

Optimal Hamiltonian Simulation by Quantum Signal Processing

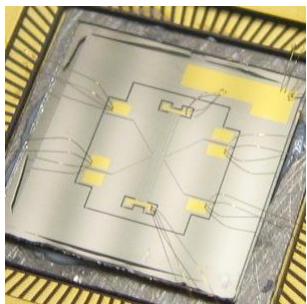
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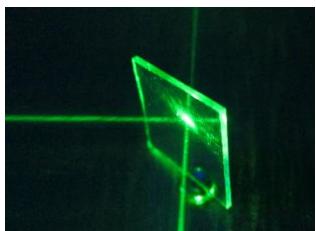
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Physics \leftrightarrow Computation



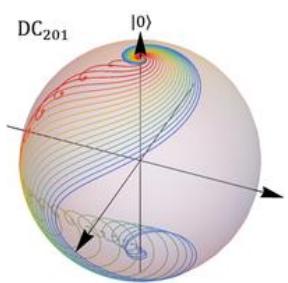
Physical concepts inspire problem-solving techniques

- Adiabatic evolution, Quantum walks
- Universal quantum computation
- Compiling to specific architecture has large overheads



Physical systems solve problems

- Boson sampler, Analog quantum simulator
- Very elegant, simple, fast; Near future prospects
- Generally not fault-tolerant



Quantum Signal Processing: Abstraction of physical dynamics into a computing module ‘**Quantum Signal Processor**’ that

- Preserve intuition of how original physical system computes
- Preserve simplicity, optimality of original physical system
- In a FTQC compatible manner

Outline

1. Hamiltonian Simulation

- A Brief History of Simulation

2. Overview of Algorithm

- Quantum Signal Processing
- Optimal Hamiltonian Simulation

3. Conclusion

1. Hamiltonian Simulation

$$i \frac{d|\psi\rangle}{dt} = \hat{H}|\psi\rangle \rightarrow e^{-i\hat{H}t}$$

Simulate nature:

Chemistry, materials, etc.



Run quantum algorithms:

Quantum walks, span programs etc.

“... nature isn’t classical, dammit, and if you want to make a simulation of nature, you’d better make it quantum mechanical, and by golly it’s a wonderful problem, because it doesn’t look so easy.”

Richard Feynman (1982)

Difficult for classical computers in general

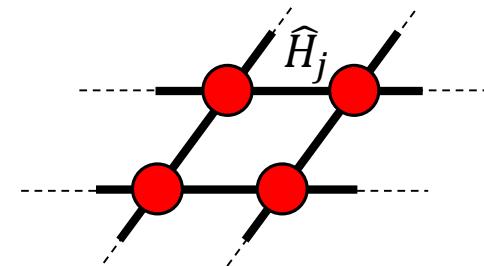
Curse of dimensionality: n qubits $\Rightarrow N = 2^n$ states

PromiseBQP-complete

Assumptions

***k*-local interactions [Lloyd 1996]**

- $\hat{H} = \sum_{j=1}^m \hat{H}_j$, \hat{H}_j on $k = O(1)$ qubits
 - Naturally describes physical systems
 - $e^{-i\hat{H}_j t}$ are easy; $e^{-i(\hat{A}+\hat{B})} = (e^{-i\hat{A}/r} e^{-i\hat{B}/r})^r + O(1/r)$



d -sparse matrices [Aharanov & Ta-Shma 2002]

- d non-zero entries per row, e.g. $d = 2^k m$
 - Oracles for positions & values of non-zero elements
 - Exponential speedups claimed when $d = \text{polylog}(N)$
 - Popular in quantum algorithm design by quantum walk

$$\left(\begin{array}{cccccc} 0 & -2 & 0 & 0 & 0 & \cdots \\ -2 & 0.333 & 0 & 1 & -i & \cdots \\ 0 & 0 & -1 & 1 & 1 & \cdots \\ 0 & 1 & 1 & 2 & 0 & \cdots \\ 0 & i & 1 & 0 & 0 & \cdots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots \end{array} \right) \quad N$$

Unitary access model [Childs & Wiebe 2012]

