

Introduction to Quantum Information Processing

Lecture 19

Richard Cleve

Overview of Lecture 19

- Approximately universal sets of gates
- More on complexity classes
 - **NP**: definitions and examples of problems therein
 - **FACTORING** versus **NP** and **co-NP**
 - quantum speed-up for **NP**-complete problems
- Optimality of Grover's search algorithm

approximately
universal sets
of gates

A universal set of gates

Theorem: any unitary operation U acting on k qubits can be decomposed into $O(4^k)$ CNOT and one-qubit gates

Thus, the set of ***all*** one-qubit gates and the CNOT gate are ***universal*** in that they can simulate any other gate set

Question: is there a ***finite*** set of gates that is universal?

Answer 1: strictly speaking, ***no***, because this results in only countably many quantum circuits, whereas there are uncountably many unitary operations on k qubits (for any k)

Approximately universal gate sets

Answer 2: yes, for universality in an *approximate* sense

As an illustrative example, any rotation can be approximated within any precision by repeatedly applying

$$R = \begin{pmatrix} \cos(\sqrt{2}\pi) & -\sin(\sqrt{2}\pi) \\ \sin(\sqrt{2}\pi) & \cos(\sqrt{2}\pi) \end{pmatrix}$$

some number of times

In this sense, R is *approximately universal* for the set of all one-qubit rotations: any rotation S can be approximated within precision ε by applying R a suitable number of times

It turns out that $O((1/\varepsilon)^c)$ times suffices (for a constant c)

Approximately universal gate sets

Theorem: the gates **CNOT**, **H** , and $S = \begin{pmatrix} 1 & 0 \\ 0 & e^{i\pi/4} \end{pmatrix}$

are ***approximately universal***, in the sense that any unitary operation on k qubits can be simulated within precision ε by applying $O(4^k \log^c(1/\varepsilon))$ of them (c is a constant)

**more on
complexity
classes**

Complexity classes

Recall from Lecture 6:

- **P (polynomial time):** problems solved by $O(n^c)$ -size classical circuits (decision problems and uniform circuit families)
- **BPP (bounded error probabilistic polynomial time):** problems solved by $O(n^c)$ -size *probabilistic* circuits that err with probability $\leq 1/4$
- **BQP (bounded error quantum polynomial time):** problems solved by $O(n^c)$ -size *probabilistic* circuits that err with probability $\leq 1/4$
- **PSPACE (polynomial space):** problems solved by algorithms that use $O(n^c)$ memory.

Summary of previous containments

$$\mathbf{P} \subseteq \mathbf{BPP} \subseteq \mathbf{BQP} \subseteq \mathbf{PSPACE} \subseteq \mathbf{EXP}$$

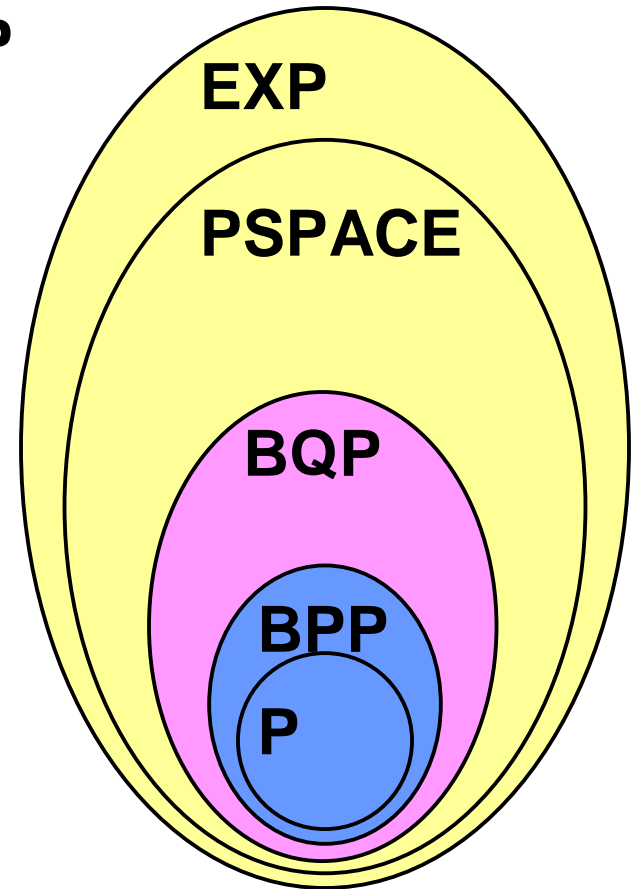
We now consider further structure between **P** and **PSPACE**

