# Quantum Walks 

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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 $\sim 1 / \sqrt{T}$, i.e., the distribution is roughly uniform on this interval.


## Quantum walk on the line



## 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.

## Distribution after $T$ steps - upon measurement



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## Main differences

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


## Szegedy quantum walks

## Discrete-time quantum / random walks

## Discrete-time Markov-chain on a weighted graph

Transition probability in one step (stochastic matrix)

$$
P_{v u}=\operatorname{Pr}(\text { step to } v \mid \text { being at } u)=\frac{w_{v u}}{\sum_{v^{\prime} \in U} w_{v^{\prime} u}}
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## A unitary implementing the update

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U:|0\rangle|u\rangle \mapsto \sum_{v \in V} \sqrt{P_{v u}}|v\rangle|u\rangle
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## Generic coin operator

$$
\begin{aligned}
& S:=\text { SWAP } \\
& C:=U((2|0 \times 0| \otimes I)-I) U^{\dagger}
\end{aligned}
$$

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How to erase history? The Szegedy quantum walk operator:

$$
\begin{aligned}
W^{\prime} & : \\
W: & U^{\dagger} \cdot \operatorname{SWAP} \cdot U \\
W & U^{\dagger} \cdot \operatorname{SWAP} \cdot U((2|0 X 0| \otimes I)-I)
\end{aligned}
$$

## Understanding Szegedy's quantum walk operator

For simplicity let us assume $P_{u v}=P_{v u}$, i.e., the total weight of vertices is constants.

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

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

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## Are we happy with Chebyshev polynomials?

Linear combination of (non-)unitary mat. [Childs \& Wiebe '12, Berry et al. '15]
Suppose that $U=\sum_{i}|i X i| \otimes U_{i}$, and $Q:|0\rangle \mapsto \sum_{i} \sqrt{q_{i}}| \rangle$ 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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(\langle 0| \otimes I) \vee(|0\rangle \otimes I) \stackrel{\varepsilon}{\approx} P^{t}
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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 $O(\sqrt{H T})$ (Szegedy 2004)
- find a marked vertex in time $O\left(\frac{1}{\sqrt{\delta \varepsilon}}\right)$ (Magniez, Nayak, Roland, Sántha 2006)
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## Starting from arbitrary distributions

Starting from distribution $\sigma$ on some vertices we can

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


## Walks on the Johnson graph (Sántha arXiv:0808.0059)

 Vertices: $\{S \subset N:|S|=K\}$; Edges: $\left\{\left(S, S^{\prime}\right):\left|S \Delta S^{\prime}\right|=2\right\}$
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## 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: $O\left(n^{2 / 3}\right)$ (Ambainis 2003) $\Omega\left(n^{2 / 3}\right)$ (Aaronson \& Shi 2001)


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## Triangle Finding

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

- Black box: For any pair $u, v \in V \times V$ tells whether there is an edge $u v$
- Question: Is there any triangle in $G$ ?
- Query complexity: $O\left(n^{13 / 10}\right)$ (Magniez, Sántha, Szegedy 2003)


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## Matrix Product Verification

- Black box: Tells any entry of the $n \times n$ matrices $A, B$ or $C$.
- Question: Does $A B=C$ hold?
- Query complexity: $O\left(n^{5 / 3}\right)$ (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 $w: E \rightarrow \mathbb{R}_{+}$. The Laplacian is defined as

$$
u \neq v: L_{u v}=w_{u v} \text {, and } L_{u u}=-\sum_{v} w_{u v} .
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## Continuous-time walks

Evolution of the state:

$$
\frac{d}{d t} p_{u}(t)=\sum_{v \in V} L_{u v} p_{v}(t) \quad \Longrightarrow \quad p(t)=e^{t L} p(0)
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i \frac{d}{d t} \psi_{u}(t)=\sum_{v \in V} L_{u v} \psi_{v}(t) & \Longrightarrow & \psi(t)=e^{-i t L} \psi(0)
\end{array}
$$

## 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

$$
A=\frac{1}{3}\left(\begin{array}{cccccccc}
1 & \sqrt{2} & 0 & 0 & 0 & 0 & 0 & 0 \\
\sqrt{2} & 0 & \sqrt{2} & 0 & 0 & 0 & 0 & 0 \\
0 & \sqrt{2} & 0 & \sqrt{2} & 0 & 0 & 0 & 0 \\
0 & 0 & \sqrt{2} & 0 & 2 & 0 & 0 & 0 \\
0 & 0 & 0 & 2 & 0 & \sqrt{2} & 0 & 0 \\
0 & 0 & 0 & 0 & \sqrt{2} & 0 & \sqrt{2} & 0 \\
0 & 0 & 0 & 0 & 0 & \sqrt{2} & 0 & \sqrt{2} \\
0 & 0 & 0 & 0 & 0 & 0 & \sqrt{2} & 1
\end{array}\right)
$$

Show that the bottom left corner of $T_{m}(A)$ is $1 / \operatorname{poly}(n)$ large for some $m=\operatorname{poly}(n)$.