- $\hat{H} = \sum_{j=1}^m \alpha_j \hat{U}_j$, \hat{U}_j are Unitary, Oracles for α_j & \hat{U}_j
 - New promising approach to quantum physics simulations

A Brief History of Simulation

Parameters:

t time

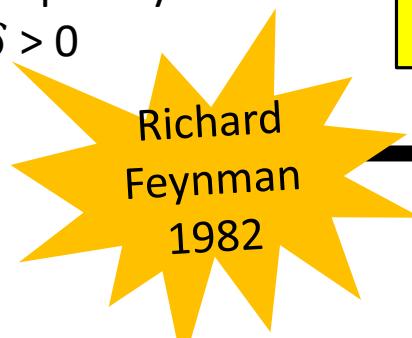
ε error

N dimension

m terms

d sparsity

$\delta > 0$



$$[\text{BACS07}] O\left(m \frac{(mt\|H\|)^{1+\delta}}{\varepsilon^\delta}\right)$$

$$[\text{PZ10}] O\left(m \frac{(mt\|H_1\|)(mt\|H_2\|)^\delta}{\varepsilon^\delta}\right)$$

k -local era
 $[\text{L96}] O\left(m \frac{(mt\|H\|)^2}{\varepsilon}\right)$

Unitary access era [CW12]
 $O\left(m^2\|H_j\|t \exp\left(1.6\sqrt{\log\left(\frac{m\|H_j\|t}{\varepsilon}\right)}\right)\right)$
Better than product formulas

$$[\text{CK11}] O(d^3 \log^* N \|H\| t)^{1+\delta} / \varepsilon^\delta$$

d -sparse Graph coloring era
 $[\text{AT02}] \text{poly}(d, \log N) (\|H\| t)^{3/2} / \sqrt{\varepsilon}$

$$[\text{C04}] O(d^4 \log^4 N \|H\| t)^{1+\delta} / \varepsilon^\delta$$

d -sparse Quantum walk era
 $[\text{C10}][\text{BC12}] O(d \|H\|_{\max} t / \varepsilon)$
Optimal time & sparsity

$$[\text{BACS07}] O(d^4 \log^* N \|H\| t)^{1+\delta} / \varepsilon^\delta$$

Modern History

Parameters:

t time

ε error

N dimension

m terms

d sparsity

[BCCKS14B] $O\left(\tau \frac{\log(\tau/\varepsilon)}{\log \log(\tau/\varepsilon)}\right)$

Qubitization² Era?

$$\hat{H} \propto \langle G | \hat{U} | G \rangle$$

[LC16B]²

$$O\left(\tau + \frac{\log(1/\varepsilon)}{\log \log(1/\varepsilon)}\right)$$

[NB16]

$$O\left(\tau \frac{\log \tau}{\log \log \tau} + \log\left(\frac{1}{\varepsilon}\right)\right)$$

[BCK15] $O\left(\tau \frac{\log(\tau/\varepsilon)}{\log \log(\tau/\varepsilon)}\right)$

Lower bound $\Omega\left(\tau + \frac{\log(1/\varepsilon)}{\log \log(1/\varepsilon)}\right)$

[LC16A]¹

$$O\left(\tau + \frac{\log(1/\varepsilon)}{\log \log(1/\varepsilon)}\right)$$

Optimal

[BN16]

$$O\left(\tau \frac{\log \log \tau}{\log \log \log \tau} + \log\left(\frac{1}{\varepsilon}\right)\right)$$

Our results!

[BCCKS14A]

$$O\left(\tau d \frac{\log(\tau/\varepsilon)}{\log \log(\tau/\varepsilon)}\right)$$

Optimal error

Legend

Unitary access $\tau = t \sum_{j=1}^m \alpha_j \geq \|H\|t$

d -sparse Quantum walk $\tau = d\|H\|_{\max}t$

d -sparse Graph coloring $\tau = d\|H\|_{\max}t$

¹Low, Chuang, arXiv:1606.02685 | Phys. Rev. Lett. 118, 010501

²Low, Chuang, arXiv:1610.06546

2. Overview of Algorithm

Input: d -sparse oracles

$$\text{COLUMN}|i\rangle|l\rangle = |i\rangle|f(l,j)\rangle$$

row
|

$$\text{ELEMENT}|i\rangle|j\rangle|z\rangle = |i\rangle|j\rangle|z \oplus \hat{H}_{ij}\rangle$$

row
|
column
|

Output: $\approx e^{-i\hat{H}t}$

Discrete-time quantum walk for any Hamiltonian [Childs 2010]

