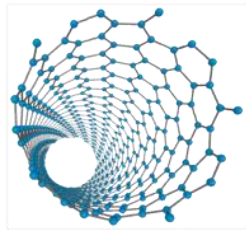


# A new light on a family of hybrid algorithms

Miguel Murça

19 April 2023



@

# Simplifying a classical-quantum algorithm interpolation with quantum singular value transformations

Duarte Magano and Miguel Murça

Phys. Rev. A **106**, 062419 – Published 16 December 2022

# Outline

- Crash course in Quantum Computing (as seen by someone doing algorithms who was previously in physics and for a physics audience)
- Quantum Phase Estimation
- “ $\alpha$ -Eigenvalue Estimation”
- Conclusion

# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

# Crash course in Quantum Computing

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→ What is a quantum computation?

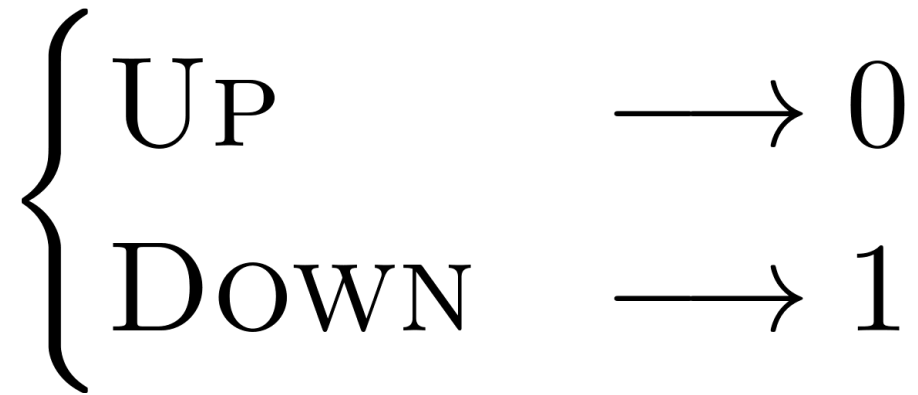
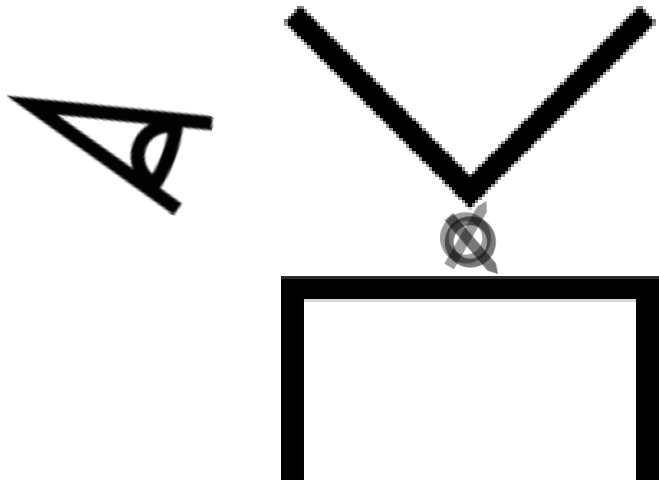
1.



2.



3.



# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

→ What is a quantum computation?

$U$   $\longleftrightarrow$  Yes/No question

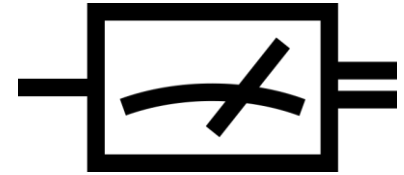
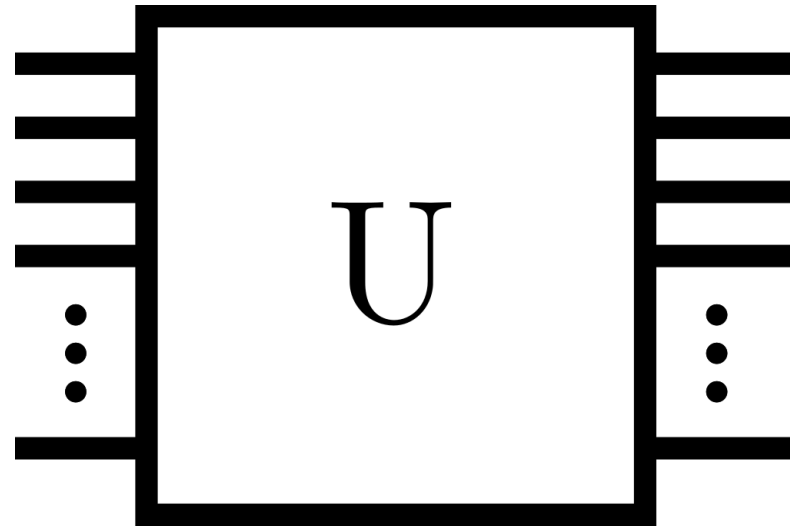
$\left\{ \begin{array}{l} \text{UP} \longrightarrow 0 \\ \text{DOWN} \longrightarrow 1 \end{array} \right.$   $\longleftrightarrow$  Yes/No answer

# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

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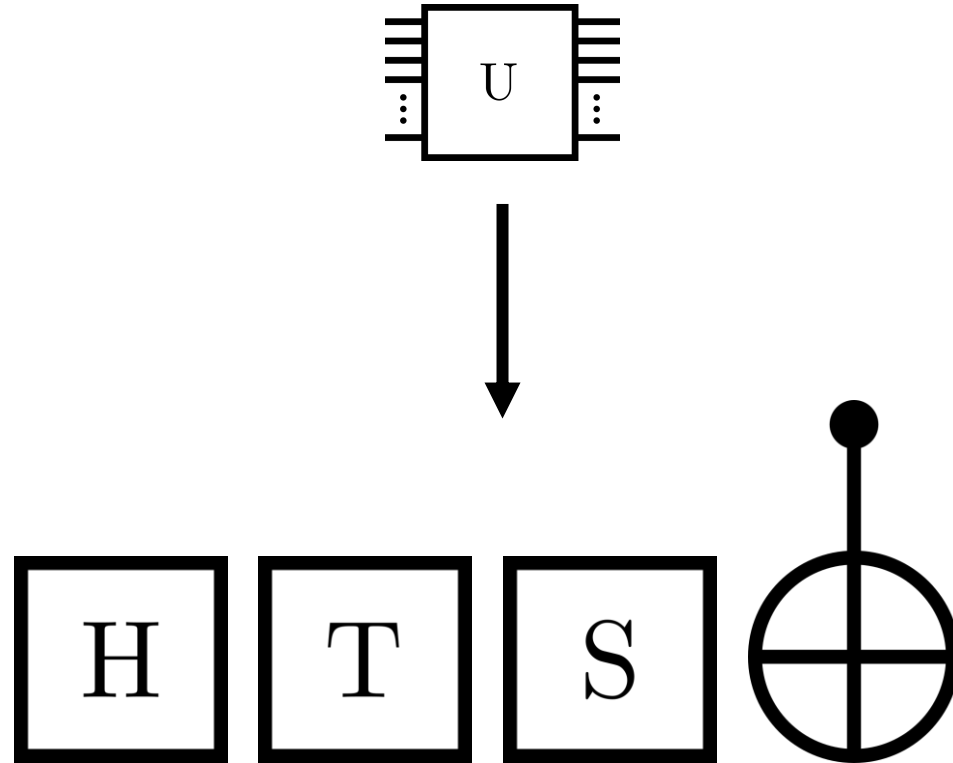
$|0\rangle, |1\rangle$



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→ What is a quantum computation?

