

Lecture 7

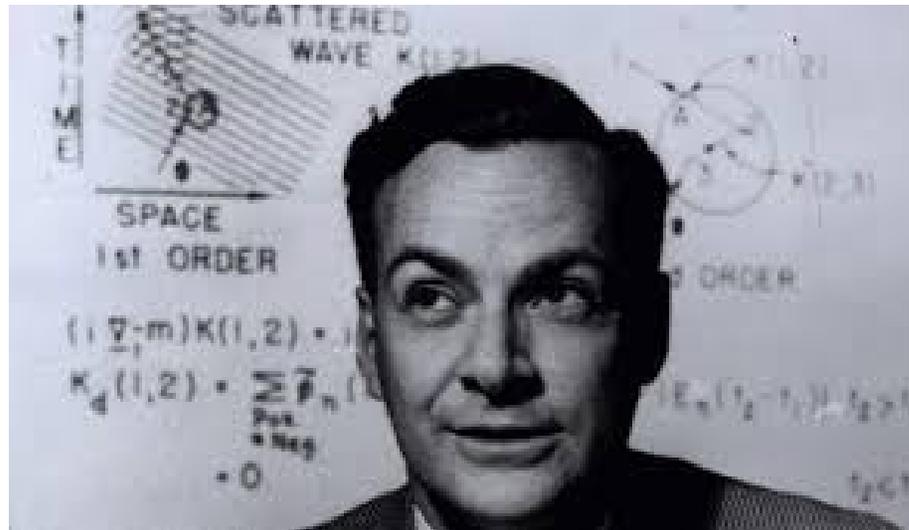
Quantum simulation– The HHL algorithm

February 27, 2020

Plan

1. Quantum simulation
2. The HHL algorithm

1. Quantum simulation; the dream of Feynman



Feynman (1982) “Can physics be simulated by a quantum computer? [. . .] the full description of quantum mechanics for a large system with R particles has too many variables, it **cannot be simulated with a normal computer** with a number of elements proportional to R [. . . but it can be simulated with] quantum computer elements. ”

The Hamiltonian

- ▶ Dynamical behavior of quantum systems governed by Schrödinger's equation

$$i\hbar \frac{d}{dt} |\psi\rangle = \mathbf{H} |\psi\rangle$$

- ▶ Key challenge in simulating quantum systems: **exponential** number of differential equations that have to be solved: a system of n qubits \Rightarrow solving 2^n differential equations . . .

Applications of quantum simulation

- ▶ Condensed-matter physics
 - solving problems that are classically intractable: Hubbard model (simplest model of interacting particles on a lattice), spin systems..;
 - understanding phase transitions, disordered systems, high-temperature superconductivity,...
- ▶ High-energy physics
- ▶ Cosmology
- ▶ Atomic and nuclear physics

Application to quantum chemistry

- calculating the thermal rate constant
- obtaining the energy spectrum of a molecular system
- simulate the static and dynamical chemical properties of molecules
- simulate chemical reactions

Exponentiating the Hamiltonian

$$\begin{aligned} i \frac{d}{dt} |\psi\rangle &= \mathbf{H} |\psi\rangle \\ &\Downarrow \\ |\psi(t)\rangle &= e^{-i\mathbf{H}t} |\psi(0)\rangle \end{aligned}$$

Hamiltonian simulation

Problem 1. [Hamiltonian simulation]

Input: Hamiltonian \mathbf{H} acting on n qubits, time $t \in \mathbb{R}^+$, accuracy $\varepsilon \in \mathbb{R}^+$

Output: a quantum circuit/algorithm implementing a unitary \mathbf{U} which is such that

$$\left\| e^{it\mathbf{H}} - \mathbf{U} \right\| \leq \varepsilon$$

Cost: the number of gates implementing \mathbf{U}

- ▶ \mathbf{H} is **can be efficiently simulated** if the quantum circuit consists of $\text{poly}(n, t, \frac{1}{\varepsilon})$ gates

Exercise: some simple principles

1. Show that if the unitary transform \mathbf{U} can be efficiently implemented and the Hamiltonian \mathbf{H} be efficiently simulated, then $\mathbf{U}\mathbf{H}\mathbf{U}^*$ can be efficiently simulated
2. if \mathbf{H} is diagonalizable in a basis corresponding to a unitary \mathbf{U} that can be efficiently implemented and if its eigenvalues can be efficiently computed, show that such \mathbf{H} can be efficiently simulated