Implement $e^{i \arcsin(\hat{H}/d\|H\|_{\max})}$ with 2 COLUMN & 4 ELEMENT queries

Linearize with ‘quantum signal processing’

Linearize spectrum with eigenphase transformation $e^{i\lambda} \mapsto e^{-id\|H\|_{\max}\sin(\lambda)}$

$O\left(td\|H\|_{\max} + \frac{\log(1/\varepsilon)}{\log \log(1/\varepsilon)}\right)$ single qubit gates & queries to controlled-walk

1 ancilla qubit

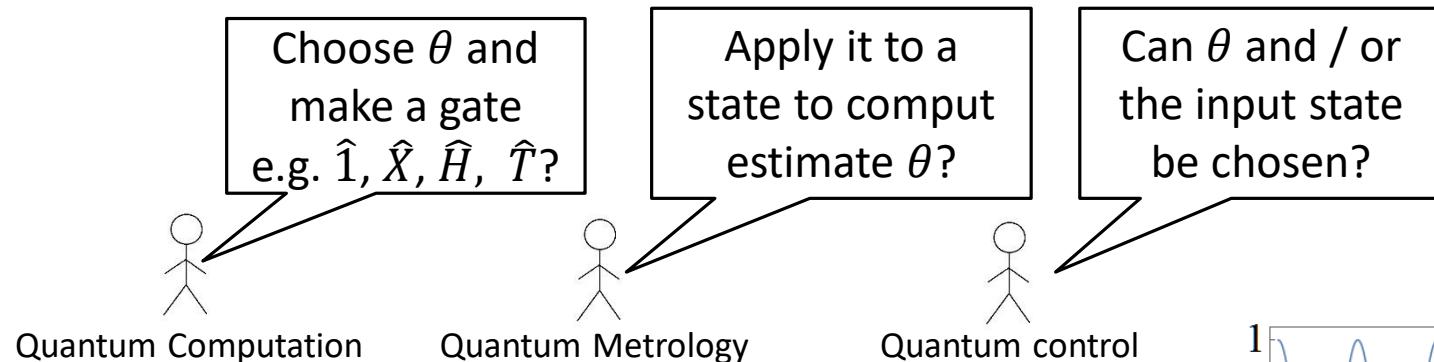
Extremely small overhead:
Candidate for small quantum computer!

Quantum Physics 8.01

Single-qubit unitary

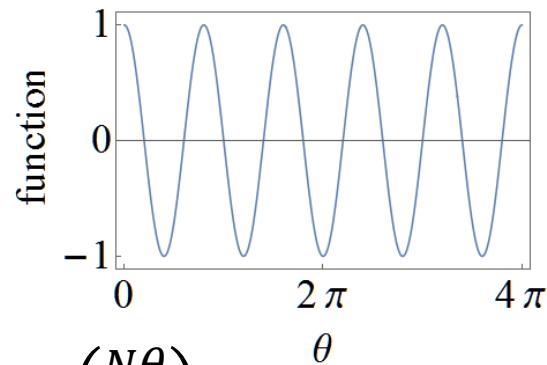
$$\hat{R}[\theta] \equiv e^{-i\frac{\theta}{2}\hat{X}} \equiv \cos\left(\frac{\theta}{2}\right)\hat{1} - i \sin\left(\frac{\theta}{2}\right)\hat{X} \equiv \boxed{\hat{R}}$$

Q: How does this compute?



Combination of Ideas: $\hat{R}[\theta]$ is a computational module that computes some function depending on input parameter θ , input state and measurement basis.

