

# Quantum Computation

(Lecture 6)

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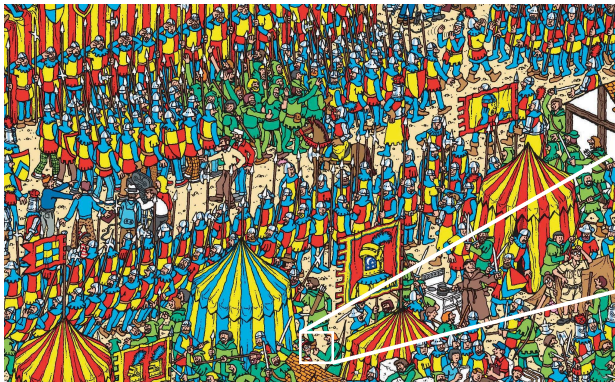
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## Quantum Computing Course Unit

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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  $hash(x) = c9b93f3f0682250b6cf8331b7ee68fd8$   
( $f$  **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  $f : 2^n (= N) \rightarrow 2$  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  $f$

## An oracle for $f$

... provides a means to **recognize** a solution 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.

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

## An oracle for $f$

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

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

which form 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 just

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

Since  $|-\rangle (= \frac{|0\rangle - |1\rangle}{\sqrt{2}})$  is an **eigenvector** of  $X$ , this corresponds to a **single qubit oracle** which encodes the answer of  $U_f$  as a **phase shift**:

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

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

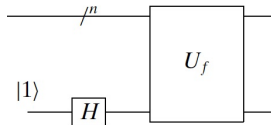
which can be expressed as

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

## 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

This entails the need for a mechanism to **boost the probability of retrieving the solution**.

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

$P$  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

Prepare a state in uniform superposition:

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

and define an operator  $W = H^{\otimes n} P H^{\otimes n}$ , which

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

Then,

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

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

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

## 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

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

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

$$2 \underbrace{\left[ \begin{array}{c} \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \\ \frac{1}{4} \end{array} \right]}_{\alpha \sum_k |k\rangle} - \underbrace{\left[ \begin{array}{c} \frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \\ \frac{1}{2} \end{array} \right]}_{\sum_k \alpha_k |k\rangle} = \left[ \begin{array}{c} \frac{2}{4} \\ \frac{2}{4} \\ \frac{2}{4} \\ \frac{2}{4} \end{array} \right] + \left[ \begin{array}{c} -\frac{1}{2} \\ \frac{1}{2} \\ -\frac{1}{2} \\ \frac{1}{2} \end{array} \right] = \left[ \begin{array}{c} 0 \\ 1 \\ 0 \\ 0 \end{array} \right]$$

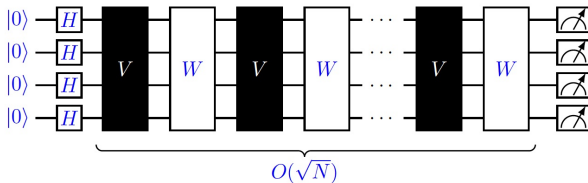
Measuring returns the solution with probability 1!



# The Grover iterator

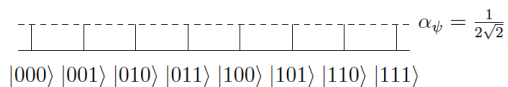
$$\begin{aligned}
 G &= WV \\
 &= H^{\otimes n} P H^{\otimes n} V \\
 &= (2|\psi\rangle\langle\psi| - I)(I - 2|a\rangle\langle a|)
 \end{aligned}$$

## The Grover circuit

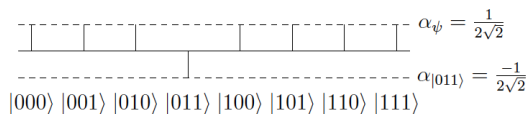


# Example: $N = 8, a = 3$

Starting point:



After the oracle



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

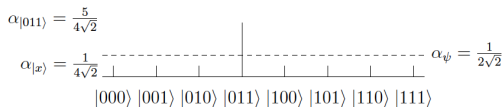
Inversion about the average

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

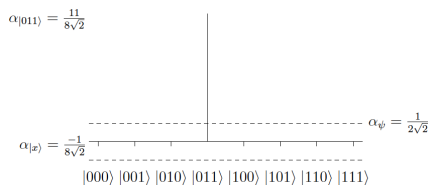
As  $|\psi\rangle = \frac{1}{2\sqrt{2}} \sum_{k=0}^7 |k\rangle$ , we end up with

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

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



Making a second iteration yields



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

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

## A geometric perspective on $G$

**Initial state:**  $|\psi\rangle = \frac{1}{\sqrt{N}}|a\rangle + \sqrt{\frac{N-1}{N}}|r\rangle$

The repeated application of  $G$  leaves the system in the 2-dimensional subspace of the original  $N$ -dimensional space, spanned by  $|a\rangle$  and  $|r\rangle$ .