Solution

1.

$$\begin{aligned}
 e^{it\mathbf{U}\mathbf{H}\mathbf{U}^*} &= \sum_{i=0}^{\infty} \frac{(it\mathbf{U}\mathbf{H}\mathbf{U}^*)^i}{i!} \\
 &= \sum_{i=0}^{\infty} \mathbf{U} \frac{(it\mathbf{H})^i}{i!} \mathbf{U}^* \\
 &= \mathbf{U} e^{i\mathbf{H}t} \mathbf{U}^*
 \end{aligned}$$

2. Assume first that \mathbf{H} is diagonalizable in the computational basis. Then \mathbf{H} can be efficiently simulated by performing the following steps

$$\begin{aligned}
 |a, 0\rangle &\mapsto |a, \lambda(a)\rangle \\
 &\mapsto e^{it\lambda(a)} |a, \lambda(a)\rangle \\
 &\mapsto e^{it\lambda(a)} |a, 0\rangle \\
 &= e^{it\mathbf{H}} |a\rangle |0\rangle
 \end{aligned}$$

The general case is handled by using the previous principle

The Hamiltonian encountered in practice

$$\mathbf{H} = \sum_{j=1}^m \mathbf{H}_j$$

where the \mathbf{H}_j only involve a few qubits

Method 1: the Lie-Suzuki-Trotter method

- ▶ If \mathbf{H}_1 and \mathbf{H}_2 can be efficiently simulated, then \mathbf{H}_1 and \mathbf{H}_2 can also be efficiently simulated

Theorem 1.

$$e^{\mathbf{A}+\mathbf{B}} = e^{\mathbf{A}}e^{\mathbf{B}} + O(\|\mathbf{A}\| \cdot \|\mathbf{B}\|)$$

Zassenhaus formula

$$e^{\mathbf{A}+\mathbf{B}} = e^{\mathbf{A}}e^{\mathbf{B}}e^{-\frac{1}{2}[\mathbf{A},\mathbf{B}]} \dots$$

The method

$$\begin{aligned}
 e^{i\mathbf{H}t} &= (e^{i\mathbf{H}t/r})^r \\
 &= \left(e^{i\mathbf{H}_1 t/r + i\mathbf{H}_2 t/r} \right)^r \\
 &= \left(e^{i\mathbf{H}_1 t/r} e^{i\mathbf{H}_2 t/r} + E \right)^r \\
 &= \left(e^{i\mathbf{H}_1 t/r} e^{i\mathbf{H}_2 t/r} \right)^r + O(r \|E\|) \\
 \|E\| &= O(\|i\mathbf{H}_1 t/r\| \cdot \|i\mathbf{H}_2 t/r\|) \\
 &= O\left(\|\mathbf{H}_1\| \cdot \|\mathbf{H}_2\| \frac{t^2}{r^2} \right) \\
 \text{choose } r &= O\left(\frac{t^2}{\varepsilon \|\mathbf{H}_1\| \cdot \|\mathbf{H}_2\|} \right) \\
 \Rightarrow e^{i\mathbf{H}t} &= \left(e^{i\mathbf{H}_1 t/r} e^{i\mathbf{H}_2 t/r} \right)^r + O(\varepsilon)
 \end{aligned}$$

► In general for a sum of m hamiltonians : $O(mt^2/\varepsilon)$ simulations of individual hamiltonians

Method 2: quantum walk approach

- ▶ Dependency in t of the previous approach $O(t^2)$
- ▶ Optimal dependency in t : $O(t)$
- ▶ Can be obtained by a quantum walk approach

Block encoding

- ▶ A recent and flexible approach with a logarithmic dependency in ε
- ▶ Assume that \mathbf{A} acts on n qubits with operator norm $\|\mathbf{A}\| \leq 1$ and we know how to implement an $(n + a)$ -qubit unitary operator

$$\mathbf{U} = \begin{pmatrix} \mathbf{A} & \cdot \\ \cdot & \cdot \end{pmatrix}$$

Definition 1. [**block-encoding**] \mathbf{U} is said to be an a -block encoding of \mathbf{A} . If

$$\mathbf{U} = \begin{pmatrix} \frac{\tilde{\mathbf{A}}}{s} & \cdot \\ \cdot & \cdot \end{pmatrix} \quad \text{with} \quad \|\mathbf{A} - \tilde{\mathbf{A}}\| \leq \varepsilon$$

then \mathbf{U} is an (s, a, ε) approximate block encoding of \mathbf{A} .

