

Grover's Algorithm and its application

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Problem

Given a function $f(x): \{0,1\}^n \rightarrow \{0,1\}$, find a n -bit string x^* such that $f(x^*) = 1$.

Let $N = 2^n$.

Requires $O(N)$ function calls in the classical model.

Grover's Algorithm (1996)

Requires $\Theta(\sqrt{N})$ function calls in the quantum model.

Recap

Recap – complex number

Complex number. $z = a + bi$ where a and b are real numbers.

- $a = \operatorname{Re}(z)$ is the *real part* of z

- $b = \operatorname{Im}(z)$ is the *imaginary part* of z

- $z^* := a - bi$ is the *conjugate* of z .

- $|z| = \sqrt{\operatorname{Re}(z)^2 + \operatorname{Im}(z)^2} = \sqrt{a^2 + b^2}$ is the *magnitude* of z .

Observation. $|z|^2 = (a + bi)(a - bi) = z^*z$.

Recap – qubit

The *Qubit* (short for *quantum bit*). $|\phi\rangle = \alpha|0\rangle + \beta|1\rangle$

where α and β are complex numbers such that $|\alpha|^2 + |\beta|^2 = 1$.

Superposition. Measuring $|\phi\rangle$ will yield either zero w/ probability $|\alpha|^2$ or one w/ probability $|\beta|^2$.

The state of the qubit $|\phi\rangle$ is two-dimensional complex vector $\begin{pmatrix} \alpha \\ \beta \end{pmatrix}$.

$\langle\phi| := (\alpha \ \beta)^* = (\alpha^* \ \beta^*)$, i.e., the conjugate transpose of $\begin{pmatrix} \alpha \\ \beta \end{pmatrix}$.

Recap – systems of qubit

Systems of Qubit. (Tensor product or Kronecker product)

$$|\phi_1\phi_2\rangle = |\phi_1\rangle \otimes |\phi_2\rangle = \begin{pmatrix} \alpha_1 \\ \beta_1 \end{pmatrix} \otimes \begin{pmatrix} \alpha_2 \\ \beta_2 \end{pmatrix} = \begin{pmatrix} \alpha_1\alpha_2 \\ \alpha_1\beta_2 \\ \beta_1\alpha_2 \\ \beta_1\beta_2 \end{pmatrix} = \alpha_1\alpha_2|00\rangle + \alpha_1\beta_2|01\rangle + \beta_1\alpha_2|10\rangle + \beta_1\beta_2|11\rangle$$

The state below is *entangled* i.e., not separable.

$$|\phi\rangle = \frac{1}{\sqrt{2}}(|00\rangle + |11\rangle)$$

The state $|\phi'\rangle = \frac{1}{\sqrt{2}}(|01\rangle + |11\rangle)$ is not entangled. Since

$$|\phi'\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 \\ 1 \\ 0 \\ 1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \frac{|0\rangle + |1\rangle}{\sqrt{2}} \otimes |1\rangle$$

Recap – qubits are just vectors

Inner product. $\langle \phi_1 | \phi_2 \rangle = (\alpha_1 \ \beta_1)^* \begin{pmatrix} \alpha_2 \\ \beta_2 \end{pmatrix} = \alpha_1^* \alpha_2 + \beta_1^* \beta_2$ where $(\alpha_1 \ \beta_1)^* = (\alpha_1^* \ \beta_1^*)$.

Outer product. $|\phi_1\rangle\langle\phi_2| = \begin{pmatrix} \alpha_1 \\ \beta_1 \end{pmatrix} (\alpha_2 \ \beta_2)^* = \begin{pmatrix} \alpha_1 \alpha_2^* & \alpha_1 \beta_2^* \\ \beta_1 \alpha_2^* & \beta_1 \beta_2^* \end{pmatrix}$ $|\phi_1\rangle\langle\phi_2|$: a matrix with mapping $|\phi_1\rangle \rightarrow |\phi_2\rangle$

Exercise (expressing matrix).

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

= Mapping $\{|0\rangle \rightarrow |0\rangle, |1\rangle \rightarrow -|1\rangle\}$

$$-|00\rangle\langle 00| + |01\rangle\langle 01| + |10\rangle\langle 11| + |11\rangle\langle 10| = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

= Mapping $\{|00\rangle \rightarrow |00\rangle, |01\rangle \rightarrow |01\rangle, |10\rangle \rightarrow |11\rangle, |11\rangle \rightarrow |10\rangle\}$

Recap – unitary matrix

Unitary Matrix. The matrix U is unitary if $UU^\dagger = U^\dagger U = I$ where U^\dagger is the transposed conjugate of U .

- $U^\dagger := U^{*T}$ is sometimes called Hermitian conjugate matrix or adjoint matrix.

Unitary Transformation. Change of the state is done by a series of unitary transformations.

Basic unitary transformations are called *gates*.

Unitarity implies

1. #input qubits = #output qubits
2. Reversible

Recap – common gates (one-qubit gates)

Not. NOT = $X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$

- $\alpha|0\rangle + \beta|1\rangle \rightarrow \beta|0\rangle + \alpha|1\rangle$

Hadamard. $H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$

Exercise.

