

Lecture: Quantum query complexity

What is the true source of quantum power?

- Amplitude interference?
- Size of Hilbert space?
- Entanglement?

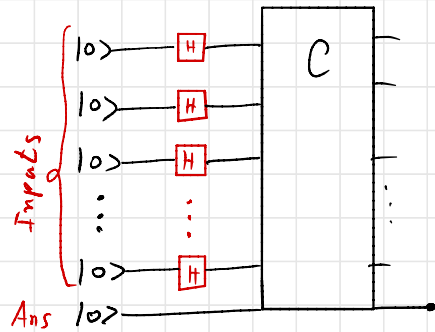
Pairwise classically simulatable

My take: Query in superposition!

Let C be a quantum circuit

implementing $f: \{0,1\}^n \rightarrow \{0,1\}$

i.e.; $C(|x\rangle \otimes |0\rangle) = |x\rangle \otimes |f(x)\rangle$.



$$\left(\frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle\right) \otimes \dots \otimes \left(\frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle\right) \otimes |0\rangle = \frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} |x\rangle \otimes |0\rangle$$



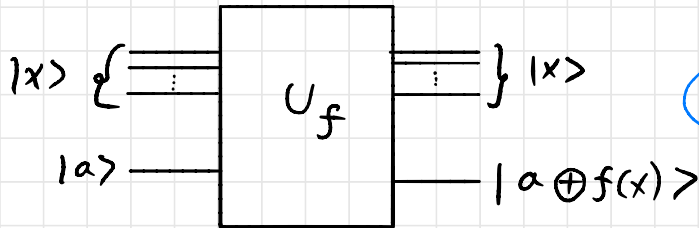
$$\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} |x\rangle \otimes |f(x)\rangle$$

Deutsch-Simon

Quantum Query Complexity

- Guest
- Assignment
- Approach

We are given $f: \{0,1\}^n \rightarrow \{0,1\}$ as a black-box

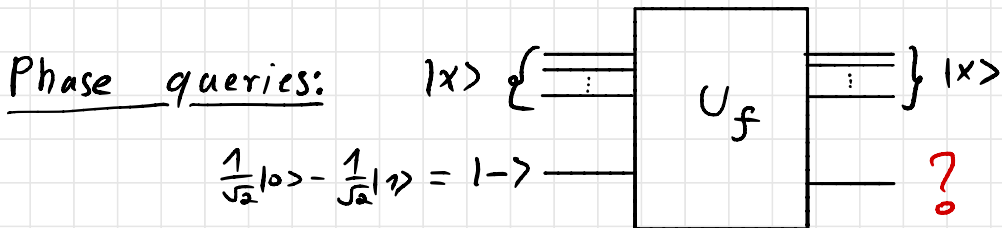


Superposition!

Usually $|a\rangle = |0\rangle \Rightarrow |a \oplus f(x)\rangle = |f(x)\rangle$

Goal: Learn about f (formally decide $f \in S$)
using as little as possible queries.

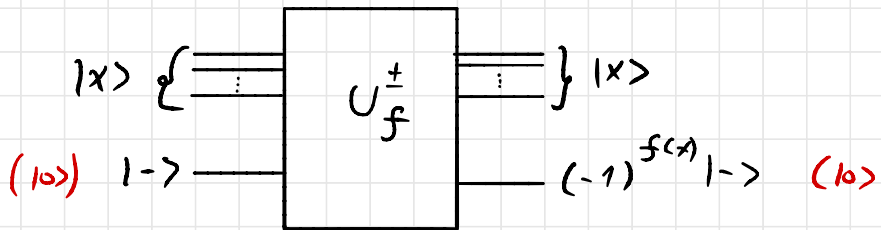
why? implicit representation, sublinear-time



$$\begin{aligned} \frac{1}{\sqrt{2}}|0 \oplus f(x)\rangle - \frac{1}{\sqrt{2}}|1 \oplus f(x)\rangle &= \frac{1}{\sqrt{2}}|f(x)\rangle - \frac{1}{\sqrt{2}}|1 \oplus f(x)\rangle \\ &= (-1)^{f(x)} |1\rangle \end{aligned}$$