Sparse access

- We assume that \mathbf{A} is s -sparse and have sparse access to \mathbf{A}

$$\begin{aligned} O_A : |i, j\rangle |0^b\rangle &\mapsto |i, j\rangle |A_{ij}\rangle \\ O_r : |i, \ell\rangle &\mapsto |i, r(i, \ell)\rangle \\ O_c : |\ell, j\rangle &\mapsto |c(\ell, j), j\rangle \end{aligned}$$

where

- A_{ij} is a b -bit description of A_{ij} ,
- $r(i, \ell)$ denotes the location of the ℓ -th nonzero entry of the i -th row of \mathbf{A} ,
- $c(\ell, j)$ denotes the location of the ℓ -th nonzero entry of the j -th column of \mathbf{A}

Exercise

Let

$$W_1 : |0\rangle |0^n\rangle |j\rangle \mapsto \frac{1}{\sqrt{s}} |0\rangle \sum_{k:A_{kj} \neq 0} |k, j\rangle$$

$$W_2 : |0\rangle |k, j\rangle |0^b\rangle \mapsto \left(A_{kj} |0\rangle + \sqrt{1 - |A_{kj}^2|} |1\rangle \right) |k, j\rangle |0^b\rangle$$

$$W_3 : |0\rangle |0^n\rangle |i\rangle \mapsto \frac{1}{\sqrt{s}} |0\rangle \sum_{\ell:A_{i\ell} \neq 0} |i, \ell\rangle$$

We assume $s = 2^m$.

1. Show how to implement W_1 and W_3 using an O_c and O_r queries and a few other A -independent gates.
2. Show how to implement W_2 using an O_A -query, an O_A^{-1} -query, and a few other A -independent gates.
3. Show that the $(0^{n+1}i, 0^{n+1}j)$ -entry of $W_3^{-1}W_1$ is exactly $1/s$ if $A_{ij} \neq 0$, and is 0 if $A_{ij} = 0$
4. Show that the $(0^{n+1}i, 0^{n+1}j)$ -entry of $W_3^{-1}W_2W_1$ is exactly A_{ij}/s

Solution

1.

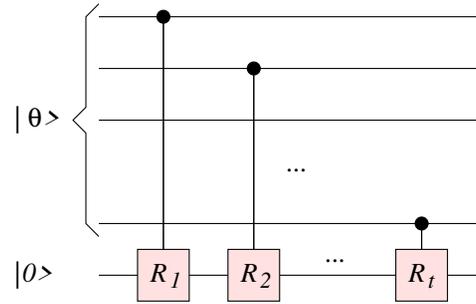
$$\begin{aligned}
|0\rangle |0^n\rangle |j\rangle &\xrightarrow{\text{Id} \otimes \mathbf{H}^{\otimes m} \otimes \text{Id}} \sum_{\ell} \frac{1}{\sqrt{s}} |0\rangle |\ell\rangle |j\rangle \\
&\xrightarrow{\text{Id} \otimes 0_c} \sum_{\ell} \frac{1}{\sqrt{s}} |0\rangle |c(\ell, j)\rangle |j\rangle \\
&= \frac{1}{\sqrt{s}} |0\rangle \sum_{k: A_{kj} \neq 0} |k, j\rangle
\end{aligned}$$

2.

$$R(\theta) \stackrel{\text{def}}{=} \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

$$\theta = \cos^{-1}(A_{kj})$$

$$R_j \stackrel{\text{def}}{=} R(2\pi/2^j)$$



$$\begin{aligned}
 |0\rangle |k, j\rangle |0^b\rangle &\xrightarrow{\text{Id} \otimes O_A} |0\rangle |k, j\rangle |A_{kj}\rangle \\
 &\xrightarrow{\text{Id} \otimes \cos^{-1}} |0\rangle |k, j\rangle |\theta_{kj}\rangle \\
 &\xrightarrow{c-R} (\cos \theta_{kj} |0\rangle + \sin \theta_{kj} |1\rangle) |k, j\rangle |\theta_{kj}\rangle \\
 &\xrightarrow{\text{Id} \otimes \cos} \left(A_{kj} |0\rangle + \sqrt{1 - A_{kj}^2} |1\rangle \right) |k, j\rangle |A_{kj}\rangle \\
 &\xrightarrow{\text{Id} \otimes O_A^{-1}} \left(A_{kj} |0\rangle + \sqrt{1 - A_{kj}^2} |1\rangle \right) |k, j\rangle |0^b\rangle
 \end{aligned}$$