- $|0\rangle \rightarrow \frac{|0\rangle + |1\rangle}{\sqrt{2}}$

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

Recap – common gates (multi-qubit gates)

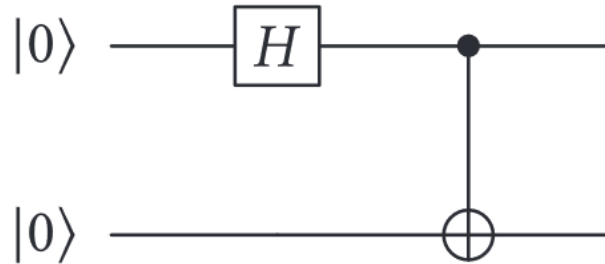
$$\text{Controlled-NOT, CNOT} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$- \alpha_1 \alpha_2 |00\rangle + \alpha_1 \beta_2 |01\rangle + \beta_1 \alpha_2 |10\rangle + \beta_1 \beta_2 |11\rangle \rightarrow \alpha_1 \alpha_2 |00\rangle + \alpha_1 \beta_2 |01\rangle + \beta_1 \beta_2 |10\rangle + \beta_1 \alpha_2 |11\rangle$$

$$\text{Observe. } \text{CNOT} = |00\rangle\langle 00| + |01\rangle\langle 01| + |10\rangle\langle 11| + |11\rangle\langle 10| = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

Recap – applying transformations

CNOT($H \otimes I$)



$$1. (H \otimes I)(|0\rangle \otimes |0\rangle) = \frac{|0\rangle + |1\rangle}{\sqrt{2}} \otimes |0\rangle = \frac{|00\rangle + |10\rangle}{\sqrt{2}}$$

$$2. \text{CNOT} \frac{|00\rangle + |10\rangle}{\sqrt{2}} = \frac{1}{\sqrt{2}} (\text{CNOT}|00\rangle + \text{CNOT}|10\rangle) = \frac{1}{\sqrt{2}} (|00\rangle + |11\rangle) = \frac{|00\rangle + |11\rangle}{\sqrt{2}}$$

Bell state $|\beta_{00}\rangle = \frac{|00\rangle + |11\rangle}{\sqrt{2}}$. (Note that its state is entangled.)

Quantum Oracle

Problem

Given a function $f(\mathbf{x}): \{0,1\}^n \rightarrow \{0,1\}$, find a n -bit string \mathbf{x}^* such that $f(\mathbf{x}^*) = 1$.

Let $N = 2^n$.

Requires $O(N)$ function calls in the classical model.

Grover's Algorithm

Requires $\Theta(\sqrt{N})$ function calls in the quantum model.

“Function” should be something like a quantum gate... and note that it must be **unitary**...

Quantum Oracle for Unary Function

Suppose we are given a black box unary function $f(x): \{0,1\}^n \rightarrow \{0,1\}$.

Using this black box, one can build a new (unitary) gate that

- takes $(n + 1)$ -bits in and $(n + 1)$ -bits out,
- computes f when proper input is given, and
- has the same computational complexity.

How?

$$|x\rangle|y\rangle \xrightarrow{U_f} |x\rangle|f(x) \oplus y\rangle$$

where \oplus is integer mod-2.

x	y	$x \oplus y$
0	0	0
0	1	1
1	0	1
1	1	0

Quantum Oracle for Unary Function

Suppose we are given a constant function $f(x): \{0,1\} \rightarrow 1$.

$$|x\rangle|y\rangle \xrightarrow{U_f} |x\rangle|1 \oplus y\rangle$$

$ x\rangle y\rangle$	$U_f(x\rangle y\rangle)$
$ 0\rangle 0\rangle$	$ 0\rangle 1\rangle$
$ 0\rangle 1\rangle$	$ 0\rangle 0\rangle$
$ 1\rangle 0\rangle$	$ 1\rangle 1\rangle$
$ 1\rangle 1\rangle$	$ 1\rangle 0\rangle$

How does U_f look?

$$|00\rangle\langle 01| + |01\rangle\langle 00| + |10\rangle\langle 11| + |11\rangle\langle 10| = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{pmatrix} \sigma_x: \text{Pauli matrix}$$

Unitary?

- A square matrix is unitary if it can be broken down into smaller unitary matrices along its diagonal.

Quantum Oracle for Unary Function

Suppose we are given a unary function $f(x): \{0,1\}^n \rightarrow \{0,1\}$.

Consider when $x = 00 \dots 0$.

$$|x\rangle|y\rangle \xrightarrow{U_f} |x\rangle|f(x) \oplus y\rangle$$

$ x\rangle y\rangle$	$f(x)$	$U_f(x\rangle y\rangle)$
$ 00 \dots 0\rangle 0\rangle$	0	$ 00 \dots 0\rangle 0\rangle$
$ 00 \dots 0\rangle 1\rangle$	0	$ 00 \dots 0\rangle 1\rangle$
$ 00 \dots 0\rangle 0\rangle$	1	$ 00 \dots 0\rangle 1\rangle$
$ 00 \dots 0\rangle 1\rangle$	1	$ 00 \dots 0\rangle 0\rangle$

Quantum Oracle for Unary Function

Suppose we are given a unary function $f(x): \{0,1\}^n \rightarrow \{0,1\}$.

Consider when $x = 00 \dots 0$.

$$|x\rangle|y\rangle \xrightarrow{U_f} |x\rangle|f(x) \oplus y\rangle$$

$ x\rangle y\rangle$	$f(x)$	$U_f(x\rangle y\rangle)$
$ 00 \dots 0\rangle 0\rangle$	0	$ 00 \dots 0\rangle 0\rangle$
$ 00 \dots 0\rangle 1\rangle$	0	$ 00 \dots 0\rangle 1\rangle$
$ 00 \dots 0\rangle 0\rangle$	1	$ 00 \dots 0\rangle 1\rangle$
$ 00 \dots 0\rangle 1\rangle$	1	$ 00 \dots 0\rangle 0\rangle$

When $f(x) = 0$, U_f looks like

$$\left(\begin{array}{cc|c} 1 & 0 & ? \\ 0 & 1 & \\ \hline 0 & 0 & \\ \vdots & \vdots & ? \\ 0 & 0 & \end{array} \right)$$

Quantum Oracle for Unary Function

Suppose we are given a unary function $f(x): \{0,1\}^n \rightarrow \{0,1\}$.

