

Practical Quantum Computing: Grover's algorithm

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14/04/2026 Pasteura 5, Warszawa

Plan

- 1 Quantum circuit model of computation
- 2 Quantum algorithm: formulation
- 3 Implementation of Grover's algorithm

Sources of quantum algorithms:

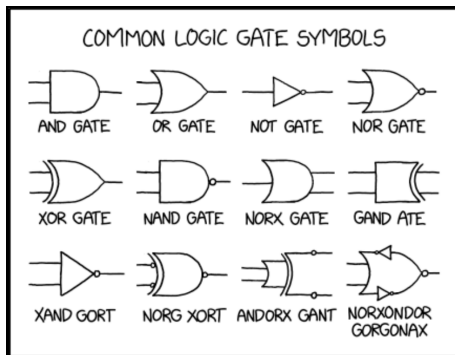
- **Search algorithms** — e.g., Grover's algorithm
- **Fourier-Transform-Based algorithms** — e.g., Shor's algorithm, Quantum Phase Estimation
- **Simulation algorithms** — e.g., quantum simulation of physical systems

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Classical circuits

- Turing machine is equivalent to some digital circuit (with memory)
- classical digital circuit consists of: wires, connectors, ancillas, logic gates (NOT, OR, AND, XOR)
- NAND gate is universal, but not reversible
- CNOT gate is reversible, but not universal

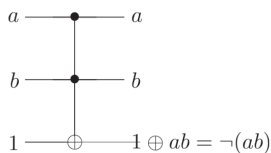


Classical reversible circuits

- every classical digital circuit can be run as a reversible circuit (with ancillas)
- universal reversible computation needs: reversible NAND, FANOUT
- Toffoli gate is universal

Reversible gates

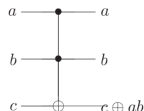
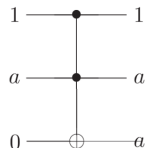
NAND



Toffoli

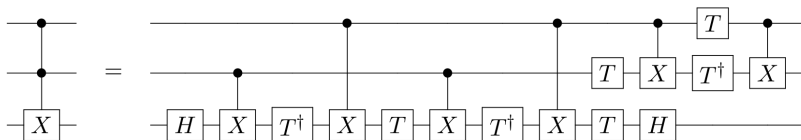
Inputs			Outputs		
a	b	c	a'	b'	c'
0	0	0	0	0	0
0	0	1	0	0	1
0	1	0	0	1	0
0	1	1	0	1	1
1	0	0	1	0	0
1	0	1	1	0	1
1	1	0	1	1	1
1	1	1	1	1	0

FANOUT



Quantum Toffoli gate

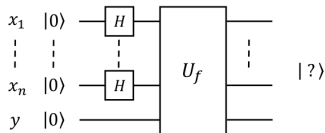
- classical C-NOT gate has a quantum counterpart CX ; CC-NOT (Toffoli) is quantum CCX
- CCX is a complex gate (provided by Qiskit)
- quantum Toffoli can be build with CX gates, which is not possible classically (derivation in *Nielsen & Chaung chap. 11.4.3*)



Quantum parallelism

Let $U_f|x\rangle|y\rangle = |x\rangle|y \oplus f(x)\rangle$, we can achieve quantum parallelism:

Quantum parallelism



$$\begin{aligned} |0\rangle^{\otimes n} \otimes |0\rangle &\xrightarrow{H^{\otimes n} \otimes I} \left[\frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) \right]^{\otimes n} \otimes |0\rangle \\ &= \left(\frac{1}{\sqrt{2}} \right)^n \left(\sum_{x_1, \dots, x_n=0}^1 |x_1, \dots, x_n\rangle \right) \otimes |0\rangle \\ &\xrightarrow{U_f} \left(\frac{1}{\sqrt{2}} \right)^n \sum_{x_1, \dots, x_n=0}^1 |x_1, \dots, x_n\rangle \otimes |f(x_1, \dots, x_n)\rangle \end{aligned}$$

"the most democratic state" $|s\rangle$
- superposition of all possible input states

superposition of function values for all possible inputs, with a single evaluation of U_f

But what now? If we just measure the resulting state,
the output will be a randomly drawn input and its function value.

