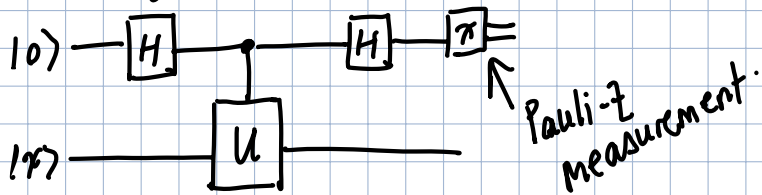


① Recap: Hadamard Test.

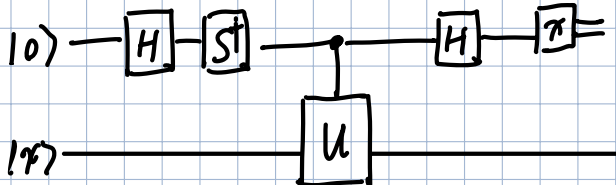
Estimating $\langle \psi | U | \psi \rangle$, where U is a unitary and $|\psi\rangle$ is a state vector.



$$\Pr[0] = \frac{1}{2} (1 + \text{Re}(\langle \psi | U | \psi \rangle)), \quad \Pr[1] = \frac{1}{2} (1 - \text{Re}(\langle \psi | U | \psi \rangle)).$$

$$S = \begin{pmatrix} 1 & 0 \\ 0 & i \end{pmatrix} \Rightarrow S^\dagger = \begin{pmatrix} 1 & 0 \\ 0 & -i \end{pmatrix}$$

For the imaginary part:



$$\Pr[0] = \frac{1}{2} (1 + \text{Im}(\langle \psi | U | \psi \rangle))$$

$$\Pr[1] = \frac{1}{2} (1 - \text{Im}(\langle \psi | U | \psi \rangle)).$$

⊛ In both cases, we have $\Pr[0] = \frac{1}{2} (1 + \alpha)$, $\Pr[1] = \frac{1}{2} (1 - \alpha)$, where α is unknown. To estimate α , we do the following procedure:

For $i = 1, 2, \dots, N$ ($N \equiv \#$ of samples / $\#$ of times we run the quantum circuit).

- Each time we get outcome "0" \rightarrow record $x_i = 1$
 - Each time we get outcome "1" \rightarrow record $x_i = -1$
- ⊛ This defines a random variable X :

- Do this N times, then take the sample mean/average:

$$\hat{x}_N = \frac{1}{N} \sum_{i=1}^N x_i$$

Sample mean

$$\Pr[X = \pm 1] = \frac{1}{2} (1 \pm \alpha)$$

This defines a random variable $\hat{X}_N = \frac{1}{N} \sum_{i=1}^N X_i$. \hat{X}_N is an unbiased estimator of X :

$$\mathbb{E}[\hat{X}_N] = \frac{1}{N} \sum_{i=1}^N \mathbb{E}[X_i] = \frac{1}{N} \sum_{i=1}^N \mathbb{E}[X] = \mathbb{E}[X].$$

True mean

$= \mathbb{E}[X] \forall i$, b/c all samples are independent and identical.

$$\mathbb{E}[X] = (+1) \cdot \Pr[X=+1] + (-1) \cdot \Pr[X=-1] = \frac{1}{2}(1+d) - \frac{1}{2}(1-d) = d.$$

• As $N \rightarrow \infty$, $\hat{X}_N \rightarrow \mathbb{E}[X] = d$ (law of large numbers).

⊛ So the sample average approaches the true (unknown) value of d !

② Motivation: Expectation values

• Why do we care about quantities of the form $\langle \psi | U | \psi \rangle$?

B/c they can be used to compute expectation values of observables.

• Observable: Any Hermitian operator (recall: M Hermitian $\Leftrightarrow M^\dagger = M$)

⊛ Fact from linear algebra: Every Hermitian operator can be diagonalized.

$$M = \sum_{i=1}^d \lambda_i |v_i\rangle\langle v_i|, \text{ where: } \lambda_i: \text{eigenvalues (real numbers). } z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

$= |0\rangle\langle 0| - |1\rangle\langle 1|$

$|v_i\rangle$: orthonormal eigenvectors.

$\sum_{i=1}^d |v_i\rangle\langle v_i| = \mathbb{1}$

⊛ Axiom of quantum mechanics: every physical quantity corresponds to a Hermitian operator. (e.g., position, momentum, energy...).

The eigenvalues are the possible values of the quantity

We get the value of the quantity through measurement.