Consider when $x = 00 \dots 0$.

$$|x\rangle|y\rangle \xrightarrow{U_f} |x\rangle|f(x) \oplus y\rangle$$

$ x\rangle y\rangle$	$f(x)$	$U_f(x\rangle y\rangle)$
$ 00 \dots 0\rangle 0\rangle$	0	$ 00 \dots 0\rangle 0\rangle$
$ 00 \dots 0\rangle 1\rangle$	0	$ 00 \dots 0\rangle 1\rangle$
$ 00 \dots 0\rangle 0\rangle$	1	$ 00 \dots 0\rangle 1\rangle$
$ 00 \dots 0\rangle 1\rangle$	1	$ 00 \dots 0\rangle 0\rangle$

When $f(x) = 1$, U_f looks like

$$\left(\begin{array}{cc|c} 0 & 1 & ? \\ 1 & 0 & \\ \hline 0 & 0 & \\ \vdots & \vdots & ? \\ 0 & 0 & \end{array} \right)$$

Quantum Oracle for Unary Function

Suppose we are given a unary function $f(x): \{0,1\}^n \rightarrow \{0,1\}$.

$$|x\rangle|y\rangle \xrightarrow{U_f} |x\rangle|f(x) \oplus y\rangle$$

Then U_f looks like

$$\begin{pmatrix} [I \text{ or } \sigma_x] & & & & \\ & [I \text{ or } \sigma_x] & & & \\ & & \ddots & & \\ & & & [I \text{ or } \sigma_x] & \\ & & & & [I \text{ or } \sigma_x] \end{pmatrix}$$

and it is unitary.

Quantum Algorithm's Complexity

Relativized Time Complexity

- The time complexity **without** knowledge of the oracle's design. *(if we allow small error, no speed-up.)*
- Deutsch-Jozsa's algorithm offers a deterministic exponential speed-up *relative to the oracle*.
- Bernstein-Vazirani's algorithm offers a polynomial speed-up *relative to the oracle*. *(even if small error is allowed)*

Absolute Time Complexity

- The time complexity **with** knowledge of the oracle's design
- Shor's algorithm provides *absolute* speed-up.

Problem

Given a function $f(x): \{0,1\}^n \rightarrow \{0,1\}$, find the **n -bit target string** x^* such that $f(x^*) = 1$.

Let $N = 2^n$.

Requires $O(N)$ function calls in the classical model.

Given a quantum oracle O_f , find a (n -bit) target string x^* .

Grover's Algorithm. Requires $\Theta(\sqrt{N})$ calls to the **quantum oracle**.

Grover's Algorithm

Grover operator G

Let $|\psi\rangle := \frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + |00 \dots 10\rangle + \dots + |11 \dots 11\rangle)$ be the uniform superposition.

(Shorthand) $|\psi\rangle = \frac{1}{\sqrt{N}} \sum_{i=0}^{N-1} |i\rangle = \frac{1}{\sqrt{N}} \sum_i |i\rangle$

Grover operator $G := ((2|\psi\rangle\langle\psi| - I_N) \otimes I_2) O_f$

$$|\psi\rangle\langle\psi| = \frac{1}{N} \begin{pmatrix} 1 & 1 & 1 & \dots & 1 \\ 1 & 1 & 1 & \dots & 1 \\ 1 & 1 & 1 & \dots & 1 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & 1 & 1 & \dots & 1 \end{pmatrix}$$

The action of $2|\psi\rangle\langle\psi| - I_N$ on an arbitrary state $|\phi\rangle = \sum_i a_i |i\rangle = \begin{pmatrix} a_0 \\ a_1 \\ \vdots \\ a_{N-1} \end{pmatrix}$

$$I_N = \begin{pmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 1 \end{pmatrix}$$

$$(2|\psi\rangle\langle\psi| - I_N)|\phi\rangle = \sum_i \left(2 \frac{a_0 + \dots + a_{N-1}}{N} - a_i \right) |i\rangle$$

$$I_2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Grover's Algorithm

Step 1. Perform state initialization

- (n qubits) $|00 \dots 0\rangle \rightarrow \frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + |00 \dots 10\rangle + \dots + |11 \dots 11\rangle)$

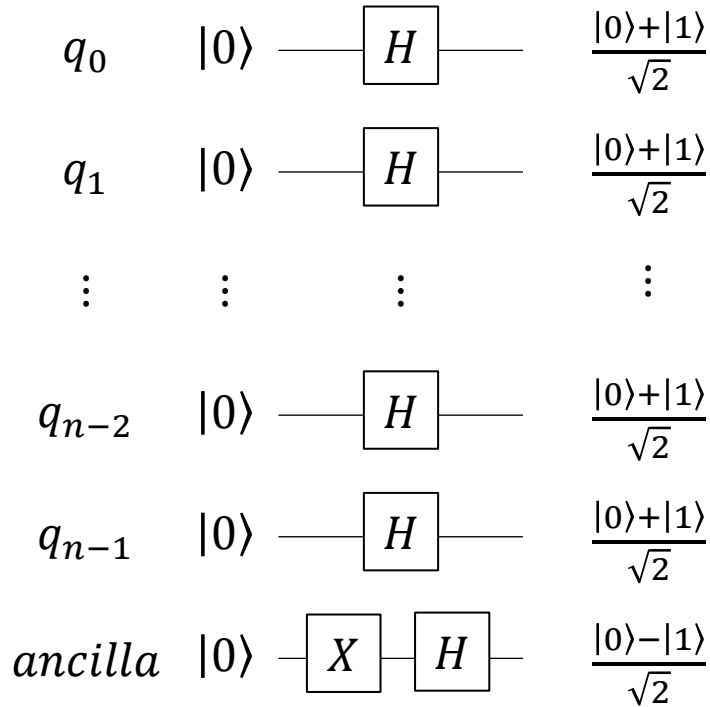
- (ancillary qubit) $|0\rangle \rightarrow \frac{|0\rangle - |1\rangle}{\sqrt{2}}$

Step 2. Apply Grover operator $\left\lceil \frac{\pi\sqrt{N}}{4} \right\rceil$ times

Step 3. Perform measurement on all qubit (except the ancillary qubit)

