Lecture Topic: Quantum Walks and Quantum Replacements of Monte Carlo Sampling

Introduction to Quantum Walks

- Quantum (Random) Walks: fundamental concept in the realm of quantum computing
- Distinct perspective on random processes compared to their classical counterparts
- Quantum walks, and algorithms that utilize them, have several important features

Speedups

- Quantum walks often show quadratic speedups
- Sometimes show exponential speedups
- Hidden Flat Problem
- Quantum walks form a model of universal (quantum) computation

Definitions

- Quantum walk: process on a graph G = (V, E)
- Basis states $|x\rangle, x \in V$

In what follows, for simplicity, let $V = \mathbb{Z}$ in what follows.

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The "naive" unitaries

Consider $U \in U(N)$ such that

$$egin{aligned} & U: \mathcal{H}_G
ightarrow \mathcal{H}_G \ & |x
angle \mapsto a|x-1
angle + b|x
angle + c|x+1
angle \end{aligned}$$

which conveys the information for the potential that $|x\rangle$

- **(**) moves left with some amplitude $a \in \mathbb{C}$,
- **2** stays at the same place with amplitude $b \in \mathbb{C}$,
- **③** moves right with amplitude $c \in \mathbb{C}$.

The quantum walk process needs exhibit consistent behavior across all vertices:

That is, a, b and c should be independent of $x \in V$ (similarly to how the probabilities of moving left/right are independent of x in the classical random walk). Unfortunately, this definition does not work.

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The coin space

Solution: additional "coin" space: $|i, x\rangle$ for $i \in \{0, 1\}$, $x \in \mathbb{Z}$, with Hilbert spaces $\mathcal{H}_{\mathrm{C}}, \mathcal{H}_{\mathrm{W}}$. At each step, we perform two unitary operations:

- (1) A coin flip operation $C : \mathcal{H}_{C} \to \mathcal{H}_{C}$ which "puts" the walker in superposition, so it walks all possible paths simultaneously.
- (2) Followed by a **shift** operation $S : \mathcal{H}_W \to \mathcal{H}_W$ the operator responsible for the actual walk on G.

$$C|i, x\rangle = \begin{cases} a|0, x\rangle + b|1, x\rangle & \text{if } i = 0, \\ c|0, x\rangle + d|1, x\rangle & \text{if } i = 1. \end{cases}$$

$$S|i, x\rangle = \begin{cases} |0, x+1\rangle & \text{if } i = 0, \\ |1, x-1\rangle & \text{if } i = 1. \end{cases}$$
(1.2)
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Hadamard walker

If we choose for C the Hadamard matrix:

$$H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1\\ 1 & -1 \end{pmatrix}, \qquad (1.4)$$

we have a 'Hadamard walker' while S can be explicitly described as follows:

$$S = \left(|0
angle \langle 0| \otimes \sum_{x=-\infty}^{\infty} |x+1
angle \langle x|
ight) + \left(|1
angle \langle 1| \otimes \sum_{x=-\infty}^{\infty} |x-1
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 (1.5)

The walker's unitary step and asymmetry

A step of a quantum walk amounts to the unitary $U = SC \in U(N)$.



Figure: Probability distribution of quantum walk, starting at $|-,0\rangle,$ after different numbers of steps.

The quantum walker's initial state is the product of the coin state and the position state. The former state controls the direction in which the walker moves. Therefore, the choice of coin operator leads to vastly different constructive and destructive interference patterns.

This behavior is in stark contrast to a classical random walk, where the walker has equal probability of moving left or right at each step, and there is no preference or bias for either direction. The bias in a quantum walk is a unique characteristic of the underlying physics.

Example on a bounded subset of the integer line with C = H. It is common to assume that the walker starts at position x = 0 with the coin state being the $|0\rangle$ or $|1\rangle$ state.



Figure: Beginning a quantum walk, after the coin operator has been applied, at $|+,0\rangle$, by applying C = H on $|0,0\rangle$, on the \mathbb{Z} -line.

