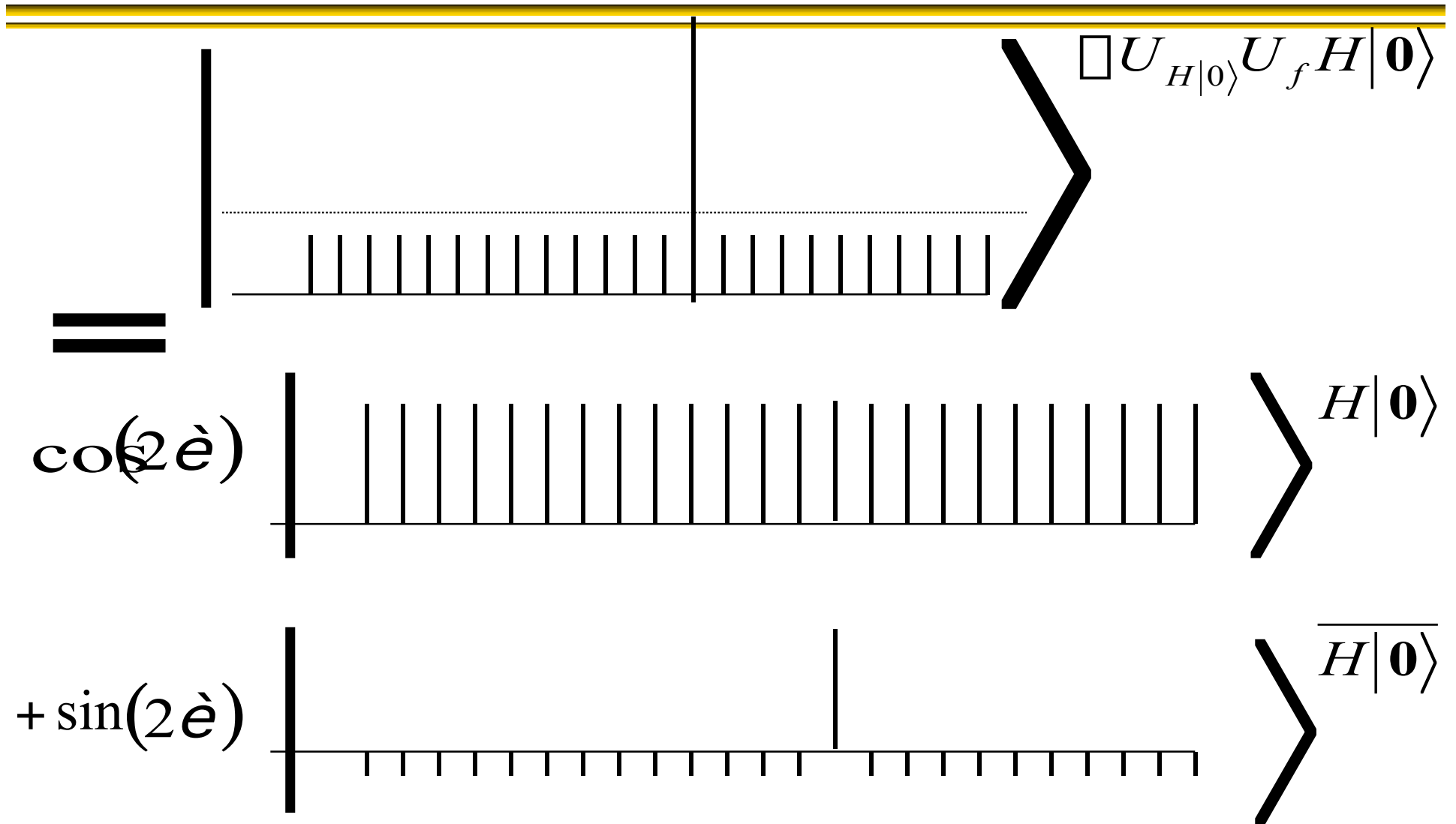
Verify that

$$\frac{1}{\sqrt{2}} \sin(\theta) |w\rangle + \cos(\theta) |X_0\rangle$$

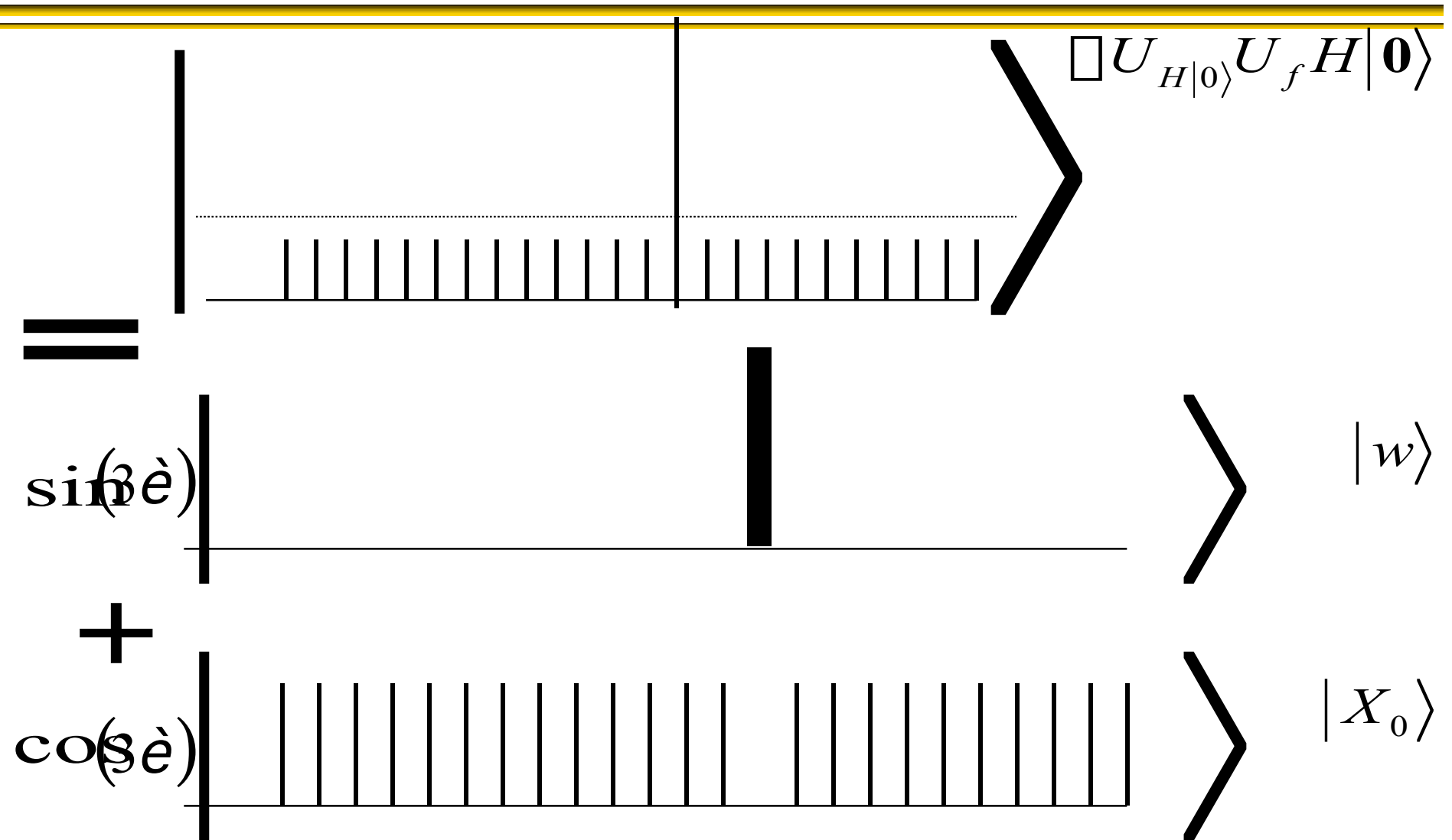
