

4. Cook's algorithm

For many hard computational problems, it is possible to check a solution efficiently, but we don't know how to find it.

So-called "NP problems" (non-deterministic polynomial)

Many interesting problems are of this type:

- graph coloring
- factoring
- 3-SAT
- tiling problems
- Hamiltonian path
- travelling salesman (markedly phrased)
- ⋮
- ⋮

Reformulation of NP problems:

We have an efficiently computable function

$$f(x) \in \{0, 1\}; \quad x \in \{0, 1, \dots, N-1\}$$

efficient =

— where $f(x)$ is a "verifier" for a given "solution" x ,
i.e. $f(x) = 1$ if & only if x is a correct solution —

and we want to find an x_0 s.t. $f(x_0) = 1$
(i.e. a correct solution).

(Can be interpreted as "database search": Want to
find a "marked element" x_0 in an un-
structured database.)

We assume for now that $x_0 : f(x_0) = 1$ is unique.
(Generalization: later/homework)

Classically: Will need $\Theta(N)$ queries to f for an
unstructured search (i.e., without using properties
of f).

Quantumly: Will see that $O(\sqrt{N})$ queries are enough.

(Note: This is only a quadratic speedup, but it applies
to a very large class of very relevant problems.)

Consider $f: \{0, \dots, N-1\} \rightarrow \{0, 1\}$
 \uparrow
 $N = 2^n$: n qubits

Algorithm 1:

Oracle

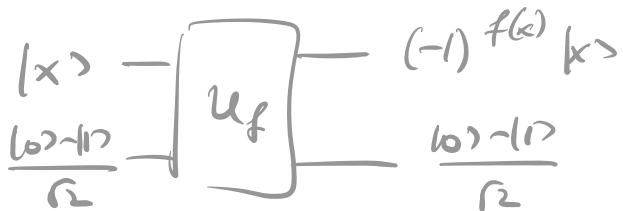
$$O_f : |x\rangle \mapsto (-1)^{f(x)} |x\rangle = (-1)^{\delta_{x,x_0}} |x\rangle$$

i.e. O_f flips amplitude of "marked" element.

Can also write as

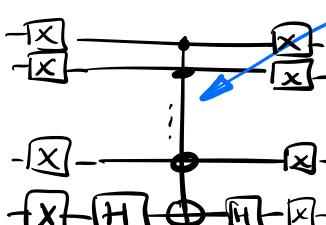
$$\boxed{O_f = I - 2|x_0\rangle\langle x_0|}$$

(Note: Can build O_f from U_f via
phase kick-back technique:



Ingredient 2:

Unitary $O_0 : |x\rangle \mapsto (-1)^{\delta_{x,0}} |x\rangle$

Corresponds to 

→ can be realized efficiently.

Again, write $O_0 = I - 2|0\rangle\langle 0|$

Define

$$\boxed{O_\omega := H^{\otimes n} O_0 H^{\otimes n} = I - 2|\omega\rangle\langle\omega|},$$

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

Grover's algorithm:

① Start from $|y_0\rangle = |\omega\rangle = H^{\otimes n} |0\rangle$

② Repeat: Apply Grover iteration

$$G = -H^{\otimes n} O_0 H^{\otimes n} O_f = (-\omega) O_f$$

$$|\psi_k\rangle \xrightarrow{C} |\psi_{k+1}\rangle = C|\psi_k\rangle = (-D_\omega)O_f|\psi_k\rangle.$$

(How many times? — Soon!)

How to analyze "trajectory" $|\psi_0\rangle \rightarrow |\psi_1\rangle \rightarrow |\psi_2\rangle \rightarrow \dots$?

Recall: $O_f = I - 2|\psi_0\rangle\langle\psi_0|$

$$-D_\omega = 2(\omega X_\omega) - I$$

and moreover, $|\psi_0\rangle = |\omega\rangle$.

\Rightarrow Only two special vectors in $|\psi_0\rangle, O_f, -D_\omega$:

$|\psi_0\rangle$ and $|\omega\rangle$. The identity I will

not change those vectors.