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Database search

- $N = 2^n$ objects, labeled $x \in \{0, 1, 2, \dots, 2^n - 1\}$;
one marked item w (the winner)
- the oracle: black box function

$$f(x) = 0 \text{ for } x \neq w$$

$$f(w) = 1$$

- “inverting a function”: find a unique input that satisfies the constrain

Classical complexity: “brute force” search in $\mathcal{O}(N)$ time

Quantum oracle

All operations: * are unitary * performed on a quantum register $|x\rangle$

- imagine we have U_w , which implements our $f(x)$

$$U_w |x\rangle = -|x\rangle \quad \text{for } x = w$$

$$U_w |x\rangle = |x\rangle \quad \text{for } x \neq w$$

- construction... check it works(!)

$$U_w = \mathbb{I} - 2|w\rangle\langle w|$$

- time of computation: number of U_w operations

Grover operator

- amplitude amplification operator – useful trick
- "most democratic" state $|s\rangle = \frac{1}{\sqrt{N}} \sum_x |x\rangle$
- unitary Grover operator

$$U_s = 2|s\rangle\langle s| - \mathbb{I}$$

simple calculation gives $|\langle w|s\rangle|^2 = \frac{1}{N}$, $|\langle w|U_s U_w|s\rangle|^2 = \frac{1}{N}(3 - \frac{4}{N})^2$

with more tricky algebra

$$|\langle w|(U_s U_w)^t|s\rangle|^2 = \sin^2((2t+1)\theta),$$

where $\theta = \arcsin \frac{1}{\sqrt{N}}$.

Derivation

ALGEBRAIC DERIVATION OF Grover's algorithm

$$U_w = \mathbb{1} - 2 |w\rangle\langle w| \quad U_s = 2 |s\rangle\langle s| - \mathbb{1}$$

$$\text{with } |s\rangle = \frac{1}{\sqrt{N}} \sum_x |x\rangle; \text{ where } N = 2^n; x = 0, 1, \dots, 2^n - 1$$

check how U_w, U_s acts on these states:

$$U_w |s\rangle = |s\rangle - 2 |w\rangle \underbrace{\langle w | s \rangle}_{1/\sqrt{N}} = |s\rangle - \frac{2}{\sqrt{N}} |w\rangle$$

$$U_w |w\rangle = -|w\rangle$$

$$U_s |s\rangle = |s\rangle$$

$$U_s |w\rangle = \frac{2}{\sqrt{N}} |s\rangle - |w\rangle$$

we can express U_w, U_s in basis $\text{span}(|s\rangle, |w\rangle)$

Derivation continued

$$U_w \xrightarrow{\text{equiv.}} \begin{pmatrix} 1 & 0 \\ -\frac{2}{\sqrt{N}} & -1 \end{pmatrix} \quad U_s \xrightarrow{\text{equiv.}} \begin{pmatrix} 1 & \frac{2}{\sqrt{N}} \\ 0 & -1 \end{pmatrix}$$

using this matrix representation:

$$U_s U_w |s\rangle = \left(1 - \frac{4}{N}\right) |s\rangle + \frac{2}{\sqrt{N}} |w\rangle$$

$$|\langle w | U_s U_w |s\rangle|^2 = \frac{1}{N} \left(3 - \frac{4}{N}\right)^2$$

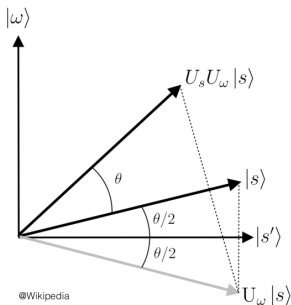
to calculate $(U_s U_w)^t$ (after t -iterations) one needs to diagonalize $(U_s U_w)$; ... result is

$$|\langle w | (U_s U_w)^t |s\rangle|^2 = \sin^2(\theta(2t+1))$$

$$\theta = \arcsin \frac{1}{\sqrt{N}}$$

Geometric interpretation

Geometric intuition for how it works:



The n -qubit system starts in state $|s\rangle$. It lives in a subspace spanned by $|\omega\rangle$ and $|s'\rangle = \frac{1}{\sqrt{N-1}} \sum_{x \neq \omega} |x\rangle$: $|s\rangle = \frac{1}{\sqrt{N}} |\omega\rangle + \frac{\sqrt{N-1}}{\sqrt{N}} |s'\rangle$.

It can be verified that the application of the oracle U_ω and the diffusion operator $U_s = 2|s\rangle\langle s| - \mathbb{I}$ rotates the n -qubit state within the subspace by an angle $\theta = 2 \arcsin \frac{1}{\sqrt{N}}$.