$$=$$

$$\cos(2\theta) |H|0\rangle - \frac{1}{\sqrt{2}} \sin(2\theta) \overline{|H|0\rangle}$$

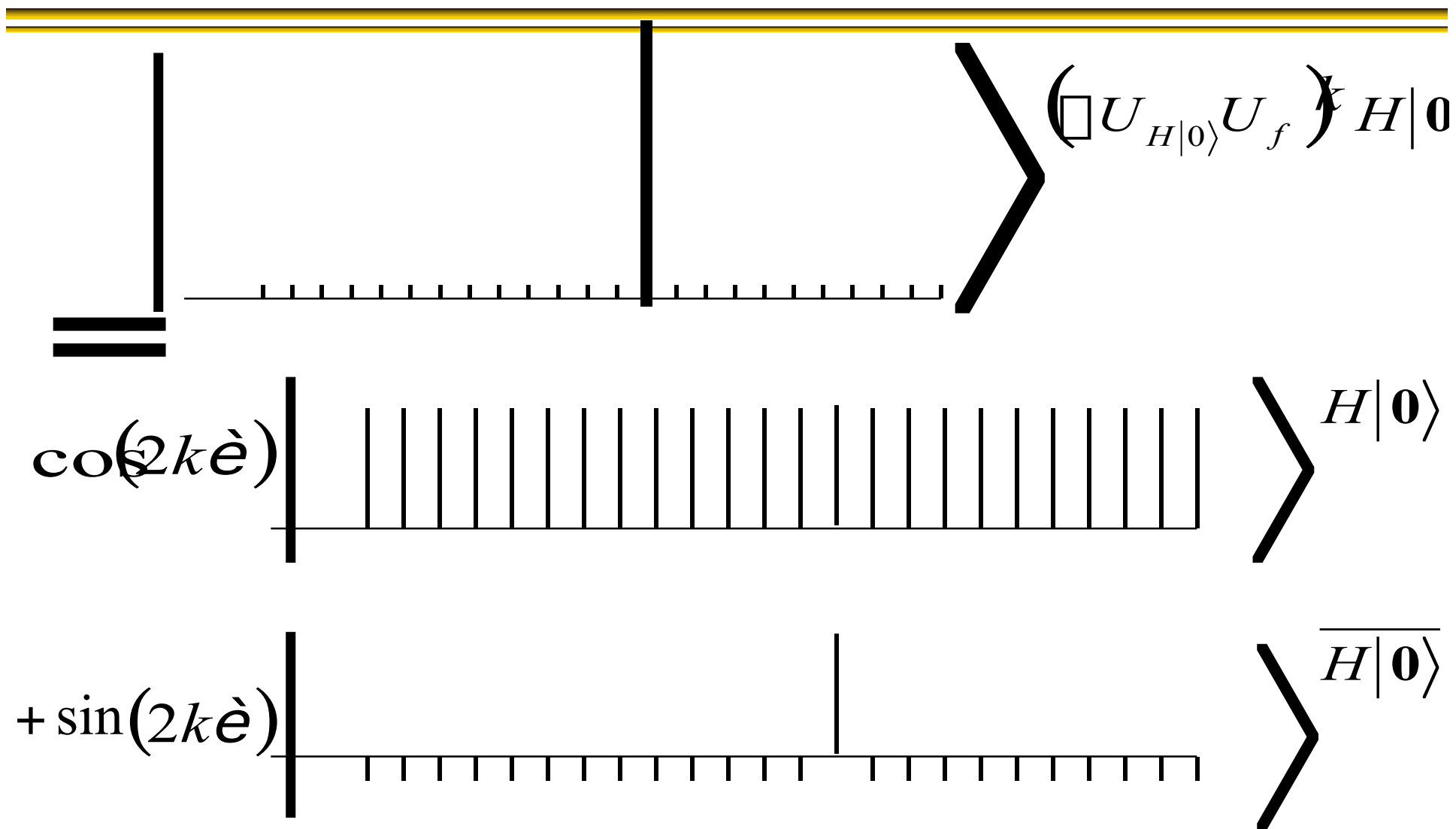
After "inversion"



