## MATH 217A. Introduction to Quantum Algorithms

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## Lecture 3. Continuous and discrete time classical random walks

In the previous lecture we discussed classical and quantum random walks on $\mathbb{Z}_{N}$, in continuous time, and took continuum limits in space to get the heat and Schrödinger equations.

Problem 3.1. The classical and quantum random walk evolutions preserve the $\ell^{1}$ - and $\ell^{2}$-norms, respectively, of the state vector. Can you construct a system of linear ODEs, local on $\mathbb{Z}_{N}$, that preserves the $\ell^{3}$-norm? If not, can you prove it is impossible? Or can you construct a nonlocal, or nonlinear system that does so? For context and hints see [1] and [2].

The physical systems Feynman considered in his original paper [3], however, are discrete in time, as well as in space. So we want to consider discrete time random walks, starting with classical ones.

## Discrete time integrated random walk

The continuous time dynamics from the last lecture is $\dot{p}=b L p$, which has solution

$$
\begin{equation*}
p(t)=e^{b L t} p(0)=M(t) p(0) \tag{1}
\end{equation*}
$$

where $M(t)$ is a one-parameter family of Markov matrices, i.e., matrices with non-negative entries summing to 1 in each column, which thus preserve the $\ell^{1}$-norm.

A Markov matrix $P$ is embeddable if there is an intensity matrix $Q$, such that $P=e^{Q}$ and $e^{Q t}$ is a Markov matrix for $t \geq 0 . Q$ has non-negative offdiagonal entries, and each column sums to 0 . For each $t, M(t)$ defined by (1) is embeddable, since it is embedded in the one-parameter family $M(t)=e^{b L t}$. See Davies [4] for a recent survey of results on this topic.

Integrating the continuous time evolution of (1) for time $\Delta t$ will give a discrete time evolution matrix $M(\Delta t)=e^{b L \Delta t}$.

Since the dynamics is translation invariant: $L=-2 I+X+X^{-1}$, where $X_{i j}=\delta_{i, j+1}$, $i, j \in \mathbb{Z}_{N}$, we can compute $M(\Delta t)$ by diagonalizing $L$ using the discrete Fourier transform:

$$
F_{k x}=\frac{1}{\sqrt{N}} e^{-2 \pi i k x / N}=\frac{1}{\sqrt{N}} \omega^{-k x}, \text { where } \omega^{N}=1, \text { so }\left(F^{-1}\right)_{y k}=\frac{1}{\sqrt{N}} \omega^{y k}
$$

For practice let's check that this is really $F^{-1}$ :

$$
\left(F^{-1} F\right)_{y x}=\frac{1}{N} \sum_{k} \omega^{y k} \omega^{-k x}=\frac{1}{N} \sum_{k} \omega^{(y-x) k}= \begin{cases}1 & \text { if } y-x=0 \\ \frac{1}{N} \frac{1-\omega^{(y-x) N}}{1-\omega^{y-x}}=0 & \text { otherwise }\end{cases}
$$

The calculation for $F F^{-1}$ is similar. Now let's see that the discrete Fourier transform diagonalizes $X$ :

$$
\left(F^{-1} X F\right)_{y x}=\frac{1}{N} \sum_{k l} \omega^{y k} \delta_{k, l+1} \omega^{-l x}=\frac{1}{N} \sum_{k} \omega^{y k} \omega^{-(k-1) x}=\omega^{x} \frac{1}{N} \sum_{k} \omega^{k(y-x)}=\omega^{x} \delta_{y x} .
$$

Similarly, $\left(F^{-1} X^{-1} F\right)_{y x}=\omega^{-x} \delta_{y x}$. Thus

$$
\left(F^{-1} L F\right)_{y x}=-2+\left(\omega^{x}+\omega^{-x}\right) \delta_{y x}=-\left|1-\omega^{x}\right|^{2} \delta_{y x}=: D_{y x}
$$

Now $M(\Delta t)=F e^{b D \Delta t} F^{-1}$. The first column of $e^{b D \Delta t} F^{-1}$ is just the diagonal of $e^{b D \Delta t}$, so the zeroth column of $M(\Delta t)$ is the discrete Fourier transform of this:

$$
\begin{equation*}
(M(\Delta t))_{y 0}=\sum_{x} F_{y x} e^{-b\left|1-\omega^{x}\right|^{2} \Delta t}=\frac{1}{\sqrt{N}} \sum_{x} \omega^{-y x} e^{-b\left|1-\omega^{x}\right|^{2} \Delta t} \tag{2}
\end{equation*}
$$

$M(\Delta t)$ is a circulant matrix, so this suffices to determine it completely.
Problem 3.2. The elements of $M$ are real since the $x$ and $-x$ terms in the sum (2) are complex conjugates. Prove from the formula (2) that they are non-negative, and sum to 1 in each column.