3.

$$\langle 0^{n+1} | \langle i | W_3^* W_1 | 0^{n+1} \rangle |j\rangle = \frac{1}{s} \text{ if } A_{ij} \neq 0 \text{ and } 0 \text{ otherwise}$$

4.

$$\langle 0^{n+1} | \langle i | W_3^* W_2 W_1 | 0^{n+1} \rangle |j\rangle = \frac{A_{ij}}{s}$$

Low-degree approximation

- We want to implement

$$\mathbf{V} = \begin{pmatrix} \mathbf{A} & \cdot \\ \cdot & \cdot \end{pmatrix}$$

given a block encoding of \mathbf{A} ,

$$U : |0^a\rangle |\psi\rangle \mapsto |0^a\rangle \mathbf{A} |\psi\rangle + \sum_{j>0} |j\rangle |\psi_j\rangle$$

examples of interest:

- $f(x) = e^{ixt}$
- $f(x) = \frac{1}{x}$

- We have a **degree polynomial** P approximating f

Theorem 2. Let $P : [-1, 1] \rightarrow \{z \in \mathbb{C} : |z| \leq 1/4\}$ be a **degree- d** polynomial and let U an (s, a, ε) approximate block encoding of \mathbf{A} . We can implement an $(1, a + 2, 4d\sqrt{\varepsilon/s})$ approximate block encoding of $P(\mathbf{A}/s)$ with d applications of \mathbf{U} and \mathbf{U}^* , a single application of $c\mathbf{U}$ and $O(ad)$ other 1 and 2-qubit gates

Exercise

Show that a sufficiently large constant c can be chosen such that for all hermitian \mathbf{A} with operator norm $\|\mathbf{A}\| \leq 1$, we have

$$\left\| e^{it\mathbf{A}} - \sum_{k=0}^{c(t+\log(1/\varepsilon))-1} \frac{(it\mathbf{A})^k}{k!} \right\| \leq \varepsilon$$

Solution

$$\begin{aligned}
\left\| e^{it\mathbf{A}} - \sum_{k=0}^{c(t+\log(1/\varepsilon))-1} \frac{(it\mathbf{A})^k}{k!} \right\| &= \left\| \sum_{k=c(t+\log(1/\varepsilon))}^{\infty} \frac{(it\mathbf{A})^k}{k!} \right\| \\
&\leq \sum_{k=c(t+\log(1/\varepsilon))}^{\infty} \left(\frac{te}{k} \right)^k \\
&\leq \sum_{k=c(t+\log(1/\varepsilon))}^{\infty} \left(\frac{e}{c} \right)^k \\
&\leq \frac{\left(\frac{e}{c} \right)^{c(t+\log(1/\varepsilon))}}{1 - \frac{e}{c}}
\end{aligned}$$

Let $c = e^2$

$$\left\| e^{it\mathbf{A}} - \sum_{k=0}^{c(t+\log(1/\varepsilon))-1} \frac{(it\mathbf{A})^k}{k!} \right\| \leq \frac{\left(\varepsilon \frac{1}{e^{2t}}\right)}{1 - \frac{1}{e}}$$
$$\leq \varepsilon$$

Hamiltonian simulation via transforming block-encoded matrices

- ▶ Approximate $f(x) = e^{ixt}$ with a degree $d = O(t + \log(1/\varepsilon))$ polynomial P
- ▶ If \mathbf{H} is s -sparse \Rightarrow block encoding \mathbf{U} of \mathbf{H}/s using $O(1)$ queries to \mathbf{H} and $O(n)$ other gates evolving \mathbf{H} for time $t =$ evolving \mathbf{H}/s for time st
- ▶ Previous theorem \Rightarrow block-encoding \mathbf{V} of $P(x) \approx \frac{e^{it\mathbf{H}}}{4}$ by invoking \mathbf{U} an \mathbf{U}^{-1} $O(st + \log(1/\varepsilon))$ times and mapping

$$\mathbf{V} : |0\rangle |\psi\rangle \mapsto |0\rangle P(\mathbf{H}) |\psi\rangle + |\phi\rangle$$

where $|\phi\rangle$ has no support on basis states starting with $|0\rangle$

- ▶ Complexity of ε -precise Hamiltonian simulation of s -sparse \mathbf{H} is $O(st + \log(1/\varepsilon))$ queries to \mathbf{H} and $O(n(st + \log(1/\varepsilon)))$ 2-qubit gates