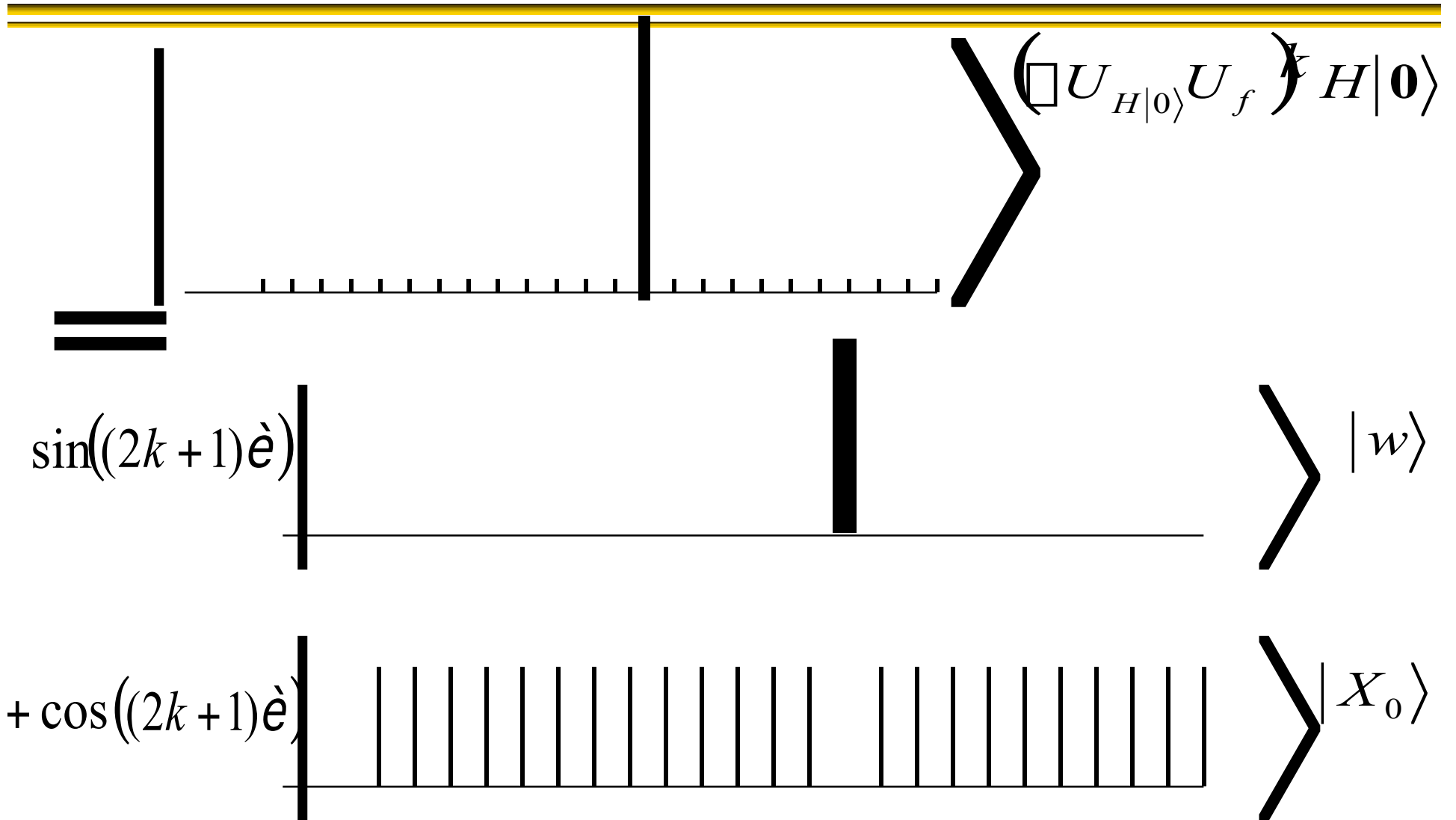
Alternatively



After k iterations



(*formula found by BBHT)
 Alternatively



Selecting parameters

So we need

$$\sin((2k+1)\epsilon) \approx 1$$

$$k \approx \frac{\partial}{4\epsilon} \approx \frac{1}{2} \approx \frac{\partial \sqrt{2^n}}{4}$$

Square root speed-up! What if we don't know k ? See [BBHT] (or [M98]) for a protocol that works in this case as well

Generalization: Amplitude Amplification (BBHT, BH, BHT, G, BHMT, ...)