For large $N, \omega=e^{2 \pi i / N}=: e^{2 \pi i \epsilon} \approx 1+2 \pi i \epsilon$, so $1-\omega^{x} \approx-2 \pi i x \epsilon$, for $x \ll N$. In this approximation,

$$
M(\Delta t)_{y 0} \approx \sum_{x} e^{-2 \pi i y x} e^{-4 \pi^{2} x^{2} \epsilon^{2} b \Delta t}
$$

The zeroth column of $M(\Delta t)$ is thus the discrete Fourier transform of a discretization of a Gaussian; thus it is also a discretization of a Gaussian. In particular, $M(\Delta t)_{y 0}$ is not supported only on $y \in\{0, \pm 1\}$, so the Markov matrix for this discrete time integrated random walk is not local in the way that the intensity matrix is.

An alternate (microscopic) perspective
The ODE

$$
\dot{p}_{x}=b p_{x+1}-2 b p_{x}+b p_{x-1}
$$

describes the change in the probability of a particle to be at position $x$ when it is hopping away from $x$, to $x-1$ or to $x+1$, each with equal probability, at a rate of $2 b$ hops per unit time. The adjacent figure shows a typical path, or history, for the particle; the times of the hops constitute a Poisson process. In time $\Delta t$ any number $n$ of hops is possible, with probability given by

$$
\frac{e^{-2 b \Delta t}(2 b \Delta t)^{n}}{n!}
$$

Thus

$$
\begin{aligned}
\operatorname{Pr}(\text { move } y \text { to right in time } \Delta t) & =\sum_{n=0}^{\infty} \operatorname{Pr}(n \text { hops }) \operatorname{Pr}(\text { hops right }- \text { hops left }=y) \\
& =\sum_{n=0}^{\infty} \frac{e^{-2 b \Delta t}(2 b \Delta t)^{n}}{n!}\binom{n}{(n-y) / 2} \frac{1}{2^{n}}
\end{aligned}
$$

The latter term in the sum is approximately a Gaussian around $y=0$, so the result is a weighted sum of such Gaussians, with the largest weight for $n \approx 2 b \Delta t$. Again, we can see that the probability distribution is not supported only on $y \in\{0, \pm 1\}$.

Discrete time local random walk
Can we find a local discrete time random walk? It should be defined by $p_{t+1}=M p_{t}$, where $M$ is a Markov matrix with the same sparsity pattern as the intensity $b L$. If we assign probability $q$ to hopping in each time step, with the probability split equally between left and right, we get:

$$
\begin{align*}
p_{x}(t+1) & =\frac{q}{2} p_{x+1}(t)+(1-q) p_{x}(t)+\frac{q}{2} p_{x-1}(t)  \tag{3}\\
\Longrightarrow p_{x}(t+1)-p_{x}(t) & =\frac{q}{2}\left(p_{x+1}(t)-2 p_{x}(t)+p_{x-1}(t)\right) .
\end{align*}
$$

Assuming $p(t, x)$ is a sufficiently smooth function of $t$ and $x$, this becomes

$$
\begin{aligned}
\Delta t \frac{\partial p}{\partial t} & =\frac{q}{2}(\Delta x)^{2} \frac{\partial^{2} p}{\partial x^{2}}+\text { higher order terms } \\
\Longrightarrow \frac{\partial p}{\partial t} & =\frac{q}{2} \frac{(\Delta x)^{2}}{\Delta t} \frac{\partial^{2} p}{\partial x^{2}}+\text { higher order terms. }
\end{aligned}
$$

Now, taking the $\Delta x, \Delta t \rightarrow 0$ limit with $\frac{q}{2} \frac{(\Delta x)^{2}}{\Delta t}=\kappa$ gives

$$
\frac{\partial p}{\partial t}=\kappa \frac{\partial^{2} p}{\partial x^{2}} .
$$

The stochastic process defined by (3) is the usual (discrete time, classical) random walk; we have just shown that despite being local, it has the heat equation as its continuum limit, just as does the nonlocal, discrete time integrated random walk (2).

## References

[1] R. D. Sorkin, "Quantum mechanics as quantum measure theory", Mod. Phys. Lett. A 9 (1994) 3119-3127; arXiv:gr-qc/9401003.
[2] S. Aaronson, "Is quantum mechanics an island in theoryspace?", in A. Khrennikov, ed., Proceedings of the Växjö Conference Quantum Theory: Reconsideration of Foundations (2004); arXiv:quant-ph/0401062.
[3] R. P. Feynman, "Simulating physics with computers", Int. J. Theor. Phys. 21 (1982) 467-488.
[4] E. B. Davies, "Embeddable Markov matrices", Electronic J. Prob. 15 Art. 47, http://www.emis.de/journals/EJP-ECP/article/view/733.html; arXiv:1001.1693 [math.PR].