Technically, we will restrict our attention to *languages* (essentially $\{0,1\}$ -problems)

Many problems of interest can be cast in terms of languages

For example,

FACTORING = $\{(x,y) : \exists 2 \leq z \leq y, \text{ such that } z \text{ divides } x\}$



NP

Define **NP (non-deterministic polynomial time)** as the class of languages whose *positive* instances have “witnesses” that can be verified in polynomial time

Example: Let **3-CNF-SAT** be the language consisting of all **3-CNF** formulas that are satisfiable

3-CNF formula:

$$f(x_1, \dots, x_n) = (x_1 \vee \bar{x}_3 \vee x_4) \wedge (\bar{x}_2 \vee x_3 \vee \bar{x}_5) \wedge \dots \wedge (\bar{x}_1 \vee x_5 \vee \bar{x}_n)$$

$f(x_1, \dots, x_n)$ is **satisfiable** iff there exists $b_1, \dots, b_n \in \{0, 1\}$ such that $f(b_1, \dots, b_n) = 1$

No sub-exponential-time algorithm is known for **3-CNF-SAT**

But poly-time verifiable witnesses exist (namely, b_1, \dots, b_n)

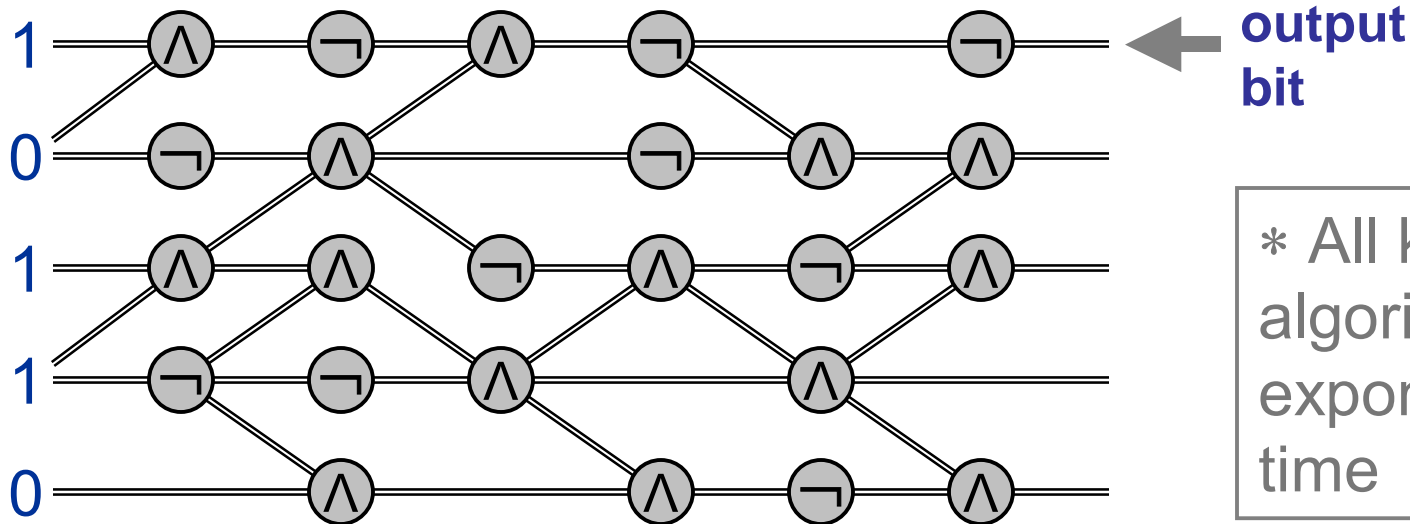
Other “logic” problems in NP

- ***k*-DNF-SAT:**

$$f(x_1, \dots, x_n) = (x_1 \wedge \bar{x}_3 \wedge x_4) \vee (\bar{x}_2 \wedge x_3 \wedge \bar{x}_5) \vee \dots \vee (\bar{x}_1 \wedge x_5 \wedge \bar{x}_n)$$