Grover's Algorithm

Step 1. Initialization



$$\begin{aligned}
 & \frac{|0\rangle+|1\rangle}{\sqrt{2}} \otimes \frac{|0\rangle+|1\rangle}{\sqrt{2}} \otimes \dots \otimes \frac{|0\rangle+|1\rangle}{\sqrt{2}} \\
 &= \frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + |00 \dots 10\rangle + \dots + |11 \dots 11\rangle)
 \end{aligned}$$

$$X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

$$- |0\rangle \rightarrow \frac{|0\rangle+|1\rangle}{\sqrt{2}}, |1\rangle \rightarrow \frac{|0\rangle-|1\rangle}{\sqrt{2}}$$

$$\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + |00 \dots 10\rangle + \dots + |11 \dots 11\rangle) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}}$$

Grover's Algorithm

Step 2. Apply $G := ((2|\psi\rangle\langle\psi| - I_N) \otimes I_2) O_f$

$$|x\rangle|q\rangle \rightarrow |x\rangle|f(x) \oplus q\rangle$$

$$\begin{aligned} & O_f \left(\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + |x^*\rangle + \dots + |11 \dots 11\rangle) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \\ &= \frac{1}{\sqrt{N}} \left(O_f \left(|00 \dots 00\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) + \dots + \underbrace{O_f \left(|x^*\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right)} + \dots + O_f \left(|11 \dots 11\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \right) \\ &= O_f \frac{|x^*\rangle|0\rangle - |x^*\rangle|1\rangle}{\sqrt{2}} = \frac{O_f(|x^*\rangle|0\rangle) - O_f(|x^*\rangle|1\rangle)}{\sqrt{2}} \\ &= \frac{|x^*\rangle|1\rangle - |x^*\rangle|0\rangle}{\sqrt{2}} \\ &= |x^*\rangle \otimes \frac{|1\rangle - |0\rangle}{\sqrt{2}} \end{aligned}$$

Grover's Algorithm

Step 2. Apply $G := ((2|\psi\rangle\langle\psi| - I_N) \otimes I_2) O_f$

$$|x\rangle|q\rangle \rightarrow |x\rangle|f(x) \oplus q\rangle$$

$$\begin{aligned} & O_f \left(\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + |x^*\rangle + \dots + |11 \dots 11\rangle) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \\ &= \frac{1}{\sqrt{N}} \left(O_f \left(|00 \dots 00\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) + \dots + O_f \left(|x^*\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) + \dots + O_f \left(|11 \dots 11\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \right) \\ &= \frac{1}{\sqrt{N}} \left(|00 \dots 00\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} + \dots + |x^*\rangle \otimes \frac{-|0\rangle + |1\rangle}{\sqrt{2}} + \dots + |11 \dots 11\rangle \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \\ &= \frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + (-\mathbf{1})|x^*\rangle + \dots + |11 \dots 11\rangle) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \end{aligned}$$

Grover's Algorithm

Step 2. Apply $G := ((2|\psi\rangle\langle\psi| - I_N) \otimes I_2)O_f$

$$\begin{aligned} & ((2|\psi\rangle\langle\psi| - I_N) \otimes I_2) \left(\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + (-1)|x^*\rangle + \dots + |11 \dots 11\rangle) \right) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \\ &= (2|\psi\rangle\langle\psi| - I_N) \left(\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + (-1)|x^*\rangle + \dots + |11 \dots 11\rangle) \right) \otimes \left(I_2 \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \\ &= (2|\psi\rangle\langle\psi| - I_N) \left(\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + (-1)|x^*\rangle + \dots + |11 \dots 11\rangle) \right) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \end{aligned}$$

Grover's Algorithm

Step 2. Apply $G := (2|\psi\rangle\langle\psi| - I_N) \otimes I_2 O_f$

$$(2|\psi\rangle\langle\psi| - I_N)|\phi\rangle = \sum_i \left(\frac{2}{N} (a_0 + \dots + a_{N-1}) - a_i \right) |i\rangle$$

$$(2|\psi\rangle\langle\psi| - I_N) \left(\frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + \dots + (-1)|x^*\rangle + \dots + |11 \dots 11\rangle) \right)$$

$$\frac{N-2}{\sqrt{N}}$$

$$= \left(\frac{2N-2}{N} \frac{1}{\sqrt{N}} - \frac{1}{\sqrt{N}} \right) |00 \dots 00\rangle + \dots + \left(\frac{2N-2}{N} \frac{1}{\sqrt{N}} + \frac{1}{\sqrt{N}} \right) |x^*\rangle + \dots + \left(\frac{2N-2}{N} \frac{1}{\sqrt{N}} - \frac{1}{\sqrt{N}} \right) |11 \dots 11\rangle$$

$$= \frac{1}{\sqrt{N}} \left(\frac{N-4}{N} |00 \dots 00\rangle + \dots + \frac{3N-4}{N} |x^*\rangle + \dots + \frac{N-4}{N} |11 \dots 11\rangle \right)$$

amplified

Grover's Algorithm

Step 2. Apply $G := ((2|\psi\rangle\langle\psi| - I_N) \otimes I_2) O_f$ again

$$|x\rangle|q\rangle \rightarrow |x\rangle|f(x) \oplus q\rangle$$

$$\begin{aligned} & O_f \left(\frac{1}{\sqrt{N}} \left(\frac{N-4}{N} |00 \dots 00\rangle + \dots + \frac{3N-4}{N} |x^*\rangle + \dots + \frac{N-4}{N} |11 \dots 11\rangle \right) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \right) \\ &= \frac{1}{\sqrt{N}} \left(\frac{N-4}{N} |00 \dots 00\rangle + \dots + \underbrace{(-1) \frac{3N-4}{N} |x^*\rangle}_{\text{flipped}} + \dots + \frac{N-4}{N} |11 \dots 11\rangle \right) \otimes \frac{|0\rangle - |1\rangle}{\sqrt{2}} \end{aligned}$$

