

Lecture 12: Revealing XOR patterns II

(This is perhaps the most important lecture for the practice of Q.C. We'll see how Q.C. lets you)

find patterns in implicitly-represented data

today: XOR

(common story in data processing....)

Data vector
length N

$$|g\rangle \in \mathbb{C}^N$$

Fourier transform \rightarrow
based on N

"pattern vectors"

$$|x_0\rangle, |x_1\rangle, \dots, |x_{N-1}\rangle$$

Length- N vector,
 s^{th} entry is
"strength of
 $|x_s\rangle$ pattern
in the data"

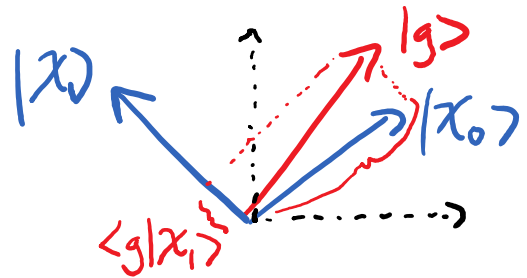
Classically: N is of "physical size" (e.g. 1000)
· vectors explicitly rep'd

Quantum: $N = 2^n$, n of "physical size" (e.g. N is 2^{1000})
· vecs. implicitly rep'd by n -qubit state.

Pattern vecs $|X_0\rangle, \dots, |X_{N-1}\rangle$ can be any orthonormal basis for \mathbb{C}^N .

"Strength of patterns in $|g\rangle$ ":
coeffs when $|g\rangle$ rep'd in $|X_s\rangle$ basis.

"strength" of $|X_s\rangle$:
 $\langle X_s | g \rangle$



(aka $\hat{g}(s)$) (← but won't dwell on this notation much)

Let $U \in \mathbb{C}^{N \times N}$ be matrix w/ cols. $|X_0\rangle, \dots, |X_{N-1}\rangle$.

U is unitary (like a "rotation") since \nearrow orthonormal

U : standard basis U^{-1} : $|X_s\rangle$ basis \rightarrow std basis

↓
 basis of $|X_s\rangle$

U^\dagger (← often same as U , or nearly same)

$$|g\rangle = \begin{bmatrix} d \\ a \\ t \\ a \end{bmatrix} \xrightarrow{U^\dagger, \text{Fourier transf.}} \begin{bmatrix} \hat{g}(0) \\ \hat{g}(1) \\ \vdots \\ \hat{g}(N-1) \end{bmatrix}, \quad \hat{g}(s) = \text{"strength of } |X_s\rangle \text{ pattern"}$$

(This is a very general setup, but we'll be focusing on just a couple of particular cases.)

Particular cases for patterns $|x_0\rangle, \dots, |x_{N-1}\rangle$:

- Want:
- ① "Interesting/useful"
 - ② Associated change of basis $U \in \mathbb{C}^{N \times N}$ easy to compute by quantum gates.

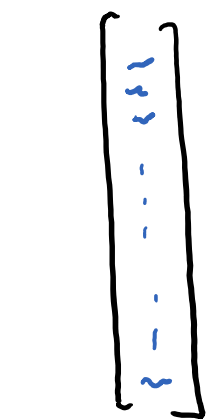
- Today:
- ① "XOR patterns"
 - ② $U = H^{\otimes n}$: just Hadamard on each qubit

- Later (Shor):
- ① discrete sines/cosines
 - ② U : classic DFT matrix; a little bit harder to quantumly compute.

XOR Functions?

"N" will be 2^n (the beauty of quantum)

Crucial mental perspective:



vector
in \mathbb{C}^N ,
 $N = 2^n$

x	f(x)
00...0	~
00...1	~
...	...
11...1	~

truth-table
of function
 $f: \{0,1\}^n \rightarrow \mathbb{C}$

\equiv
n-qubit.
quantum state
 $\sim |00\dots 0\rangle +$
 $\sim |00\dots 1\rangle +$
 \dots
 $+ \sim |11\dots 1\rangle$

(There's a notational hassle here, regarding normalization.)

Unit vec.

(Need to be a unit vec. Not so if Boolean-valued.)