$1 \rightarrow$ if $f(x) = 0$
 $-1 \rightarrow$ if $f(x) = 1$

In summary:



So $|x\rangle \otimes |1-\rangle \xrightarrow{(10)} (-1)^{f(x)} |x\rangle \otimes |1-\rangle$

The $|1-\rangle$ is just an ancilla - ignore.

$$|x\rangle \xrightarrow{U_f^\pm} (-1)^{f(x)} |x\rangle$$

Finally, in superposition:

- Standard queries: $\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} |x\rangle \otimes |f(x)\rangle$

- Phase queries: $\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle$

This is where the magic is happening!

But how can we use it? info is stuck in the amplitudes...

Rediscovering the first quantum algorithm

$$|0\rangle = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

Typically we teach: Problem, algorithm, analysis

This isn't how research always works...

Instead, we will try to understand the notion!

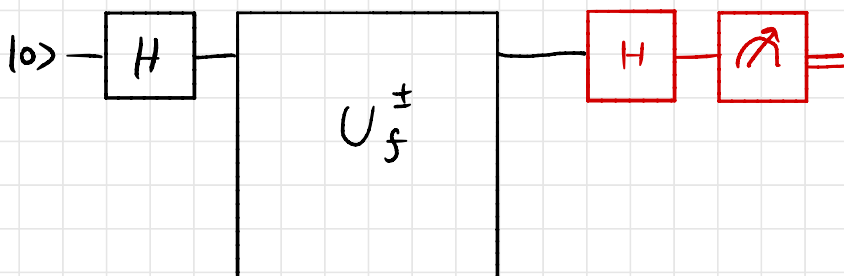
Goal: Understand the power of computing in superposition.

This is
amazing:

$$\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle$$

But
complicated...

Simplify: Consider $n=1$. Let $f: \{0,1\} \rightarrow \{0,1\}$



$$|0\rangle \rightarrow \frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle \rightarrow \frac{1}{\sqrt{2}} \left[(-1)^{f(0)} |0\rangle + (-1)^{f(1)} |1\rangle \right]$$

Cool, but how is that useful?

$$= \frac{(-1)^{f(0)}}{\sqrt{2}} \left[|0\rangle + (-1)^{f(1)-f(0)} |1\rangle \right]$$

Observation:

Global phase is indisting.

- If $f(0) = f(1) \leadsto \frac{(-1)^{f(0)}}{\sqrt{2}} (|0\rangle + |1\rangle) \simeq |+\rangle$
- If $f(0) \neq f(1) \leadsto \frac{(-1)^{f(0)}}{\sqrt{2}} (|0\rangle - |1\rangle) \simeq |-\rangle$

Apply Hadamard: If $f(0) = f(1)$, we get $|0\rangle$

If $f(0) \neq f(1)$, we get $|1\rangle$

This is same as $\text{XOR}(f(0), f(1))$

Deutsch's algorithm: computes the parity (XOR) of two bits using 1 query. "Sublinear" effect

Classically, this requires 2 queries.

The Deutsch-Jozsa Algorithm

Recall that Deutsch's algorithm computes the parity (XOR) of two bits using 1 query.

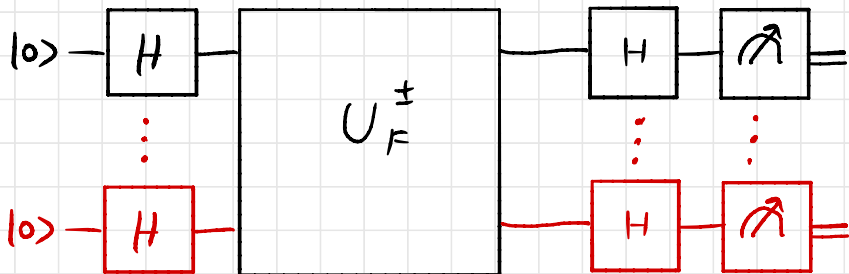
We looked at 2 bits for simplicity.

Can we generalise to n bits?