• For an observable/Hermitian operator M , the corresponding measurement is given by the eigenvectors. ($\{|v_i\rangle\langle v_i|\}$ is the POVM.)

• For a state ρ , the probability that it has value λ_i is

$$P(i) = \text{Tr}[|v_i\rangle\langle v_i| \rho] \quad (\text{Born rule}) \rightarrow = \langle v_i | \rho | v_i \rangle$$

The expected value of the observable is:

$$\langle M \rangle_\rho = \sum_{i=1}^n \lambda_i P(i) = \sum_{i=1}^n \lambda_i \text{Tr}[|v_i\rangle\langle v_i| \rho] = \text{Tr}\left[\underbrace{\left(\sum_{i=1}^n \lambda_i |v_i\rangle\langle v_i|\right)}_M \rho\right] = \text{Tr}[M\rho].$$

Definition
of expectation value!

- If we don't know the eigenvectors (b/c they are hard to get for large matrices!), but we know that M can be written as

$$M = \sum_{j=1}^k c_j U_j, \quad \text{then} \quad \langle M \rangle_\rho = \text{Tr}[M(\gamma X \gamma)]$$

$$= \sum_{j=1}^k c_j \text{Tr}[U_j(\gamma X \gamma)]$$

$$= \sum_{j=1}^k c_j \langle \gamma | U_j | \gamma \rangle$$

Estimate using
Hadamard test!

So we estimate each term individually using the Hadamard test and then add to get an overall estimate of the expectation value.

③ Quantum Fourier Transform

(a) Recap: discrete Fourier transform (DFT).

- Used in a lot of places!

- Signal processing — identifying frequency components of a signal. (audio and video processing)

- Machine learning — feature extraction

- Scientific Computing — solving differential equations numerically.

- Definition: for a signal $\{x_k\}_{k=0}^{N-1}$, its DFT is

$$y_k = \frac{1}{\sqrt{N}} \sum_{k'=0}^{N-1} x_{k'} e^{\frac{2\pi i k' k}{N}} \quad \left(e^{2\pi i x} = \cos(2\pi x) + i \sin(2\pi x) \right)$$

- Example: consider a signal of two sine waves put together:

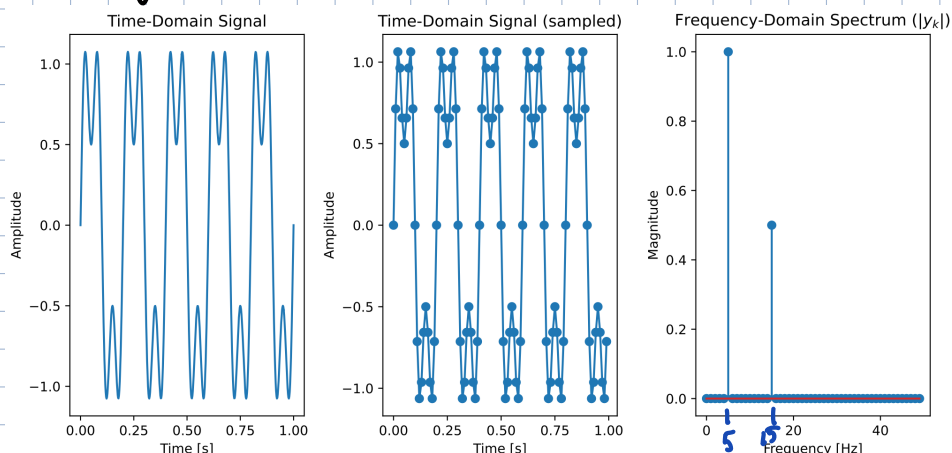
$$x(t) = \sin(2\pi f_1 t) + \frac{1}{2} \sin(2\pi f_2 t)$$

↑ ↑
frequencies

- $f_1 = 5 \text{ Hz}$

- $f_2 = 15 \text{ Hz}$

- Sampling rate: $f_s = 100 \text{ Hz} \Rightarrow 100 \text{ samples total (in 1 sec.)}$



(b) The Quantum Version! (Quantum Fourier Transform \equiv QFT).

- $y_k = \frac{1}{\sqrt{N}} \sum_{k'=0}^{N-1} x_{k'} e^{\frac{2\pi i k' k}{N}} \rightarrow Q|k\rangle = \frac{1}{\sqrt{d}} \sum_{k'=0}^{d-1} e^{\frac{2\pi i k k'}{d}} |k'\rangle$ (action on basis vectors).