Composite gate: $\langle 0 | \hat{R}[\theta] \hat{R}[\theta] \cdots \hat{R}[\theta] | 0 \rangle = \cos\left(\frac{N\theta}{2}\right)$



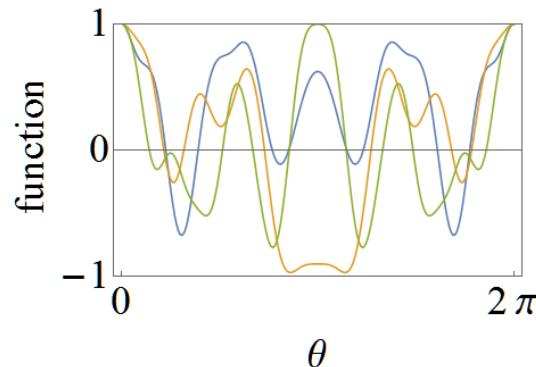
The Single-Qubit Processor

Q: Can this compute anything interesting?

Idea: Less trivial primitive unitary by with discrete quantum control

$$\hat{R}_\phi[\theta] \equiv e^{-i\frac{\phi}{2}\hat{Z}} \hat{R}[\theta] e^{i\frac{\phi}{2}\hat{Z}} \equiv - \boxed{\hat{Z}_\phi} \boxed{\hat{R}} \boxed{\hat{Z}_{-\phi}} -$$

Composite gate $\langle 0 | \hat{R}_{\phi_N}[\theta] \cdots \hat{R}_{\phi_2}[\theta] \hat{R}_{\phi_1}[\theta] | 0 \rangle$ for some choice $\vec{\phi}$



Very non-trivial
functions for random
choice of ϕ

Idea: $\vec{\phi}$ assembly code for single-qubit processor

Compute unitary function $A[\theta] \hat{1} + iB[\theta]\hat{Z} + iC[\theta]\hat{X} + iD[\theta]\hat{Y}$

Select desired components with choice of input state and measurement

Qubit Unitary Function Synthesis

Q: Why is this hard? Analogy:

- Single-qubit gate synthesis from {H, T}
- Single-qubit function synthesis from $\{\hat{R}_\phi[\theta], \vec{\phi} \in \mathbb{R}\}$

Exponential time for finding best-fit $\vec{\phi}$ via gradient descent

- $\hat{R}_{\phi_N}[\theta] \cdots \hat{R}_{\phi_2}[\theta] \hat{R}_{\phi_1}[\theta] = A[\theta]\hat{1} + iB[\theta]\hat{Z} + iC[\theta]\hat{X} + iD[\theta]\hat{Y}$

Necessary & sufficient constraints on achievable $A[\theta], C[\theta]$ ¹

1. $A[0] = 1$
 2. $A[\theta]^2 + C[\theta]^2 \leq 1$
 3. $A[\theta] = \sum_{k=0}^{N/2} a_k \cos(k\theta)$
 4. $C[\theta] = \sum_{k=1}^{N/2} c_k \sin(k\theta)$
- Fourier series

- Given $A[\theta], C[\theta]$, can compute $\vec{\phi}$ in classical poly(N) time

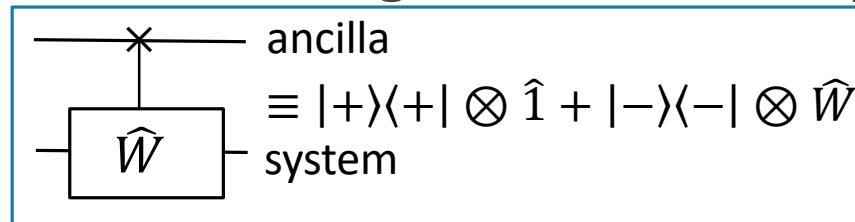
¹Low, Yoder, Chuang, Phys. Rev. X 6, 041067 (2016)

1-qubit Quantum Signal Processing

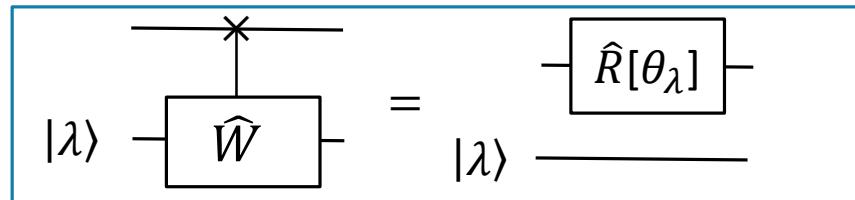
Q: How relevant is this to multi-qubit problems?