Unit vec.

For $F: \{0,1\}^n \rightarrow \{0,1\}$,

$$\begin{array}{ccc}
 \begin{bmatrix} 0 \\ \vdots \\ 0 \\ \vdots \end{bmatrix} & \rightsquigarrow & \begin{bmatrix} +1 \\ -1 \\ -1 \\ +1 \\ \vdots \end{bmatrix} & \rightsquigarrow & \frac{1}{\sqrt{2}} \begin{bmatrix} +1 \\ -1 \\ -1 \\ +1 \\ \vdots \end{bmatrix} \\
 F & & f = (-1)^F & & \text{unit!} \dots
 \end{array}$$

def: (Non-standard.) If $g: \{0,1\}^n \rightarrow \mathbb{C}$,

$$|g\rangle \text{ denotes } \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} g(x) |x\rangle$$

rem: A quantum state/unit vec. iff

$$\frac{1}{N} \sum_x |g(x)|^2 = 1$$

avg $\{ |g(x)|^2 \}$. E.g., if $g(x) \in \{\pm 1\}$

E.g. if $g = f = (-1)^F$

for some $F: \{0,1\}^n \rightarrow \{0,1\}$

recall:

(we'll recall later)

$$\text{unif. superpos } \boxed{Q_F^\pm} |f\rangle$$

"Pattern vectors" $|\chi_s\rangle$ also thought of as functions $\{0,1\}^n \rightarrow \mathbb{C}$.

Today: $|\chi_s\rangle$ given by "XOR function"

$$\chi_s : \{0,1\}^n \rightarrow \{-1,+1\}$$

$$x \mapsto (-1)^{\text{XOR}_s(x)},$$

$$\text{where } s \in \{0,1\}^n, \text{ XOR}_s(x) = \sum_{i: s_i=1} x_i \pmod 2$$

(We'll remember why these are "orthonormal" later.)

Unusual / special property: these pattern vectors are Boolean-valued!

(This is the beauty & simplicity of the Hadamard / Boolean Fourier transform. Typically not like this.)

cf: "usual DFT": $\chi_s(x) = e^{\frac{2\pi i}{N} sx}$,
(for Shor)

$$s, x \in \{0, 1, 2, 3, \dots, N-1\}.$$

"Strength" of pattern $|\chi_s\rangle$ in $|g\rangle$?

(Recall, it's just the coefficient of $|g\rangle$ on $|\chi_s\rangle$ in the χ -basis....) " $\hat{g}(s)$ " = $\langle \chi_s | g \rangle$

U^{-1} transform maps $|g\rangle \mapsto \sum_{\substack{\text{pattern} \\ \text{indices} \\ s \in \{0,1\}^n}} \hat{g}(s) |s\rangle$.
↑ "strengths"

In function notation:

$$\hat{g}(s) = \langle \chi_s | g \rangle = \frac{1}{\sqrt{N}} [\chi_s(00\dots 0)^* \dots \chi_s(11\dots 1)^*] \cdot \frac{1}{\sqrt{N}} \begin{matrix} |g\rangle \\ g(00\dots 0) \\ \vdots \\ g(11\dots 1) \end{matrix}$$