Another basis is given by  $|\psi\rangle$  and the state **orthogonal** to  $|\psi\rangle$ :

$$|\bar{\psi}\rangle = -\frac{1}{\sqrt{N}}|a\rangle + \sqrt{\frac{N-1}{N}}|r\rangle$$

Define an angle  $\theta$  st  $\sin \theta = \frac{1}{\sqrt{N}}$  (and, of course,  $\cos \theta = \sqrt{\frac{N-1}{N}}$ ), and express both basis as

$$\begin{aligned} |\psi\rangle &= \sin \theta |a\rangle + \cos \theta |r\rangle & |\bar{\psi}\rangle &= \cos \theta |a\rangle - \sin \theta |r\rangle \\ |a\rangle &= \sin \theta |\psi\rangle + \cos \theta |\bar{\psi}\rangle & |r\rangle &= \cos \theta |\psi\rangle - \sin \theta |\bar{\psi}\rangle \end{aligned}$$

## A geometric perspective on $G$

$G$  has two components:

- $V$  which applies a phase shift to  $|a\rangle$ : reflection over  $|r\rangle$ .
- $W$  which applies a phase shift to all vectors in the subspace orthogonal to  $|\psi\rangle$ : reflection over  $|\psi\rangle$ .

Thus, one should express the action of  $V$  in the basis  $|\psi\rangle, |\bar{\psi}\rangle$  to perform afterwards the second reflection:

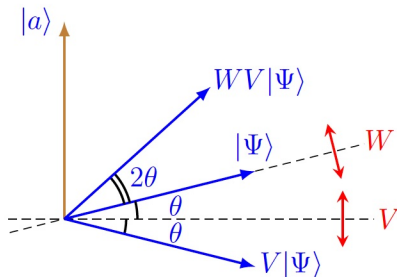
$$\begin{aligned}
 V|\psi\rangle &= -\sin\theta|a\rangle + \cos\theta|r\rangle \\
 &= -\sin\theta(\sin\theta|\psi\rangle + \cos\theta|\bar{\psi}\rangle) + \cos\theta(\cos\theta|\psi\rangle - \sin\theta|\bar{\psi}\rangle) \\
 &= -\sin^2\theta|\psi\rangle - \sin\theta\cos\theta|\bar{\psi}\rangle + \cos^2\theta|\psi\rangle - \cos\theta\sin\theta|\bar{\psi}\rangle \\
 &= (-\sin^2\theta + \cos^2\theta)|\psi\rangle - 2\sin\theta\cos\theta|\bar{\psi}\rangle \\
 &= \cos 2\theta|\psi\rangle - \sin 2\theta|\bar{\psi}\rangle
 \end{aligned}$$

## A geometric perspective on $G$

Then, the second reflection over  $|\psi\rangle$  yields the effect of the Grover iterator:

$$G|\psi\rangle = \cos 2\theta|\psi\rangle + \sin 2\theta|\bar{\psi}\rangle$$

which boils down to  $2\theta$  rotation:

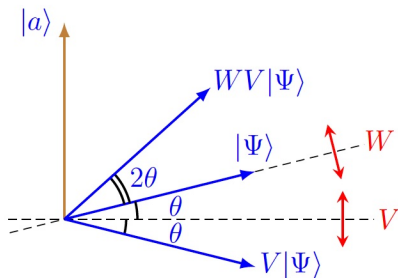


## What's behind the scenes?

- The key is the selective shifting of the phase of one state of a quantum system, one that satisfies some condition, at each iteration.
- Performing a phase shift of  $\pi$  is equivalent to multiplying the amplitude of that state by  $-1$ : the amplitude for that state changes, but the probability of being in that state remains the same
- Subsequent transformations take advantage of that difference in amplitude to single out that state and increase the associated probability.
- This would not be possible if the amplitudes were probabilities, not holding extra information regarding the phase of the state in addition to the probability — it's a **quantum feature**.