Consider functions with t solutions

$$X_1 = f^{-1}(1) \quad X_0 = f^{-1}(0) \quad t = |X_1|$$

Consider any algorithm that works with non-zero probability

$$A|0\rangle = |\emptyset\rangle \quad |\emptyset\rangle = \sin(\epsilon)|\emptyset_1\rangle + \cos(\epsilon)|\emptyset_0\rangle$$

$$|\emptyset_1\rangle = \sum_{x \in X_1} \alpha_x |x\rangle \quad \sum_{x \in X_1} |\alpha_x|^2 = 1$$

$$|\emptyset_0\rangle = \sum_{y \in X_0} \beta_y |y\rangle \quad \sum_{y \in X_0} |\beta_y|^2 = 1$$

Amplitude Estimation

- Given operators

$$A|0\rangle = |\emptyset\rangle = \sin(\theta)|\emptyset_1\rangle + \cos(\theta)|\emptyset_0\rangle$$

$$U_f : \begin{array}{l} |\emptyset_1\rangle \rightarrow -|\emptyset_1\rangle \\ |\emptyset_0\rangle \rightarrow |\emptyset_0\rangle \end{array}$$

$$\sin^2(\theta)$$

- Estimate

Application: Counting

- E.g. $A|0\rangle = \sum_{x \in \{0,1\}^n} \frac{1}{\sqrt{2^n}} |x\rangle$

$$|\alpha_1\rangle = \sum_{x \in X_1} \frac{1}{\sqrt{t}} |x\rangle \quad |\alpha_0\rangle = \sum_{y \in X_0} \frac{1}{\sqrt{N-t}} |y\rangle$$

- So $A|0\rangle = \sqrt{\frac{t}{N}} |\alpha_1\rangle + \sqrt{\frac{N-t}{N}} |\alpha_0\rangle$

- So $\sin(\theta) = \sqrt{\frac{t}{N}}$

Eigenvectors of Q

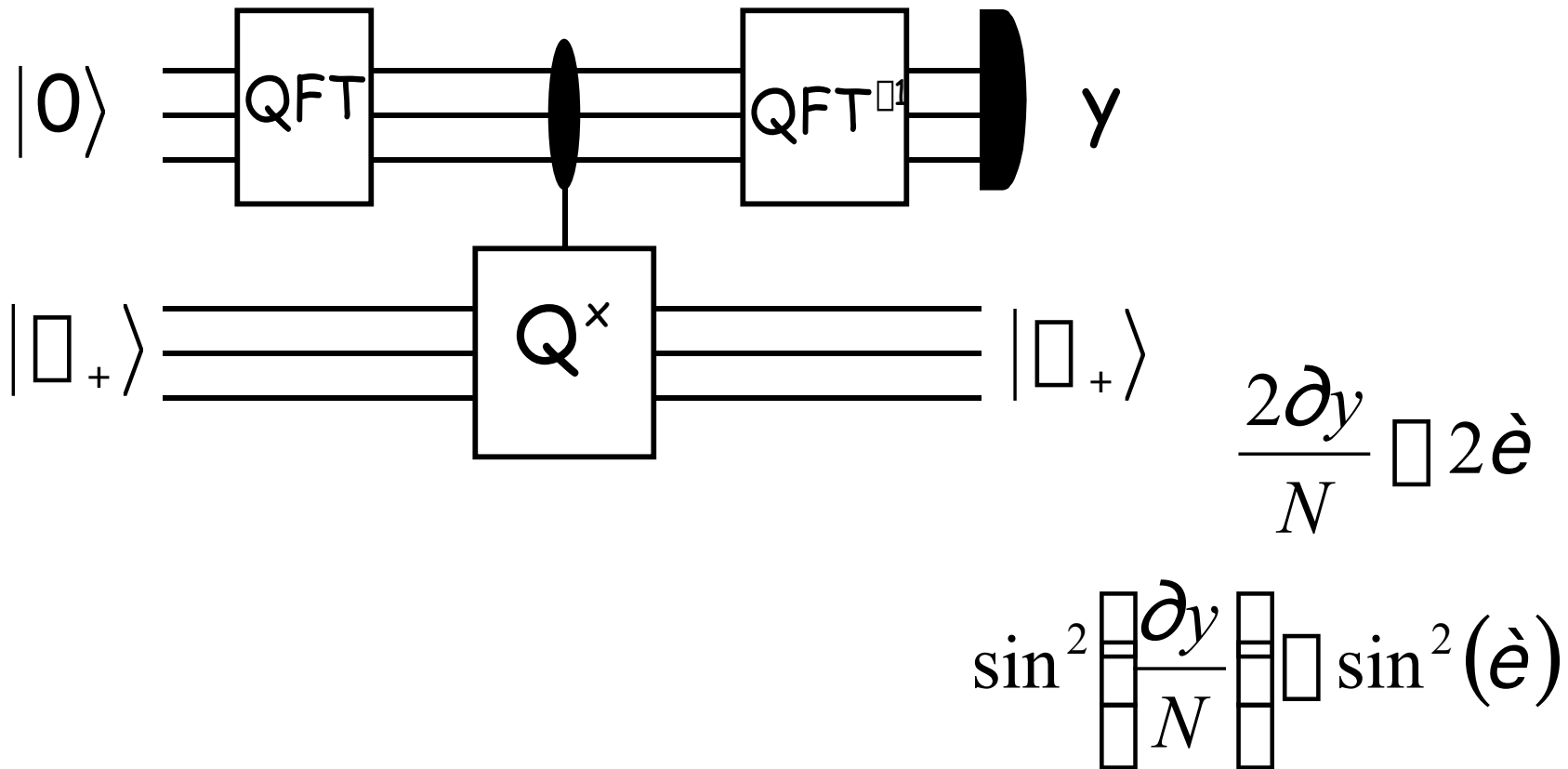
$$|\emptyset_+\rangle = \frac{1}{\sqrt{2}}|\emptyset_0\rangle + \frac{i}{\sqrt{2}}|\emptyset_1\rangle$$

$$|\emptyset_-\rangle = \frac{1}{\sqrt{2}}|\emptyset_0\rangle - \frac{i}{\sqrt{2}}|\emptyset_1\rangle$$

$$Q|\emptyset_+\rangle = e^{i2\hat{e}}|\emptyset_+\rangle \quad Q|\emptyset_-\rangle = e^{-i2\hat{e}}|\emptyset_-\rangle$$