The HHL algorithm

$$N \stackrel{\text{def}}{=} 2^n$$

$$\mathbf{A} \in \mathbb{C}^{N \times N}$$

$$\mathbf{b} \in \mathbb{C}^N$$

Find \mathbf{x} s.t. $\mathbf{Ax} = \mathbf{b}$

Problem 2. [quantum linear system problem (QLSP)] Find an n -qubit state $|\hat{\mathbf{x}}\rangle$ such that

$$(i) \quad \|\mathbf{x}\rangle - |\hat{\mathbf{x}}\rangle\| \leq \varepsilon$$

$$(ii) \quad \mathbf{Ax} = \mathbf{b}$$

Assumptions

1. \mathbf{A} is non-singular
2. $|b\rangle$ can be prepared using a circuit of B 2-gates
3. \mathbf{A} is s -sparse
4. $\lambda_i \in (0, 1]$ for all i where $\lambda_1, \dots, \lambda_N$ are the singular values of \mathbf{A}

Complexity

$$\kappa \stackrel{\text{def}}{=} \frac{\max_j \mu_j}{\min_j \mu_j}$$

Problem	Algorithm	Complexity
LSP	Conjugate Gradient	$O(Ns\kappa \log(1/\epsilon))$
QLSP	HHL 2009	$O\left(\frac{s^2\kappa^2 \log N}{\epsilon}\right)$
QLSP	VTAA-HHL (Ambainis 2010)	$O\left(\frac{s^2\kappa \log N}{\epsilon}\right)$
QLSP	Childs et al 2017	$O(s\kappa \text{polylog}(s\kappa/\epsilon))$
QLSP	QLSA 2018	$O\left(\frac{\kappa^2 \text{polylog}(n) \sqrt{\text{Tr}(AA^*)}}{\epsilon}\right)$

Application: recommendation system

Problem 3. [Recommendation system] *An unknown (hidden) $m \times n$ binary matrix \mathbf{P} modelling customers preferences and \mathbf{P} is of low rank k . For a customer i one should output columns j such that it is likely that $P_{ij} = 1$.*

- ▶ Quantum algorithm based on HHL that is of complexity $O(\text{poly}(k)\text{polylog}(mn))$ (we do not use all the entries of \mathbf{P} !)

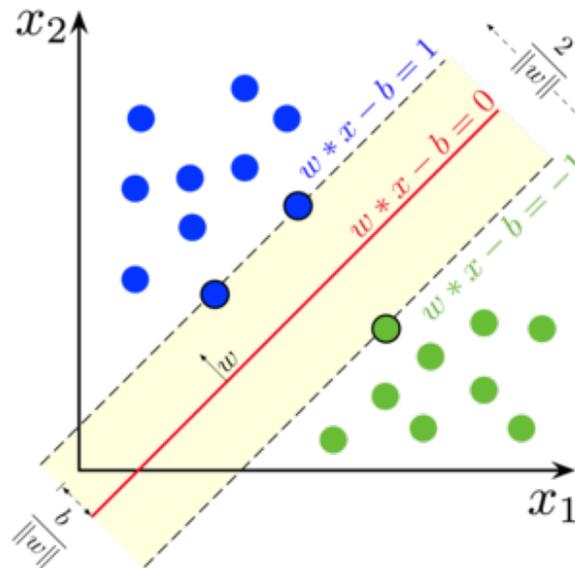
Application: support vector machine

Problem 4. [Support Vector Machine]

Input: M training data points of the form $\{(\mathbf{x}_i, y_i), \mathbf{x}_i \in \mathbb{R}^N, y_i = \pm 1\}_{i=1 \dots M}$

Output: $\mathbf{w} \in \mathbb{R}^N$ and b that minimizes $\|\mathbf{w}\|$ under the constraint $y_i(\mathbf{w} \cdot \mathbf{x}_i + b) \geq 1$

- ▶ best classical algorithm takes $\text{poly}(M, N)$ whereas quantum complexity $O(\log(MN))$