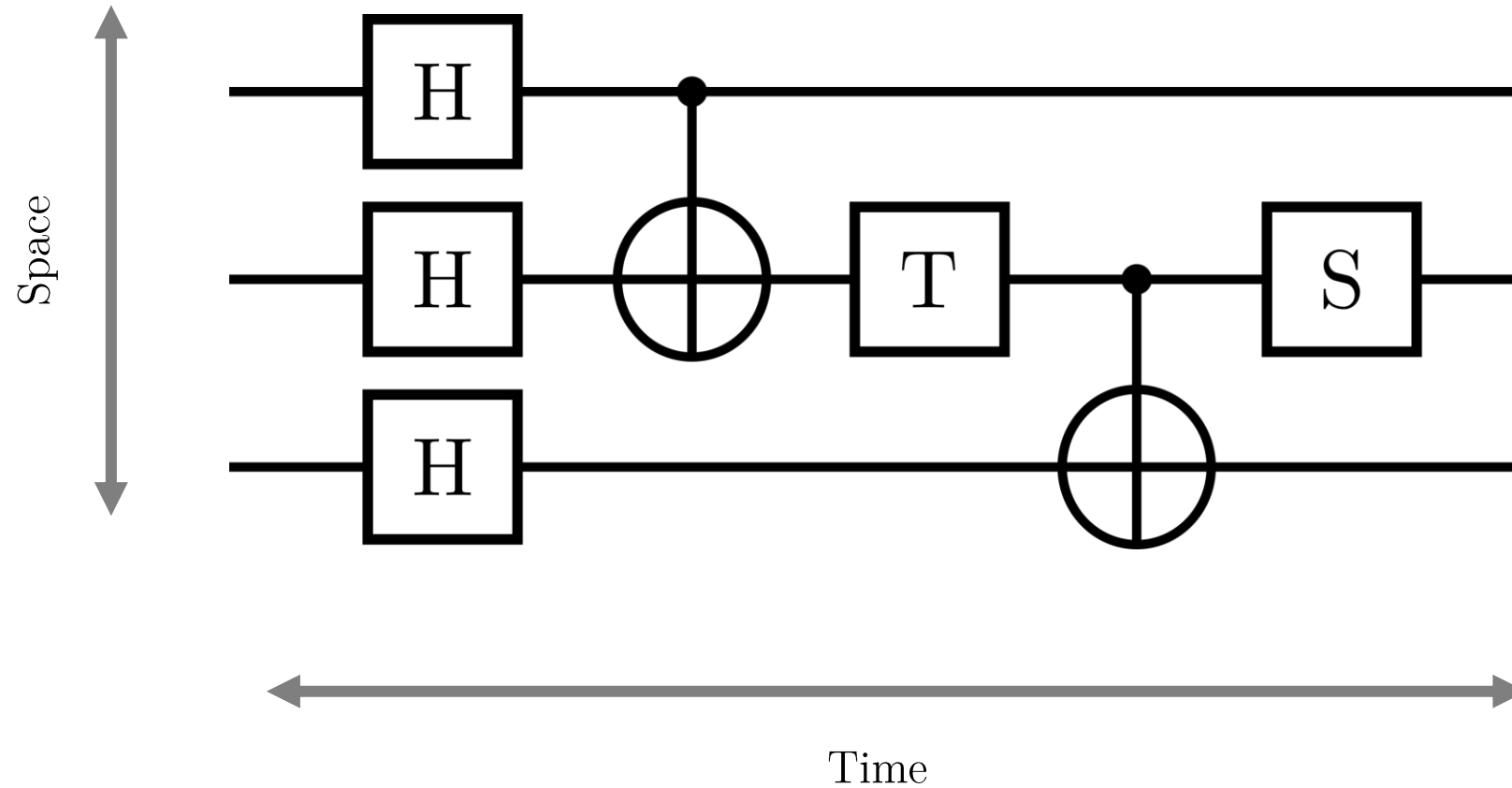




# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

→ What is a quantum computation?



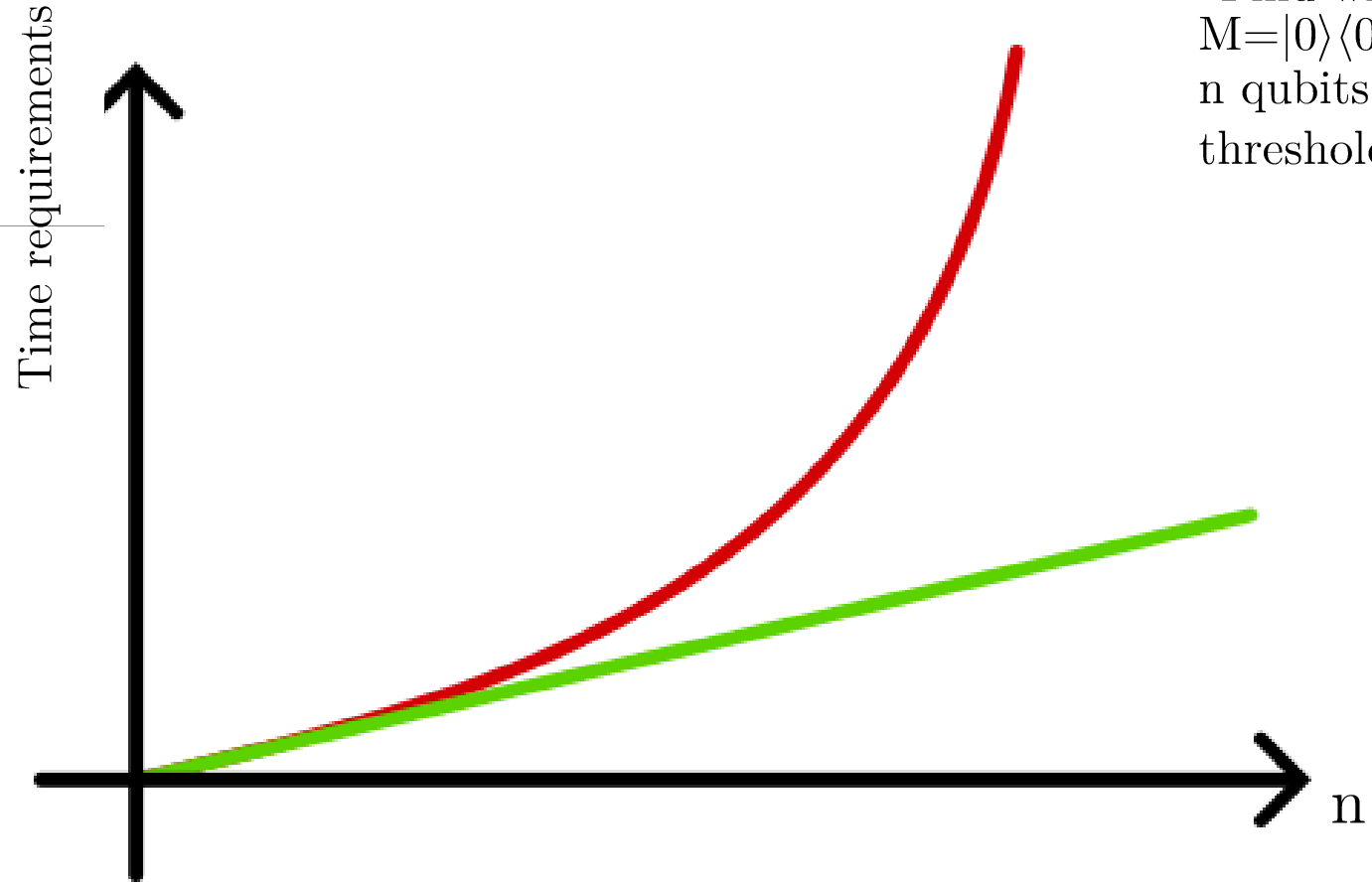
# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

→ What is a quantum computation?

$U \longleftrightarrow$  Yes/No question

$\begin{cases} \text{UP} & \rightarrow 0 \\ \text{DOWN} & \rightarrow 1 \end{cases} \longleftrightarrow$  Yes/No answer

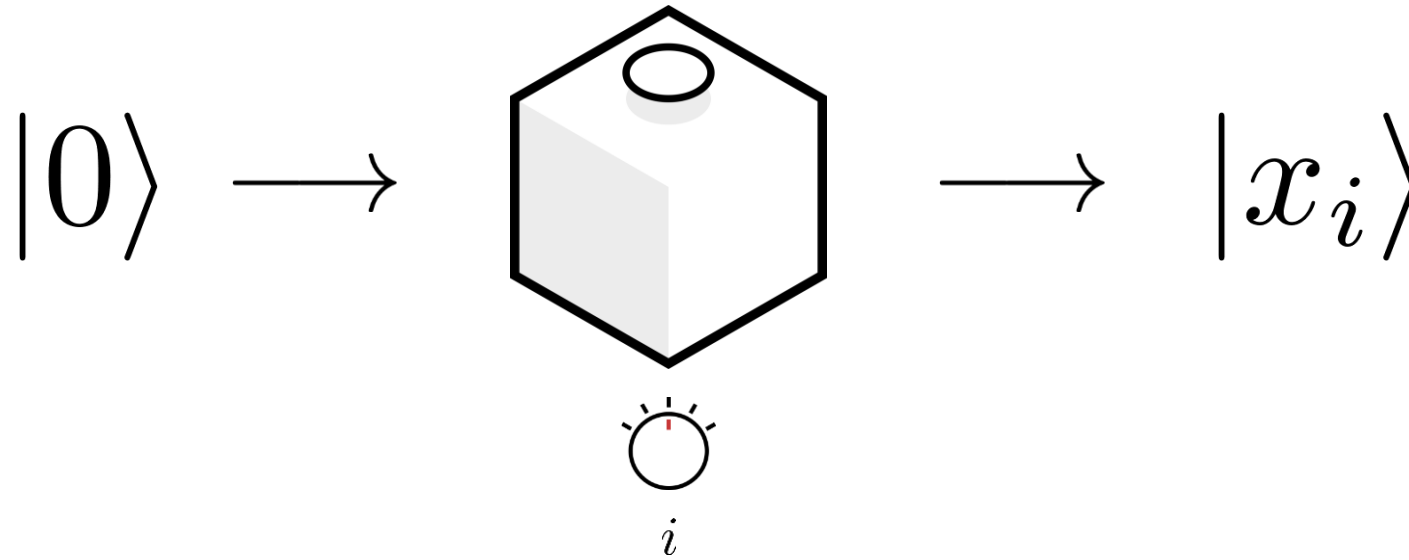
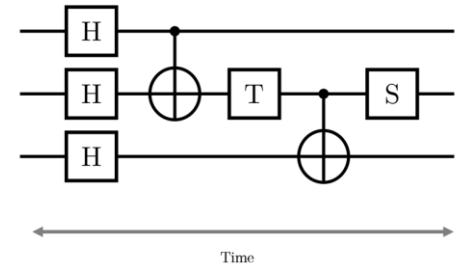


“Find whether the expectation of  $M=|0\rangle\langle 0|$  under preparable  $|\psi\rangle$  of  $n$  qubits is above or below a threshold for given  $n$ .”

# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

→ Oracle Model



$$O |0\rangle |i\rangle = |x_i\rangle |i\rangle$$

# Crash course in Quantum Computing

(as seen by someone doing algorithms who was previously in physics and for a physics audience)

→ Oracle Model

$$U = U_T O U_{T-1} \cdots O U_1 \quad \& \quad \langle 0 | U | 0 \rangle \text{ gives you the answer}$$

Then  $T$  is the complexity of the algorithm.

- Good for proofs!
- May be a good description of your problem

# Quantum Phase Estimation

# Quantum Phase Estimation

Inputs:

$U$   $2^n \times 2^n$  unitary operator

$U_\psi$  to prepare  $|\psi\rangle$ , i.e.,  $U_\psi |0\rangle = |\psi\rangle$  (Formally: access to controlled  $U_\psi, U_\psi^\dagger$ )

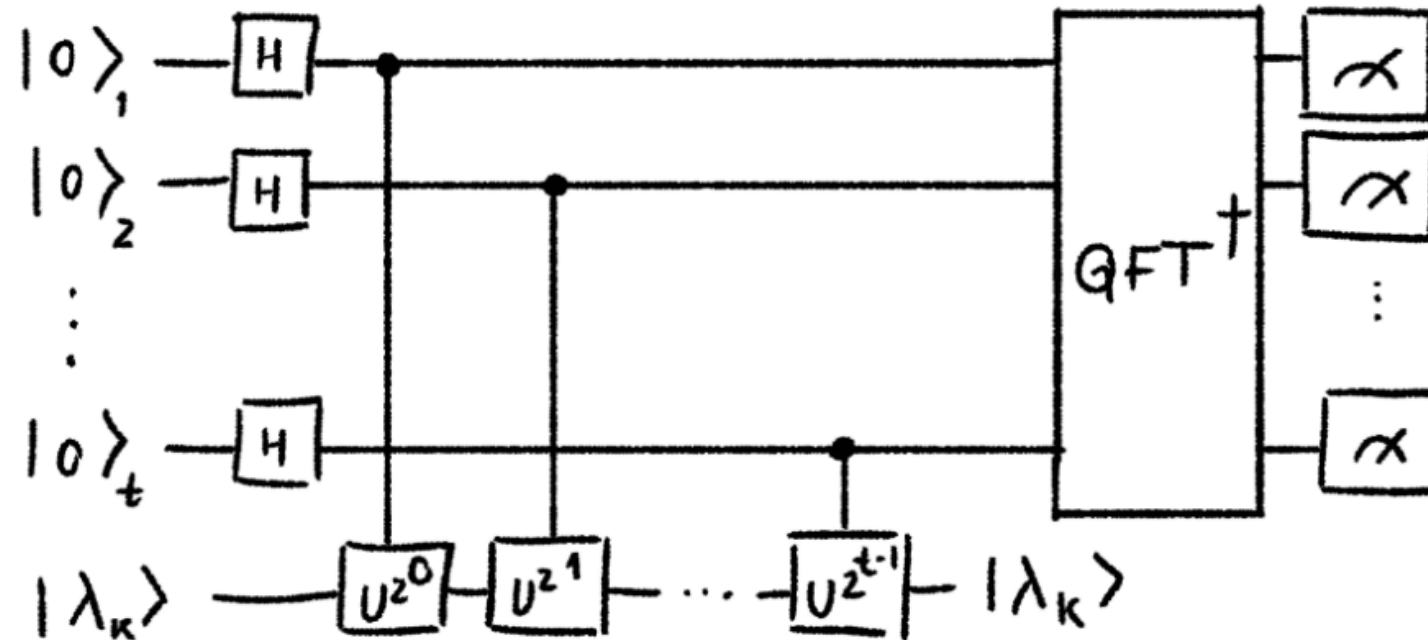
$|\psi\rangle \in \mathbb{C}^{2^n}$  such that  $U |\psi\rangle = e^{i\phi} |\psi\rangle$   $\phi \in [0, 2\pi)$

$\epsilon > 0$  precision parameter

Outputs:  $\phi$  up to precision  $\epsilon$  with bounded error probability

# Quantum Phase Estimation

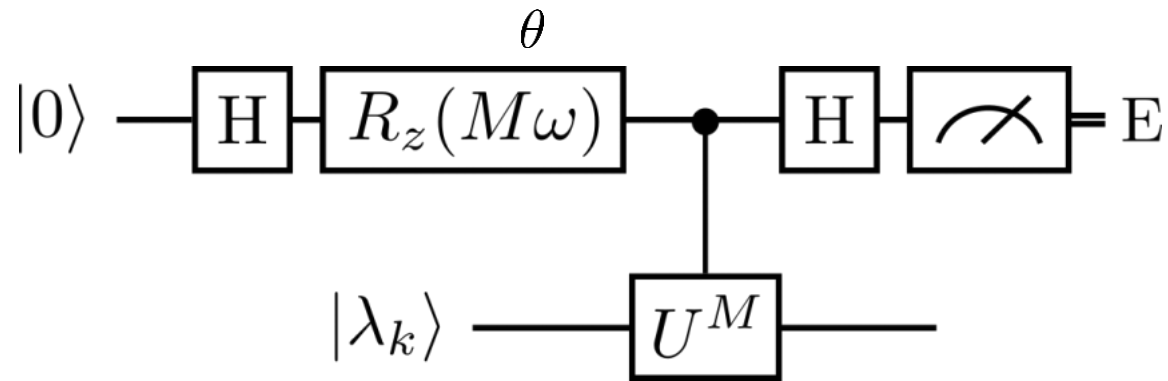
→ Textbook QPE Algorithm



(Quadratic oracle advantage)

# Quantum Phase Estimation

→ Kitaev's/Iterative Phase Estimation Algorithm



$$(M, \theta)_1 = (2^{m-1}, 0)$$

$$(M, \theta)_2 = (2^{m-2}, -\pi \cdot 2^{-m} \cdot \phi_m)$$

$$(M, \theta)_3 = (2^{m-3}, -\pi \cdot (2^{-m} \cdot \phi_m + 2^{-m+1} \cdot \phi_{m-1}))$$

⋮

$$(M, \theta)_m = (1, -\pi \cdot (2^{-m} \cdot \phi_m + \dots + 2^{-2} \cdot \phi_2))$$

$$\phi = 0.\phi_1\phi_2 \dots \phi_m \dots$$

⌋

$$\phi = 0.\phi_m \dots$$

⌋

$$\phi = 0.\phi_{m-1}\phi_m \dots - 0.0\phi_m$$

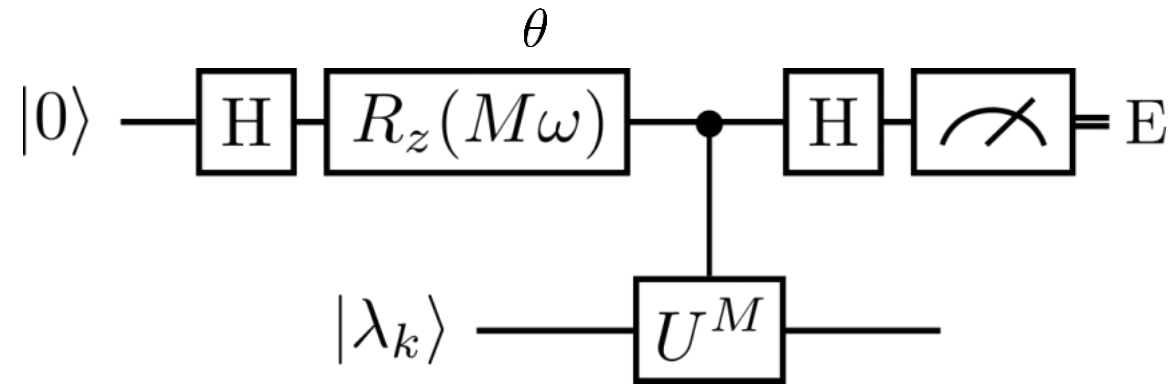
⌋

⋮



# Quantum Phase Estimation

→ Faster Phase Estimation



$$P(0|\phi; \theta, M) = \frac{1 + \cos(M[\phi + \theta])}{2},$$
$$P(1|\phi; \theta, M) = \frac{1 - \cos(M[\phi + \theta])}{2}.$$