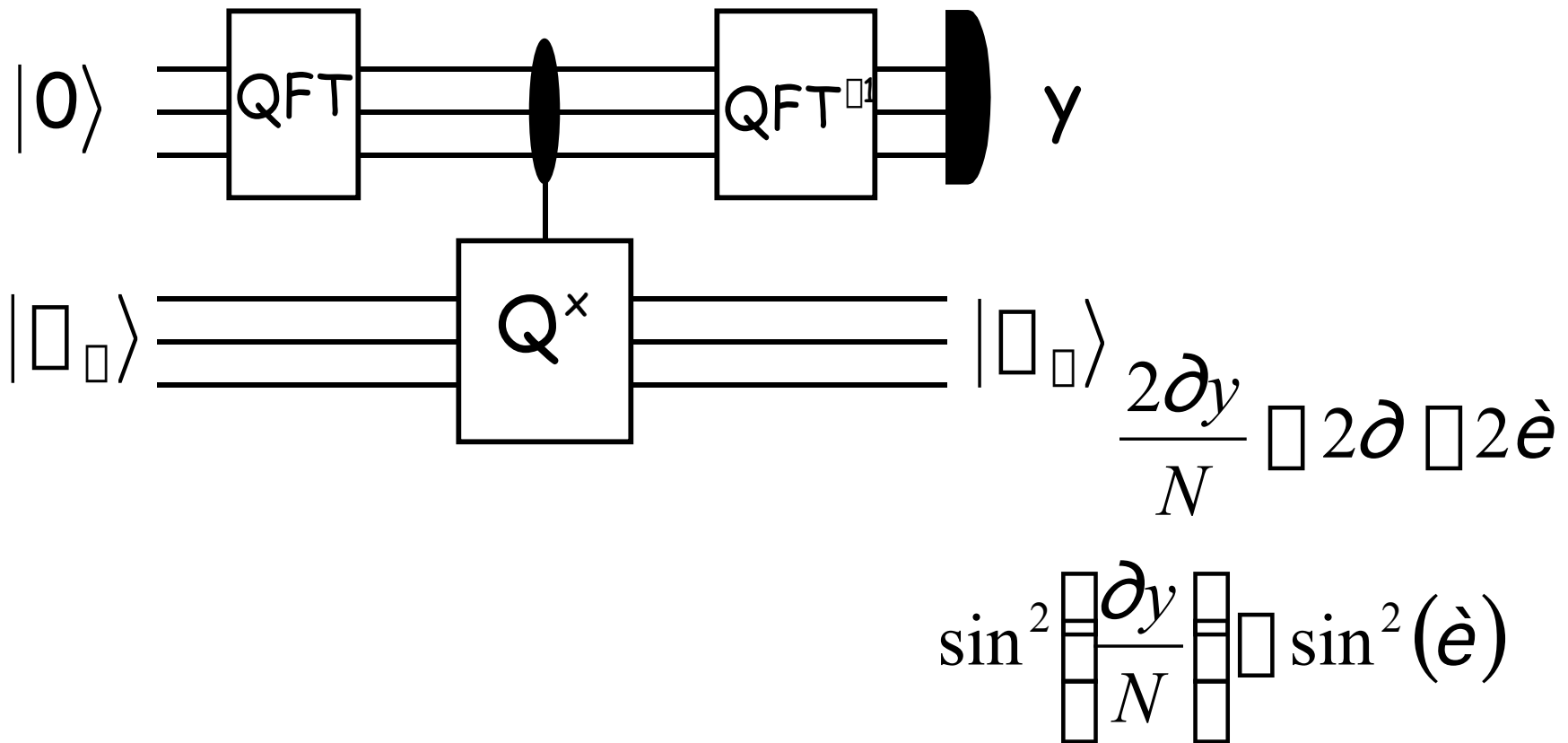
Amplitude Estimation

Eigenvalue Estimation



Amplitude Estimation

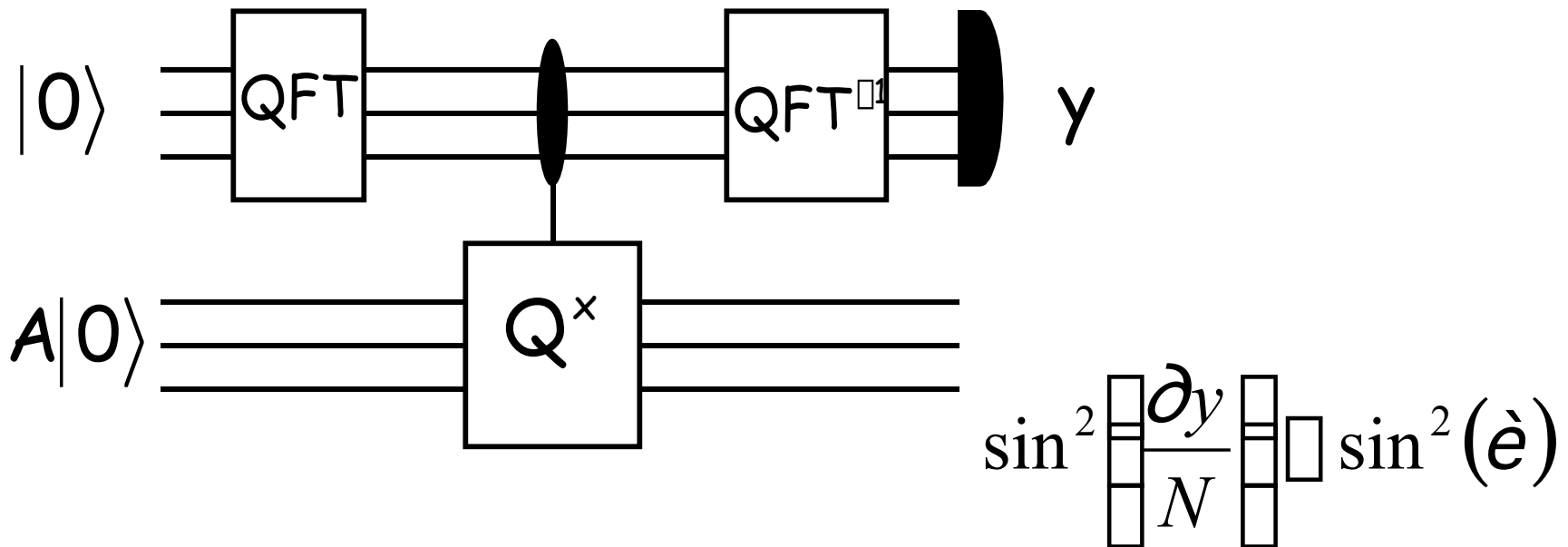
□ Eigenvalue Estimation



Amplitude Estimation

□ Eigenvalue Estimation

$$A|0\rangle = \frac{1}{\sqrt{2}} e^{i\theta} |\theta_+\rangle + \frac{1}{\sqrt{2}} e^{-i\theta} |\theta_-\rangle$$



(BBHT discovered this in the Shor picture)

Application: Tight exact counting (BBHT, BHT, M, BHMT)

Using $A|0\rangle = \sum_{x \in \{0,1\}^n} \frac{1}{\sqrt{2^n}} |x\rangle$

we have $\sin(\hat{e}_t) = \sqrt{\frac{t}{N}}$

To count exactly requires us to distinguish \hat{e}_t from \hat{e}_k , $k \neq t$

This requires precision $\hat{E} \frac{1}{\sqrt{(t+1)(2^n - t + 1)}}$

Application: Tight exact counting

QFT eigenvalue estimation techniques will give us this precision using $\tilde{O}\left(\sqrt{(t+1)(2^n - t + 1)}\right)$ applications of Q