$$= \frac{1}{N} \sum_{x \in \{0,1\}^n} \chi_s(x)^* g(x)$$

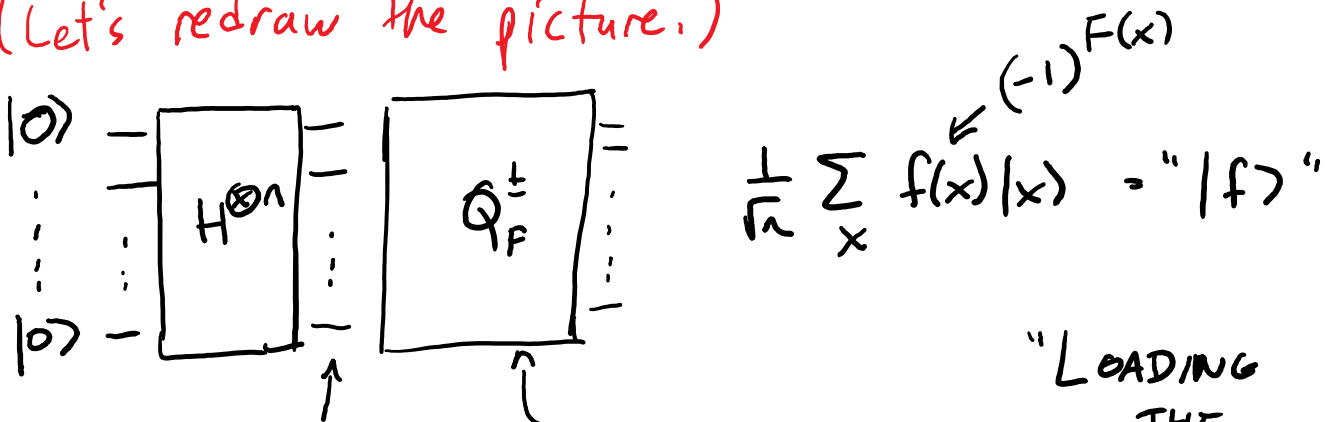
$$= \text{avg}_{x \in \{0,1\}^n} \{ \chi_s(x)^* g(x) \} = \text{"correlation of } \chi_s \text{ and } g \text{"}$$

If $g: \{0,1\}^n \rightarrow \{\pm 1\}$, and $\chi_s: \{0,1\}^n \rightarrow \{\pm 1\}$ (as in XOR patterns)

it's $\text{avg}_x \left\{ \begin{matrix} +1 & \text{if } g(x) = \chi_s(x) \\ -1 & \text{if } g(x) \neq \chi_s(x) \end{matrix} \right\}$

$$= \Pr_{x \in \{0,1\}^n} [g(x) = \chi_s(x)] - \Pr_x [g(x) \neq \chi_s(x)], \text{ in } [-1, +1]$$

(Let's redraw the picture.)



uniform superpos

$$\frac{1}{\sqrt{N}} \sum_x |x\rangle$$

sign-implementation of a classical circuit computing

$$F: \{0,1\}^n \rightarrow \{0,1\}$$

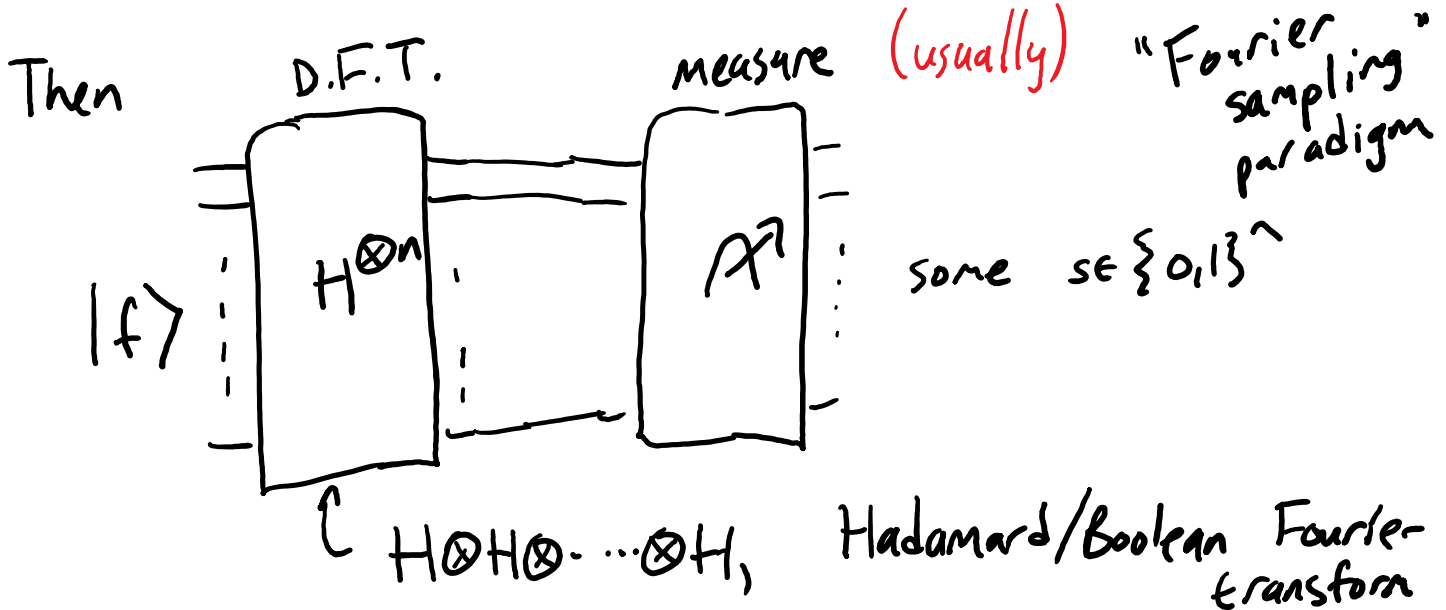
"LOADING THE DATA"