*[Submitted on 2 Apr 2013]*

## **Faster Phase Estimation**

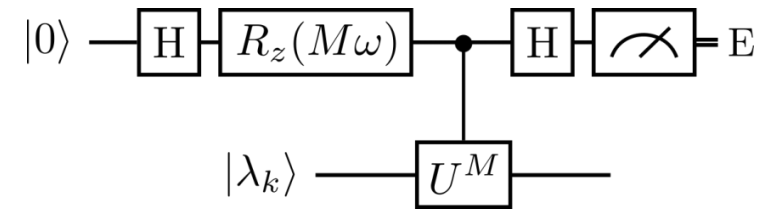
[Krysta M. Svore](#), [Matthew B. Hastings](#), [Michael Freedman](#)

Idea: new “schedules” for  $M$  and  $\omega$

Plus: Informational perspective

# Quantum Phase Estimation

→  $\alpha$ -Quantum Phase Estimation



## Efficient Bayesian Phase Estimation

Nathan Wiebe and Chris Granade  
Phys. Rev. Lett. 117, 010503 – Published 30 June 2016

## Accelerated Variational Quantum Eigensolver

Daochen Wang, Oscar Higgott, and Stephen Brierley  
Phys. Rev. Lett. 122, 140504 – Published 12 April 2019

## Low depth algorithms for quantum amplitude estimation

Tudor Giurgica-Tiron, Iordanis Kerenidis, Farrokh Labib,  
Anupam Prakash, and William Zeng  
Quantum 6, 745 (2022)

$$M \sim 1/\epsilon, \theta = \mu$$

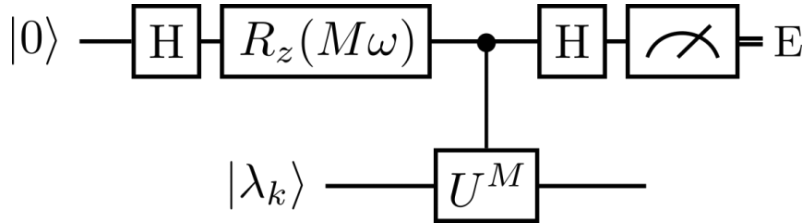
$$(M, \theta) = \left( \frac{1}{\epsilon^\alpha}, \mu - \epsilon \right)$$

$$N(\alpha) = \begin{cases} \frac{2}{1-\alpha} \left( \frac{1}{\epsilon^{2(1-\alpha)}} - 1 \right) & \text{if } \alpha \in [0, 1) \\ 4 \log(1/\epsilon) & \text{if } \alpha = 1 \end{cases}$$

$$D(\alpha) = \mathcal{O}(1/\epsilon^\alpha)$$

# Quantum Phase Estimation

→ Hybridization/Query Perspective

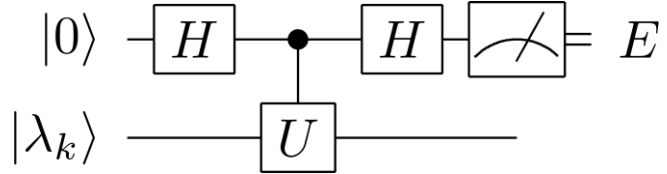


$$P(0|\phi; \theta, M) = \frac{1 + \cos(M[\phi + \theta])}{2},$$

$$P(1|\phi; \theta, M) = \frac{1 - \cos(M[\phi + \theta])}{2}.$$

$$P(x|\phi; \theta, M) = \frac{1 + (-1)^x \cos(M[\phi + \theta])}{2}$$

### Classical Sampling

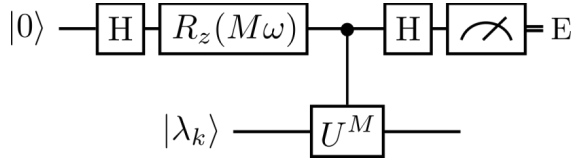


$D = \mathcal{O}(1)$       trivially

$N = \mathcal{O}(1/\epsilon^2)$       use Cramér-Rao bound

$T = N \times D = \sum D = \mathcal{O}(1/\epsilon^2)$

### Fully Coherent



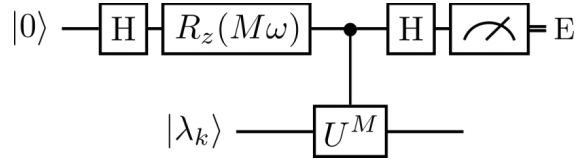
Plus Kitaev schedule

$D = \mathcal{O}(1/\epsilon)$

$T = \tilde{\mathcal{O}}(1/\epsilon)$

By inspection of the schedule

### Hybrid ( $\alpha$ -QPE)

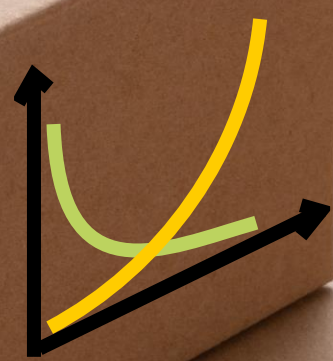
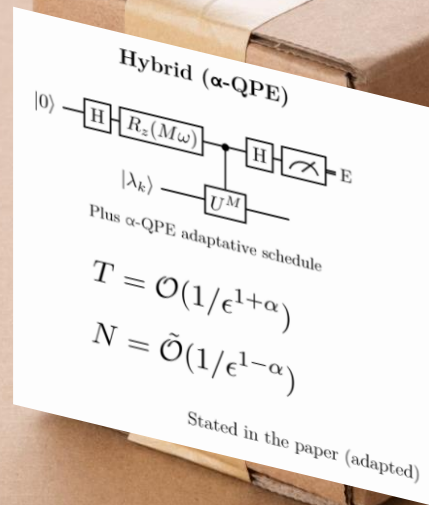


Plus  $\alpha$ -QPE adaptive schedule

$D = \mathcal{O}(1/\epsilon^{1-\alpha})$

$T = \tilde{\mathcal{O}}(1/\epsilon^{1+\alpha})$

Stated in the paper (adapted)



# $\alpha$ -Eigenvalue Estimation

# $\alpha$ -Eigenvalue Estimation

→ Quantum Singular Value Transformations

1. Block-Encode a Matrix

$$U = \begin{bmatrix} A & * \\ * & * \end{bmatrix} = A \otimes |0\rangle\langle 0| + \dots$$

2. Choose a suitable polynomial

$$\begin{aligned} p &\in \mathbb{R}[x^d] \\ |p(x)| &\leq 1 \text{ for } x \in [-1, 1] \\ p &\text{ has parity } d \end{aligned}$$

3. Then with  $d$  applications of  $U$  you can produce

$$U' = \begin{bmatrix} p(A) & * \\ * & * \end{bmatrix} = \begin{bmatrix} \sum_{\lambda} p(\lambda) |\lambda\rangle\langle\lambda| & * \\ * & * \end{bmatrix}$$

$$A = \sum_{\lambda} \lambda |\lambda\rangle\langle\lambda|$$

# $\alpha$ -Eigenvalue Estimation

→ “Phase Estimation is a weaker form of Eigenvalue Estimation”

**Plan:** PE  $\preceq$  AE  $\preceq$  EE

**Define:**

## Amplitude Estimation

$$A|0^m\rangle = \sqrt{p}|good\rangle + \sqrt{1-p^2}|bad\rangle$$

$$O_A|good/bad\rangle = \pm|good/bad\rangle$$

Input:  $A, A^\dagger, O_A, \epsilon > 0$

Output:  $|p|$  up to  $\epsilon$  with bounded error probability

## Eigenvalue Estimation

$$\mathcal{H} \in \mathbb{H}_N \quad \mathcal{H}|\psi\rangle = E|\psi\rangle$$

$U_H$  is a  $(\gamma, m)$ -BE of  $\mathcal{H}$

$$U_\psi|0^m\rangle = |\psi\rangle$$

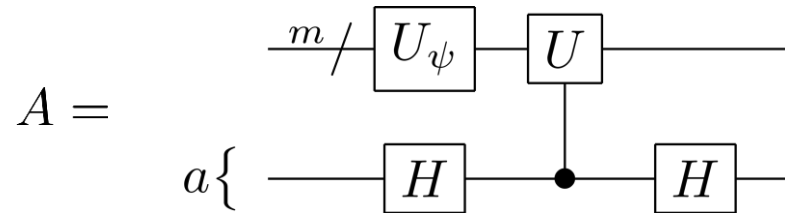
Input:  $U_\psi, U_H, U_H^\dagger, \gamma, \epsilon > 0$

Output:  $E$  up to  $\epsilon$  with bounded error probability

# $\alpha$ -Eigenvalue Estimation

→ “Phase Estimation is a weaker form of Eigenvalue Estimation”

PE  $\preceq$  AE

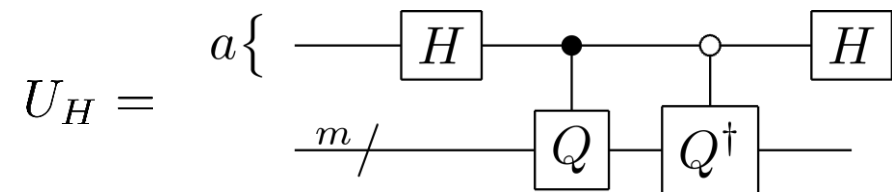


$$A |0^m\rangle |0\rangle = \cos(\phi/2) |\psi\rangle |0\rangle - i \sin(\phi/2) |\psi\rangle |1\rangle$$