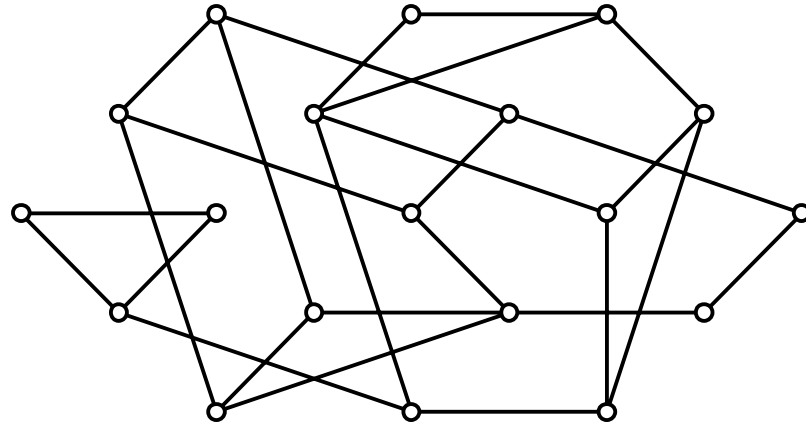
* But, unlike with *k*-CNF-SAT, this one is known to be in P

- **CIRCUIT-SAT:**



* All known algorithms exponential-time

“Graph theory” problems in NP



- **k -COLOR:** does G have a k -*coloring*?
- **k -CLIQUE:** does G have a *clique* of size k ?
- **HAM-PATH:** does G have a *Hamiltonian path*?
- **EUL-PATH:** does G have an *Eulerian path*?

“Arithmetic” problems in NP

- **FACTORING** = $\{(x, y) : \exists 2 \leq z \leq y, \text{ such that } z \text{ divides } x\}$
- **SUBSET-SUM**: given integers x_1, x_2, \dots, x_n, y , do there exist $i_1, i_2, \dots, i_k \in \{1, 2, \dots, n\}$ such that $x_{i_1} + x_{i_2} + \dots + x_{i_k} = y$?
- **INTEGER-LINEAR-PROGRAMMING**: linear programming where one seeks an *integer-valued* solution (its existence)

P vs. NP

All of the aforementioned problems have the property that they *reduce* to **3-CNF-SAT**, in the sense that a polynomial-time algorithm for **3-CNF-SAT** can be converted into a polynomial-time algorithm for the problem

Example:



If a polynomial-time algorithm is discovered for **3-CNF-SAT** then there is a polynomial-time algorithm for **3-COLOR**

In fact, this holds for *any* problem $X \in \mathbf{NP}$, hence **3-CNF-SAT** is *NP-hard* ... and so are **CIRCUIT-SAT**, k -**COLOR**, ...

FACTORING vs. NP

Is **FACTORING** NP-hard too?