\Rightarrow The dynamics $|\psi_0\rangle \rightarrow |\psi_1\rangle \rightarrow |\psi_2\rangle \rightarrow \dots$

takes place in $\text{span}\{|\psi_0\rangle, |\omega\rangle\}$, i.e.,
a two-dimensional space!

Two natural ONSs for Heis space:

$$|x_0\rangle$$

$$|x_0^\perp\rangle := \frac{1}{\sqrt{N-1}} \sum_{x \neq x_0} |x\rangle$$

$$\propto |\omega\rangle - |x_0\rangle \underbrace{\langle x_0 | \omega \rangle}_{= \frac{1}{\sqrt{N}}}$$

$$\langle x_0 | x_0^\perp \rangle = 0$$

And another basis

$$|\omega\rangle \quad \text{with} \quad \langle \omega | \omega^\perp \rangle = 0$$

$$|\omega^\perp\rangle$$

Of course, any vector in $\text{span}\{|x_0\rangle, |\omega\rangle\}$ can be expanded in either basis:

$$|\psi\rangle = a|x_0\rangle + b|\omega\rangle = x|\omega\rangle + y|\omega^\perp\rangle$$

What is the effect of O_f and $(-\partial_\omega)$ on $|\psi\rangle$?

$$O_f |\psi\rangle = O_f (a|x_0\rangle + b|\omega^\perp\rangle) = \underbrace{-a|x_0\rangle + b|x_0^\perp\rangle}_{?}$$

$$O_f = I - 2|x_0\rangle\langle x_0|$$

$\Rightarrow \underline{O_f \text{ reflects } |\psi\rangle \text{ about } |x_0^\perp\rangle!}$

$$(-O_\omega)|\psi\rangle = (-O_\omega)(x|\omega\rangle + y|\omega^\perp\rangle)$$

$$\stackrel{\vec{\uparrow}}{=} -(-x|\omega\rangle + y|\omega^\perp\rangle) = x|\omega\rangle - y|\omega^\perp\rangle.$$

$$O_\omega = I - 2(\omega X\omega)$$

$\Rightarrow \underline{(-O_\omega) \text{ reflects } |\psi\rangle \text{ about } |\omega\rangle!}$

Thus: each Grover iteration consists of two steps:

