Quantum Walks

András Gilyén

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Outline

- Coined quantum walks
- Szegedy quantum walks
- Continuous-time quantum walks

Coined quantum walks

(Classical) Random walk on the line

In each step we toss a coin:

- ▶ if heads we go one step left
- ▶ if tails we go one step right

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- The probability of being further than $\sim \sqrt{T}$ is small.
- ▶ Being at some specific point of the interval $[-\sqrt{T}, \sqrt{T}]$ has chance ~ 1/ \sqrt{T} , i.e., the distribution is roughly uniform on this interval.

Quantum walk on the line

$$S = \begin{bmatrix} |n, 0\rangle \rightarrow |n-1, 1\rangle & |0, 1\rangle & |1, 1\rangle & |2, 1\rangle & |3, 1\rangle \\ |-3, 0\rangle & |-2, 0\rangle & |-1, 0\rangle & |0, 0\rangle & |1, 0\rangle & |2, 0\rangle \\ & & & \\ START \\ |n, 1\rangle \rightarrow |n+1, 1\rangle & C = \begin{bmatrix} |n, 0\rangle \rightarrow \frac{1}{\sqrt{2}} |n, 0\rangle + \frac{1}{\sqrt{2}} |n, 1\rangle \\ |n, 1\rangle \rightarrow \frac{1}{\sqrt{2}} |n, 0\rangle - \frac{1}{\sqrt{2}} |n, 1\rangle \\ |n, 1\rangle \rightarrow \frac{1}{\sqrt{2}} |n, 0\rangle - \frac{1}{\sqrt{2}} |n, 1\rangle \end{bmatrix}$$

Quantum walk on the line



One step of the walk: SC i.e., first apply C and then apply S

- Quantum walk is deterministic and reversible!
- Only the measurement introduces probability.

Distribution after *T* **steps** – upon measurement



The probability distribution of the quantum random walk with Hadamard coin starting in $|0,0\rangle$ after T = 100 steps.

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Main differences

- Peak around $T/\sqrt{2} \rightarrow$ ballistic spreading
- Between -T/2 and T/2 the distribution is ~ uniform \rightarrow quadratically faster mixing

Szegedy quantum walks

Discrete-time Markov-chain on a weighted graph

Transition probability in one step (stochastic matrix)

$${\sf P}_{{\sf v}{\sf u}}={\sf Pr}({\sf step to } {\sf v}\,|\,{\sf being at }{\sf u})=rac{{\sf W}_{{\sf v}{\sf u}}}{\sum_{{\sf v}'\in {\sf U}}{\sf w}_{{\sf v}'{\sf u}}}$$

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A unitary implementing the update

$$U: |0\rangle |u\rangle \mapsto \sum_{v \in V} \sqrt{P_{vu}} |v\rangle |u\rangle$$

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$$U: |0\rangle|u
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Generic coin operator

S := SWAP $C := U((2|0\rangle 0| \otimes I) - I)U^{\dagger}$

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How to erase history? The Szegedy quantum walk operator:

 $W' := U^{\dagger} \cdot \text{SWAP} \cdot U$ $W := U^{\dagger} \cdot \text{SWAP} \cdot U((2|0\rangle 0| \otimes I) - I)$

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$$\langle 0|\langle u|W'|0\rangle|v\rangle = \langle 0|\langle u|U^{\dagger} \cdot \mathrm{SWAP} \cdot U|0\rangle|v\rangle = \left(\sum_{v' \in V} \sqrt{P_{v'u}}|v'\rangle|u\rangle\right)^{\dagger} \mathrm{SWAP}\left(\sum_{u' \in V} \sqrt{P_{u'v}}|u'\rangle|v\rangle\right)$$

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Multiple steps of the quantum walk: $(\langle 0 | \otimes l \rangle W^k (| 0 \rangle \otimes l) = T_k(P)$ $[T_k(x) = \cos(k \arccos(x))$ Chebyshev polynomials: $T_{k+1}(x) = 2xT_k(x) - T_{k-1}(x)]$

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Linear combination of (non-)unitary mat. [Childs & Wiebe '12, Berry et al. '15] Suppose that $U = \sum_{i} |i| \langle i| \otimes U_i$, and $Q : |0\rangle \mapsto \sum_{i} \sqrt{q_i} |i\rangle$ for $q_i \in [0, 1]$.