## How many times should $G$ be applied?



From this picture, we may also conclude that

- the **angular distance to cover** is

$$\frac{\pi}{2} - \theta = \frac{\pi}{2} - \arcsin\left(\frac{1}{\sqrt{N}}\right)$$

## How many times should $G$ be applied?

Thus, the ideal number of iterations is

$$t = \left\lfloor \frac{\frac{\pi}{2} - \arcsin \frac{1}{\sqrt{N}}}{2\theta} \right\rfloor$$

A lower bound for  $\theta$  gives an upper bound for  $t$   
 — for  $N$  large  $\theta \approx \sin \theta = \frac{1}{\sqrt{N}}$ . Thus,

$$t \approx \frac{\frac{\pi\sqrt{N}}{2\sqrt{N}}}{\frac{2}{\sqrt{N}}} = \frac{\pi}{4}\sqrt{N}$$

So,  $G$  applied  $t$  times leaves the system within an angle  $\theta$  of  $|a\rangle$ . Then, a measurement in the computational basis yields the correct solution with probability

$$\|\langle a|G^t|\psi\rangle\| \geq \cos^2 \theta = 1 - \sin^2 \theta = \frac{N-1}{N}$$

which, for large  $N$ , is very close to 1.

## How many times should $G$ be applied?

For an alternative computation, recall

$$G|\psi\rangle = \cos 2\theta|\psi\rangle + \sin 2\theta|\bar{\psi}\rangle$$

By induction, after  $k$  iterations,

$$\begin{aligned} G^k|\psi\rangle &= \cos(2k\theta)|\psi\rangle + \sin(2k\theta)|\bar{\psi}\rangle \\ &= \sin(2k+1)\theta|a\rangle + \cos(2k+1)\theta|r\rangle \end{aligned}$$

Thus, to maximize the probability of obtaining  $|a\rangle$ ,  $k$  is selected st

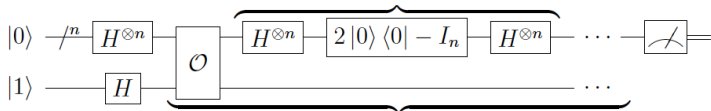
$$\sin((2k+1)\theta) \approx 1 \quad \text{i.e.} \quad (2k+1)\theta \approx \frac{\pi}{2}$$

which leads to

$$k \approx \frac{\pi}{4\theta} - \frac{1}{2} \approx \frac{\pi}{4}\sqrt{N} \approx t$$

# Grover's algorithm ( $\mathcal{O}(\sqrt{N})$ )

- Prepare the initial state:  $|0\rangle^{\otimes n}|1\rangle$
- Apply  $H^{\otimes n} \otimes H$  to yield  $\frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle|-\rangle$
- Apply the Grover iterator  $G$  to  $\frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle|-\rangle$ ,  $t \approx \frac{\pi}{4} \sqrt{N}$  times, leading approximately to state  $|a\rangle|-\rangle$
- Measure the first  $n$  qubits to retrieve  $|a\rangle$



## Multiple solutions

There  $M$  (out of  $2^n = N$ ) input strings evaluating to 0 by  $f$

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

where

$$|s\rangle = \frac{1}{\sqrt{M}} \sum_{x \text{ solution}} |x\rangle \quad \text{and} \quad |r\rangle = \frac{1}{\sqrt{N-M}} \sum_{x \text{ no solution}} |x\rangle$$

## Multiple solutions

$$t = \left\lfloor \frac{\frac{\pi}{2} - \arcsin \sqrt{\frac{M}{N}}}{2\theta} \right\rfloor$$

which, for  $N$  large,  $M \ll N$  (thus  $\theta \approx \sin \theta$ ), yields

$$t \approx \frac{\pi}{4} \sqrt{\frac{N}{M}}$$

The probability to retrieve a correct solution is

$$\|\langle s | G^t | \psi \rangle\| \geq \cos^2 \theta = 1 - \sin^2 \theta = \frac{N - M}{N}$$

which, for  $M = \frac{N}{2}$  yields  $\frac{1}{2}$ , but for  $M \ll N$ , is again close to 1.

## Multiple solutions

Computing the effect of  $G$ :  $2\theta$

$$\sin 2\theta = 2\sqrt{\frac{N-M}{N}} = 2\frac{\sqrt{M(N-M)}}{N}$$

$$2\theta = \arcsin\left(2\frac{\sqrt{M(N-M)}}{N}\right)$$

$M$ (out of 100)	$\arcsin \theta$
0	0
1	0.198
20	0.8
40	0.979
50	1
60	0.979
80	0.8
99	0.198
$M$	0

# Multiple solutions

Surprisingly, the rotation in each iteration decreases from  $M = \frac{N}{2}$  to  $N$ , and the number of iterations consequently increases, although one would expect to be easier to find a correct solution if their number increases!

## Solution

To double the number of elements in the search space, by adding  $N$  extra elements, none of which being a solution.



# The technique: Amplitude amplification

Grover's algorithm made use of

$$H^{\otimes n}|00\dots 0\rangle$$

to prepare a **uniform** superposition of potential solutions.