(i) reflect about $|x_0^\perp\rangle$

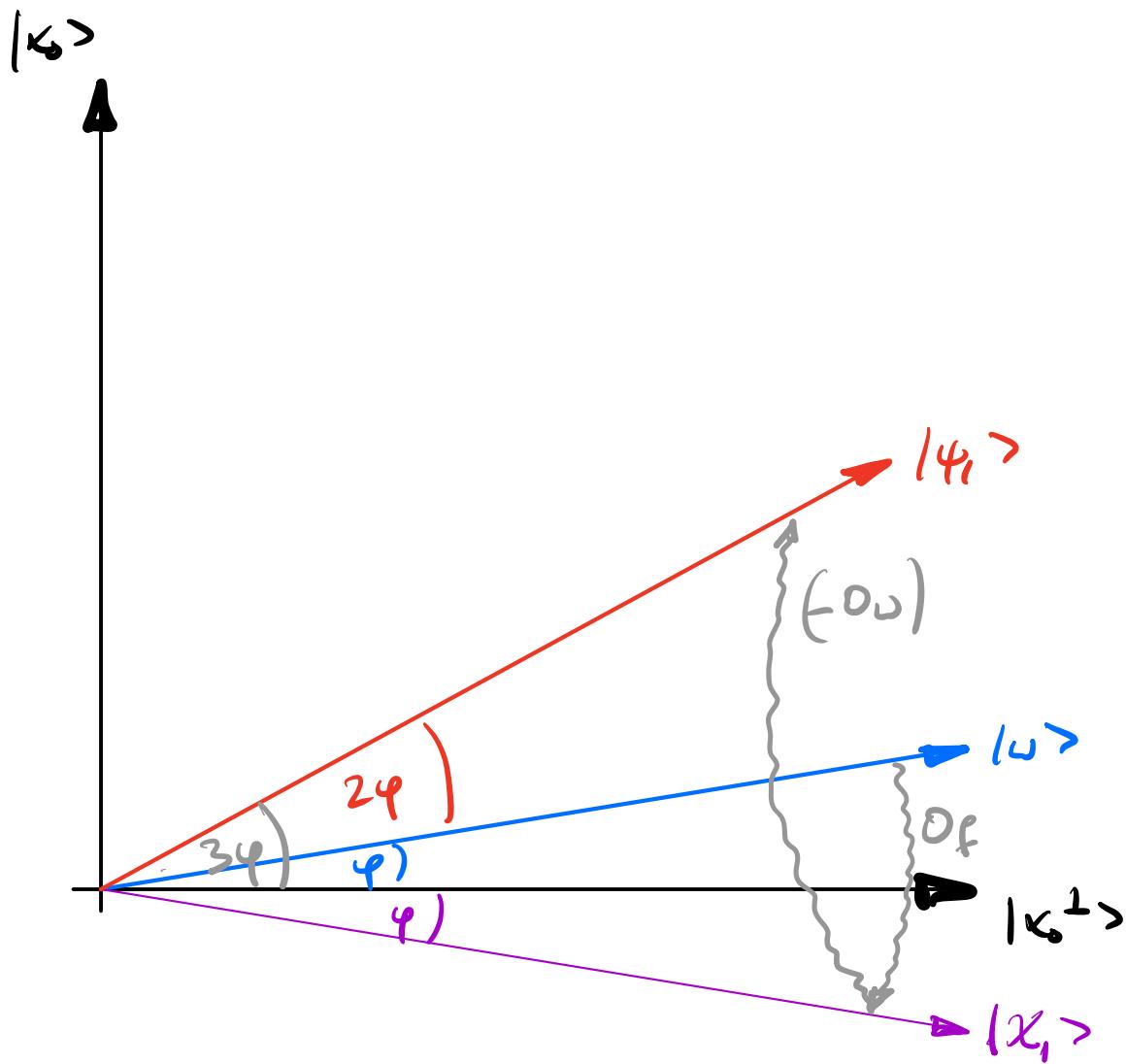
(ii) reflect about $|\omega\rangle$

What happens now if we start with $|\psi_0\rangle = |\omega\rangle$ and apply one iteration?