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For ease of notation, we denote the *r*-th application of the quantum walk operator U by $U^{(r)}|\psi_{r-1}\rangle$.

The quantum walk amounts to the following set of operations:

- Select coin operator C = H
- Initialize the state (position of the walker): $|\bm{0}\rangle=|0\rangle_{\rm C}\otimes|0\rangle_{\rm W}=|0,0\rangle$ (or $|1,0\rangle)$
- for $r \in \mathbb{N}$ repeat $U^r | \mathbf{0}
 angle$ as:
 - Apply the coin operator: $C|\mathbf{0}\rangle$
 - Apply the shift operator: $S(C|\mathbf{0}\rangle)$
- Measure $U^r | \mathbf{0} \rangle$

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Therefore, the initial state is $|\mathbf{0}\rangle\equiv|\psi_{\mathbf{0}}\rangle$ and we obtain

$$\begin{aligned} |\psi_{1}\rangle &= \frac{|0,-1\rangle + |0,1\rangle}{\sqrt{2}} \\ |\psi_{2}\rangle &= \frac{|0,-2\rangle + |1,0\rangle + |0,0\rangle - |1,2\rangle}{2} \\ |\psi_{3}\rangle &= \frac{|1,-3\rangle - |0,-1\rangle + 2(|0\rangle + |1\rangle)|1\rangle + |0,3\rangle}{2\sqrt{2}} \end{aligned}$$
(1.6)

This state is not symmetric around the origin and the probability distributions will not be centered at the origin. This is clear from Fig. 1.1. As a matter of fact the standard deviation of the walker, after r iterations of U is:

$$\sigma(r) \approx 0.54r \tag{1.9}$$

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Standard deviation



Figure: The standard deviation of a classical versus quantum walk as a function of the steps.

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The walker's efficiency



Aspman/Korpas/Mareček (CTU)

Quantum Computing

Quantum Walk on a Complete Graph



Figure: An asymmetric non-complete graph G = (8, 10) and its symmetric completion $\overline{G} = K_8$.

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Easy-to-work-with graph, the complete graph K_4 with 4 vertices and 6 edges and perform such *search*.

Let us commence with a classical random walk on K_4 wherein we are looking to "find" the marked vertex #2 (but we do not know it). In the Fig. next we display the success probability after 1 and 2 steps.

Classical walk on $G = K_4$



Figure: Left: At step 1 the probability that the walker "lands" on vertex #2 is 1/4. Right: At step 2 the probability that the walker "lands" on vertex #2 is 1/2. The loop in vertex #2 denotes that this vertex is a trap: it allows us to know the walker landed on the marked vertex and the walker is not allowed to attain any other state.

Success Probability

Overall, the trend for the success probability continues, and we observe the behavior of the walker. For large N, the success probability of 1/2 is reached after $\mathcal{O}(N)$ steps.



We have to implement the coin and shift operators. Diagrammatically at step 0 we are back at the initial state of the classical walk. In total we have 12 amplitudes to consider;



Figure: Left: the state of the quantum walk is a superposition of the amplitudes $a_{ij} \in \mathbb{C}$, for all $i, j \in V(K_4)$. Once the oracle is applied the marked state's amplitudes obtain a negative sign (marked with blue and in analogy with Grover's operator).

Initially, we have $a_{ij} = \frac{1}{\sqrt{12}}$ for all i, j.

Then, the coin flip operator C, which here is taken to be Grover's diffusion operator, amounts to marking the state we look for, assigning a negative sign to the corresponding amplitudes. The marking is done by assuming access to an oracle O (essentially the same oracle found in Grover's operator) that is able to perform this operation. Then, it changes the direction of adjacent red-blue pair vertices.

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Then S reverses the amplitude values along their mean at each vertex. For example, the mean of the vertex #1 after application of C is

$$\mu_{12} = \frac{a_{21} + a_{13} + a_{14}}{3}.$$
 (1.10)

Therefore, S amounts to a map $S : a_{ij} \mapsto a'_{ij} = \mu_{12} - a_{ij}$, for the three pairs $\{21, 13, 14\}$. Of course, this is applied to all amplitudes for all vertices.