Idea: Single-qubit rotation angle θ controlled by system

- Controlled- \hat{W}
 $\hat{W}|\lambda\rangle = e^{i\theta_\lambda}|\lambda\rangle$



- On input $|\lambda\rangle$



- Replace all \hat{R} in $\hat{R}_{\phi_N}[\theta] \cdots \hat{R}_{\phi_2}[\theta] \hat{R}_{\phi_1}[\theta]$ with controlled- \hat{W}

Idea: Project ancilla to apply coefficient on eigenstate

$$\langle + | \hat{R}_{\phi_N}[\theta] \cdots \hat{R}_{\phi_2}[\theta] \hat{R}_{\phi_1}[\theta] | + \rangle = A[\theta] + iC[\theta]$$



$$\langle + | \hat{U} | + \rangle | \lambda \rangle = A[\theta_\lambda] + iC[\theta_\lambda] | \lambda \rangle$$

Quantum Signal Processing (cont.)

Q: What can this do?

Approximate $e^{i\theta} \rightarrow e^{i h(\theta)}$ with degree $N/2$ Fourier series

- **Input:** Objective function $h: [0, 2\pi) \rightarrow [0, 2\pi)$
- $\|A[\theta] + iC[\theta] - e^{i h(\theta)}\|_{\infty} \leq \epsilon \Rightarrow$ can find achievable $A_1[\theta] + iC_1[\theta]$ s.t.
 $\|A_1[\theta] + iC_1[\theta] - e^{i h(\theta)}\|_{\infty} \leq 8\epsilon$

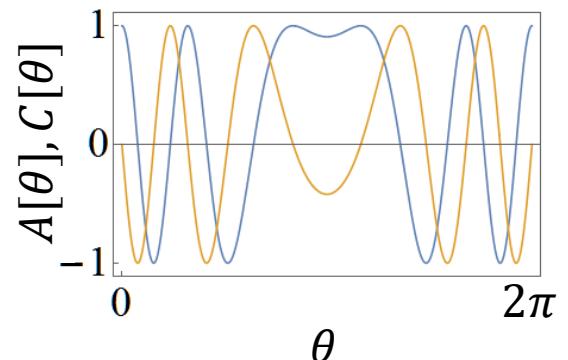
Eigenphase transformation

- **Output:** $\langle + | \hat{U} | + \rangle |\lambda\rangle = A_1[\theta_\lambda] + iC_1[\theta_\lambda] |\lambda\rangle \approx e^{i h(\theta_\lambda)} |\lambda\rangle$
 - Query Complexity N vs. ϵ depends on smoothness / analyticity of $e^{i h(\theta)}$
 - Trace distance $\leq 8\epsilon$
 - Success probability $\geq 1 - 16\epsilon$
- k -smooth: $\epsilon^{-1/k}$
Analytic: $\log(1/\epsilon)$
Entire: super-logarithmic

Optimal Hamiltonian Simulation

Eigenphase transformation with QSP

- Objective function $e^{i h(\theta)} = e^{-i \tau \sin(\theta)}$
- Linearize quantum walk spectrum to $e^{-i \lambda t}$



Degree N Fourier approximation error

- Truncate Jacobi-Anger expansion $e^{-i \tau \sin(\theta)} = \sum_{k=-\infty}^{\infty} J_k(\tau) e^{-ik\theta}$
- Super-exponential scaling $\varepsilon \leq \sum_{k>N}^{\infty} 2|J_k(\tau)| \leq O\left((e\tau/2N)^N\right)$

Query complexity by solving for $N = O\left(\tau + \frac{\log(1/\varepsilon)}{\log \log(1/\varepsilon)}\right)$

d -sparse PARITY lower bound [Berry Childs Cleve Kothari Somma 2015]

