

A Quantum Speed-Up for Approximating the Top Eigenvector of a Matrix via Improved Tomography

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Goal: finding the top eigenvector of a Hermitian M

- ▶ Used for: computing PageRank, ground states, dimensionality reduction, etc.
- ▶ Assume we can query matrix elements of the (dense) matrix $M \in \mathbb{C}^{d \times d}$.
- ▶ Full diagonalization takes $\sim d^\omega$ time where $\omega \in [2, 2.37 \dots]$; in practice $\omega \approx 3$.
- ▶ A matrix-vector multiplication takes merely d^2 time \Rightarrow use power method!

Power method

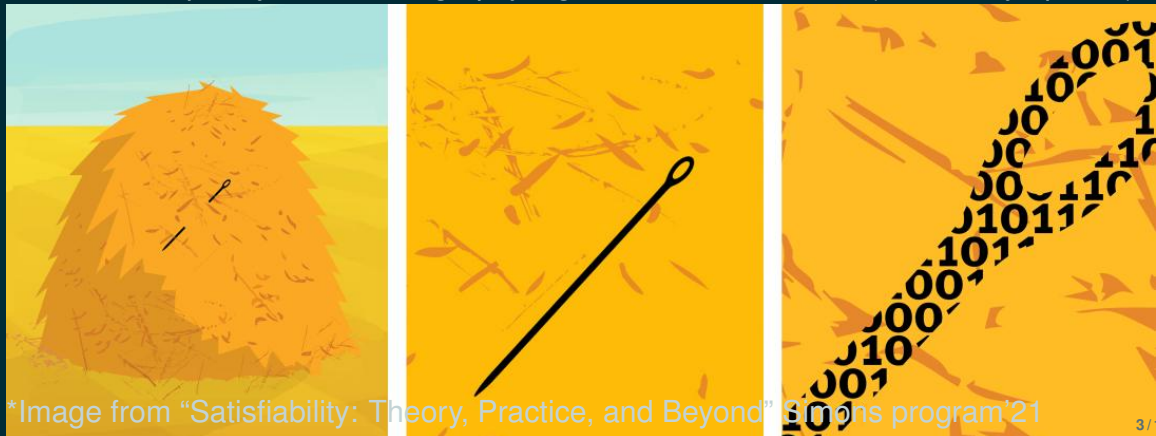
- ▶ For simplicity assume $\|M\| = 1$, and let Δ be the spectral gap of M
- ▶ Sample a random vector v
- ▶ Iterate $\sim \frac{\log d}{\Delta}$ times:
 - compute and update $v \leftarrow Mv$
- ▶ The final vector is close to the top eigenvector $\psi^{(1)}$ with high probability
- ▶ Initial overlap is $|\langle \psi^{(1)}, v \rangle| \approx \frac{1}{\sqrt{d}}$:
 - gets magnified by $\sim (1 + \Delta)$ per step.
- ▶ The overall time complexity is $\sim d^2 / \Delta$; this is provably optimal for $\Delta = \Theta(1)$
- ▶ The Lánczos algorithm improves the gap dependence quadratically $\sim d^2 / \sqrt{\Delta}$

Speed up using quantum linear algebra?

- ▶ Preparing $|v\rangle$ takes time $O(\log d)$ assuming QRAM (Kerenidis-Prakash'17)
- ▶ Applying M takes time $d^{0.5+o(1)}$ (Low'18 + G, Su, Low, Wiebe'18)
- ▶ Applying $\Pi = |\psi^{(1)}\rangle\langle\psi^{(1)}|$ takes time $d^{0.5+o(1)}/\Delta$ (QSVT – G, Su, Low, Wiebe'18)
- ▶ Amplifying Πv to $\Pi v/\|\Pi v\|$ has overhead $\sim \sqrt{d}$
- ▶ Tomography of Πv has overhead $\sim d/\varepsilon$ (Apeldoorn, Cornelissen, G, Nannicini'22)
- ▶ Combining everything the total complexity is $\sim d^2/(\Delta\varepsilon)$
- ▶ We prove a quantum query lower bound $\sim d^{1.5}$ when $\Delta, \varepsilon = \Theta(1)$
- ▶ There is some hope, the running time is $\sim d^{1.5}/(|\langle\psi^{(1)}, v\rangle|\Delta\varepsilon)$