QFT matrix

$|x\rangle = \sum_{k=0}^{d-1} x_k |k\rangle \Rightarrow Q|x\rangle = \sum_{k=0}^{d-1} x_k Q|k\rangle = \sum_{k=0}^{d-1} x_k \frac{1}{\sqrt{d}} \sum_{k'=0}^{d-1} e^{\frac{2\pi i k k'}{d}} |k'\rangle$

$= \sum_{k'=0}^{d-1} \left(\frac{1}{\sqrt{d}} \sum_{k=0}^{d-1} x_k e^{\frac{2\pi i k k'}{d}} \right) |k'\rangle = |y\rangle$

$\equiv y_{k'}$

- The QFT matrix is unitary! Let $\omega = e^{\frac{2\pi i}{d}}$, $\bar{\omega} = e^{-\frac{2\pi i}{d}} = \omega^{-1}$

$Q = \frac{1}{\sqrt{d}} \sum_{k, k'=0}^{d-1} \omega^{k k'} |k\rangle\langle k'|$

$\cos\left(\frac{2\pi}{d}\right) + i \sin\left(\frac{2\pi}{d}\right)$

$\omega = e^{\frac{2\pi i}{d}} = \cos\left(\frac{2\pi}{d}\right) + i \sin\left(\frac{2\pi}{d}\right)$

$\bar{\omega} = \cos\left(\frac{2\pi}{d}\right) - i \sin\left(\frac{2\pi}{d}\right)$

$= e^{-\frac{2\pi i}{d}}$

$= \omega^{-1}$

$Q Q^\dagger = \left(\frac{1}{\sqrt{d}} \sum_{k, k'=0}^{d-1} \omega^{k k'} |k\rangle\langle k'| \right) \left(\frac{1}{\sqrt{d}} \sum_{j, j'=0}^{d-1} \bar{\omega}^{j j'} |j'\rangle\langle j| \right)$

$= \frac{1}{d} \sum_{k, k'=0}^{d-1} \sum_{j, j'=0}^{d-1} \omega^{k k'} \omega^{-j j'} |k\rangle\langle k'| |j'\rangle\langle j|$

$= \delta_{j', k'}$

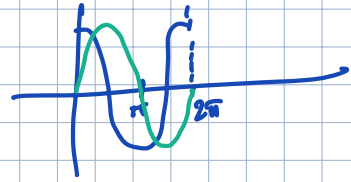
$= \frac{1}{d} \sum_{k, k'=0}^{d-1} \sum_{j=0}^{d-1} \omega^{k k'} \omega^{-j k'} |k\rangle\langle j|$

$= \sum_{k, j=0}^{d-1} \left(\frac{1}{d} \sum_{k'=0}^{d-1} \omega^{k'(k-j)} \right) |k\rangle\langle j|$

$$\rightarrow \delta_{j,k}$$

$$= \frac{1}{d} \sum_{k'=0}^{d-1} e^{\frac{2\pi i}{d} k'(k-j)}$$

$\underbrace{k'(k-j)}_{\equiv x}$



⊛ If $x=0 \rightarrow \frac{1}{d} \sum_{k'=0}^{d-1} (1) = 1$

⊛ If $x \neq 0 \rightarrow \frac{1}{d} \sum_{k'=0}^{d-1} \omega^{k'x}$ $\rightarrow (\omega^x)^{k'}$

$(\omega^x)^d = e^{\frac{2\pi i x}{d} \cdot d} = e^{2\pi i x} = \underbrace{\cos(2\pi x)}_{=1 \forall x} + i \underbrace{\sin(2\pi x)}_{=0 \forall x}$

$\omega = e^{\frac{2\pi i}{d}}$ $\omega^x = e^{\frac{2\pi i x}{d}}$ $\omega^{xd} = e^{2\pi i x} = 1$

$\rightarrow = \frac{1}{d} \sum_{k'=0}^{d-1} (\omega^x)^{k'} = \frac{1}{d} \left(\frac{1 - (\omega^x)^d}{1 - \omega^x} \right) = 0.$

Finite geometric

Series!

$$\sum_{k=0}^{d-1} r^k = \frac{1-r^d}{1-r}$$

⊛ So $\frac{1}{d} \sum_{k'=0}^{d-1} \omega^{k'(k-j)} = \delta_{j-k,0} = \delta_{jk}$

$$\Rightarrow Q^T = \sum_{k_{ij}=0}^{d-1} \underbrace{\left(\frac{1}{d} \sum_{k'=0}^{d-1} \omega^{k'(k-j)} \right)}_{\delta_{k_{ij}}} |kX_j\rangle = \mathbb{1} \checkmark$$