Grover's Algorithm

Step 2. Apply $G := (2|\psi\rangle\langle\psi| - I_N) \otimes I_2 O_f$ again

$$(2|\psi\rangle\langle\psi| - I_N)|\phi\rangle = \sum_i \left(\frac{2}{N} (a_0 + \dots + a_{N-1}) - a_i \right) |i\rangle$$

$$(2|\psi\rangle\langle\psi| - I_N) \left(\frac{1}{\sqrt{N}} \left(\frac{N-4}{N} |00\dots 00\rangle + \dots + (-1)^i \frac{3N-4}{N} |x^*\rangle + \dots + \frac{N-4}{N} |11\dots 11\rangle \right) \right)$$
$$= \frac{1}{\sqrt{N}} \left(\frac{N^2 - 12N + 16}{N^2} |00\dots 00\rangle + \dots + \frac{5N^2 - 20N + 16}{N^2} |x^*\rangle + \dots + \frac{N^2 - 12N + 16}{N^2} |11\dots 11\rangle \right)$$

more amplified

Grover's Algorithm

Step 2. Apply G fixed amount

(informally) $\frac{1}{\sqrt{N}} (\epsilon |00 \cdots 00\rangle + \cdots + \underbrace{(\sqrt{N} - \epsilon')}_{\text{amplified a lot}} |x^*\rangle + \cdots + \epsilon |11 \cdots 11\rangle)$

for some small ϵ, ϵ' .

Grover's Algorithm

Step 3. Measurement

$$\text{(informally)} \quad \frac{1}{\sqrt{N}} (\epsilon |00 \cdots 00\rangle + \cdots + (\sqrt{N} - \epsilon') |x^*\rangle + \cdots + \epsilon |11 \cdots 11\rangle)$$

Obtain $|x^*\rangle$ with probability close to 1.

Analysis

Presented by Changyeol Lee

Grover's Algorithm, Geometric Explanation

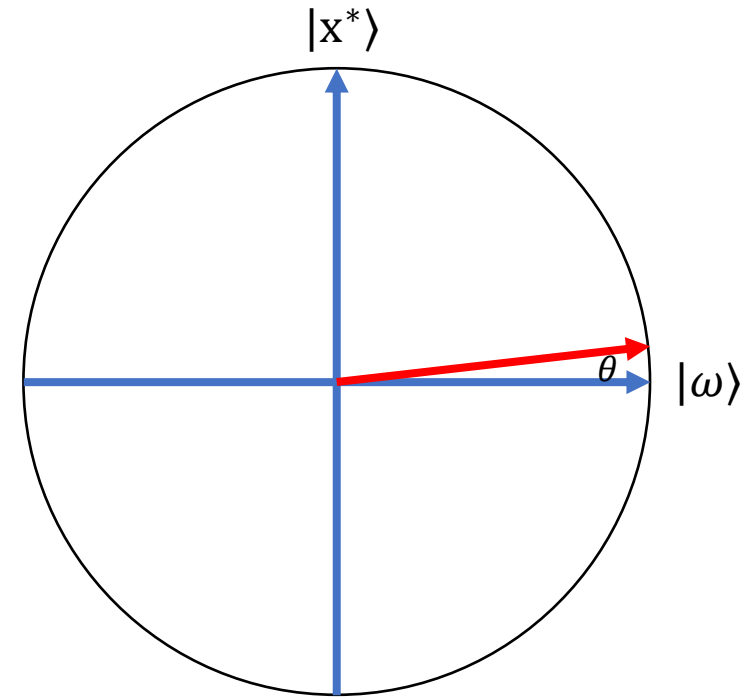
Why applying Grover operator (exactly) $\left\lceil \frac{\pi\sqrt{N}}{4} \right\rceil$ times?

$$\text{Let } |\omega\rangle = \frac{1}{\sqrt{N-1}} (\sum_i |i\rangle - |x^*\rangle)$$

Note that $|\omega\rangle$ and $|x^*\rangle$ are orthonormal.

Note. After the step 1, the state is

$$\begin{aligned} & \frac{1}{\sqrt{N}} (|00 \dots 00\rangle + |00 \dots 01\rangle + |00 \dots 10\rangle + \dots + |11 \dots 1\rangle) \\ &= \frac{\sqrt{N-1}}{\sqrt{N}} |\omega\rangle + \frac{1}{\sqrt{N}} |x^*\rangle \\ &= \cos \theta |\omega\rangle + \sin \theta |x^*\rangle \end{aligned}$$

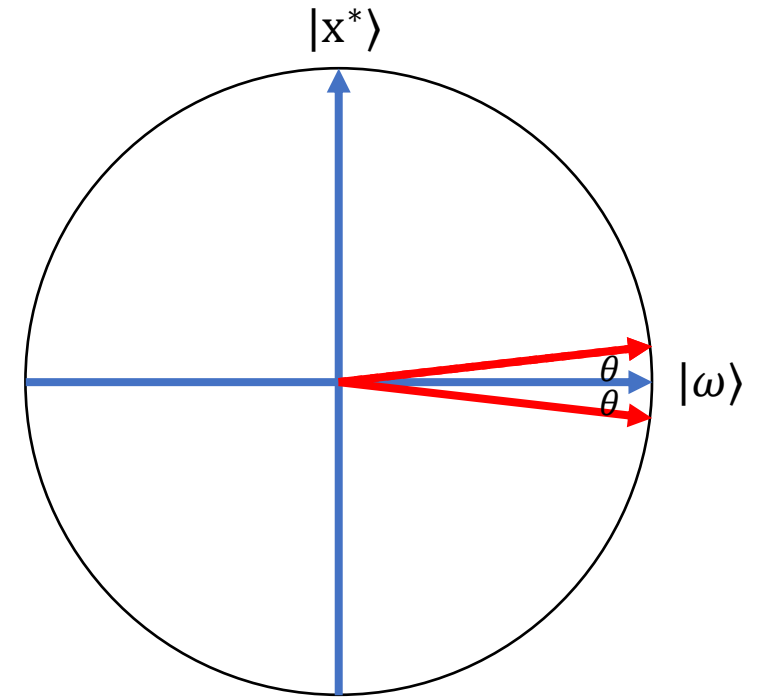


Grover's Algorithm, Geometric Explanation

What happens we apply O_f ?

$$\cos \theta |\omega\rangle + \sin \theta |x^*\rangle \rightarrow \cos \theta |\omega\rangle - \sin \theta |x^*\rangle$$

Applying O_f = Reflection about $|\omega\rangle$



Grover's Algorithm, Geometric Explanation

What happens we apply $(2|\psi\rangle\langle\psi| - I_N)$?