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Figure: Left: Coin operator is applied and reverses the relevant amplitudes. The shift operator reverses these amplitudes along their means.

In the second step, we already get the amplitude asymmetry resulting from the oracle flipping the signs of the marked vertex followed by C and then S. As a result, one observes that:

probability of success at step
$$1 = \frac{11}{108} \approx 0.1$$
 (1.11)
probability of success at step $2 = \frac{25}{36} \approx 0.7$ (1.12)
probability of success at step $3 = \dots$ (1.13)

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Quantum Computing

Overall, for a large number of vertices N, the probability that the walker lands on the marked vertex is 1/2 is given after $\pi\sqrt{N}$ steps and therefore the run-time is $\mathcal{O}(\sqrt{N})$. This marks another example in which quantum walks portray a quadratic speedup over classical random walks.
Consider an undirected and unweighted graph G. Szegedy's quantum walk occurs on the edges of the bipartite double cover of the original graph.

If the original graph is G, then its bipartite double cover is the graph tensor product $G \times K_2$ which duplicates the vertices into two partite sets X and Y.

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Figure: Left: A graph G. Right: The bipartite double cover of G. The double cover contains double the number of edges.

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The Hilbert space of a Szegedy walk: $\mathbb{C}^{2|E|}$.

Let us denote a walker on the edge connecting $x \in X$ with $y \in Y$ as $|x, y\rangle$. Then the computational basis is:

$$|x,y\rangle, \qquad x \in X, y \in Y, x \sim y$$
 (1.14)

where $x \sim y$ denotes that the vertices x and y are adjacent.

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Szegedy's walk is defined by repeated applications of the unitary

$$U = R_2 R_1, \tag{1.15}$$

where

$$R_{1} = 2 \sum_{x \in X} |\phi_{x}\rangle \langle \phi_{x}| - \mathbf{1}$$

$$R_{2} = 2 \sum_{y \in Y} |\psi_{y}\rangle \langle \psi_{y}| - \mathbf{1},$$
(1.16)
(1.17)

are reflection operators and

$$\begin{split} |\phi_{x}\rangle &= \frac{1}{\sqrt{\deg(x)}} \sum_{y \sim x} |x, y\rangle \tag{1.18} \\ |\psi_{y}\rangle &= \frac{1}{\sqrt{\deg(y)}} \sum_{x \sim y} |x, y\rangle \tag{1.19}$$

Before:

- deg(x) is the degree of vertex x
- $y \sim x$ denotes the sums over the neighbors of x

Observe that $|\phi_x\rangle$ is the equal superposition of edges incident to $x \in X$, and $|\psi_y\rangle$ is the equal superposition of edges incident to $y \in Y$.

Here, there is an equivalent of the "inversion about the mean" operation of Grover's algorithm, which we also saw previously in the context of walks over K_4 . The reflection R_1 goes through each vertex in X and reflects the amplitude of its incident edges about their average amplitude, and R_2 similarly does this for the vertices in Y.

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Classically: search for a marked vertex on G amounts to randomly walks until a marked vertex is found (then walker freezes)

Quantumly: Szegedy's quantum walk searches by quantizing this random walk with absorbing vertices and the resulting bipartite double cover. Search is performed by repeatedly applying the unitary

$$\widetilde{U} = \widetilde{R}_2 \widetilde{R}_1, \tag{1.20}$$

where the tilde distinguishes in that we are searching for absorbing vertices. At unmarked vertices they act as $\widetilde{R}_j = R_j$ simply by inverting the amplitudes of the edges around their average at each vertex. At the marked vertices, similarly to the K_4 case, they act by flipping the signs of the amplitudes of all incident edges. A similar search can be performed using Grover's diffusion operator.

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Figure: The marked state corresponds to vertex #2 which is an absorbing vertex: $\langle 2_Y | 2_X \rangle = \langle 2_X | 2_Y \rangle = 0.$

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Continuous-time random walk but quantum..