Needle in the haystack: quantum-classical conversions

- ▶ Classical amplification is for free — skip expensive quantum amplification
- ▶ Do tomography directly on $|0\rangle|\Pi v\rangle + |1\rangle|(I - \Pi)v\rangle$
- ▶ Power method is robust to small errors $\Pi v + \zeta$ (Hardt and Price'14)
- ▶ Suffices to ensure the error term is $\|\zeta\| \leq \varepsilon$ and $|\langle \psi^{(1)}, \zeta \rangle| \leq \frac{\varepsilon}{\sqrt{d}}$ in each iteration
- ▶ We develop a new pure-state tomography procedure with such guarantees
- ▶ The complexity of our tomography algorithm remains $\sim d/\varepsilon$ (essentially optimal)



*Image from "Satisfiability: Theory, Practice, and Beyond" Simons program'21

Tomography by computational basis measurements | · |

Okamoto-Hoeffding Bound

Let $0 \leq X \leq 1$ be a bounded random variable, and $s := \frac{X^{(1)} + X^{(2)} + \dots + X^{(n)}}{n}$ be the empirical mean of n i.i.d. samples. Then for $p := \mathbb{E}[X]$ we have that

$$\mathbb{P}(\sqrt{s} \geq \sqrt{p} + \varepsilon) \leq \exp(-2\varepsilon^2 n),$$

$$\mathbb{P}(\sqrt{s} \leq \sqrt{p} - \varepsilon) \leq \exp(-\varepsilon^2 n).$$

Gate-efficient estimation of absolute amplitudes

Given $\frac{1}{\varepsilon^2} \ln\left(\frac{2d}{\delta}\right)$ samples of the pure quantum state $|\varphi\rangle := |0\rangle|\psi\rangle + |1\rangle|\cdot\rangle \in \mathbb{C}^{2d}$, measure each copy and let s_i be the frequency of 0, i outcomes and define

$$\bar{\psi}_i := \sqrt{s_i}.$$

With probability at least $1 - \delta$ it gives an ε - ℓ_∞ approximation of the absolute values $|\psi|$.

Tomography using conditional samples (\mathbb{R})

Gate-efficient tomography of “real” states using reference state $|u\rangle = \sum_{i=0}^{d-1} \frac{1}{\sqrt{d}}|i\rangle$

Given $\frac{4d}{\varepsilon^2} \ln\left(\frac{4d}{\delta}\right)$ samples of the state $|\varphi\rangle := (|+\rangle(|0\rangle|\psi\rangle + |1\rangle|\cdot\rangle) + |-\rangle|0\rangle|u\rangle) / \sqrt{2} \in \mathbb{R}^{4d}$, measure each copy and let $s_{b,i}$ be the frequency of $b, 0, i$ outcomes and define

$$\tilde{\psi}_i := \begin{cases} \max\{0, 2\sqrt{s_{0,i}} - \frac{1}{\sqrt{d}}\} & \text{if } s_{0,i} > s_{1,i} \\ 0 & \text{if } s_{0,i} = s_{1,i} \\ \min\{0, -(2\sqrt{s_{1,i}} - \frac{1}{\sqrt{d}})\} & \text{if } s_{0,i} < s_{1,i} \end{cases}$$

With probability at least $1 - \delta$ it gives an $\frac{\varepsilon}{\sqrt{d}}\text{-}\ell_\infty$ (and thus $\varepsilon\text{-}\ell_2$) approximation of ψ .

- ▶ Idea can be extended for $\psi \in \mathbb{C}^d$, see (Apeldoorn, Cornelissen, **G**, Nannicini'22).