Black-box approach: Compute $\bigoplus_{x \in \{0,1\}^n} F(x)$ by breaking into $\frac{2^n}{2}$ blocks.

Nice, but not great...

White-box approach:



Time 0: $|0\rangle^{\otimes n} = |0\rangle \otimes \dots \otimes |0\rangle$ (Notation)

Time 1: $\left(\frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle\right) \otimes \dots \otimes \left(\frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle\right) = \frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} |x\rangle$

($H^{\otimes n}$ as uniform superposition)

Time 2: $\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} |x\rangle$ (loading the func.)

Time 3: $\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} \underline{\underline{H^{\otimes n} |x\rangle}}$

We need a Fourier analysis interlude...

Boolean Fourier Transform (Hadamard transform)

What's $H^{\otimes n} |x\rangle$? (we know $H^{\otimes n} |0\rangle$)

Recall: $H|0\rangle = |+\rangle = \frac{1}{\sqrt{2}}|0\rangle + \frac{1}{\sqrt{2}}|1\rangle$

$$H|1\rangle = |-\rangle = \frac{1}{\sqrt{2}}|0\rangle - \frac{1}{\sqrt{2}}|1\rangle$$

Trick: For $b \in \{0, 1\}$, write $H|b\rangle = \frac{|0\rangle + (-1)^b |1\rangle}{\sqrt{2}}$.

$$\text{So, } H^{\otimes n} |x\rangle = \frac{|0\rangle + (-1)^{x_1} |1\rangle}{\sqrt{2}} \otimes \dots \otimes \frac{|0\rangle + (-1)^{x_n} |1\rangle}{\sqrt{2}}$$

$$= \frac{1}{2^{n/2}} \sum_{y \in \{0,1\}^n} (-1)^{x \cdot y} |y\rangle$$

where $x \cdot y = \sum x_i y_i \pmod{2} = \bigoplus_{i: x_i=1} y_i = \text{XOR}_x(y)$

Later: Why is that a Fourier transform?

Now we can go back to Deutsch-Jozsa

Time 3: $\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} \underline{\underline{H^{\otimes n} |x\rangle}}$

$$\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} \cdot \frac{1}{2^{n/2}} \sum_{y \in \{0,1\}^n} (-1)^{x \cdot y} |y\rangle$$

How to deal with that?

Shortcut: what is the amplitude of $|00 \dots 0\rangle$?

$$\alpha := \frac{1}{2^n} \sum_{x \in \{0,1\}^n} (-1)^{F(x)}$$

Observation: If F is const ($F(x) = b \ \forall x$) $\Rightarrow \alpha \in \{-1, 1\}$.

$$\text{If } F \text{ is balanced } \left(\sum F(x) = \frac{2^n}{2} \right) \Rightarrow \alpha = 0$$

Thus, if F const, we'll always measure $|0\rangle$

, if F bal., we'll never measure $|0\rangle$

Deutsch-Jozsa problem: distinguish between constant and balanced functions using 1 query (w.p. 1!)

Digest (Deutsch-Jozsa)

Recall that Deutsch's algorithm computes the parity (XOR) of two bits using 1 query.

We wanted to generalise to parity of n bits.

Blackbox doesn't scale gracefully.

Whitebox gave us something entirely different:

The Deutsch-Jozsa alg. distinguishes between constant and balanced functions using 1 query (w.p. 1!)

Classically, this requires $\frac{2^n}{2} + 1$ (det) queries.

But with randomness, suffices $O(1)$ queries to distinguish w.h.p. we want more!

What about XORing n -bits: requires $\frac{n}{2}$ queries!

The Bernstein-Vazirani Algorithm

We can't use superposition to query n bits
using 1-query (or even $o(n)$ queries)

Next: The first quantum alg. that does exp
better than any (randomised) classical alg.

Problem: Let $F: \{0,1\}^n \rightarrow \{0,1\}$ s.t. $F(x) = S \cdot x \pmod{2}$
for $S \in \{0,1\}^n$.