Will allow us later to understand the universality of quantum walks.

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Continuous-time random walk on a graph G = (V, E) with adjacency matrix A defined as:

$$A_{i,j} = \begin{cases} 1, & (i,j) \in E \\ 0, & (i,j) \notin E \end{cases}$$
(1.21)

for every pair $i, j \in V$. In this definition we do not allow self-loops therefore the diagonal of A is zero.

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There is another matrix associated with G that is of equal importance, the Laplacian of G defined as:

$$L_{i,j} = \begin{cases} -\deg(i), & i = j \\ 1, & (i,j) \in E \\ 0, & \text{otherwise.} \end{cases}$$
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Let $p_i(t)$ denote the probability associated with the vertex *i* at time *t*. The continuous-time random walk on *G* is defined as the solution of the differential equation

$$\frac{\mathrm{d}}{\mathrm{d}t}p_i(t) = \sum_{j \in V} L_{jk} p_j(t).$$
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Observe that

$$\frac{\mathrm{d}}{\mathrm{d}t}\sum_{j\in V}p_j(t) = \sum_{j,k\in V}L_{jk}p_k(t) = 0$$
(1.24)

You have to prove it ;)

An initially normalized distribution remains normalized; The solution of the differential equation can be given in closed form as:

$$p(t) = e^{Lt} p(0).$$
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$$i \frac{\mathrm{d}}{\mathrm{d}t} |\psi\rangle = H |\psi\rangle,$$
 (1.26)

Instead of probabilities of Eq. (1.24) we can insert the amplitudes $q_j(t) = \langle j | \psi(t) \rangle$ where $\{|j\rangle : j \in V\}$ is an orthonormal basis \mathcal{H} . Then, we obtain the equation:

$$\imath \frac{\mathrm{d}}{\mathrm{d}t} q_j(t) = \sum_{k \in V} L_{jk} q_k(t), \qquad (1.27)$$

where the Hamiltonian is given by the Laplacian L.

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 (1.26)

Instead of probabilities of Eq. (1.24) we can insert the amplitudes $q_j(t) = \langle j \mid \psi(t) \rangle$ where $\{ \mid j \rangle : j \in V \}$ is an orthonormal basis \mathcal{H} . Then, we obtain the equation:

$$i\frac{\mathrm{d}}{\mathrm{d}t}q_j(t) = \sum_{k\in V} L_{jk}q_k(t), \qquad (1.27)$$

where the Hamiltonian is given by the Laplacian L.

The solution of reads:

$$U(t) = e^{-iHt} = e^{-iLt},$$
 (1.28)

and the evolution of an initial state from t = 0 to some arbitrary time t is given by:

$$|\psi(t)\rangle = U(t)|\psi(0)\rangle. \tag{1.29}$$

This is an algorithm that aims to find hidden nonlinear structures over Galois fields \mathbb{F}_p , for p prime.

Galois fields over primes are also called prime fields. For each prime number p, the prime field \mathbb{F}_p of order p is constructed as the integers modulo p, that is $\mathbb{Z}/p\mathbb{Z}$.

Hidden Flat Problem



Figure: Equidistant circles of various radii over \mathbb{F}_q^2 that lie on an unknown flat H on which the radii sit at. Note that the density of points in each sphere is approximately the same since they live on a Galois field.

Let $\mathbb{S}_r^t(\mathbb{F}_q^d)$ denote the sphere of radius r with center t over \mathbb{F}_q^d . Additionally, for a finite set S, we denote by

$$|S\rangle \coloneqq rac{1}{\sqrt{|S|}} \sum_{s \in S} |s
angle$$
 (1.30)

the normalized uniform superposition over elements of S.