In general, one may resort to any module  $K$  to map the solution space to any **superposition of guesses**, plus some extra qubits to be used as **draft paper**:

$$K|00\dots 0\rangle = \sum_x \alpha_x |x\rangle |\text{draft}(x)\rangle$$

## The technique: Amplitude amplification

$$|\psi\rangle = \sum_{x \text{ solution}} \alpha_x |x\rangle |\text{draft}(x)\rangle + \sum_{x \text{ no solution}} \alpha_x |x\rangle |\text{draft}(x)\rangle$$

yielding the following probabilities:

$$p_s = \sum_{x \text{ solution}} \|\alpha_x\|^2 \quad \text{and} \quad p_{ns} = \sum_{x \text{ no solution}} \|\alpha_x\|^2 = 1 - p_s$$

Of course, amplification has no use if  $p_s \in \{0, 1\}$ .

## The technique: Amplitude amplification

Otherwise ( $0 < p_s < 1$ ), the amplitudes of **solution** inputs should be amplified.

First, express

$$|\psi\rangle = \sqrt{p_s}|\psi_s\rangle + \sqrt{p_{ns}}|\psi_{ns}\rangle$$

for the **normalised** components

$$|\psi_s\rangle = \sum_{x \text{ solution}} \frac{\alpha_x}{\sqrt{p_s}} |x\rangle |\text{draft}(x)\rangle$$

$$|\psi_{ns}\rangle = \sum_{x \text{ solution}} \frac{\alpha_x}{\sqrt{p_{ns}}} |x\rangle |\text{draft}(x)\rangle$$

which rewrites to

$$|\psi\rangle = \sin \theta |\psi_s\rangle + \cos \theta |\psi_{ns}\rangle$$

for  $\theta \in \{0, \frac{\pi}{2}\}$  such that  $\sin^2 \theta = p_s$ .

# The technique: Amplitude amplification

A generic search iterator is built as

$$S = KPK^{-1}V = W_KV$$

where

$$W_K|\psi\rangle = |\psi\rangle$$

$$W_K|\phi\rangle = -|\phi\rangle \quad \text{for all states orthogonal to } |\psi\rangle$$

The sets  $\{|\psi_s\rangle, |\psi_{ns}\rangle\}$  and  $\{|\psi\rangle, |\bar{\psi}\rangle\}$  are bases for the relevant 2-dimensional subspace.

## The technique: Amplitude amplification

As expected, starting in  $|\psi\rangle$ , the oracle produces

$$-\sin\theta|\psi_s\rangle + \cos\theta|\psi_{ns}\rangle = \cos(2\theta)|\psi\rangle - \sin(2\theta)|\bar{\psi}\rangle$$

which, followed by the amplifier, yields

$$\cos(2\theta)|\psi\rangle + \sin(2\theta)|\bar{\psi}\rangle$$

i.e. the effect of iterator  $S$  is

$$S|\psi\rangle = \cos(2\theta)|\psi\rangle + \sin(2\theta)|\bar{\psi}\rangle$$

which can be expressed in the basis  $\{|\psi_s\rangle, |\psi_{ns}\rangle\}$  as

$$S|\psi\rangle = \sin(3\theta)|\psi_s\rangle + \cos(3\theta)|\psi_{ns}\rangle$$

## The technique: Amplitude amplification

The repeated application of  $S$  a total of  $k$  times rotates the initial state  $|\psi\rangle$  to

$$S^k|\psi\rangle = \sin((2k+1)\theta)|\psi_s\rangle + \cos((2k+1)\theta)|\psi_{ns}\rangle$$

For the correct number of iterations, this procedure reaches a state such that a measurement will return an element of the **subspace spanned by  $|\psi_s\rangle$**  with a probability close to 1.

## The technique: Amplitude amplification

As before, to get that high probability, the smallest value for  $k$  one can choose is such that

$$(2k + 1)\theta \approx \frac{\pi}{2}$$

For a **small**  $\theta$ , as

$$\sin \theta = \sqrt{p_s} \approx \theta$$

the magnitude of the right number of iterations is

$$\Theta\left(\sqrt{\frac{1}{\theta}}\right)$$

because

$$(2k + 1)\sqrt{p_s} = \theta \Leftrightarrow k = \frac{\pi}{4\sqrt{p_s}} - \frac{1}{2}$$

## To follow

The algorithm requires that one knows **in advance** how many times iterator  $S$  is to be applied:

- For  $K = H$  (uniform sampling the input) this boils down to know the number of solutions of the search problem.
- For a generic  $K$  this amounts to know the probability with which  $K$  guesses a solution to the problem, i.e.  $\sin(\theta)$ .

To see ...

- **blind** search
- **estimate the amplitude** with which  $K$  maps  $|00 \dots 0\rangle$  to the subspace of solutions