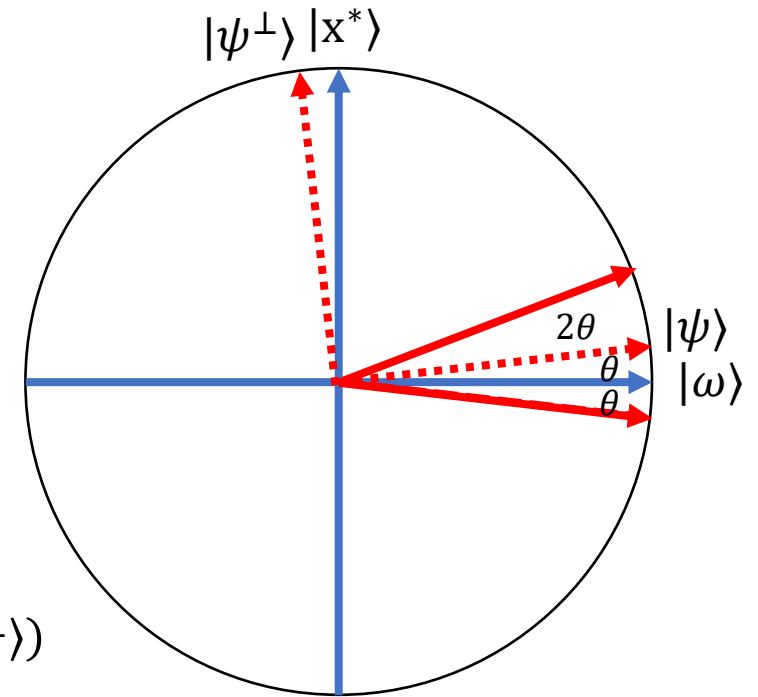
Any state $|\phi\rangle$ of this plane can be decomposed into

$$|\phi\rangle = \alpha|\psi\rangle + \beta|\psi^\perp\rangle$$

Then,

$$\begin{aligned} & (2|\psi\rangle\langle\psi| - I_N)|\phi\rangle \\ &= 2|\psi\rangle\langle\psi|(\alpha|\psi\rangle + \beta|\psi^\perp\rangle) - (\alpha|\psi\rangle + \beta|\psi^\perp\rangle) \\ &= 2\alpha|\psi\rangle\langle\psi|\psi\rangle + 2\beta|\psi\rangle\langle\psi|\psi^\perp\rangle - (\alpha|\psi\rangle + \beta|\psi^\perp\rangle) \\ &= 2\alpha|\psi\rangle - (\alpha|\psi\rangle + \beta|\psi^\perp\rangle) \\ &= \alpha|\psi\rangle - \beta|\psi^\perp\rangle \end{aligned}$$

Applying $(2|\psi\rangle\langle\psi| - I_N) =$ Reflection about $|\psi\rangle$



$|\psi\rangle\langle\psi|$: a mapping $|\psi\rangle \rightarrow |\psi\rangle$

inner product $\langle\psi||\psi^\perp\rangle = 0$

Grover's Algorithm, Geometric Explanation

After first iteration,

$$\cos \theta |\omega\rangle + \sin \theta |x^*\rangle \rightarrow \cos 3\theta |\omega\rangle + \sin 3\theta |x^*\rangle$$

After each iteration,

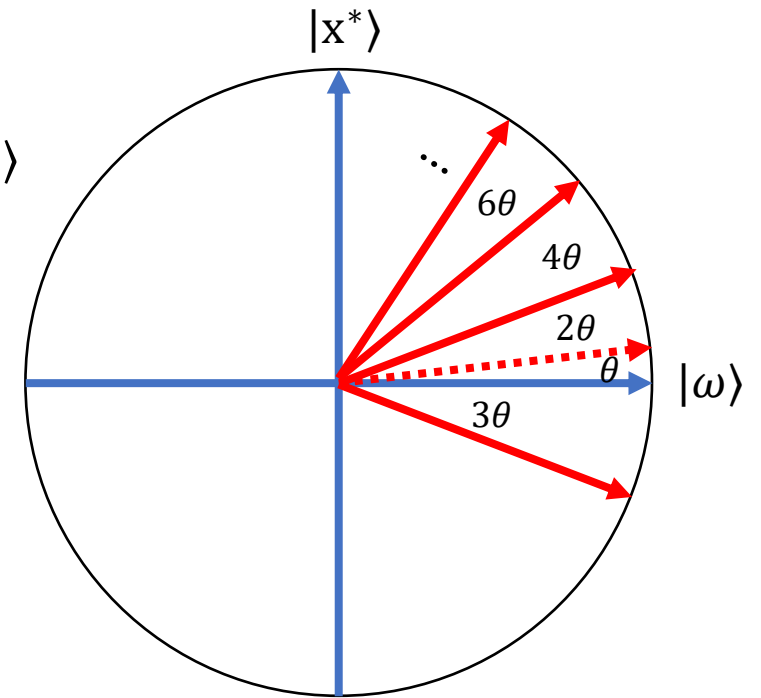
$$\cos 5\theta |\omega\rangle + \sin 5\theta |x^*\rangle$$