Hidden Flat Problem

Using f_1, f_{-1} – let us assume they exist indeed – it is possible to construct the state

$$\rho_r := \frac{1}{q^d} \sum_{t \in \mathbb{F}_q} |\mathbb{S}_r + t\rangle \, \langle \mathbb{S}_r + t| \tag{1.31}$$

t = 0 implied. The flat we are looking for is such a discrete set $H \subseteq \mathbb{F}_q$ allowing us to construct

$$\rho_1 := \frac{1}{|H|} \sum_{h \in H} |\mathbb{S}_1 + h\rangle \langle \mathbb{S}_1 + h|$$
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Goal: determine H by making measurements on this state.

How? By making measurements on this state. Implement a quantum walk is that moves the amplitude from $|S_1 + h\rangle$ to $|h\rangle$.

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If a sufficiently large fraction of the amplitude is moved, then the hidden flat can be determined by (classically) solving a noisy linear algebra problem.

To move amplitude from unit spheres to their centers, we will use a continuous-time quantum walk on the Winnie-Li graph.



Figure: A Winnie-Lie graph over \mathbb{F}_p^2 centered at x = 0. The edges are not shown.

This graph has vertex set \mathbb{F}_q^d , and edges between points $x, x' \in \mathbb{F}_q^d$ with $\Delta(x - x') = 1$.

Thus, its adjacency matrix (that serves as a Hamiltonian) is

$$A := \sum_{x \in \mathbb{F}_q^d} \sum_{s \in \mathbb{S}_1} |x + s\rangle \langle x|$$
(1.33)

Continuous-time quantum walk for time *t*:

$$U(t) = e^{-\imath A t}$$

This unitary operator can be efficiently implemented on a quantum computer provided that we can efficiently transform into the eigenbasis of *A*, and can efficiently compute the eigenvalue corresponding to a given eigenvector.

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Eigenvectors of (1.33):

$$|\tilde{k}
angle := rac{1}{\sqrt{q^d}} \sum_{x \in \mathbb{F}_q^d} \omega_p^{k \cdot x} |x
angle,$$
 (1.34)

for $k \in \mathbb{F}_q^d$. By Fourier transform of

$$U := \frac{1}{\sqrt{q^d}} \sum_{x,k \in \mathbb{F}_q^d} \omega_p^{k \cdot x} |k\rangle \langle x|$$
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transform to the eigenbasis of A where the corresponding eigenvalues are given by the Fourier transform of a unit sphere λ_k . Almost all of these eigenvalues can be computed with complexity $\mathcal{O}\left(\sqrt{q^{d-1}}\right)$.

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• Require ρ_H

• for $t = 1/\sqrt{q^{d-1}\log q}$: Perform a continuous-time quantum walk with

$$U = e^{-\imath At}$$

Measure in the computational basis

Each point in H occurs with probability $|H|^{-1} \left(1/\log q + \mathcal{O}\left(1/\log^{3/2}q\right)\right)$, and any point not on H occurs with probability $\mathcal{O}\left(q^{-d}\right)$.

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- We will show this using the concept of the "universal computation graph".
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Consider a (continuous) walker on \mathbb{Z} where the basis states are $|x\rangle$.

The eigenstates of the adjacency matrix are the (normalized) momentum states $|k\rangle$, that is, the states that satisfy

$$\langle x|k\rangle = e^{-\imath kx},\tag{1.36}$$

with $\langle k|k'\rangle \sim \delta(k-k')$. Intuition: $|k\rangle$ are the momentum eigenstates which are used to understand how scattering (particle interactions) works in quantum mechanics (and quantum field theory).

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In momentum space, with orthogonal states $|\phi_k
angle\equiv|k
angle$, we know that

$$|k\rangle = \sum_{x \in \mathbb{Z}} e^{-\imath k x} |x\rangle.$$
 (1.37)

These are also referred to as *momentum states* however, they are not normalizable (think as maps $E(G) \to \mathbb{C}$).

Using the adjacency matrix as the Hamiltonian H, it follows that

$$H|k\rangle = 2\cos(k)|k\rangle.$$
 (1.38)

Next, let us consider a finite graph G and create out of it an infinite graph with adjacency matrix H by attaching semi-infinite lines to M of its vertices.



Figure: Universal computation graph.