(Sort of a coincidence

that we get unif. superpos via $H^{\otimes n}$.)

(There are other ways to do it; particularly for $F: \{0,1\}^n \rightarrow \{0,1\}^m$. We'll discuss later.)

Just a convenient way to do it. Tho. it is $|x_{00\dots 0}\rangle$; i.e., |constantly +1 function>.)



$$H^{\otimes n} = H \otimes H \otimes \dots \otimes H$$



$n=2$ e.g.

$$\frac{1}{\sqrt{2}} \begin{bmatrix} +1 & +1 \\ +1 & -1 \end{bmatrix} \otimes \frac{1}{\sqrt{2}} \begin{bmatrix} +1 & +1 \\ +1 & -1 \end{bmatrix}$$

$N \times N$ unitary,
 $N = 2^n$

$$= \frac{1}{\sqrt{4}} \begin{matrix} |x\rangle \backslash |s\rangle \\ \begin{matrix} |00\rangle & |01\rangle & |10\rangle & |11\rangle \\ |00\rangle & +1 & +1 & +1 & +1 \\ |01\rangle & +1 & -1 & +1 & -1 \\ |10\rangle & +1 & +1 & -1 & -1 \\ |11\rangle & +1 & -1 & -1 & +1 \end{matrix} \end{matrix}$$

From last time:

$$H^{\otimes n} |s\rangle = \frac{1}{\sqrt{2}} (|0\rangle + (-1)^{s_1} |1\rangle) \otimes \frac{1}{\sqrt{2}} (|0\rangle + (-1)^{s_2} |1\rangle) \otimes \dots$$

$$= \frac{1}{\sqrt{N}} \sum_{x \in \{0,1\}^n} (-1)^{\text{XOR}_s(x)} |x\rangle$$

$$= |\chi_s\rangle, \quad \text{where } \chi_s: \{0,1\}^n \rightarrow \{-1, +1\}$$

$$\chi_s(x) = (-1)^{\text{XOR}_s(x)}$$

$$\text{XOR}_s(x) = \sum_{i: s_i=1} x_i \pmod{2} = \sum_{i=1}^n s_i x_i \pmod{2}$$

$$= s \cdot x \quad (\text{in } \mathbb{F}_2^n)$$

(symmetric in s, x : $(H^{\otimes n})^T = H^{\otimes n}$.)

(Stare at 4×4 example! — cols. of $H^{\otimes n}$ are the $|\chi_s\rangle$ vectors)

$$H^{\otimes n} |s\rangle = |\chi_s\rangle = \frac{1}{\sqrt{2^n}} \sum_x (-1)^{s \cdot x} |x\rangle$$

e.g. $\therefore H^{\otimes n} |00 \dots 0\rangle = \frac{1}{\sqrt{2^n}} \sum_x |x\rangle$ (unif superpos. \checkmark)

$\cdot H^{\otimes n} |11 \dots 1\rangle = \frac{1}{\sqrt{2^n}} \sum_x (-1)^{\sum_{i=1}^n x_i \pmod 2} |x\rangle$

$\cdot H^{\otimes n} |s\rangle = \frac{1}{\sqrt{2^n}} \sum_{x: s \cdot x = 0} |x\rangle - \frac{1}{\sqrt{2^n}} \sum_{x: s \cdot x = 1} |x\rangle$