Unbiased tomography using a reference state $\bar{\psi} \in \mathbb{R}^d$

Suppose we have a reference state $\bar{\psi} \in \mathbb{R}^d$ such that $|\bar{\psi}_j|^2 \geq \max\{\frac{\varepsilon^2}{d}, \frac{2}{3}|\psi_j|^2\}$ for all $j \in [d]$. Given $\frac{12d}{\varepsilon^2} \ln\left(\frac{8d}{\delta}\right)$ copies of the pure state $|\varphi\rangle := (|+\rangle|\psi\rangle + |-\rangle|\bar{\psi}\rangle) / \sqrt{2} \in \mathbb{C}^d$, measure each copy and let $s_{b,i}$ be the frequency of b, i outcomes, then

$$\tilde{\psi}_j := \frac{s_{0,j} - s_{1,j}}{\bar{\psi}_j}$$

is an unbiased estimator of $\Re(\psi)$. Moreover, for any $B = \{|v^{(j)}\rangle : j \in [k]\}$ ONB we have

$$\Pr\left[\forall v \in B: |\langle \tilde{\psi} - \Re(\psi) | v \rangle| < \frac{\varepsilon}{\sqrt{d}}\right] \geq 1 - \delta.$$

$$|\langle 0, i | \phi \rangle|^2 - |\langle 1, i | \phi \rangle|^2 = \left| \frac{\psi_i + \bar{\psi}_i}{2} \right|^2 - \left| \frac{\psi_i - \bar{\psi}_i}{2} \right|^2 = \Re(\psi_i) \bar{\psi}_i$$

Concentration follows from the Bennett-Bernstein Bound as $\frac{1}{|\psi_i|} \leq \frac{\sqrt{d}}{\varepsilon}$ and $\|\text{Cov}\| \leq 1$.

Application: Tomography using a reflection $2|\psi\rangle\langle\psi| - I$

Algorithm – quantum noisy power method – Chen, G, de Wolf [QIP'24]

- ▶ Input: Controlled reflection $R = 2|\psi\rangle\langle\psi| - I$ or block-encoding of $\Pi = |\psi\rangle\langle\psi|$
- ▶ Init: sample Gaussian random vector $\phi^{(0)}$
- ▶ For $j = 0 \dots \log(d)$ do
 - Prepare data structure in QRAM for preparing the normalized state $|\phi^{(j)}\rangle$
 - Do tomography on $\Pi|\phi^{(j)}\rangle$ to ℓ_2 -precision ε giving $\phi^{(j+1)}$
- ▶ Initial overlap with $|\psi\rangle$ is $\sim \frac{1}{\sqrt{d}}$, error overlap is $\sim \frac{\varepsilon}{\sqrt{d}}$ in each iteration
- ▶ In each iteration the overlap doubles until it is $\Omega(1)$ – total complexity is $\sim \frac{d \log(d)}{\varepsilon}$
- ▶ Given U block-encoding of a matrix M having gap Δ turn into top-eigenvector projector using QSVT by $\sim \frac{1}{\Delta}$ iterations
- ▶ Sparse M can be block-encoded by $\sim \sqrt{s}$ queries $\Rightarrow \sim d \sqrt{s} / (\Delta \varepsilon)$ overall
- ▶ Query complexity is $\sim d^{1.5} / \Delta$ for dense case and $\varepsilon = \Theta(1)$.

Extension: Process tomography of reflections $2\Pi - I$

- ▶ Similar algorithm works if we are promised rank Π is at most r
- ▶ Sample $\sim r$ independent random vectors and apply power method for each
- ▶ The subspace spanned by the left singular vectors with s.v. $\Omega(1)$ approximate Π
- ▶ Algorithm uses $R = 2\Pi - I$ about $\sim \frac{dr}{\varepsilon}$ times
- ▶ Probably optimal, we prove $\tilde{\Omega}(dr + \frac{d}{\varepsilon})$ lower bound
- ▶ Unitary tomography has complexity $\Theta(d^2/\varepsilon)$ (Haah, Kothari, O'Donnell, Tang'23)

Application to sparse matrices – Chen, G, de Wolf [QIP'24]