Last minute addition



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The states living on the *j*-th line are labeled as $|x, j\rangle$ where $|0, j\rangle$ corresponds to the state in *G* and where x is allowed to walk along the *j*-th line. The adjacency matrix of this graph is denoted by *H* and each of its eigenstates must be a superposition of the form of Eq. (1.37) with momenta k taking any of the values:

- $\pm k$ with eigenvalues $2\cos(k)$,
- $k = \pm i\kappa$ and eigenvalue $2\cosh(\kappa)$,
- k = ±iκ + π and eigenvalue -2 cosh(κ). Here κ ∈ ℝ_{≥0}. We can truncate |k⟩ such that it has support over a finite number of vertices. Denote the truncated state supported over L vertices as

$$|k\rangle_{L} \coloneqq \frac{1}{\sqrt{L}} \sum_{x=1}^{L} e^{-\imath kx} |x\rangle.$$
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In the physics literature, such states are called *wave packets* and the sign of the exponential denotes the direction of the wave; see Fig. 1.14.



Figure: A wave packet supported over 2 vertices moving coming from the (far) left.

The infinite line in Fig. 1.14 becomes a universal computation graph by inserting a finite graph G.

Frame Title

As seen in Fig. 1.15. In principle, one can prepare a wave packet as the one with momentum k and let it propagate.



Figure: Inserting a finite graph G into the integer line, yields a one-dimensional universal computation graph.

This amounts to a dynamic scattering process. Let us denote this incoming (to G) wave packet as

$$|w(k)\rangle_{\rm L}$$
 if the wave packet comes from the left, (1.40)
 $|w(k)\rangle_{\rm R}$ if the wave packet comes from the right. (1.41)

The dynamics correspond to the following equations:

$$\langle x_{\rm L}|w_{\rm L}(k)\rangle = e^{-\imath kx} + R_{\rm L}(k)e^{\imath kx}$$
(1.42)

$$\langle x_{\rm R} | w_{\rm L}(k) \rangle = T_{\rm L}(k) e^{ikx}$$
(1.43)

$$H|w(k)\rangle = 2\cos(k)|w(k)\rangle,$$
 (1.44)

where R_L is a reflection coefficient and T_L is the transfer coefficient. Similarly, we can write down the equations for right-coming wave packets.



Figure: Part of the wave packet will be reflected and part will be transferred through *G*. The coefficients $R_{L,R}$, $T_{L,R}$ are called reflection and transfer coefficients.

For every scattering process, as the one above, there is a scattering matrix S. In this case,

$$S = \begin{pmatrix} R_{
m L} & T_{
m L} \\ R_{
m R} & T_{
m R} \end{pmatrix},$$
 (1.45)

and it is an element of U(2).

More generally, an arbitrary number of semi-infinite lines can be considered as in Fig. 1.13 with an arbitrary graph G. If there are N semi-infinite lines, then $S \in U(N)$.

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More generally, an arbitrary number of semi-infinite lines can be considered as in Fig. 1.13 with an arbitrary graph G. If there are N semi-infinite lines, then $S \in U(N)$.

It is possible to encode a qubit state by considering two universal computation diagrams in one dimension:



Figure: A single qubit can be represented by two infinite lines. Crucially the momentum must be equal to $\pi/4$. The qubit is in the $|0\rangle$ state if the wavepacket propagates in the top line and in the $|1\rangle$ state if at the bottom.

As before, we can insert a graph G with 4 semi-infinite lines as in Fig. 1.18.

Figure: A two-qubit unitary U can be encoded through G to be implemented as a quantum walk.

A unitary is implemented by inserting a graph G such that its corresponding S-matrix has the structure

$$S = \begin{pmatrix} 0 & U^{\dagger} \\ U & 0 \end{pmatrix},$$
 (1.46)

where $U \in U(2)$. Therefore, a unitary U is implemented by the scattering process of quantum walkers, through a graph G that encodes it.
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Childs showed it is possible to implement the unitaries

$$U_{\pi/4} = \begin{pmatrix} e^{-\imath \pi/4} & 0\\ 0 & 1 \end{pmatrix}, \quad U_{\rm b} = -\frac{\imath}{\sqrt{2}} \begin{pmatrix} 1 & -\imath\\ -\imath & 1 \end{pmatrix}, \quad (1.47)$$

which form a universal gate set for one-qubit operations; up to a certain precision ε , any single-qubit gate can be implemented by a string of these two unitaries.