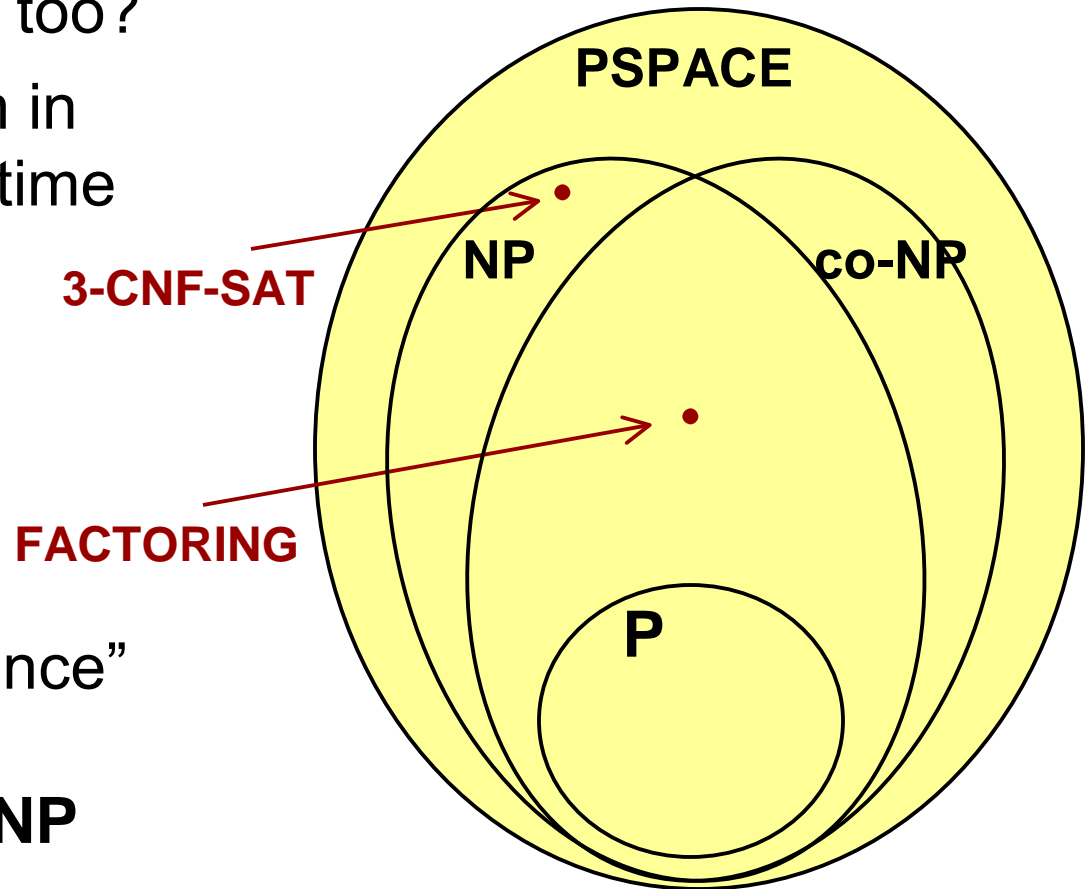
If so, then **every** problem in **NP** is solvable by a poly-time quantum algorithm!

But **FACTORING** has not been shown to be **NP-hard**

Moreover, there is “evidence” that it is not **NP-hard**:

FACTORING \in $\text{NP} \cap \text{co-NP}$

If **FACTORING** is **NP-hard** then **NP = co-NP**



FACTORING vs. co-NP

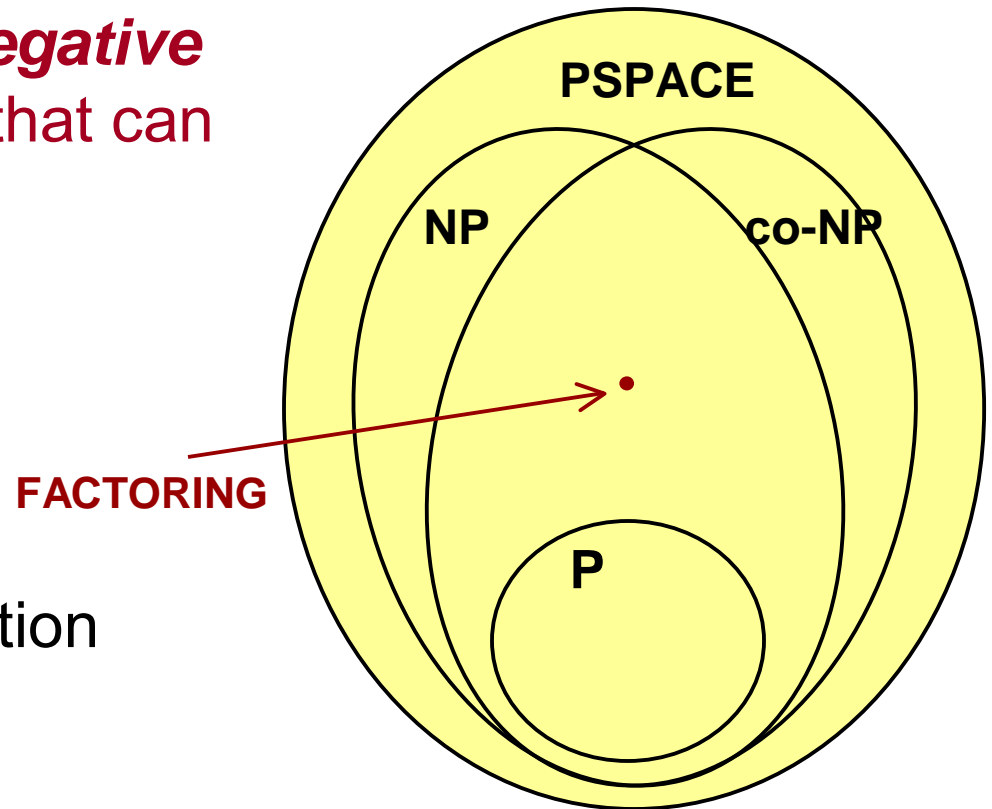
FACTORING = $\{(x, y) : \exists 2 \leq z \leq y, \text{ s.t. } z \text{ divides } x\}$