- ▶ Given U block-encoding of a matrix M having gap γ below top- r eigensubspace turn into top-eigenvector projector using QSVT by $\sim \frac{1}{\Delta}$ iterations
- ▶ Sparse M can be block-encoded by $\sim \sqrt{s}$ queries $\Rightarrow O(dr \sqrt{s}/(\Delta\varepsilon))$ overall

Iterative refinement

- ▶ Comes from early days of classical computing having limited precision numbers
- ▶ Solves a large linear equation system given such limited arithmetic precision
- ▶ Idea is to solve it only to constant precision, and then recurse
 - ▶ Compute \tilde{x} such that $\|A\tilde{x} - b\| \leq \frac{1}{2}\|b\|$
 - ▶ Set $b \leftarrow A\tilde{x} - b$ and repeat $\log_2(1/\varepsilon)$ times
 - ▶ Take the sum $\bar{x} = \sum \tilde{x}$ of the intermediate solutions
 - ▶ The result satisfies $\|A\bar{x} - b\| \leq \varepsilon\|b\|$
- ▶ In the quantum case this requires updating the state preparation unitary
- ▶ E.g., use classical write quantum read QRAM
- ▶ Could get a polynomial speedup when classical output is required (IP solvers)
- ▶ Idea pioneered by Mohammadhossein Mohammadisiahroudi, Brandon Augustino, Tamás Terlaky, et al.
- ▶ Similar ideas can be applied to pure state tomography

Iteratively refined tomography

Refinement step

- ▶ Want to learn $\psi^{(0)}$, by improving the current estimate $\psi^{(1)}$
- ▶ Input: $\varepsilon \in (0, 2]$, unitaries $U^{(0)}, U^{(1)}$ such that $\|\psi^{(0)} - \psi^{(1)}\| \leq \varepsilon$ for $\psi^{(i)} := (\langle 0^a | \otimes I) U^{(i)} |0^q\rangle$
- ▶ Let $W := |+\chi+\rangle \otimes U^{(0)} - |-\chi-\rangle \otimes U^{(1)}$, and $k \approx 1/\varepsilon$
- ▶ Let $AA(W, k)$ be the k -step amplitude amplification of $|\phi\rangle := \psi^{(0)} - \psi^{(1)} = (\langle 0^{a+1} | \otimes I) W |0^{q+1}\rangle$
- ▶ Perform tomography on $|\phi\rangle := (\langle 0^{a+1} | \otimes I) AA(W, k) |0^{q+1}\rangle$ to ℓ_2 -precision $\frac{1}{6}$ giving estimate $\tilde{\phi}$
- ▶ Output: $\psi^{(1)} \leftarrow \psi^{(1)} + \frac{2\tilde{\phi}}{2k+1}$ (satisfying $\|\psi^{(0)} - \psi^{(1)}\| \leq \varepsilon/2$ with probability $\geq 1 - \delta$)
- ▶ Complexity: $\approx \frac{1}{\varepsilon} \times d$

Quantum lower bound

Hard instance

- ▶ Hide d bits of information in $\phi \in \{-1, +1\}^d$.
- ▶ Define $M := \frac{|\phi\rangle\langle\phi|}{2d} + \mathcal{N}(0, \frac{1}{100\sqrt{d}})$ entry-wise noise.
- ▶ Because of random matrix theory with high probability $\|M\| \leq 1$ and $\Delta \geq \frac{1}{4}$.
- ▶ The top-eigenvector $\psi^{(1)}$ is $O(1)$ -close to ϕ . Learning $\psi^{(1)}$ reveals $\Omega(d)$ bits.

Quantum lower bound – using a variant of the adversary method

- ▶ To learn ϕ_i we need to extract sign $\pm\frac{1}{d}$ from noisy entries $\pm\frac{1}{d} + \mathcal{N}(0, \frac{1}{100\sqrt{d}})$.
- ▶ Requires $\sim d$ “samples” classically $\Rightarrow \sim \Omega(d^2)$ queries to learn most bits.
- ▶ Requires $\sim \sqrt{d}$ “queries” quantumly $\Rightarrow \sim \Omega(d^{1.5})$ queries to learn most bits.