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Used for Hamiltonian simulation, and much more!

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Corollary: Quantum fast-forwarding (Apers & Sarlette 2018)

We can implement a unitary V such that

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with using only $O(\sqrt{t \log(1/\varepsilon)})$ quantum walk steps.

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Szegedy quantum walk based search

Suppose we have some unknown marked vertices $M \subset V$.

Quadratically faster hitting

Hitting time: expected time to hit a marked vertex starting from the stationary distr. Starting from the quantum state $\sum_{v \in V} \sqrt{\pi_v} |v\rangle$ we can

- detect the presence of marked vertices $(M \neq 0)$ in time $O(\sqrt{HT})$ (Szegedy 2004)
- ▶ find a marked vertex in time $O\left(\frac{1}{\sqrt{\delta \varepsilon}}\right)$ (Magniez, Nayak, Roland, Sántha 2006)
- Find a marked vertex in time $\widetilde{O}(\sqrt{HT})$ (Ambainis, G, Jeffery, Kokainis 2019)

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Starting from arbitrary distributions

Starting from distribution σ on some vertices we can

- detect marked vertices in square-root commute time $O(\sqrt{C_{\sigma,M}})$ (Belovs 2013)
- Find a marked vertex in time $\widetilde{O}(\sqrt{C_{\sigma,M}})$ (Piddock; Apers, G, Jeffery 2019)

Element Distinctness

- Black box: Computes f on inputs corresponding to elements of [n]
- Question: Are there any $i \neq j \in [n] \times [n]$ such that f(i) = f(j)?
- Query complexity: $O(n^{2/3})$ (Ambainis 2003) $\Omega(n^{2/3})$ (Aaronson & Shi 2001)

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Triangle Finding

[(2014) non-walk algorithm by Le Gall: $O(n^{5/4})$]

- Black box: For any pair $u, v \in V \times V$ tells whether there is an edge uv
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Matrix Product Verification

- Black box: Tells any entry of the $n \times n$ matrices A, B or C.
- Question: Does AB = C hold?
- Query complexity: $O(n^{5/3})$ (Buhrman, Špalek 2004)

Continuous-time quantum walks

Continuous-time quantum / random walks

Laplacian of a weighted graph

Let G = (V, E) be a finite simple graph, with non-negative edge-weights $w : E \to \mathbb{R}_+$. The Laplacian is defined as

$$u \neq v$$
: $L_{uv} = w_{uv}$, and $L_{uu} = -\sum_{v} w_{uv}$.

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Evolution of the state:

$$\frac{d}{dt}\rho_u(t) = \sum_{v \in V} L_{uv} \rho_v(t) \qquad \Longrightarrow \qquad \qquad \rho(t) = e^{tL} \rho(0)$$

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Exponential speedup by a quantum walk

Childs, Cleve, Deotto, Farhi, Gutmann, and Spielman: quant-ph/0209131

Same speed-up by Szegedy walks?

Can be reduced to the line

$$A = \frac{1}{3} \begin{pmatrix} 1 & \sqrt{2} & 0 & 0 & 0 & 0 & 0 & 0 \\ \sqrt{2} & 0 & \sqrt{2} & 0 & 0 & 0 & 0 & 0 \\ 0 & \sqrt{2} & 0 & \sqrt{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & \sqrt{2} & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 0 & \sqrt{2} & 0 & 0 \\ 0 & 0 & 0 & 0 & \sqrt{2} & 0 & \sqrt{2} & 0 \\ 0 & 0 & 0 & 0 & 0 & \sqrt{2} & 0 & \sqrt{2} \\ 0 & 0 & 0 & 0 & 0 & 0 & \sqrt{2} & 1 \end{pmatrix}$$

Show that the bottom left corner of $T_m(A)$ is 1/poly(n) large for some m = poly(n).