Exercise: solving a Boolean system

1. Solve the following system over \mathbb{F}_2 :

$$\begin{cases} x_1x_2 + x_1x_3 & = & 0 \\ x_1x_3 + x_2x_3 + x_2x_4 & = & 1 \\ x_1x_2 + x_2x_3 & = & 0 \\ x_1x_2 + x_2x_4 & = & 1 \end{cases}$$

2. Outline a strategy for solving a polynomial system involving the multiplication of the polynomial equations by all monomials of degree $\leq D$
3. Can you associate to a polynomial system over \mathbb{F}_2 a polynomial system over \mathbb{C} that has as only solutions the solutions of the previous system ?
4. What happens if you apply HHL to this system over \mathbb{C} ?

Exercise : \mathbf{A} hermitian ?

1. Give a $2N \times 2N$ hermitian matrix \mathbf{A}' and $\mathbf{b}' \in \mathbb{C}^{2N}$ such that a solution \mathbf{x} of $\mathbf{A}\mathbf{x} = \mathbf{b}$ can be read off from a solution \mathbf{x}' to $\mathbf{A}'\mathbf{x} = \mathbf{b}'$
2. Relation between the condition number of \mathbf{A} and \mathbf{A}' ?

Solution

1.

$$\mathbf{A}' = \begin{pmatrix} 0 & \mathbf{A}^\top \\ \mathbf{A} & 0 \end{pmatrix}$$

$$\mathbf{b}' = \begin{pmatrix} 0 \\ \mathbf{b} \end{pmatrix}$$

$$\mathbf{x}' = \begin{pmatrix} 0 \\ \mathbf{x} \end{pmatrix}$$

2.

$$\mathbf{A}'^2 = \begin{pmatrix} \mathbf{A}^\top \mathbf{A} & 0 \\ 0 & \mathbf{A} \mathbf{A}^\top \end{pmatrix}$$

$$\Downarrow$$

$$\kappa(\mathbf{A}) = \kappa(\mathbf{A}')$$

Approach

- ▶ We assume that

$$\begin{aligned}\mathbf{x} &= \mathbf{A}^{-1}\mathbf{b} \\ \mathbf{A} &= \sum_j \lambda_j \mathbf{u}_j \mathbf{u}_j^\top \\ \mathbf{A}^{-1}\mathbf{u}_j &\mapsto \frac{1}{\lambda_j} \mathbf{u}_j \\ \mathbf{b} &= \sum_j \beta_j \mathbf{u}_j \\ \mathbf{A}^{-1}\mathbf{b} &= \sum_j \frac{\beta_j}{\lambda_j} \mathbf{u}_j\end{aligned}$$

- ▶ Problem: \mathbf{A}^{-1} is **not** unitary in general
- ▶ \mathbf{A} is **hermitian** $\Rightarrow e^{i\mathbf{A}}$ is **unitary**