Given phase-queries to F , find S .

$$\chi_S(x) = (-1)^{S \cdot x} = \text{Parity}_S^\pm(x) = \text{"character"}$$

(element of the Fourier basis for bits)

XOR-patterns

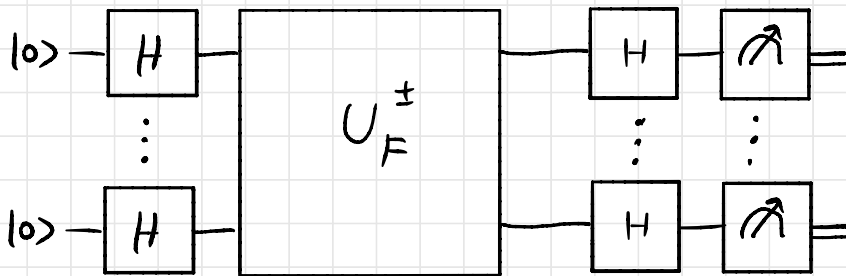
$$\text{Example: } S = (\overset{x_1}{0}, \overset{x_2}{1}, \overset{x_3}{0}, \overset{x_4}{1}, \overset{x_5}{1}) , \quad F(x) = x_2 \oplus x_4 \oplus x_5$$

Classically: requires n queries:

Since each query only provides 1-bit of info about s , we cannot do better!

$$\begin{cases} F(10000) = s_1 \\ F(01000) = s_2 \\ F(00100) = s_3 \\ F(00010) = s_4 \\ F(00001) = s_5 \end{cases}$$

The Bernstein-Vazirani alg.



Wait, what? This seems familiar...

$$H^{\otimes n} \left(\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} |x\rangle \right) = H^{\otimes n} \left(\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{s \cdot x} |x\rangle \right)$$

since $F(x) = s \cdot x$, for $s \in \{0,1\}^n$.

Let figure this out!

$$H^{\otimes n} \left(\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{s \cdot x} |x\rangle \right)$$

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

$$\text{If } s_i = 0 \Rightarrow \frac{|0\rangle + (-1)^{s_i} |1\rangle}{\sqrt{2}} = |+\rangle$$

$$H^{\otimes n} \left(\frac{|0\rangle + (-1)^{s_i} |1\rangle}{\sqrt{2}} \right) = |0\rangle$$

$$\text{If } s_i = 1 \Rightarrow \frac{|0\rangle + (-1)^{s_i} |1\rangle}{\sqrt{2}} = |-\rangle$$

$$H^{\otimes n} \left(\frac{|0\rangle + (-1)^{s_i} |1\rangle}{\sqrt{2}} \right) = |1\rangle$$

$$= |s\rangle \quad \triangleright$$

Measure, and we get $|s\rangle$ w.p. 1.

Interference.

Next: This was all Fourier analysis!

Convenient Notation

Last time

GHZ

Function-vector duality: $g: D^n \rightarrow \mathbb{R} \simeq v \in \mathbb{R}^{D^n}$

where $v = (g(\alpha_1), g(\alpha_2), \dots, g(\alpha_{|D^n|}))^+$

Example:

$x_1 x_2$	$\text{AND}(x_1, x_2)$
0 0	0
0 1	0
1 0	0
1 1	1

$$v = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$s_0 \sum |g(x)|^2 = 1$$

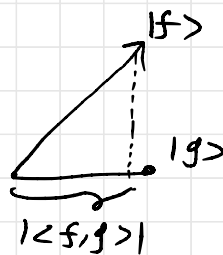
Def: Let $g: \{0,1\}^n \rightarrow \{-1,1\}$. $|g\rangle := \frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} g(x) |x\rangle$

Note: $\langle f | g \rangle = \frac{1}{2^{n/2}} (f(\bar{0}) \dots f(\bar{1}))^* \cdot \frac{1}{2^{n/2}} \begin{pmatrix} g(\bar{0}) \\ \vdots \\ g(\bar{1}) \end{pmatrix}$

$$= \frac{1}{2^n} \sum_x f(x) \cdot g(x)$$

Geometrically, this is a projection:

Next: Fourier analysis in parallel universes!