Black-box lower bounds imply that we need $\tilde{O}\left(\sqrt{(t+1)(2^n - t + 1)}\right)$ calls to U_f

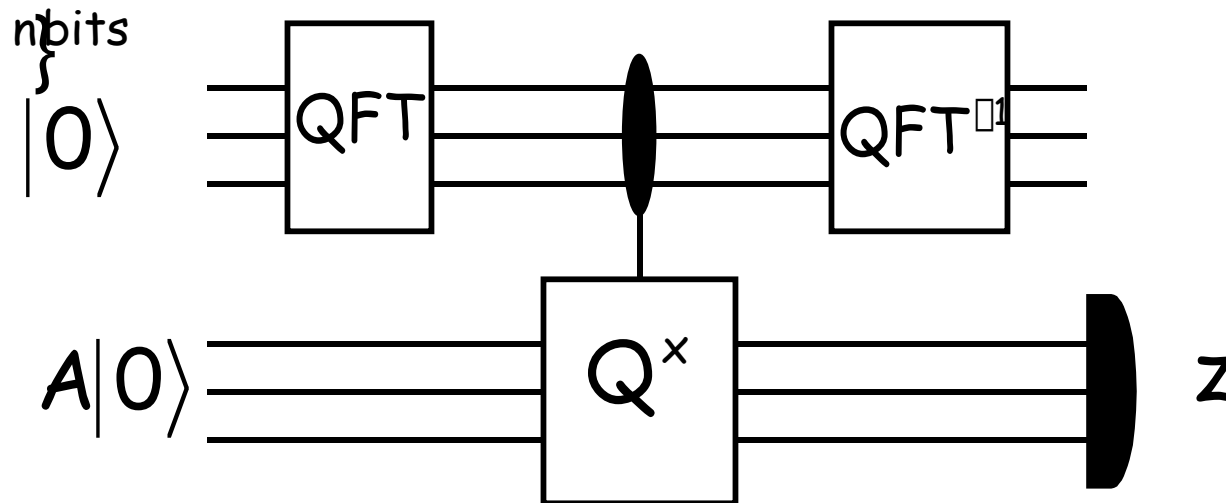
Searching when we don't know the number of solutions

Note that the amplitude estimation network produces states

$$\frac{1}{\sqrt{2}} e^{i\theta} |\tilde{e}\rangle |\emptyset_+\rangle + \frac{1}{\sqrt{2}} e^{-i\theta} |\widehat{2\theta}\tilde{e}\rangle |\emptyset_-\rangle$$

As the eigenvalue estimates become more orthogonal, the second register becomes closer and closer to an equal mixture of $\frac{1}{2} |\emptyset_+\rangle\langle\emptyset_+| + \frac{1}{2} |\emptyset_-\rangle\langle\emptyset_-| = \frac{1}{2} |\emptyset_1\rangle\langle\emptyset_1| + \frac{1}{2} |\emptyset_0\rangle\langle\emptyset_0|$