It follows that after approximately $\pi\sqrt{N}/4$ iterations of $U_s U_\omega$, the state is rotated *close* to $|\omega\rangle$, which is then likely to be measured and thus determined.

Therefore, the algorithm has complexity $\mathcal{O}(\sqrt{N})$, quadratically improving the classical algorithm. It has, however, a non-unit probability of success, with the error probability approaching 0 with growing N .

Grover's algorithm (1997)

- maximal overlap is attained for $t^* \propto \pi\sqrt{N}/4$
- quadratic speedup from quantum computation: $\mathcal{O}(\sqrt{N})$ versus $\mathcal{O}(N)$
- Grover's algorithm is optimal (no more speedup possible)
- no new complexity classification, no solution to NP-complete problems

Final remarks:

“Is quantum search practical?” explains misunderstandings and perspectives

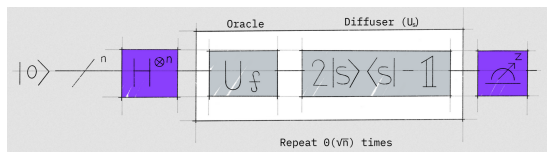
<https://arxiv.org/abs/quant-ph/0405001>

also good textbook reference: Nielsen & Chuang chap. II.6

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Building circuit for Grover's search



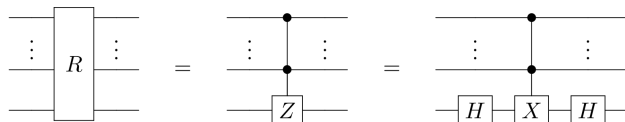
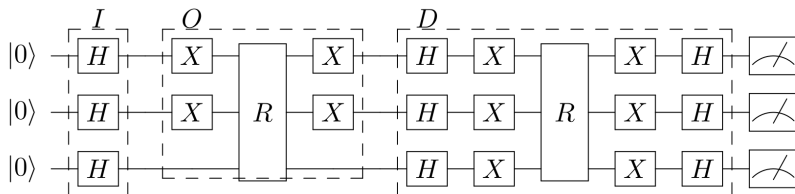
Oracle

- we need $V_w = X^{1-w_0} \otimes \dots \otimes X^{1-w_{N-1}}$, which flips a qubit only when $w_i = 0$
- $[CCZ]$ gate flips the sign for $w = [1, 1, 1]$ (all qubits are 1)
- the oracle is then $U_w = V_w [CCZ] V_w$

Grover operator

- same for $X^{(n)} [CCZ] X^{(n)}$ – flips sign only for $|0\rangle$
- we express $X^{(n)} [CCZ] X^{(n)} = \mathbb{I} - 2|0\rangle\langle 0|$
- finally: $H^{(n)} X^{(n)} [CCZ] X^{(n)} H^{(n)} = \mathbb{I} - 2|s\rangle\langle s| = -U_s$

Final diagram



- $n = 3$ qubits, marked state $w = [0, 0, 1]$
- one iteration: initialization, oracle, Grover operator (diffusor D)
- note that the core of calculation is done by Toffoli gate
- measurement in computational basis

Grover's search algorithm run on IBM quantum devices back in 2019 (by Alicja Dutkiewicz)

Tabela 4.2: Prawdopodobieństwa sukcesu

Liczba kubitów	Liczba iteracji	Średnie prawdopodobieństwo sukcesu [%]			
		Klasycznie	Teoretycznie	<i>IBM Q Tenerife</i>	<i>IBM Q Yorktown</i>
2	1	25	100	83.074 ± 0.017	84.713 ± 0.063
3	1	12.5	78.1	41.416 ± 0.017	19.617 ± 0.031
	2	25	94.5	29.873 ± 0.014	29.785 ± 0.038
4	1	6.25	47.3	6.8831 ± 0.0060	5.667 ± 0.018
			Po preselekcji	6.7354 ± 0.0075	6.293 ± 0.026
	2	12.5	90.8	6.1130 ± 0.0061	6.305 ± 0.019
			Po preselekcji	6.0601 ± 0.0082	6.299 ± 0.026
	3	25	96.1	5.6970 ± 0.0048	6.038 ± 0.019
			Po preselekcji	5.670 ± 0.007	6.005 ± 0.025

Kolorem zielonym oznaczono wyniki lepsze, a kolorem czerwonym gorsze niż najlepszy algorytm klasyczny.