co-NP: languages whose *negative* instances have “witnesses” that can be verified in poly-time

Question: what is a good witness for the negative instances?

Answer: the prime factorization p_1, p_2, \dots, p_m of x will work

Can verify primality and compare p_1, p_2, \dots, p_m with y , all in poly-time



Quantum speed-up for NP-complete problems

Can use Grover's quantum search algorithm to find a witness *quadratically* faster than with known classical algorithms

Example: for **CIRCUIT-SAT**, best classical algorithm is to search for a satisfying assignment, taking time $O(n^c 2^n)$

Quantum algorithm takes time $O(n^c 2^{n/2})$

**optimality of
Grover's search
algorithm**

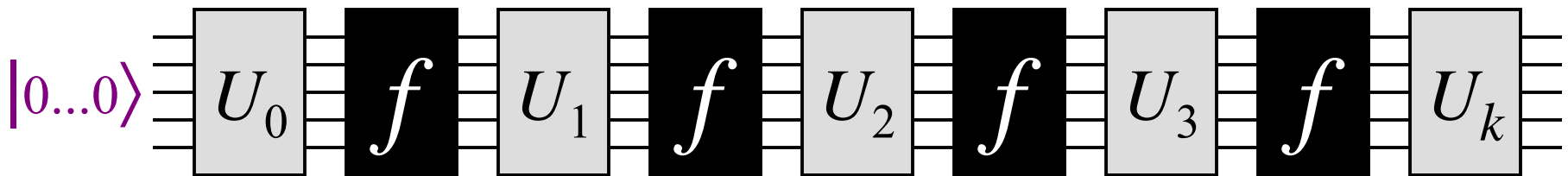
Optimality of Grover's algorithm

Theorem: any quantum search algorithm for $f: \{0,1\}^n \rightarrow \{0,1\}$ must make $\Omega(\sqrt{2^n})$ queries to f

Proof (of a slightly simplified version):

Assume queries are of the form $|x\rangle \equiv \begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{array} \begin{array}{c} \blacksquare \\ \blacksquare \\ \blacksquare \\ \blacksquare \end{array} \begin{array}{c} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{array} \equiv (-1)^{f(x)} |x\rangle$