Boolean Fourier transform

Just a change of basis: standard $\rightarrow$ XOR.

The Fourier basis over $GF(2^n)$ (Bits...)

is $\{\chi_s\}_{s \in \{0,1\}^n}$, where $\chi_s(x) = (-1)^{s \cdot x}$ (Parity)

Fourier expansion: Let $f: \{0,1\}^n \rightarrow \{-1,1\}$. Then

$$f(x) = \sum_{s \in \{0,1\}^n} \hat{f}(s) \cdot \chi_s(x).$$

$$\text{where } \hat{f}(s) = \langle f | \chi_s \rangle = \frac{1}{2^n} \sum_{x \in \{0,1\}^n} f(x) \cdot \chi_s(x).$$

(Simply the projection!)

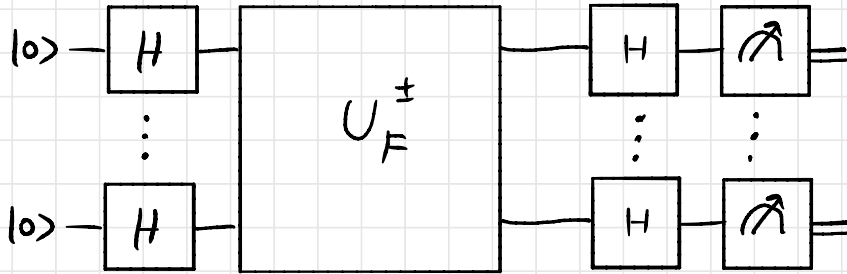
Fourier transform: $|f\rangle \xrightleftharpoons[H^{\otimes n}]{H^{\otimes n}} |\hat{f}\rangle$

Decomposing Boolean functions to XORs

Analogously to decomp. signals to sine & cosines!

Fourier Sampling

Let's distill the idea behind D-J & B-V.



$$\frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} \underline{\underline{H^{\otimes n} |x\rangle}}$$

$$= \frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} (-1)^{F(x)} \cdot \frac{1}{2^{n/2}} \sum_{s \in \{0,1\}^n} (-1)^{x \cdot s} |s\rangle$$

$$= \frac{1}{2^{n/2}} \sum_{x \in \{0,1\}^n} f(x) \cdot \frac{1}{2^{n/2}} \sum_{s \in \{0,1\}^n} \chi_s(x) |s\rangle$$

$$= \sum_{s \in \{0,1\}^n} \left(\frac{1}{2^n} \sum_{x \in \{0,1\}^n} f(x) \cdot \chi_s(x) \right) |s\rangle$$

$$= \frac{1}{2^{n/2}} \sum_{s \in \{0,1\}^n} \langle f | \chi_s \rangle |s\rangle = \sum_{s \in \{0,1\}^n} \hat{f}(s) |s\rangle$$

!!!

Since the final state is $\sum_{s \in \{0,1\}^n} \hat{f}(s) |s\rangle$,

When measuring (in comp. basis):

$$\forall s \in \{0,1\}^n \quad \Pr[|s\rangle] = |\hat{f}(s)|^2 \quad !$$

By Parseval's Theorem: $\sum_{s \in \{0,1\}^n} |\hat{f}(s)|^2 = \langle f, f \rangle = 1$

All algorithm we saw so far simply sampled from the Fourier spectrum!

Deutsch-Jozsa: Balanced vs constant

$$\text{Note } \hat{f}(0) = \langle f | \chi_0 \rangle = \frac{1}{2^n} \sum_{x \in \{0,1\}^n} f(x) \cdot \chi_0(x) = \mathbb{E}[f]$$

$$\text{- If Balanced: } \mathbb{E}[f] = 0 \Rightarrow \hat{f}(0)^2 = 0$$

$$\text{- If Constant: } \mathbb{E}[f] = \pm 1 \Rightarrow \hat{f}(0)^2 = 1$$

Berstein-Vazirani:

$$\text{If } f = \chi_s, \text{ then } \hat{f}(s) = 1 \quad (\text{since } f = \sum \hat{f}(s) \cdot \chi_s)$$