

# Quantum Computation

## (Lecture 2)

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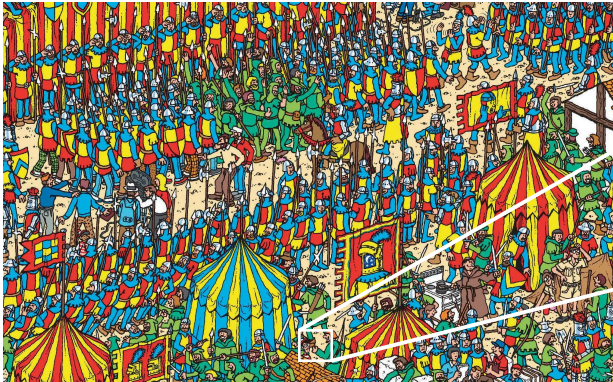
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# Search problems



# Search problems

## Search problem

- **Search space:** unstructured / unsorted
- **Asset:** a tool to efficiently recognise a solution

## Example: Searching in a sorted vs unsorted database

- find a name in a telephone directory
- find a phone number in a telephone directory

# Search problems

Note that that a procedure to **recognise** a solution does **not** need to rely on a previous knowledge of it.

## Example: password recognition

- $f(x) = 1$  iff  $x = 123456789$  ( $f$  **knows** the password)
- $f(x) = 1$  iff  $\text{hash}(x) = \text{c9b93f3f0682250b6cf8331b7ee68fd8}$   
( $f$  **recognises** a correct password, but does not know it as inverting a hash function is, in general, very hard.)

# Search problems

## A typical formulation

Given a function  $f : 2^n \rightarrow 2$  such that there exists a **unique** number, encoded by a binary string  $a$ , st

$$f(x) = \begin{cases} 1 & \Leftarrow x = a \\ 0 & \Leftarrow x \neq a, \end{cases}$$

determine  $a$ .

## A classical solution

- 0 evaluations of  $f$ : probability of success:  $\frac{1}{2^n}$
- 1 evaluation of  $f$ : probability of success:  $\frac{2}{2^n}$   
(choose a solution at random; if test fails choose another.)
- 2 evaluations of  $f$ : probability of success:  $\frac{3}{2^n}$ .
- $k$  evaluations of  $f$ : probability of success:  $\frac{k+1}{2^n}$ .

# Search problems

## Grover's algorithm (1996): A quadratic speed up

- Worst case for a classic algorithm:  $2^n$  evaluations of  $f$
- Worst case for Grover's algorithm:  $\sqrt{2^n}$  evaluations of  $f$

where  $n$  is the number of qubits necessary to represent the input (i.e. the search space)

## An oracle for $f$

As usual, an oracle encapsulates the reversible computation of  $f$  for an input  $|v\rangle$ :

$$U_f = |v\rangle|t\rangle \mapsto |v\rangle|t \oplus f(v)\rangle$$

Thus, preparing the target register with  $|0\rangle$ ,

$$U_f = |v\rangle|0\rangle \mapsto |v\rangle|f(v)\rangle$$

Measuring the target after  $U_f$  will return its answer to the given input, as (classically) expected.

**Superposition** will make the difference to take advantage of a quantum machine: Let  $N = 2^n$ , then

$$\psi = \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle$$

## An oracle for $f$

$|\psi\rangle$  can be expressed in terms of two states separating the **solution** states and **the rest**:

$$|a\rangle \text{ and } |r\rangle = \frac{1}{\sqrt{N-1}} \sum_{x \in N, x \neq a} |x\rangle$$

which forms a basis for a 2-dimensional subspace of the original  $N$ -dimensional space.

Thus,

$$|\psi\rangle = \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle = \underbrace{\frac{1}{\sqrt{N}}|a\rangle}_{\text{solution}} + \underbrace{\sqrt{\frac{N-1}{N}}|r\rangle}_{\text{the rest}}$$



## An oracle for $f$

If the target qubit is set to  $|-\rangle$ , the effect of  $U_f$  is

$$U_f = |x\rangle|-\rangle \mapsto (-1)^{f(x)}|x\rangle|-\rangle$$

Thus,  $U_f$  can be written as a **single qubit oracle** which encodes the answer of  $U_f$  as a **phase shift**:

$$V = |x\rangle \mapsto (-1)^{f(x)}|x\rangle$$

$$\text{(i.e. } V|a\rangle = -|a\rangle \text{ and } V|x\rangle = |x\rangle \text{ (for } x \neq a \text{))}$$

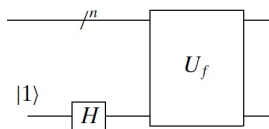
which can be expressed as

$$V = \sum_{x \neq a} |x\rangle\langle x| - |a\rangle\langle a| = I - 2|a\rangle\langle a|$$

# An oracle for $f$

$$V = \sum_{x \neq a} |x\rangle\langle x| - |a\rangle\langle a| = I - 2|a\rangle\langle a|$$

## The circuit



$V$  identifies the **solution** but does not allow for an observer to retrieve it because the square of the amplitudes for any value is always  $\frac{1}{N}$ .