↓

$$\cos(\phi/2) \rightsquigarrow |\phi|$$

AE  $\preceq$  EE



$$Q = A(2 |0^m\rangle\langle 0^m| - I)A^\dagger O_A$$

$$\langle 0|_a U_H |0\rangle_a U_\psi |0^m\rangle = (1 - 2p)U_\psi |0^m\rangle$$

↓

$$\mu = (1 - 2p) \rightsquigarrow p$$

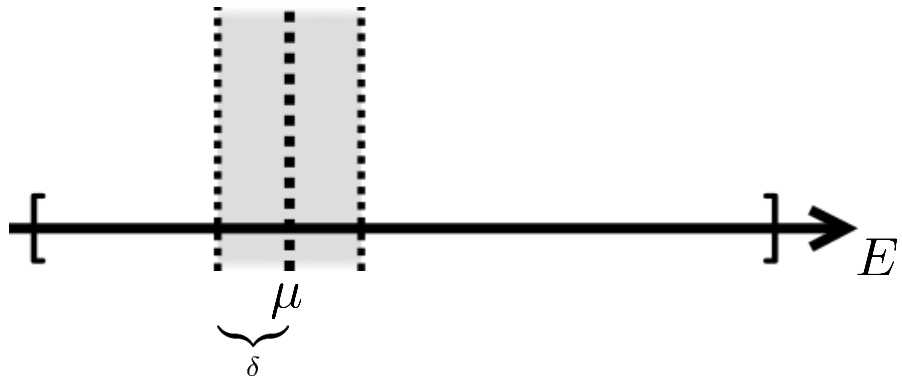


# $\alpha$ -Eigenvalue Estimation

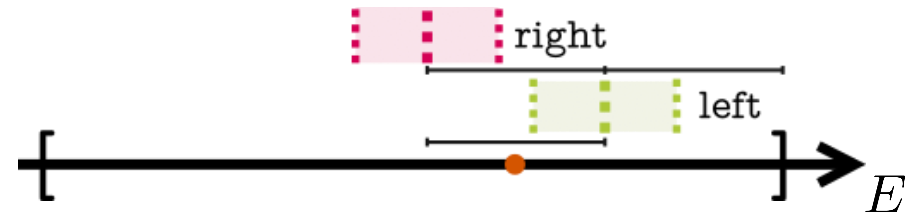
$\Rightarrow$  If we focus on EE (and find an interpolation for EE) we solve everything else

Now note:

If you can solve the decision problem



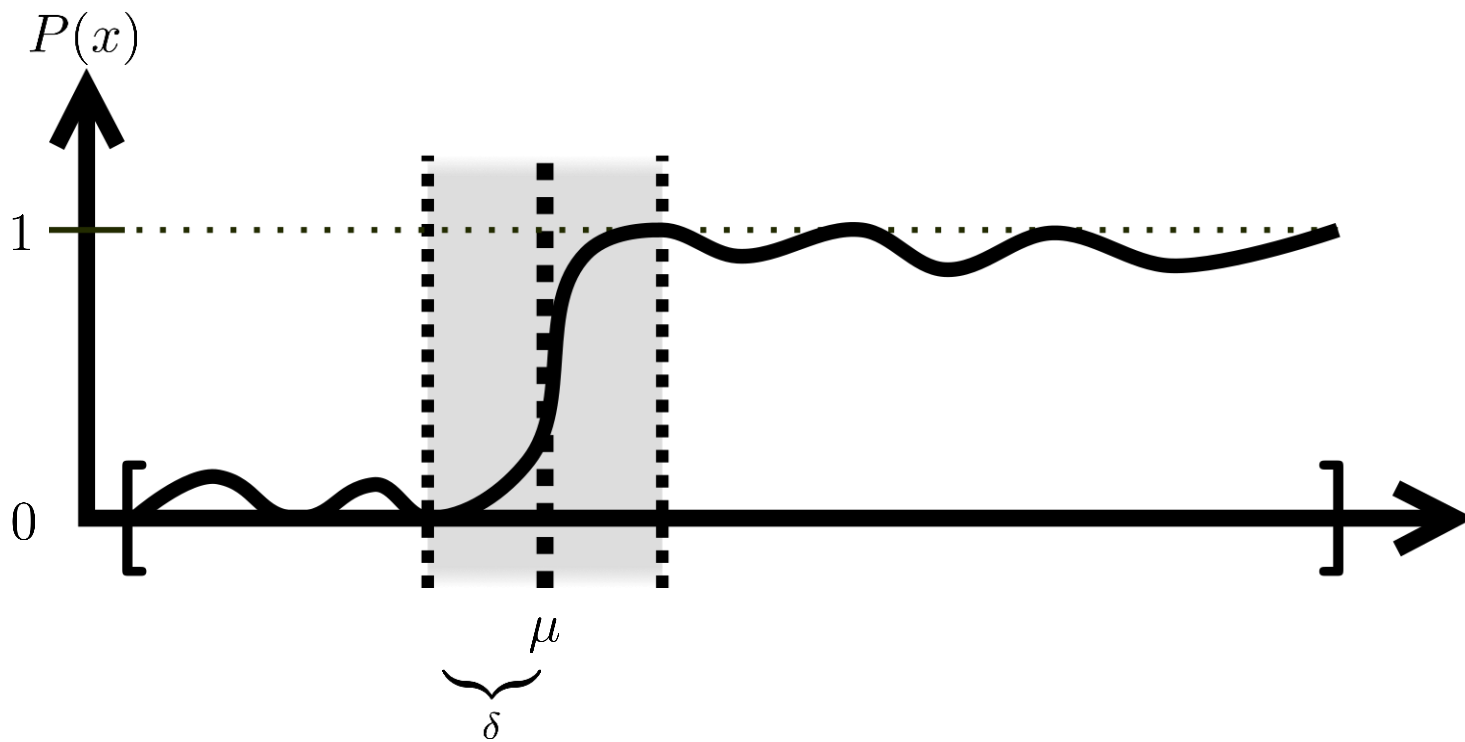
Then you can solve EE with a binary search



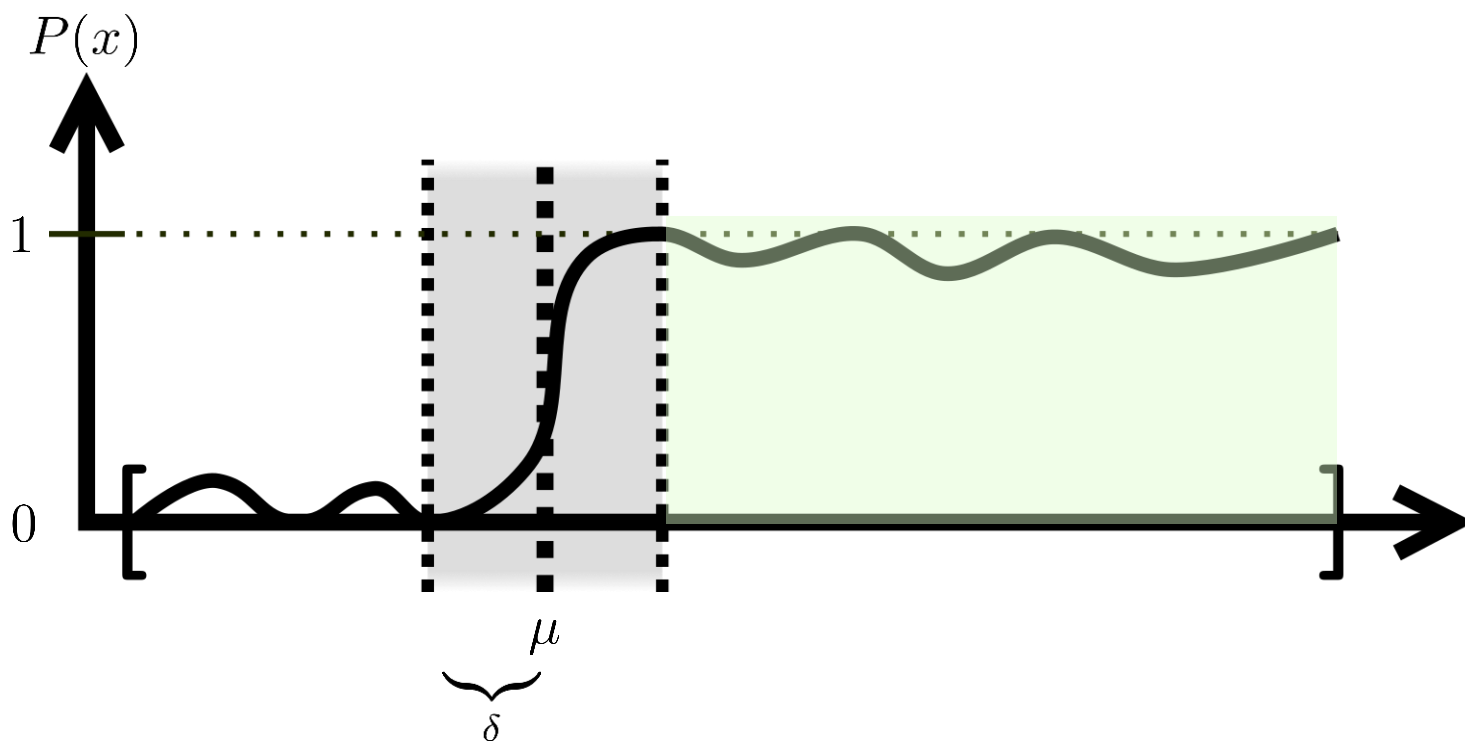
# $\alpha$ -Eigenvalue Estimation

$$\begin{matrix} & |0\rangle & |1\rangle \\ \langle 0| & P(\mathcal{H}) & \cdot \\ \langle 1| & \cdot & \cdot \end{matrix} |0\rangle \otimes |\lambda_k\rangle \rightsquigarrow P(\lambda_k) |0\rangle |\lambda_k\rangle + |1\rangle \otimes (\dots)$$