$$\cos 7\theta |\omega\rangle + \sin 7\theta |x^*\rangle$$

⋮

After applying k times,

$$\cos(\theta + 2k\theta) |\omega\rangle + \sin(\theta + 2k\theta) |x^*\rangle$$



Grover's Algorithm, Geometric Explanation

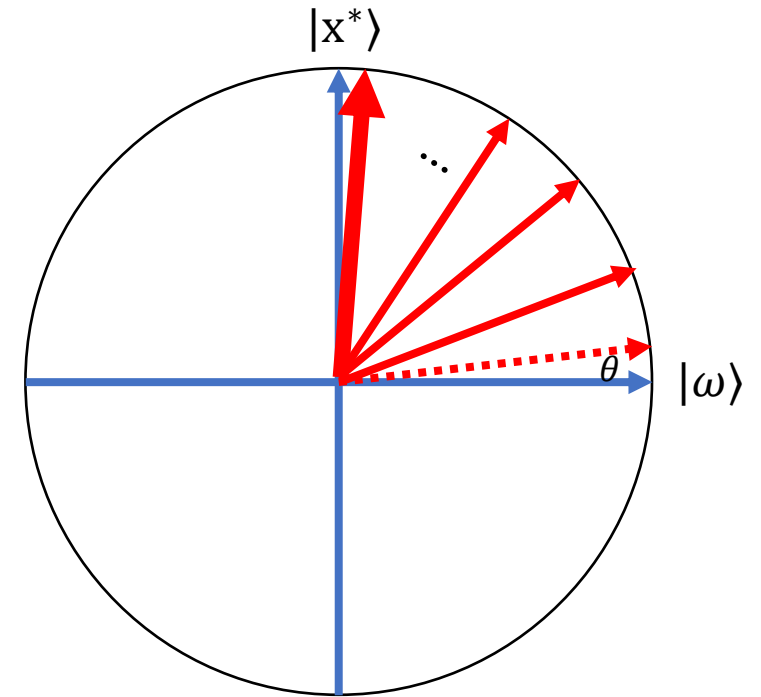
Recall $\frac{\sqrt{N-1}}{\sqrt{N}} |\omega\rangle + \frac{1}{\sqrt{N}} |x^*\rangle = \cos \theta |\omega\rangle + \sin \theta |x^*\rangle$.

$$-\theta = \arccos \sqrt{\frac{N-1}{N}}$$

Find k that maximizes $\sin^2(\theta + 2k\theta) = \sin^2\left((2k+1) \arccos \sqrt{\frac{N-1}{N}}\right)$

or find k such that $\frac{\pi}{2} \sim (2k+1) \arccos \sqrt{\frac{N-1}{N}}$

$$k_{\text{optimal}} = \frac{\pi}{4} \sqrt{N} - \frac{1}{2} - O(\sqrt{1/N})$$



Applications

Factoring

Factor an integer $M = p \times q$ into p and q where p and q are primes ($2^{2n-1} \leq M < 2^{2n}$ for some n)

WLOG, $p \leq \sqrt{M} < 2^n$.

Given a function $f(x)$ that outputs 1 if $x = p$; 0 otherwise

Given a function $f(x): \{0,1\}^n \rightarrow \{0,1\}$, find the target string p .

Grover's Algorithm.

Requires $\frac{\pi}{4} \sqrt{N} - \frac{1}{2} - O(\sqrt{1/N})$ calls to the quantum oracle. ($N = 2^n$)

Shor's algorithm runs in $O(n^3 \log n)$ times, using $O(n^2 \log n \log \log n)$ gates.

Multiple Target Strings

If there are t number of target strings, by calling $k_{optimal} = \frac{\pi}{4}\sqrt{N/t} - \frac{1}{2} - O(\sqrt{t/N})$

or $\Theta(\sqrt{N/t})$ calls to the oracle, we can find a target string.

What if the number of target strings t is unknown?

Theorem (Boyer et al., 1996).

There is an (randomized) algorithm that find a target string in expected time in $O(\sqrt{N/t})$.

Minimum Searching

Given a function $F(x): \{0,1\}^n \rightarrow \mathbb{R}$, find a n -bit string x^* that minimizes $F(x^*)$.

- We are given an oracle that "compares" F with some constant c .

E.g.) $F(0) = 9, F(1) = 5, F(2) = 6, F(3) = 7$ (Here, $n = 2$.)

Suppose $c = F(3) = 7$.

Then, $f(0) = 0, f(1) = 1, f(2) = 1, f(3) = 0$.

Construct a Grover-Search-Problem instance, i.e., given a function $f(x): \{0,1\}^n \rightarrow \{0,1\}$, find a n -bit string x^* such that $f(x^*) = 1$ where the number of target strings, say t , is unknown.

$\Rightarrow O(\sqrt{N/t})$ calls to the oracle (in expectation).

E.g.) output = 2

Minimum Searching

Given a function $F(x): \{0,1\}^n \rightarrow \mathbb{R}$, find a n -bit string x^* that minimizes $F(x^*)$.

- We are given an oracle that "compares" F with some constant c .

E.g.) $F(0) = 9, F(1) = 5, F(2) = 6, F(3) = 7$ (Here, $n = 2$.)

Suppose $c = \mathbf{F(2)} = 6$.

Then, $f(0) = 0, f(1) = 1, f(2) = 0, f(3) = 0$.

Construct a Grover-Search-Problem instance, i.e., given a function $f(x): \{0,1\}^n \rightarrow \{0,1\}$, find a n -bit string x^* such that $f(x^*) = 1$ where the number of target strings, say t , is unknown.

$\Rightarrow O(\sqrt{N/t})$ calls to the oracle (in expectation).

E.g.) output = 1

Minimum Searching

Given a function $F(x): \{0,1\}^n \rightarrow \mathbb{R}$, find a n -bit string x^* that minimizes $F(x^*)$.

- We are given an oracle that "compares" F with some constant c .

E.g.) $F(0) = 9, F(1) = 5, F(2) = 6, F(3) = 7$ (Here, $n = 2$.)

Suppose $c = F(\mathbf{1}) = 5$.