Phase estimation

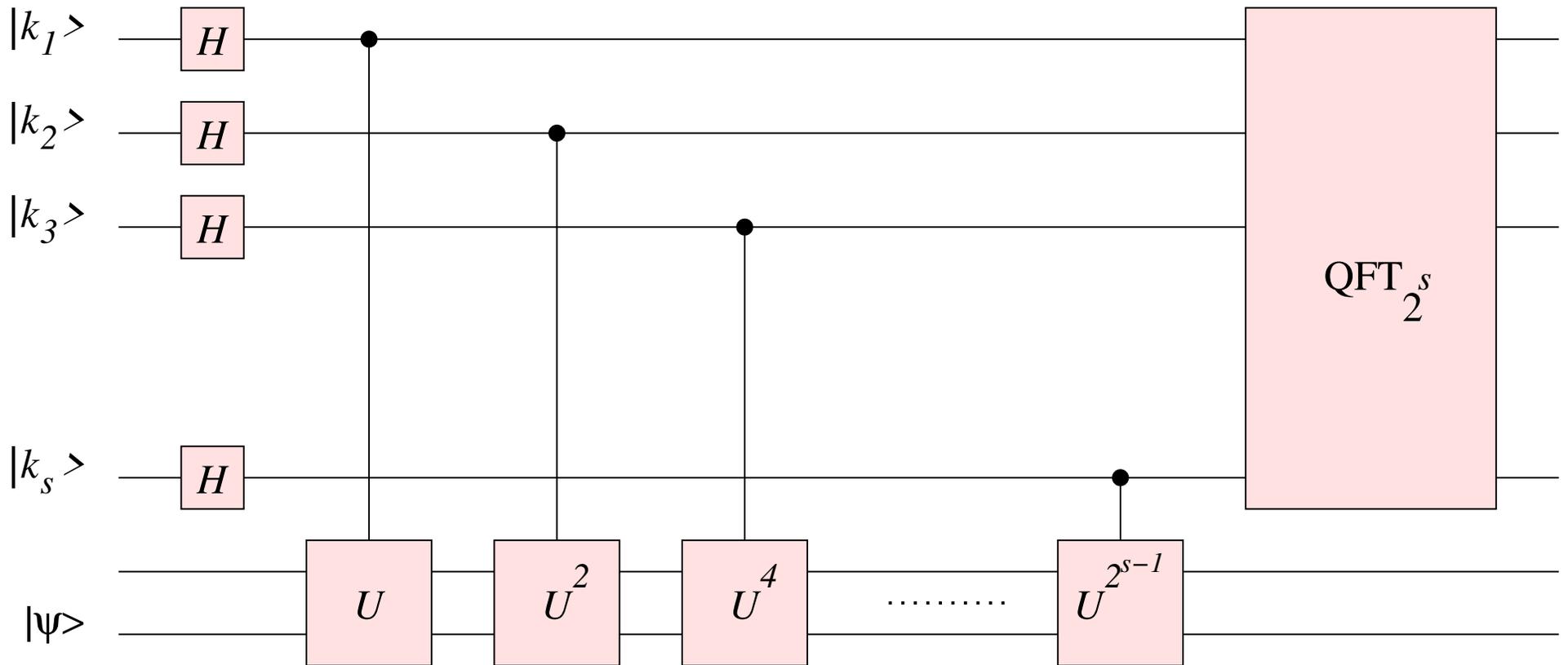
Theorem 3. For every unitary operator U acting on m qubits, there exists a quantum circuit $\mathbf{PE}(U)$ acting on $m + s$ qubits satisfying the following properties

1. the circuit $\mathbf{PE}(U)$ uses $2s$ Hadamard gates, $O(s^2)$ controlled phase rotations and makes 2^{s+1} calls to $c-U$
2. maps with probability $1 - 1/\text{poly}(n)$

$$\sum_j \alpha_j |\psi_j\rangle |0\rangle \mapsto \sum_j \alpha_j |\psi_j\rangle |\tilde{\theta}_j\rangle$$

where $|\psi_j\rangle$ are the eigenvectors of U , $e^{i\theta_j}$ is the associated eigenvalue and $|\theta_j - \tilde{\theta}_j| \leq 2^{-m}$

The circuit of phase estimation



Approach (II)

- With Hamiltonian simulation we can implement e^{iA}

$$\begin{aligned}
 |0\rangle |0\rangle |v_i\rangle &\xrightarrow{\text{PE}} |0\rangle |\lambda_j\rangle |v_j\rangle \\
 &\mapsto \left(\frac{1}{\kappa\lambda_j} |0\rangle + \sqrt{1 - \frac{1}{(\kappa\lambda_j)^2}} |1\rangle \right) |\lambda_j\rangle |v_j\rangle \\
 &\xrightarrow{\text{PE}^{-1}} \left(\frac{1}{\kappa\lambda_j} |0\rangle + \sqrt{1 - \frac{1}{(\kappa\lambda_j)^2}} |1\rangle \right) |0\rangle |v_j\rangle \\
 V : |v_i\rangle |0\rangle |0\rangle &\mapsto |v_j\rangle |0\rangle \left(\frac{1}{\kappa\lambda_j} |0\rangle + \sqrt{1 - \frac{1}{(\kappa\lambda_j)^2}} |1\rangle \right)
 \end{aligned}$$

Approach (III)

$$\begin{aligned}
 |0\rangle |b\rangle &= |0\rangle \sum_j \beta_j |v_j\rangle \\
 \xrightarrow{\mathbf{v}} &\left(\frac{1}{\kappa \lambda_j} |0\rangle + \sqrt{1 - \frac{1}{(\kappa \lambda_j)^2}} |1\rangle \right) \sum_j \beta_j |v_j\rangle \\
 &= K |0\rangle |x\rangle + |1\rangle |\phi\rangle
 \end{aligned}$$

Exercise

1. Show that the probability of measuring $|0\rangle$ in the first register is $\geq \frac{1}{\kappa^2}$
2. How can this probability be improved ?