Similarly, $Q^T Q = \mathbb{1}.$

$$Q = \frac{1}{\sqrt{d}} \begin{pmatrix} | & | & | & \dots & | \\ | & \omega & \omega^2 & \dots & \omega^{d-1} \\ | & \omega^2 & \omega^4 & \dots & \omega^{2(d-1)} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ | & \omega^{d-1} & \omega^{2(d-1)} & \dots & \omega^{(d-1)^2} \end{pmatrix}$$

Note: for $d=2$, $\omega = e^{\frac{2\pi i}{2}} = e^{\pi i} = -1 \Rightarrow Q = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \rightarrow$ Hadamard!

- Let $d=2^n$, $\omega = e^{\frac{2\pi i}{2^n}} \rightarrow$ What is a circuit representation of QFT?

Use the binary representation of $0, 1, 2, \dots, 2^n - 1$

e.g., $n=3$:

0	→	000	}	→	$k \rightarrow (k_1, k_2, k_3)$
1	→	001			
2	→	010			
3	→	011			
4	→	100			
5	→	101			
6	→	110			
7	→	111			

$e \in \{0, 1, 2, \dots, 7\}$

$k = k_1 \cdot 4 + k_2 \cdot 2 + k_3 \cdot 1$

⊛ In general: $k = \sum_{l=1}^n 2^{n-l} k_l$

$$Q|j\rangle = \frac{1}{\sqrt{2^n}} \sum_{k=0}^{2^n-1} e^{\frac{2\pi i j k}{2^n}} |k\rangle$$

$$\Rightarrow Q|j_1, j_2, \dots, j_n\rangle = \frac{1}{\sqrt{2^n}} \sum_{k_1, k_2, \dots, k_n \in \{0, 1\}} e^{\frac{2\pi i}{2^n} j (k_1 2^{n-1} + k_2 2^{n-2} + \dots + k_n)} |k_1, k_2, \dots, k_n\rangle$$

$$= \frac{1}{\sqrt{2^n}} \sum_{k_1 \in \{0, 1\}} \sum_{k_2 \in \{0, 1\}} \dots \sum_{k_n \in \{0, 1\}} e^{\frac{2\pi i}{2^n} j k_1 2^{n-1}} e^{\frac{2\pi i}{2^n} j k_2 2^{n-2}} \dots e^{\frac{2\pi i}{2^n} j k_n} |k_1, k_2, \dots, k_n\rangle$$

$$= \left(\frac{1}{\sqrt{2}} \sum_{k_1 \in \{0, 1\}} e^{\frac{2\pi i}{2^n} j k_1 2^{n-1}} |k_1\rangle \right) \left(\frac{1}{\sqrt{2}} \sum_{k_2 \in \{0, 1\}} e^{\frac{2\pi i}{2^n} j k_2 2^{n-2}} |k_2\rangle \right) \dots \left(\frac{1}{\sqrt{2}} \sum_{k_n \in \{0, 1\}} e^{\frac{2\pi i}{2^n} j k_n} |k_n\rangle \right)$$

$$= \left(\frac{1}{\sqrt{2}} \sum_{k_1 \in \{0, 1\}} e^{2\pi i k_1 \frac{j}{2}} |k_1\rangle \right) \left(\frac{1}{\sqrt{2}} \sum_{k_2 \in \{0, 1\}} e^{2\pi i k_2 \frac{j}{2^2}} |k_2\rangle \right) \dots \left(\frac{1}{\sqrt{2}} \sum_{k_n \in \{0, 1\}} e^{2\pi i k_n \frac{j}{2^n}} |k_n\rangle \right)$$