and that a k -query algorithm is of the form

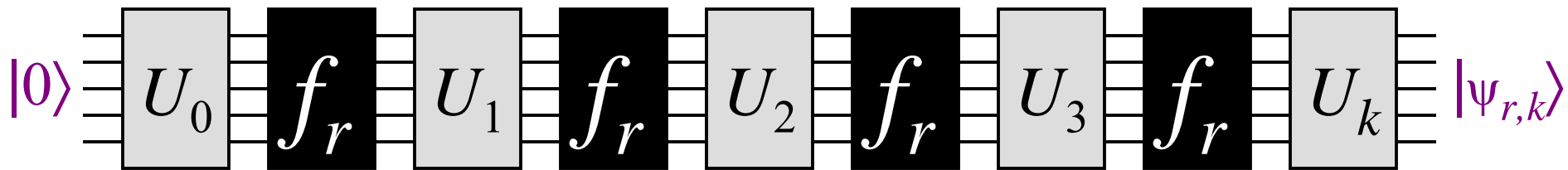


where $U_0, U_1, U_2, \dots, U_k$ are any unitary operations

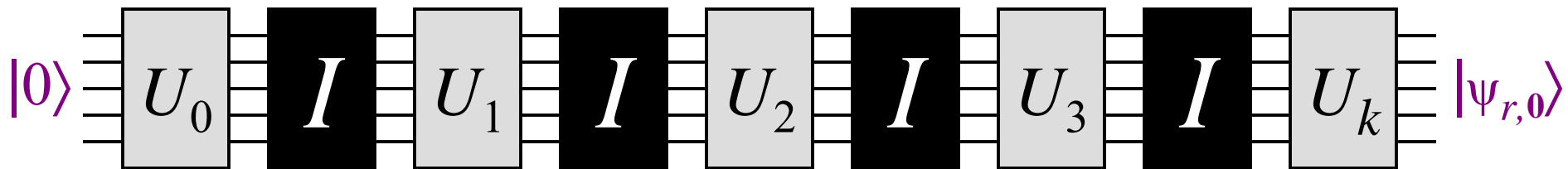
Optimality of Grover's algorithm

Define $f_r : \{0,1\}^n \rightarrow \{0,1\}$ as $f_r(x) = 1$ iff $x = r$

Consider



versus

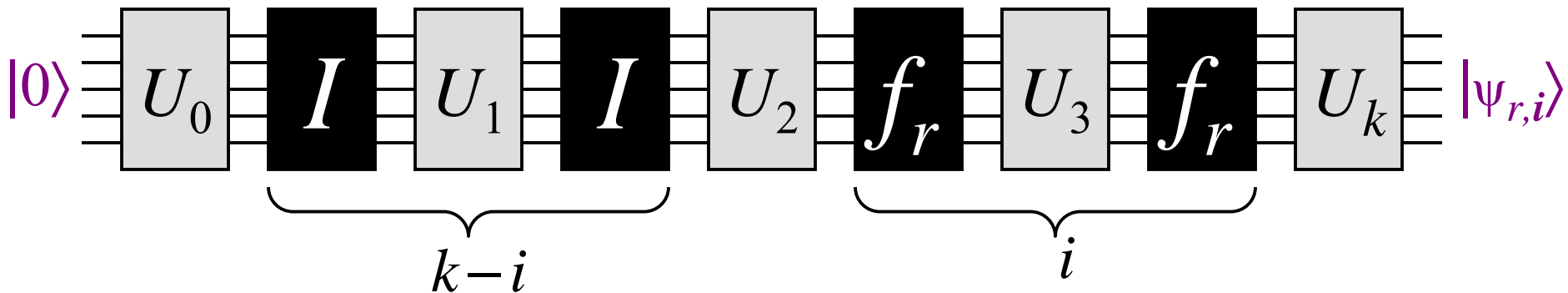


We'll show that, averaging over all $r \in \{0,1\}^n$,

$$\| |\Psi_{r,k}\rangle - |\Psi_{r,0}\rangle \| \leq 2k/\sqrt{2^n}$$

Optimality of Grover's algorithm

Consider



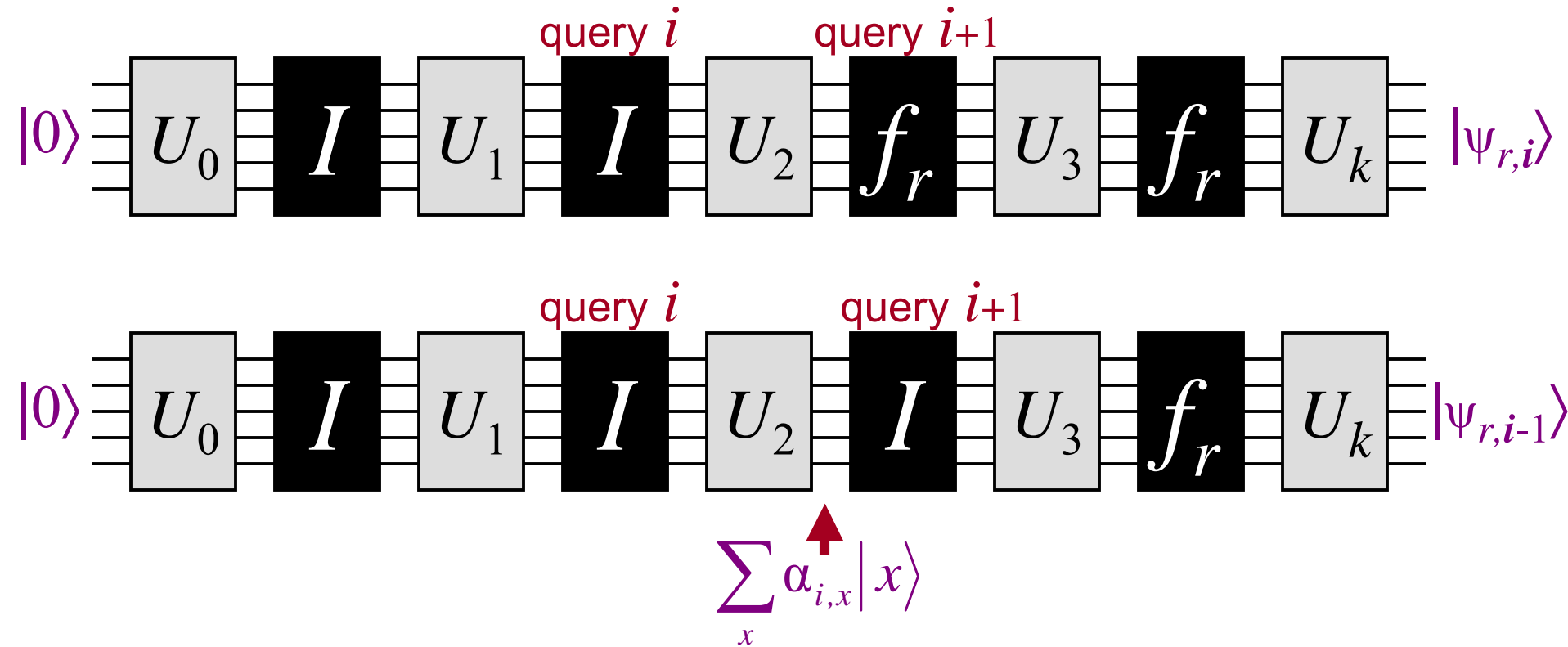
Note that

$$|\psi_{r,k}\rangle - |\psi_{r,0}\rangle = (|\psi_{r,k}\rangle - |\psi_{r,k-1}\rangle) + (|\psi_{r,k-1}\rangle - |\psi_{r,k-2}\rangle) + \dots + (|\psi_{r,1}\rangle - |\psi_{r,0}\rangle)$$

which implies

$$\| |\psi_{r,k}\rangle - |\psi_{r,0}\rangle \| \leq \| |\psi_{r,k}\rangle - |\psi_{r,k-1}\rangle \| + \dots + \| |\psi_{r,1}\rangle - |\psi_{r,0}\rangle \|$$