Searching when we don't know the number of solutions



$$\text{Prob}(f(z) = 1) \approx \frac{1}{2} \approx O\left(\frac{1}{2^n}\right)$$

$$\text{Prob}(f(z) = 1) \approx \frac{1}{2}$$

$n \approx$

Searching when we don't know the number of solutions

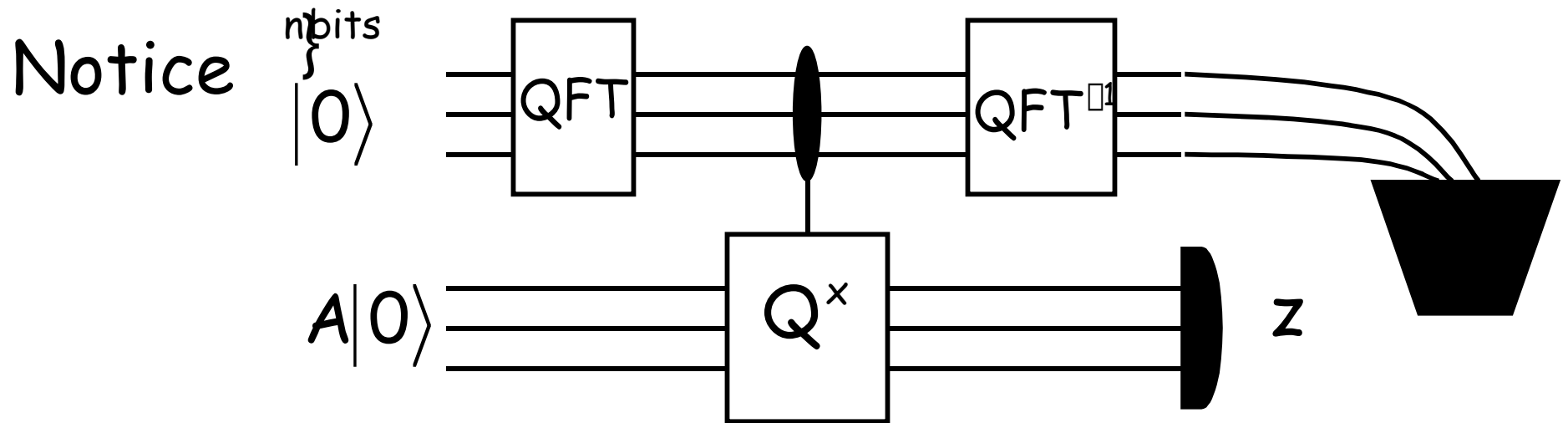
So for each $n=1,2,3,4,\dots$, we try twice to find a satisfying x

This means that once $2^n > \frac{1}{\epsilon}$ we will find a satisfying x with probability in