- $\Omega(N)$ for any N s.t. $\varepsilon = O\left((\tau/N)^N\right)$

~~Big Crunch~~ Big Rip

Many open problems in Quantum simulation

Gate optimal simulation

Time-dependent simulation

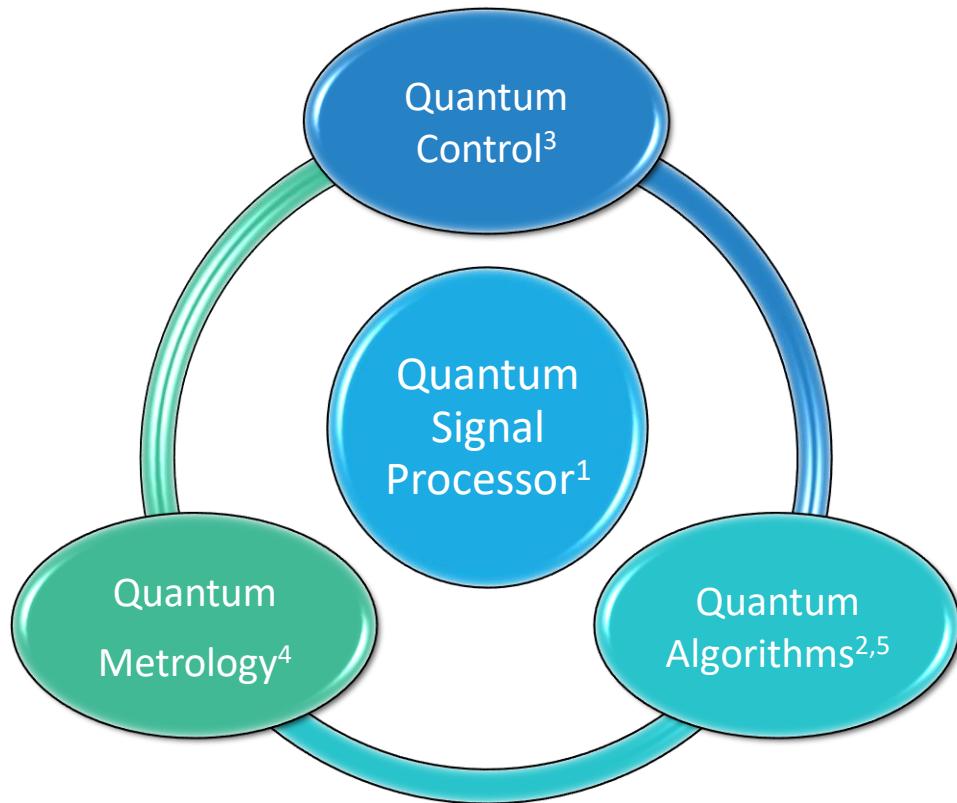
Open quantum system simulation

Hamiltonians with structured information

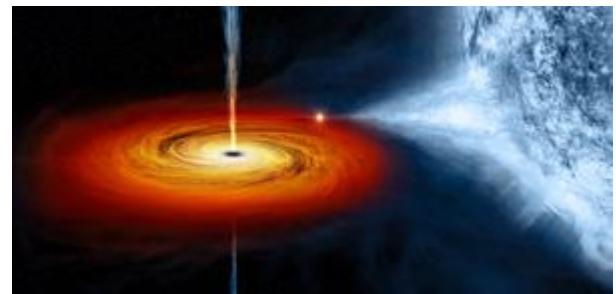
Analog simulation [Stephen Piddock 2:00-2:40 !]

Other query models

3. Conclusion



Thank you!



Work related to the single-qubit quantum signal processor

- 1) Hamiltonian simulation by qubitization, Low, Chuang, arXiv preprint arXiv:1610.06546
- 2) Optimal Hamiltonian simulation by quantum signal processing, Low, Chuang, Physical Review Letters 118 (1), 010501
- 3) Methodology of resonant equiangular composite quantum gates, Low, Yoder, Chuang, Physical Review X 6 (4), 041067
- 4) Quantum imaging by coherent enhancement, Low, Yoder Chuang, Physical Review Letters 114 (10), 100801
- 5) Fixed-point quantum search with an optimal number of queries , Yoder, Low, Chuang, Physical Review Letters 113 (21), 210501