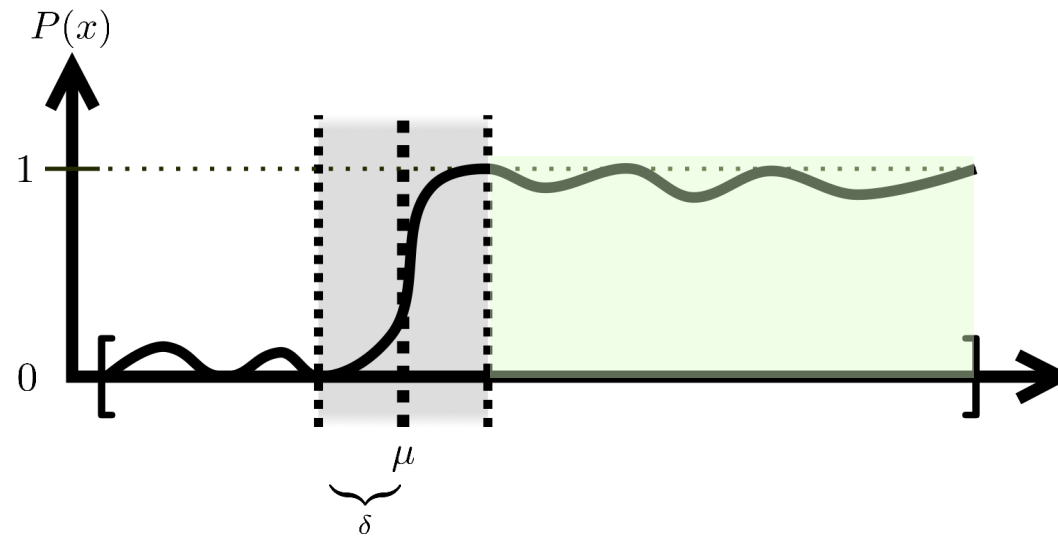
# $\alpha$ -Eigenvalue Estimation



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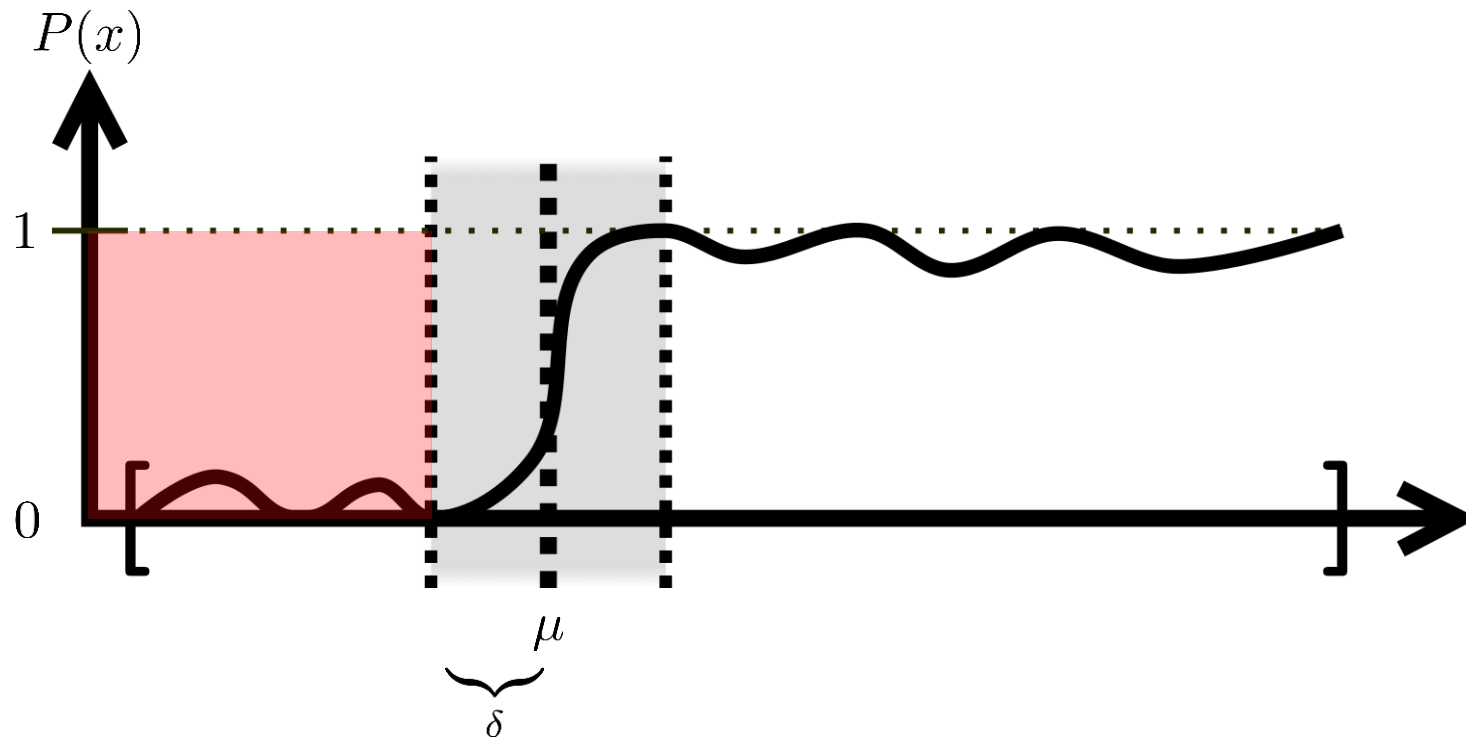
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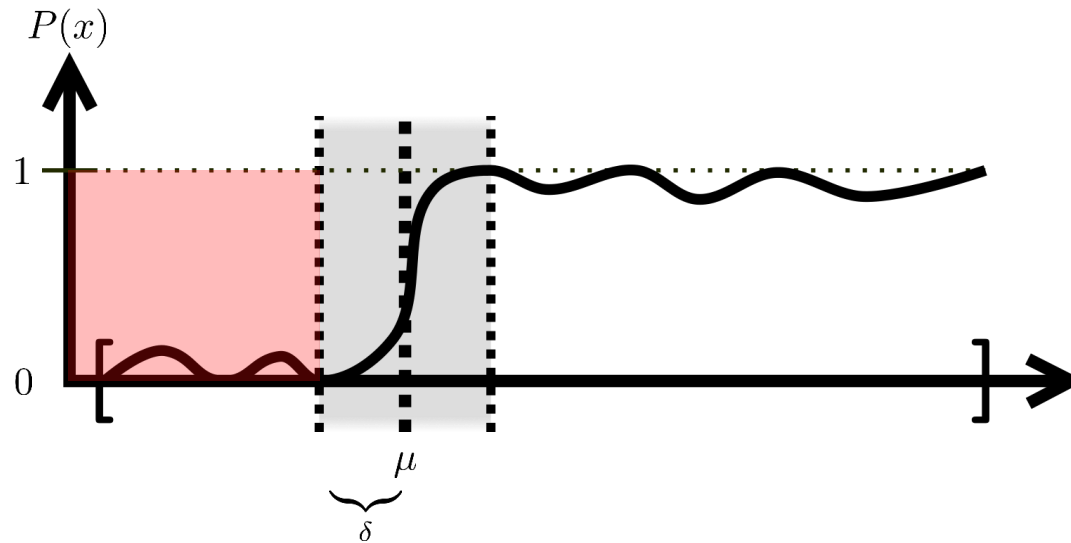
$$P(\lambda_k) |0\rangle |\lambda_k\rangle \approx |0\rangle |\lambda\rangle$$