## An amplifier

The oracle performs a phase shift over an **unknown** state. But this does not change the probability of retrieving the right answer. Thus, one needs a mechanism to **boost the probability of retrieving the solution**, which will be accomplished by another phase shift, but now applied to well-known vectors.

Consider, first the following program  $P$ :

$$\begin{aligned}
 P|x\rangle &= -(-1)^{\delta_{x,0}}|x\rangle \\
 &= |0\rangle\langle 0| + (-1) \sum_{x \neq 0} |x\rangle\langle x| \\
 &= |0\rangle\langle 0| + (-1)(I - |0\rangle\langle 0|) \\
 &= 2|0\rangle\langle 0| - I
 \end{aligned}$$

$P$  applies a **phase shift** to all vectors in the subspace spanned by all the basis states  $|x\rangle$ , for  $x \neq 0$ , i.e. all states orthogonal to  $|00 \cdots 0\rangle$ .

# An amplifier

Then, define an operator  $W = H^{\otimes n} P H^{\otimes n}$ , such that

- $W|\psi\rangle = |\psi\rangle$ , where

$$|\psi\rangle = H^{\otimes n}|00\cdots 0\rangle = |+\rangle^{\otimes n} = \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle$$

- $W|\phi\rangle = -|\phi\rangle$ , for any vector  $|\phi\rangle$  in the subspace orthogonal to  $|\psi\rangle$  (i.e. spanned by the basis vectors  $H|x\rangle$  for  $x \neq 0$ ).

$W$  applies a **phase shift** of  $-1$  to all vectors in the subspace orthogonal to  $|\psi\rangle$ .

# An amplifier

A simple calculation yields,

$$\begin{aligned}
 W &= H^{\otimes n} P H^{\otimes n} \\
 &= H^{\otimes n} (2|0\rangle\langle 0| - I) H^{\otimes n} \\
 &= 2(H^{\otimes n}|0\rangle\langle 0|H^{\otimes n}) - H^{\otimes n} I H^{\otimes n} \\
 &= 2|\psi\rangle\langle\psi| - I
 \end{aligned}$$

But does  $W$  boost the probability of finding the right solution?

## The effect of $W$ : to *invert about the average*

$$\begin{aligned}
 W \left( \sum_k \alpha_k |k\rangle \right) &= \left( 2 \left( \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle \frac{1}{\sqrt{N}} \sum_{y=0}^{N-1} \langle y| \right) - I \right) \sum_k \alpha_k |k\rangle \\
 &= \left( 2 \left( \frac{1}{N} \sum_{x=0}^{N-1} |x\rangle \sum_{y=0}^{N-1} \langle y| \right) - I \right) \sum_k \alpha_k |k\rangle \\
 &= 2 \left( \frac{1}{N} \sum_{x,y,k} \alpha_k |x\rangle \langle y|k\rangle \right) - \sum_k \alpha_k |k\rangle \\
 &= 2 \left( \underbrace{\frac{1}{N} \sum_k \alpha_k}_{\alpha - \text{mean}} \sum_x |x\rangle \right) - \sum_k \alpha_k |k\rangle \\
 &= 2 \alpha \sum_k |k\rangle - \sum_k \alpha_k |k\rangle \\
 &= \sum_k (2 \alpha - \alpha_k) |k\rangle
 \end{aligned}$$

## The effect of $W$ : to *invert about the average*

The effect of  $W$  is to transform the amplitude of each state so that it is as far above the average as it was below the average prior to its application, and vice-versa:

$$\alpha_k \mapsto 2\alpha - \alpha_k$$

$W$  inverts and boosts the “right” amplitude; slightly reduces the others.