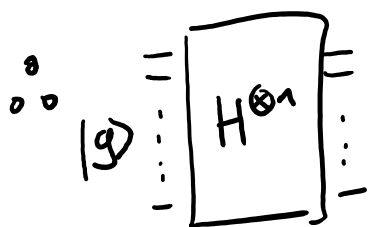
(a vector subspace)

Rec: "strength of pattern $|\chi_s\rangle$ for $|g\rangle$ "
 = coeff of $|g\rangle$ in basis of χ 's

$$\langle \chi_s | g \rangle = \hat{g}(s) = \Pr_{x \in \{0,1\}^n} [g(x) = \chi_s(x)] - \Pr_{x \in \{0,1\}^n} [g(x) \neq \chi_s(x)] \in [-1, +1]$$

$H^{\otimes n}$: std basis $\rightarrow \chi_s$ basis.

\leftarrow via $(H^{\otimes n})^{-1} = H^{\otimes n}$.



Measure: get "s" $\in \{0,1\}^n$
 w. prob. $|\hat{g}(s)|^2$
 (squared "strength" of pattern)

Last lecture (Bernstein-Vazirani)

Someone gives you quantum chip Q_F implementing $F = \text{XOR}_s$ for some unknown $s \in \{0,1\}^n$. Which?

(We're imagining the "query/oracle/black-box" model, like in HW6 #4 - we'll discuss more later.)
(Can't "look inside" Q_F ; want to apply it few times.)

Classical inputs only: $x \mapsto x \cdot s \pmod 2$.

Could do $x = (1, 0, 0, \dots, 0) \rightarrow$ get s_1

$(0, 1, 0, \dots, 0) \rightarrow$ get s_2

\vdots

$(0, 0, \dots, 0, 1) \rightarrow$ get s_n

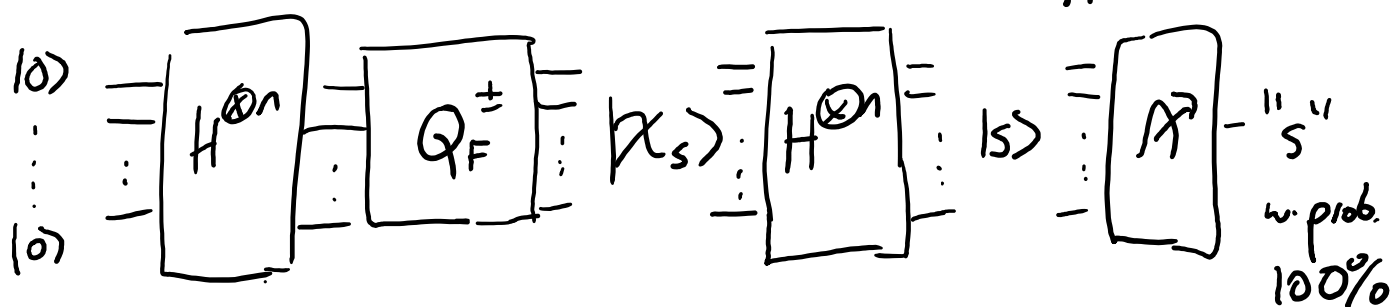
n applications. (Not bad. n is a "physically plausible amount.")

Rem: $\geq n$ necessary. Each application $Q_F(x)$ yields only 1 bit of info; need n bits to determine s .

(Indeed, each query yields one \mathbb{F}_2 -linear equation on unknown s : $x_1 \cdot s_1 + \dots + x_n \cdot s_n = Q_F(x)$.)

(For this problem, randomness doesn't really help. Still only get 1 bit of info/query.)

Quantumly: 1 use of Q_F $\rightarrow Q_F^\pm$.



Similar speedups?

Deutsch-Jozsa '92:

• Given Q_F implementing $F: \{0,1\}^n \rightarrow \{0,1\}$.

Promi set: either $F(x) = 0 \ \forall x$
 or F is "balanced": 0 for 50%
 of x 's, 1 for 50% of x 's.

• Try to decide which.

(Yes, this problem is highly contrived.)