Consider $n=3$: $|k_1, j_1, j_2, j_3\rangle = \left(\frac{1}{\sqrt{2}} \sum_{k_1 \in \{0,1\}} e^{2\pi i k_1 \frac{j_1}{2}} |k_1\rangle \right) \left(\frac{1}{\sqrt{2}} \sum_{k_2 \in \{0,1\}} e^{2\pi i k_2 \frac{j_2}{2^2}} |k_2\rangle \right) \left(\frac{1}{\sqrt{2}} \sum_{k_3 \in \{0,1\}} e^{2\pi i k_3 \frac{j_3}{2^3}} |k_3\rangle \right)$

$$j = 4j_1 + 2j_2 + j_3 \Rightarrow \frac{j}{2} = 2j_1 + j_2 + \frac{j_3}{2}$$

$$\frac{j}{2^2} = \frac{j}{4} = j_1 + \frac{j_2}{2} + \frac{j_3}{4}$$

$$\frac{j}{2^3} = \frac{j}{8} = \frac{j_1}{2} + \frac{j_2}{4} + \frac{j_3}{8}$$

$$\Rightarrow e^{2\pi i k_1 \frac{j}{2}} = \underbrace{e^{2\pi i k_1 (2j_1)}}_{=1} \underbrace{e^{2\pi i k_1 j_2}}_{=1} e^{2\pi i k_1 \frac{j_3}{2}} = e^{\pi i k_1 j_3} = (-1)^{k_1 j_3}$$

$$e^{2\pi i k_2 \frac{j}{4}} = \underbrace{e^{2\pi i k_2 j_1}}_{=1} \underbrace{e^{2\pi i k_2 (\frac{j_2}{2})}}_{=(-1)^{k_2 j_2}} \underbrace{e^{2\pi i k_2 (\frac{j_3}{4})}}_{= e^{\frac{2\pi i}{4} k_2 j_3}}$$

$$e^{2\pi i k_3 \frac{j}{8}} = \underbrace{e^{\frac{2\pi i}{2} k_3 j_1}}_{=(-1)^{k_3 j_1}} e^{\frac{2\pi i}{4} k_3 j_2} e^{\frac{2\pi i}{8} k_3 j_3}$$

Therefore: $|k_1, j_1, j_2, j_3\rangle = \left(\frac{1}{\sqrt{2}} \sum_{k_1 \in \{0,1\}} (-1)^{k_1 j_3} |k_1\rangle \right) \left(\frac{1}{\sqrt{2}} \sum_{k_2 \in \{0,1\}} (-1)^{k_2 j_2} e^{\frac{2\pi i}{4} k_2 j_3} |k_2\rangle \right) \left(\frac{1}{\sqrt{2}} \sum_{k_3 \in \{0,1\}} (-1)^{k_3 j_1} e^{\frac{2\pi i}{4} k_3 j_2} e^{\frac{2\pi i}{8} k_3 j_3} |k_3\rangle \right)$

$$= \frac{1}{\sqrt{2^3}} \sum_{k_1, k_2, k_3 \in \{0,1\}} (-1)^{k_1 j_3} (-1)^{k_2 j_2} (-1)^{k_3 j_1} e^{\frac{2\pi i}{4} k_2 j_3} e^{\frac{2\pi i}{4} k_3 j_2} e^{\frac{2\pi i}{8} k_3 j_3} |k_1, k_2, k_3\rangle$$

$$= \tilde{U} |j_3, j_2, j_1\rangle$$

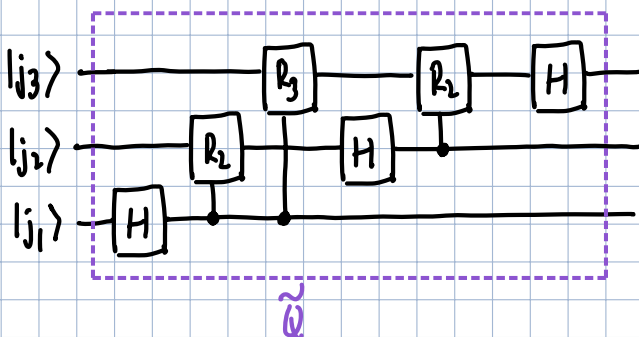
$$R_2(\theta) = e^{i\frac{\theta}{2} Z} = \begin{pmatrix} 1 & 0 \\ 0 & e^{i\theta} \end{pmatrix}$$

⊛ Define the rotation gate $R_k = \begin{pmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i}{2^k}} \end{pmatrix}$

$$R_2 = \begin{pmatrix} 1 & 0 \\ 0 & e^{\frac{2\pi i}{4}} \end{pmatrix}$$

$$\begin{pmatrix} 1 & 0 \\ 0 & i \end{pmatrix}$$

$$e^{\frac{\pi i}{2}} = i$$



$$\text{So } |j_1, j_2, j_3\rangle = \tilde{Q} |j_2, j_1, j_3\rangle = \tilde{Q} S |j_1, j_2, j_3\rangle$$

Permutation of the qubits!

⊛ In general, \tilde{Q} is given by the following circuit.