## Invert about the average: Example

Let  $N = 2^2$  and suppose the solution  $a$  is encoded as the bit string 01. The algorithm starts with a uniform superposition

$$H^{\otimes 2}|00\rangle = \frac{1}{2} \sum_{k=0}^3 |k\rangle$$

which the oracle turns into

$$\frac{1}{2}|00\rangle - \frac{1}{2}|01\rangle + \frac{1}{2}|10\rangle + \frac{1}{2}|11\rangle$$

The effect of **inversion about the average** is

$$2 \overbrace{\begin{bmatrix} \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \end{bmatrix}}^{\alpha \sum_k |k\rangle} - \overbrace{\begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{bmatrix}}^{\sum_k \alpha_k |k\rangle} = \begin{bmatrix} \frac{2}{4} & -\frac{1}{2} \\ \frac{2}{4} & +\frac{1}{2} \\ \frac{2}{4} & -\frac{1}{2} \\ \frac{2}{4} & -\frac{1}{2} \\ \frac{2}{4} & -\frac{1}{2} \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

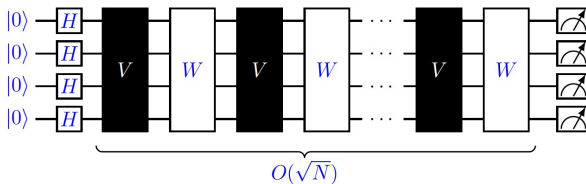
Measuring returns the solution with probability 1!



# The Grover iterator

$$\begin{aligned}
 G &= WV \\
 &= H^{\otimes n} P H^{\otimes n} V \\
 &= (2|\psi\rangle\langle\psi| - I) (I - 2|a\rangle\langle a|)
 \end{aligned}$$

## The Grover circuit



# Example: $N = 8$ , $a = 3$

Starting point:

$$\begin{array}{c}
 \text{---|---|---|---|---|---|---|---} \quad \alpha_\psi = \frac{1}{2\sqrt{2}} \\
 |000\rangle |001\rangle |010\rangle |011\rangle |100\rangle |101\rangle |110\rangle |111\rangle
 \end{array}$$

After the oracle

$$\begin{array}{c}
 \text{---|---|---|---|---|---|---|---} \quad \alpha_\psi = \frac{1}{2\sqrt{2}} \\
 \text{---|---|---|---|---|---|---|---} \quad \alpha_{|011\rangle} = \frac{-1}{2\sqrt{2}} \\
 |000\rangle |001\rangle |010\rangle |011\rangle |100\rangle |101\rangle |110\rangle |111\rangle
 \end{array}$$

## Example: $N = 8$ , $a = 3$

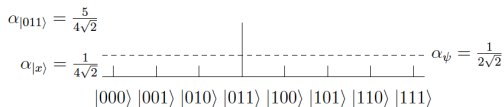
Inversion about the average

$$\begin{aligned}
 & (2|\psi\rangle\langle\psi| - I) \left( |\psi\rangle - \frac{2}{2\sqrt{2}}|011\rangle \right) \\
 &= 2|\psi\rangle\langle\psi|\psi\rangle - |\psi\rangle - \frac{2}{\sqrt{2}}|\psi\rangle\langle\psi|011\rangle + \frac{1}{\sqrt{2}}|011\rangle \\
 &= 2|\psi\rangle\langle\psi|\psi\rangle - |\psi\rangle - \frac{2}{\sqrt{2}} \frac{1}{2\sqrt{2}}|\psi\rangle + \frac{1}{\sqrt{2}}|011\rangle \\
 &= |\psi\rangle - \frac{1}{2}|\psi\rangle + \frac{1}{\sqrt{2}}|011\rangle \\
 &= \frac{1}{2}|\psi\rangle + \frac{1}{\sqrt{2}}|011\rangle
 \end{aligned}$$

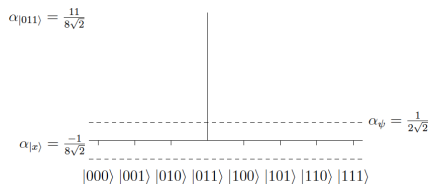
As  $|\psi\rangle = \frac{1}{2\sqrt{2}} \sum_{k=0}^7 |k\rangle$ , we end up with

$$\frac{1}{2} \left( \frac{1}{2\sqrt{2}} \sum_{k=0}^7 |k\rangle \right) + \frac{1}{\sqrt{2}}|011\rangle = \frac{1}{4\sqrt{2}} \sum_{k=0, k \neq 3}^7 |k\rangle + \frac{5}{4\sqrt{2}}|011\rangle$$

## Example: $N = 8$ , $a = 3$



Making a second iteration yields



and the probability of measuring the state corresponding to the solution is

$$\left| \frac{11}{8\sqrt{2}} \right|^2 = \frac{121}{128} \approx 94,5\%$$