Then, $f(0) = 0, f(1) = 0, f(2) = 0, f(3) = 0$.

...

Minimum Searching

Given a function $F(x): \{0,1\}^n \rightarrow \mathbb{R}$, find a n -bit string x^* that minimizes $F(x^*)$.

- We are given an oracle that "compares" F with some constant c .

Step 1. Pick an index j uniformly at random among $\{0,1, \dots, N - 1\}$

Step 2. Repeat the following until the total running time is more than $\Theta(\sqrt{N})$:

Let O_f be the comparison oracle with $c = F(j)$

Construct a Grover-Search-Problem instance and run the algorithm.

Let j' be the output. Update j to j' if $F(j') < F(j)$.

Theorem (Dürr and Høyer, 1996).

This algorithm finds the index of the minimum value of F with probability at least $\frac{1}{2}$ and runs in $\Theta(\sqrt{N})$.

Minimum Searching with Different Types

Given a function $F(\mathbf{x}): \{0,1\}^n \rightarrow \mathbb{R}$ and an onto function $\text{type}(\mathbf{x}): \{0,1\}^n \rightarrow \{1,2,\dots,d\}$,

find $\{x_1^*, \dots, x_d^*\}$ where $x_i^* = \operatorname{argmin}_{\text{type}(\mathbf{x})=i} F(\mathbf{x})$ for each type $i = 1, 2, \dots, d$.

Naïve application of the previous algorithm – run d times for each type, having $\Theta(d\sqrt{N})$ running time.

Theorem (Dürr et al., 2006).

There is an algorithm solves the problem with probability at least $\frac{1}{2}$ and runs in $O(\sqrt{dN})$.

Minimum Spanning Tree

Quantumize the classical (Boruvka's algorithm) MST algorithm

input : Adjacency list-array implementation of undirected $G = (V, E)$ with cost function $c: E \rightarrow \mathbb{R}$

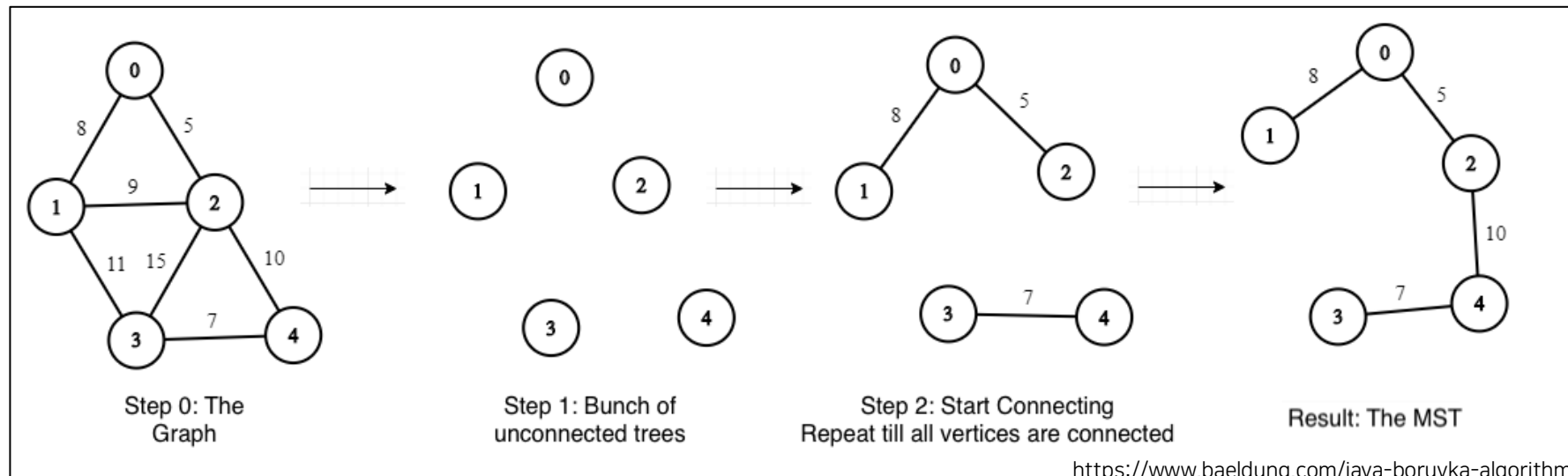
Initialization $\mathcal{T} = \{\{1\}, \{2\}, \dots, \{|V|\}\}$

Repeat until $|\mathcal{T}| = 1$:

Let $\mathcal{T} = \{T_1, \dots, T_k\}$. Find e_1, \dots, e_k where e_i is a minimum cost edge leaving T_i .

Merge T_i and e_i for each i and update \mathcal{T} .

Return T_1



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Return T_1

Consider the directed version of the graph, i.e., $(u, v) \rightarrow \langle u, v \rangle, \langle v, u \rangle$

$F(\langle u, v \rangle) = c((u, v))$ if $\langle u, v \rangle$ is leaving some tree and $F(\langle u, v \rangle) = \infty$ if not.

$\text{type}(\langle u, v \rangle) = i$ such that $u \in T_i$

Then, apply the algorithm for the **Minimum Searching with Different Types** (with *more* queries).

Minimum Spanning Tree

Let $n = |V|$ and $m = |E|$.

On ℓ^{th} iteration, queries $(\ell + 2)O(\sqrt{mk})$ times.

#of queries

- Observe that $k \leq n/2^{\ell-1}$.

$$\sum_{\ell \geq 1} (\ell + 2)O(\sqrt{mk}) \leq \sum_{\ell \geq 1} (\ell + 2)O\left(\sqrt{\frac{mn}{2^{\ell-1}}}\right) = O(\sqrt{nm})$$

Error probability at most

$$\sum_{\ell \geq 1} \frac{1}{2^{\ell+2}} \leq \frac{1}{4}$$

Other applications

Connectivity

Network flow problem. (Finding Max Flow)

Matching on graph

Graph coloring

3-SAT

⋮

approximation algorithms

Thank you