This construction was further generalized to *n*-qubit gates proving that quantum walks form a universal model of computation.



Figure: The graph G obtained by attaching N semi-infinite paths to a graph G.

By considering a finite graph G and attaching N/2 = n pairs of semi-infinite paths, we are able to encode n qubits. Eventually, it is possible to encode any *n*-qubit unitary to a graph G to obtain a quantum walk equivalent of any arbitrary circuit.



Figure: If G is chosen to encode a desired unitary $U \in U(n)$ the circuit can be implemented by a quantum walk.



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- Generalizes Grover's algorithm
- Candidate to replace Monte Carlo sampling technique (it runs quadratically faster)

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Classical Monte Carlo

Consider a one-dimensional random variable X and a function $f : \mathbb{R} \to [0, 1]$.

Assume that the mean $\mu = \mathbb{E}[f(X)] < \infty$ and the standard deviation $\sigma^2 = \mathbb{V}[f(X)] < \infty$ are well defined.

Central Limit Theorem ensures that for i.i.d. random variables (X_1, \ldots, X_N) , following the same distribution as X, for $N \to \infty$, the quantity $\sqrt{N} \frac{\hat{\mu} - \mu}{\sigma}$ converges to a mean-zero Gaussian with unit variance $\mathcal{N}(0, 1)$.

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This implies that for any $\varepsilon > 0$ we estimate that

$$\lim_{N \to \infty} \mathbb{P}\left(|\hat{\mu} - \mu| \le \varepsilon \right) = \lim_{N \to \infty} \mathbb{P}\left(|\mathcal{N}(0, 1)| \le \frac{\varepsilon \sqrt{N}}{\sigma} \right).$$
(2.1)

In turn, this implies that for any z > 0 and $\delta \in (0, 1)$, in order to obtain an estimate of the form $\mathbb{P}(|\hat{\mu} - \mu| \leq \varepsilon)$, $N = \mathcal{O}(1/\varepsilon^2)$ samples are required.

Consider a unitary operator \mathcal{A} that acts on an *n*-qubit register as follows:

$$\mathcal{A}|0\rangle^{\otimes n} = \sum_{x \in \{0,1\}^k} a_x |\psi_x\rangle |x\rangle, \tag{2.2}$$

for k < n, where $|\psi_x\rangle$ is a quantum state consisting of n - k qubits and $|x\rangle$ is a state consisting of k qubits.

Intuition: We are interested in A because it allows us to encodes a distribution of interest, with encoded data in the states $|x\rangle$, for which we want to estimate certain properties (e.g. mean or other moments).

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States $\{|\psi\rangle_x\}_{x\in\{0,1\}^k}$ are assumed to be orthogonal. Next, assume that there is a unitary W acting as follows:

$$\mathcal{W}|x\rangle|0\rangle = |x\rangle \Big(\sqrt{1-f(x)}|0\rangle + \sqrt{f(x)}|1\rangle\Big).$$
(2.3)

Intuition: This creates a quantum state that encodes the function f(x) of interest. The quantum state t creates captures the information about the properties pf f(x) in the amplitudes of the ancilla qubits $|0\rangle$ and $|1\rangle$.