# $\alpha$ -Eigenvalue Estimation

→ QSVT



# $\alpha$ -Eigenvalue Estimation

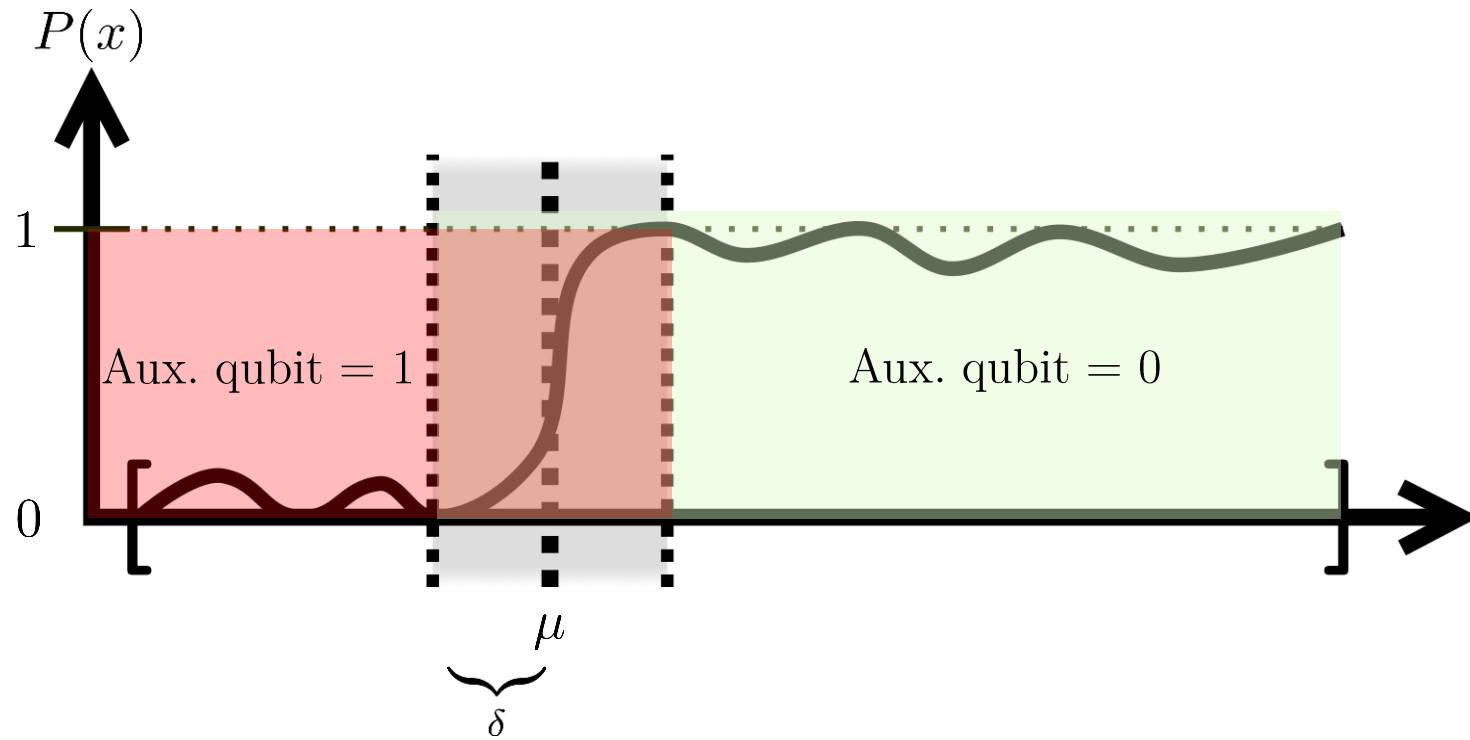


$$\begin{pmatrix} P(\mathcal{H}) & \cdot \\ \cdot & \cdot \end{pmatrix} = U_H |0\rangle |\lambda_k\rangle = P(\lambda_k) |0\rangle |\lambda_k\rangle + |1\rangle |u_j\rangle = |1\rangle |u_j\rangle$$



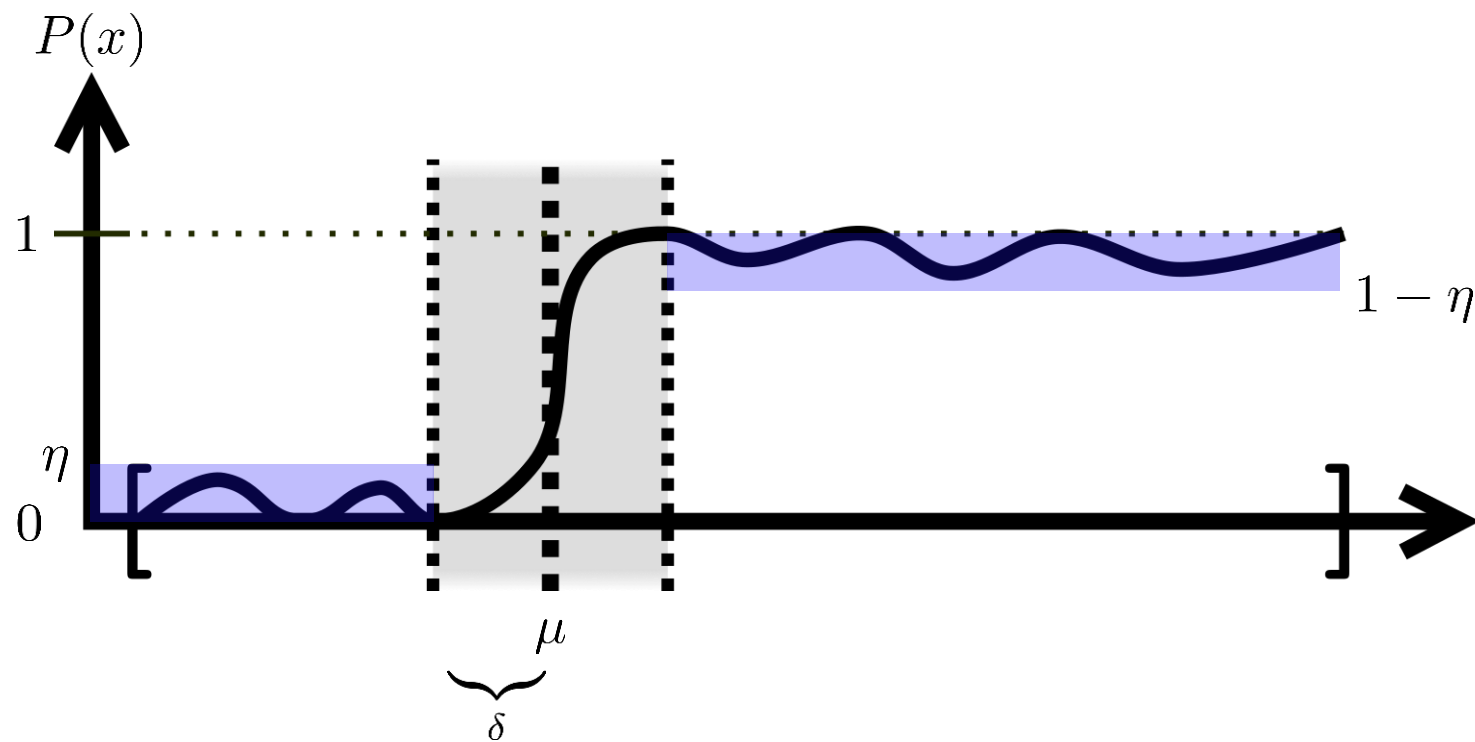
Must be unitary!

# $\alpha$ -Eigenvalue Estimation





# $\alpha$ -Eigenvalue Estimation



# $\alpha$ -Eigenvalue Estimation

→ Decision problem

⇒ If you can implement a step function, you can solve EE.

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... Can you implement a step function?

# $\alpha$ -Eigenvalue Estimation

→ Decision problem

⇒ If you can implement a step function, you can solve EE.

... Can you implement a step function?

## Near-optimal ground state preparation

Lin Lin<sup>1,2</sup> and Yu Tong<sup>1</sup>

<sup>1</sup>Department of Mathematics, University of California, Berkeley, CA 94720, USA

<sup>2</sup>Computational Research Division, Lawrence Berkeley National Laboratory, Berkeley, CA 94720, USA

Published: 2020-12-14, volume 4, page 372

Citation: Quantum 4, 372 (2020).

Yep! (Sort of)

# $\alpha$ -Eigenvalue Estimation

→ Decision problem

## Near-optimal ground state preparation

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Let  $U_H$  be a  $(\gamma, m)$ -block-encoding of a Hermitian matrix  $H$  and  $\mu_0 \in [0, \gamma]$ . Then, there is a  $(1, m + 3)$ -block-encoding of  $P\left(\frac{H - \mu_0 I}{\gamma + \mu_0}; \delta, \eta\right)$ , where  $P$  satisfies

$$\forall x \in [-1, -\delta], 0 \leq P(x; \delta, \eta) \leq \eta/2 \quad (1)$$

$$\text{and } \forall x \in [\delta, 1], 1 - \eta/2 \leq P(x; \delta, \eta) \leq 1, \quad (2)$$

using  $\mathcal{O}\left(\frac{1}{\delta} \log\left(\frac{1}{\eta}\right)\right)$  queries of  $U_H$  and  $U_H^\dagger$ .