$$\frac{3}{4} \leq O\left(\frac{1}{2^n \epsilon}\right)$$

This means the expected running time is in $O\left(\frac{1}{\epsilon}\right)$

The way BBHT do it

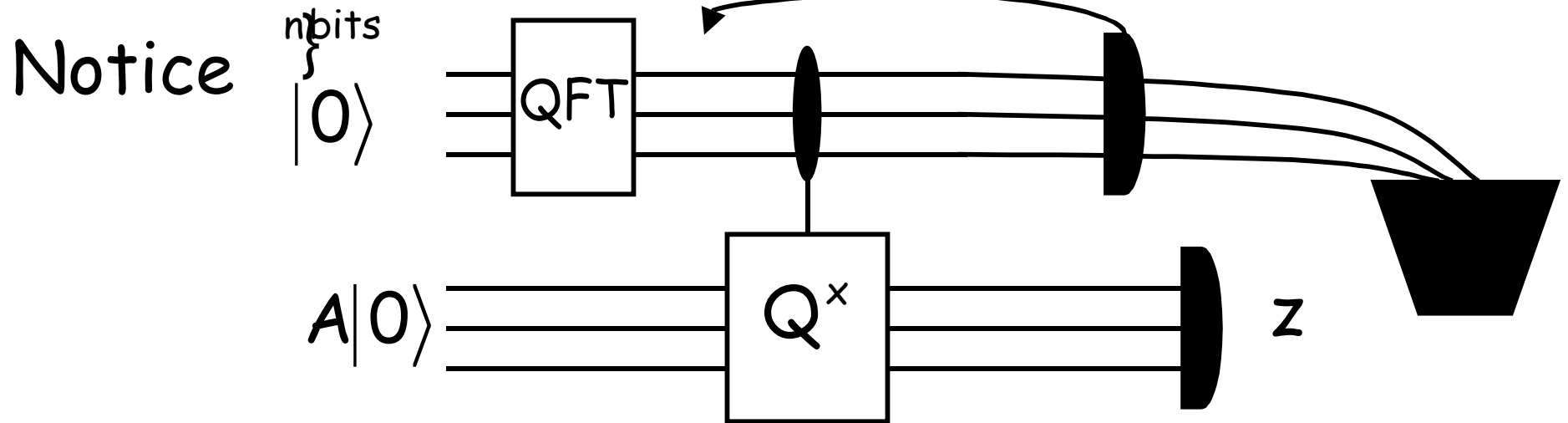


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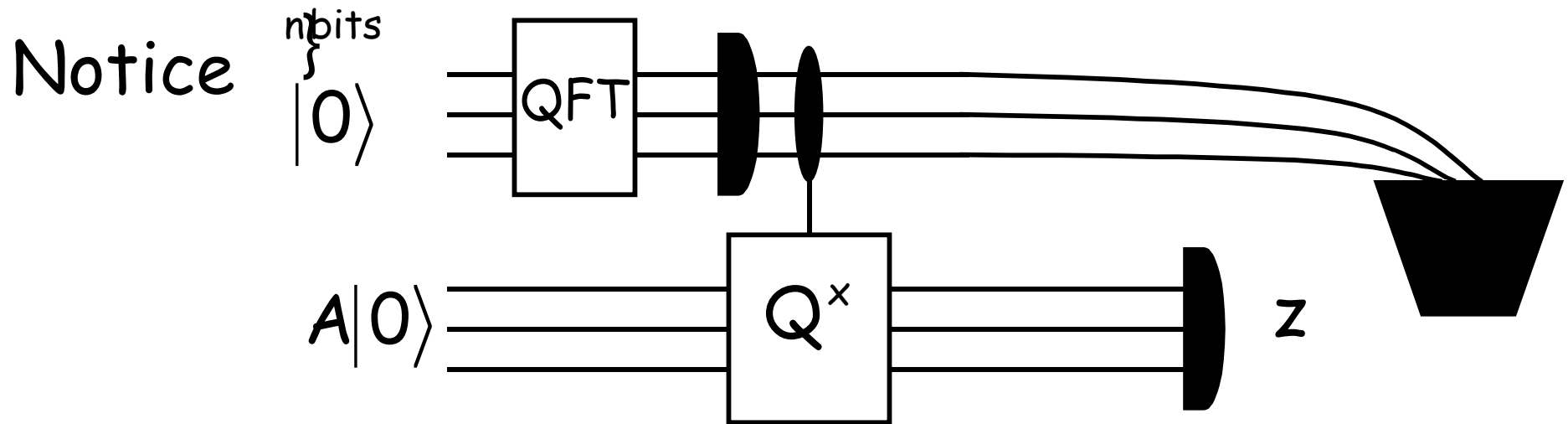


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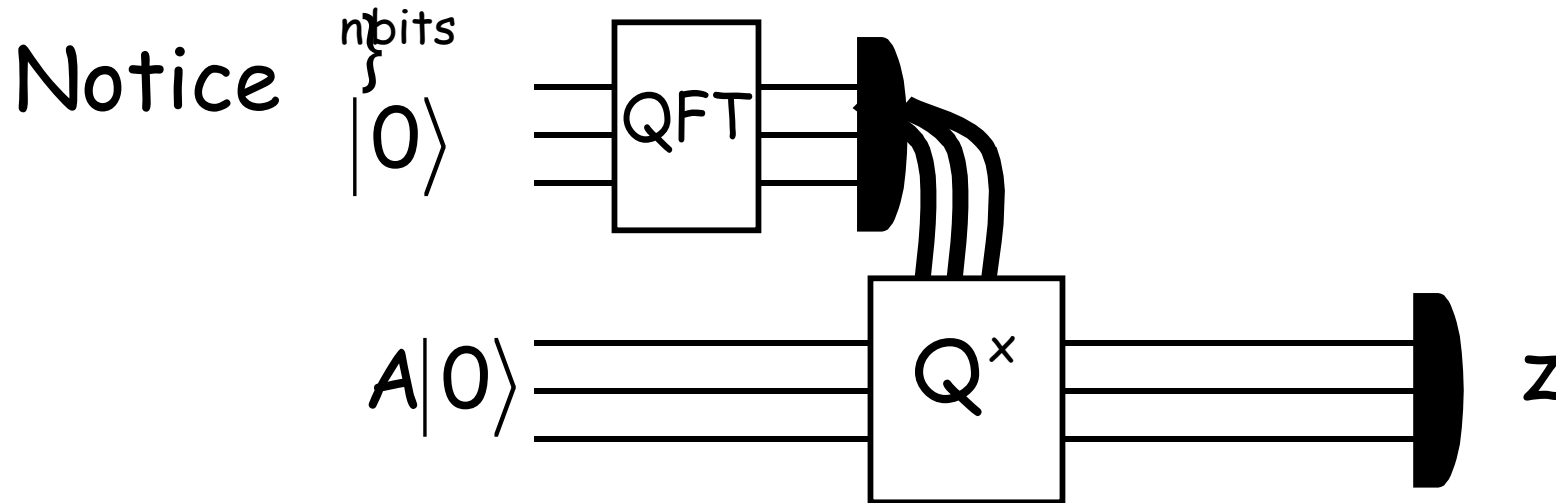


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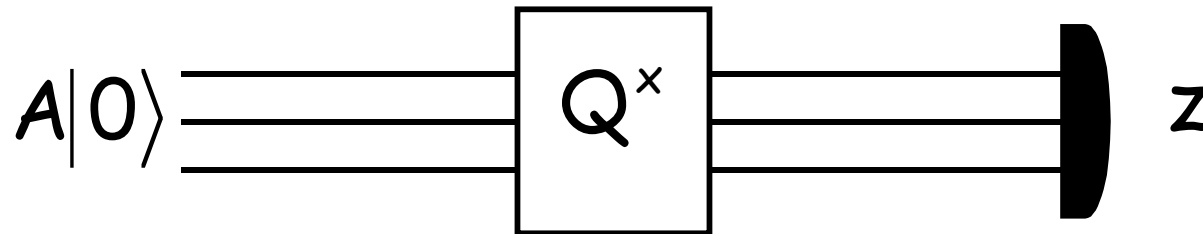
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$$\text{Prob}(f(z) = 1) \approx \frac{1}{2}$$

$n \approx$

The way BBHT do it

Pick random $x \in \{0, 1, \dots, 2^n - 1\}$



$$\text{Prob}(f(z) = 1) = \frac{1}{2} = \frac{1}{2^n} \cdot 2^{n-1}$$

$$\text{Prob}(f(z) = 1) = \frac{1}{2}$$

$n \in$