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Something interesting happens when combining:

$$\mathcal{G} \coloneqq (\mathbf{1}_{n-k} \otimes \mathcal{W})(\mathcal{A} \otimes \mathbf{1}_k). \tag{2.4}$$

Applying \mathcal{G} to a $|0\rangle^{\otimes (n+1)}$ qubit register yields the following state:

$$\begin{aligned} |\psi\rangle &= \mathcal{G}|0\rangle^{\otimes (n+1)} \\ &= \sum_{x \in \{0,1\}^k} a_x |\psi_x\rangle |x\rangle \Big(\sqrt{1-f(x)}|0\rangle + \sqrt{f(x)}|1\rangle\Big), \end{aligned} \tag{2.5}$$

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(2.5)
= $\sum_{x \in \{0,1\}^k} a_x |\psi_x\rangle |x\rangle \Big(\sqrt{1-f(x)}|0\rangle + \sqrt{f(x)}|1\rangle\Big),$ (2.6)

It is customary to refer to these two states as the "bad state":

$$|\psi_{\mathrm{bad}}\rangle \coloneqq \sum_{x \in \{0,1\}^k} a_x \sqrt{1 - f(x)} |\psi_x\rangle |x\rangle,$$
 (2.7)

and the "good state":

$$|\psi_{\text{good}}\rangle \coloneqq \sum_{x \in \{0,1\}^k} a_x \sqrt{f(x)} |\psi_x\rangle |x\rangle.$$
 (2.8)

By considering the projection operator

$$\mathcal{P} \coloneqq \mathbf{1}_n \otimes |1\rangle \langle 1|, \tag{2.9}$$

we can measure the probability that the last state is the |1
angle state,

$$\langle \psi | \mathcal{P}^{\dagger} \mathcal{P} | \psi \rangle = \dots$$
 (2.10)

$$= |\psi_{\text{good}}|^2. \tag{2.11}$$

From the definition of a good state, we can further see that

$$|\psi_{\text{good}}|^2 = \sum_{x \in \{0,1\}^k} |a_x|^2 f(x),$$
 (2.12)

which corresponds, precisely, to the mean $\mu = \mathbb{E}(f(X))$.

Estimating μ for a distribution f amounts to running the circuit \mathcal{G} , measuring the output and determining the probability of observing the state $|1\rangle$.

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Quadratic Speedup

Concretely, assume access to a unitary

$$U|0\rangle = \sqrt{1-\mu}|\psi_{\rm bad}\rangle + \sqrt{\mu}|\psi_{\rm good}\rangle.$$
(2.13)

Then, for any $N\in\mathbb{Z}_{\geq0}$, the QAE algorithm outputs the estimate $\hat{\mu}$ such that

$$|\hat{\mu} - \mu| \le 2\pi \frac{\sqrt{\mu(1-\mu)}}{N} + \frac{\pi^2}{N^2},$$
(2.14)

with probability at least $8/\pi^2$ by quering the algorithm exactly N times.

Image: A match a ma

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By using the so-called "Powering Lemma" which states (approximately) that for any $\delta \in (0, 1)$, it is sufficient to iterate with U approximately $\mathcal{O}(\log(1/\delta))$ times to obtain

$$\mathbb{P}(|\hat{\mu} - \mu| \le \varepsilon) \ge 1 - \delta. \tag{2.15}$$

Quadratic Speedup

It is required to iterate \mathcal{G} approximately $\mathcal{O}(N \log(1/\delta))$ times to obtain the guarantee of Eq. (2.15), where

$$\varepsilon = 2\pi \frac{\sqrt{\mu(1-\mu)}}{N}.$$
(2.16)

That is, for fixed δ the computational cost to obtain (2.15) is

$$\mathcal{O}(1/\varepsilon)$$
 (2.17)

which is quadratically better than the $N=\mathcal{O}(1/arepsilon^2)$ samples required by classical Monte Carlo.

Quadratic Speedup

It is required to iterate \mathcal{G} approximately $\mathcal{O}(N \log(1/\delta))$ times to obtain the guarantee of Eq. (2.15), where

$$\varepsilon = 2\pi \frac{\sqrt{\mu(1-\mu)}}{N}.$$
(2.16)

That is, for fixed δ the computational cost to obtain (2.15) is

$$\mathcal{O}(1/\varepsilon)$$
 (2.17)

which is quadratically better than the $N = O(1/\varepsilon^2)$ samples required by classical Monte Carlo.