Optimality of Grover's algorithm



$\| |\psi_{r,i}\rangle - |\psi_{r,i-1}\rangle \| = |2\alpha_{i,r}|$, since query only negates $|r\rangle$

Therefore, $\| |\psi_{r,k}\rangle - |\psi_{r,0}\rangle \| \leq \sum_{i=0}^{k-1} 2|\alpha_{i,r}|$

Optimality of Grover's algorithm

Now, averaging over all $r \in \{0,1\}^n$,

$$\begin{aligned} \frac{1}{2^n} \sum_r \left\| |\Psi_{r,k}\rangle - |\Psi_{r,0}\rangle \right\| &\leq \frac{1}{2^n} \sum_r \left(\sum_{i=0}^{k-1} 2|\alpha_{i,r}| \right) \\ &= \frac{1}{2^n} \sum_{i=0}^{k-1} 2 \left(\sum_r |\alpha_{i,r}| \right) \\ &\leq \frac{1}{2^n} \sum_{i=0}^{k-1} 2 \left(\sqrt{2^n} \right) \quad (\text{By Cauchy-Schwarz}) \\ &= \frac{2k}{\sqrt{2^n}} \end{aligned}$$

Therefore, for **some** $r \in \{0,1\}^n$, the number of queries k must be $\Omega(\sqrt{2^n})$, in order to distinguish f_r from the all-zero function

This completes the proof

THE END