(Adapted from lemma 5)

# $\alpha$ -Eigenvalue Estimation

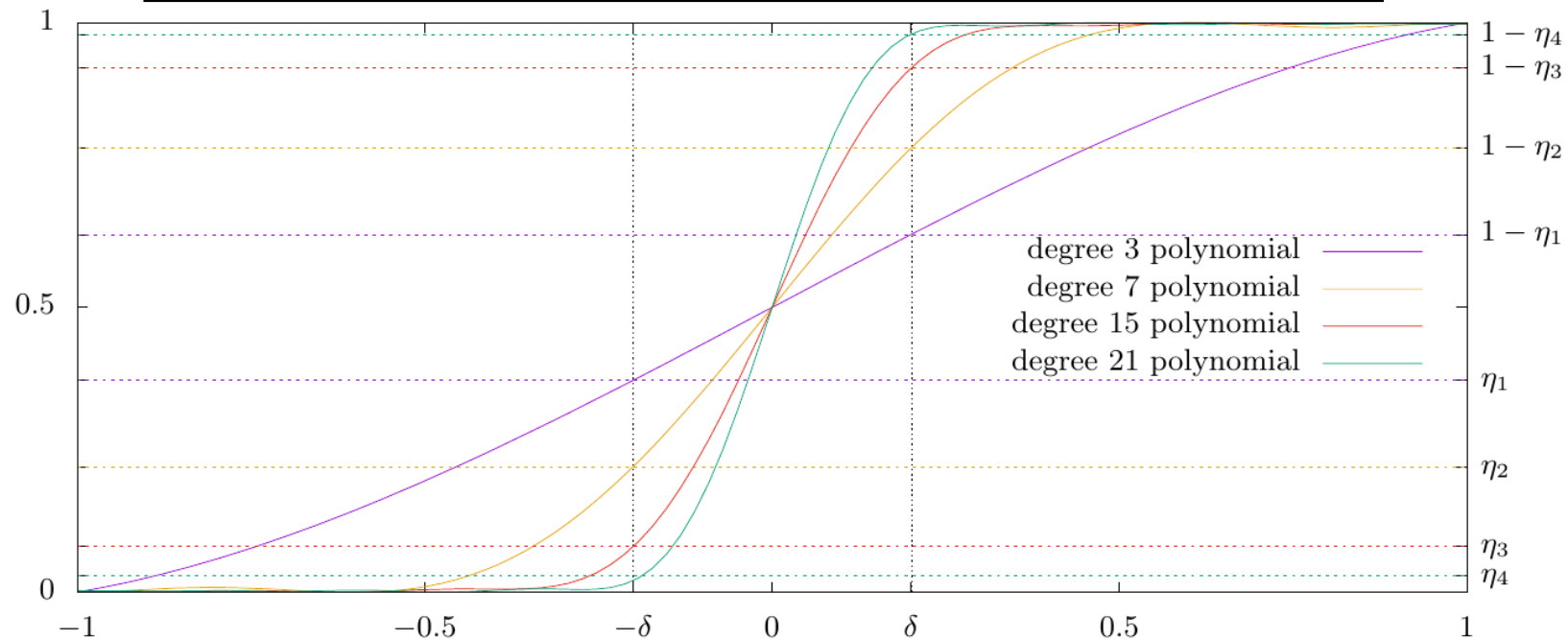
→ Decision problem

Let  $U_H$  be a  $(\gamma, m)$ -block-encoding of a Hermitian matrix  $H$  and  $\mu_0 \in [0, \gamma]$ .  
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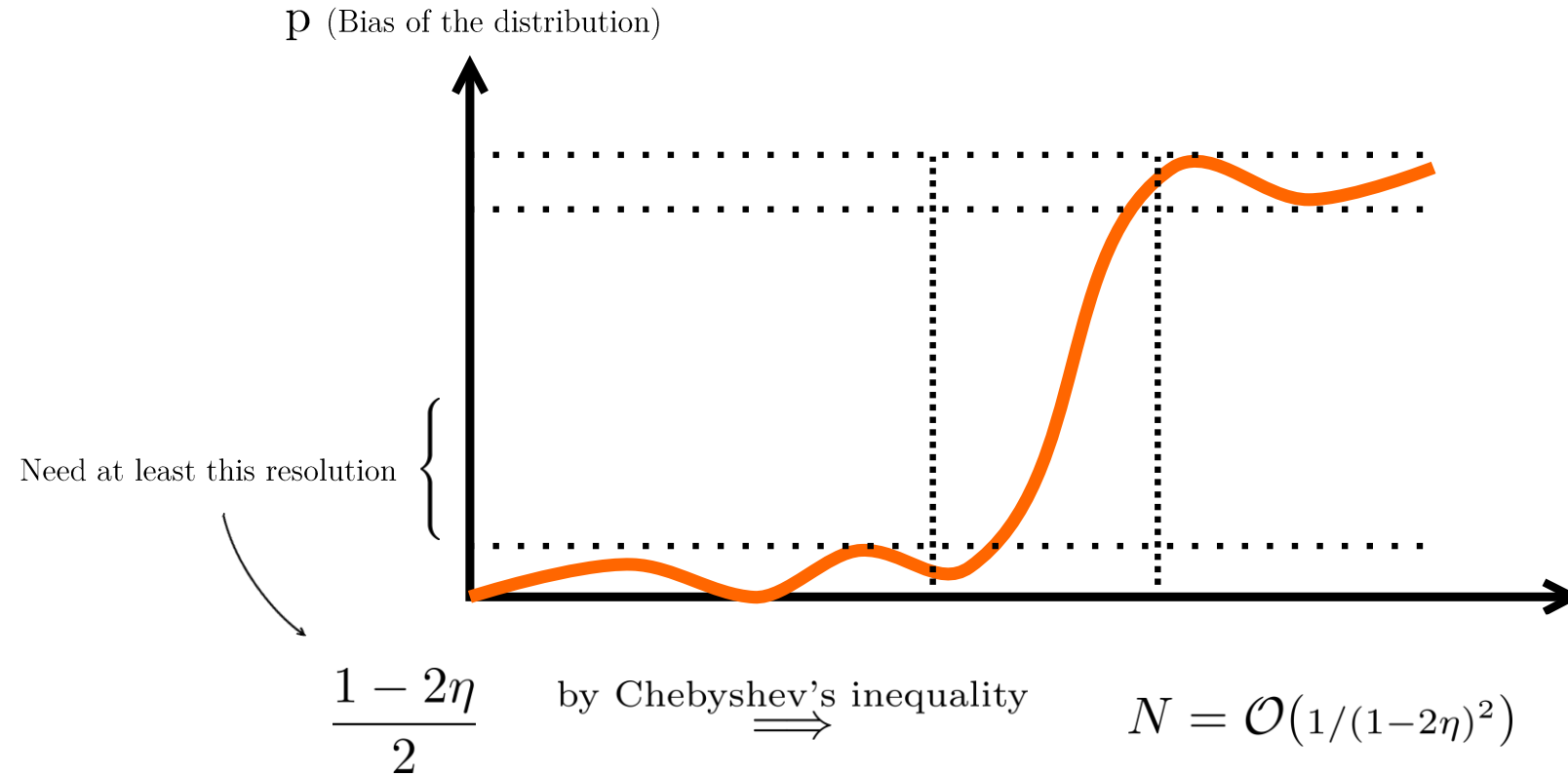
using  $\mathcal{O}\left(\frac{1}{\delta} \log\left(\frac{1}{\eta}\right)\right)$  queries of  $U_H$  and  $U_H^\dagger$ .



# $\alpha$ -Eigenvalue Estimation

→ Bringing it all together

What scalings do we get as a function of  $\varepsilon$ ?



# $\alpha$ -Eigenvalue Estimation

→ Bringing it all together

$$\eta = \frac{1}{2} - \frac{1}{4} \left(\frac{\epsilon}{4}\right)^\alpha$$

$$\delta = \epsilon/4$$

$$\Rightarrow \begin{cases} D(\alpha) = \mathcal{O}\left(\frac{1}{\epsilon} \log\left(\frac{1}{1-\epsilon^\alpha}\right)\right) = \mathcal{O}\left(\left(\frac{1}{\epsilon}\right)^{1-\alpha}\right) \\ N(\alpha) = \mathcal{O}(1/(1-2\eta)^2) = \mathcal{O}\left(D(\alpha) \left(\frac{1}{\epsilon}\right)^{2\alpha}\right) = \mathcal{O}\left(\left(\frac{1}{\epsilon}\right)^{1+\alpha}\right) \end{cases}$$



# $\alpha$ -Eigenvalue Estimation

→  $\alpha$ -Eigenvalue Estimation

$$\Rightarrow \begin{cases} D(\alpha) = \mathcal{O}\left(\left(\frac{1}{\epsilon}\right)^{1-\alpha}\right) \\ T(\alpha) = \mathcal{O}\left(\left(\frac{1}{\epsilon}\right)^{1+\alpha} \log^2\left(\frac{1}{\epsilon}\right)\right) = \tilde{\mathcal{O}}\left(\left(\frac{1}{\epsilon}\right)^{1+\alpha}\right) \end{cases}$$

...which is the  $\alpha$ -QPE scaling.

# In Conclusion

- $\alpha$ -Quantum Phase Estimation can be “upgraded” to  $\alpha$ -Eigenvalue Estimation
- You can think of the problem as “how well can I approximate a step function”
- Quantum Singular Value Transformations might be a good tool for finding hybrid algorithms
  - (Think: what do I need to do if I only have a poor approximation of my target function — and what *is* my target function)

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## A new light on a family of hybrid algorithms

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