Solution

1. Let $p = \mathbf{Prob}(\text{meas. } 0)$, then

$$p = \sum_i \frac{|\beta_i|^2}{\lambda_i^2 \kappa^2} \geq \sum_i \frac{|\beta_i|^2}{\kappa^2} \geq \frac{1}{\kappa^2}$$

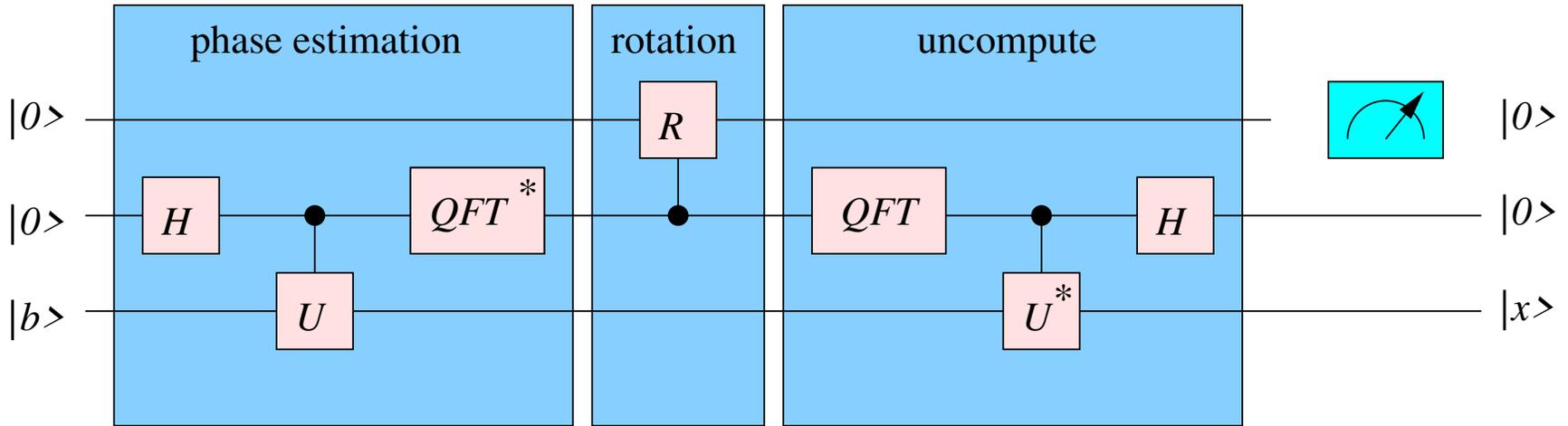
2. amplitude amplification $\Rightarrow O(\kappa)$ calls of the algorithm for having a probability of success $\Omega(1)$

Exercise

Give a quantum circuit performing

$$|0\rangle |\lambda_j\rangle \mapsto \left(\frac{1}{\kappa \lambda_j} |0\rangle + \sqrt{1 - \frac{1}{(\kappa \lambda_j)^2}} |1\rangle \right) |\lambda_j\rangle$$

Circuit



Complexity

- ▶ Leads to an algorithm that produces a state $|\tilde{x}\rangle$ that is ε -close to $|x\rangle$ using $\kappa^2 s/\varepsilon$ queries to \mathbf{A} and roughly $\kappa s(\kappa n/\varepsilon + B)$ other 2-qubit gates

Improving the efficiency of HHL

- ▶ Use the block encoding method of quantum simulation to perform $f(\mathbf{A})$ with

$$f(x) \stackrel{\text{def}}{=} \frac{1 - (1 - x^2)^b}{x}$$
$$b = \kappa^2 \ln(\kappa/\varepsilon)$$

- ▶ Complexity: $O(\kappa^2 s \log(\kappa/\varepsilon))$ queries to \mathbf{A} and $O(\kappa s (\kappa n \log(\kappa/\varepsilon) + B))$ 2-qubit gates

Exercise

1. Let $I = [-1, -1/\kappa] \cup [1/\kappa, 1]$. Give an upperbound on $|f(x) - \frac{1}{x}|$ on I .
2. Show that the polynomial $p(x) = f(x)/(4(\kappa + \varepsilon))$ meets the conditions of Theorem 1

Solution