$$|\omega\rangle = \sin \varphi |x_0\rangle + \cos \varphi |x_0^\perp\rangle.$$

$$|x_1\rangle = O_f |\omega\rangle$$

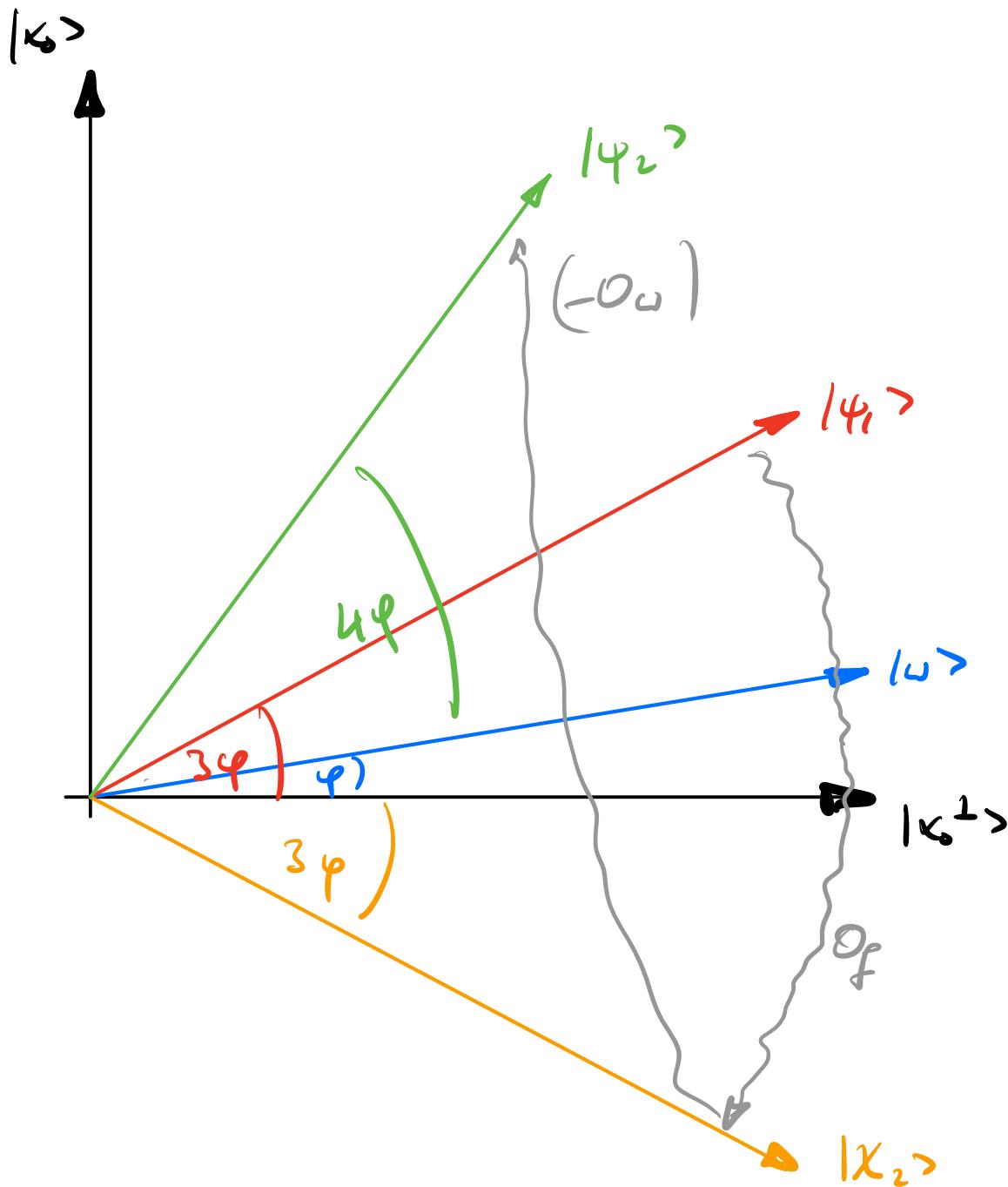
$$|\psi_1\rangle = (-O_\omega)|x_1\rangle = (-O_\omega)O_f |\omega\rangle$$



$$\rightarrow |\psi_1\rangle = \sin(3\varphi) |\kappa_0\rangle + \cos(3\varphi) |\kappa_0^\perp\rangle$$

Next Grover iteration:

$$|\psi_2\rangle = (-\omega) \underbrace{\Delta f}_{=: |\chi_2\rangle} |\psi_1\rangle$$



$$\Rightarrow |\psi_2\rangle = \sin(\sqrt{\varphi}) |\kappa\rangle + \cos(\sqrt{\varphi}) |\kappa^\perp\rangle$$

Can continue ... :

$$\Rightarrow |\psi_k\rangle = \sin((2k+1)\varphi) |\kappa\rangle + \cos((2k+1)\varphi) |\kappa^\perp\rangle$$

Want that $(2k+1)\varphi \approx \frac{\pi}{2}$: Then, measurement
in comp. basis will return $|x_0\rangle$ with high prob.!

$$\text{Since } |\omega\rangle = \frac{1}{\sqrt{N}} |x_0\rangle + \sqrt{\frac{N-1}{N}} |x_0^\perp\rangle$$

$$= \sin \varphi |x_0\rangle + \cos \varphi |x_0^\perp\rangle$$

$$\Rightarrow \frac{\sin \varphi}{\cos \varphi} = \frac{\frac{1}{\sqrt{N}}}{\sqrt{\frac{N-1}{N}}} = \frac{1}{\sqrt{N-1}}$$

$$\Rightarrow \text{for large } N, \quad \varphi \approx \frac{1}{\sqrt{N}}.$$

$$\Rightarrow \text{Need } k \approx \frac{\pi}{4} \cdot \sqrt{N} \text{ Grover iterations.}$$

$$\Rightarrow O(\sqrt{N}) \text{ calls to } f \text{ (for } O_f) \text{ sufficient!}$$

Quadratic speed-up with respect to classical
algorithms for general search problems!

Note:

- If there are $k > 1$ solutions:

Same method with $O\left(\sqrt{\frac{N}{k}}\right)$ steps works
(\rightarrow homework)

- Can also be adapted to case where number of solutions is unknown.