

INTRODUCTION TO ARTIFICIAL INTELLIGENCE

BI-ZUM / BIE-ZUM

QUANTUM ANNEALING

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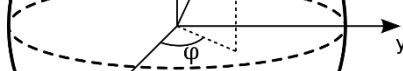
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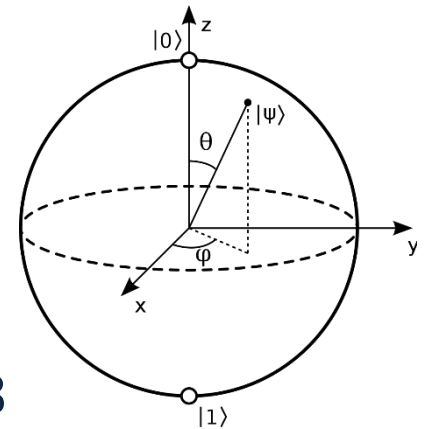
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QUANTUM COMPUTERS

- Basic concept: **qubit**
 - superposition of two basic states $|0\rangle$ and $|1\rangle$
 - qubit state $|\psi\rangle = \alpha|0\rangle + \beta|1\rangle$, where α and β are complex numbers
 - probability of $|0\rangle$ after measurement is $|\alpha|^2$, probability of $|1\rangle$ is $|\beta|^2$
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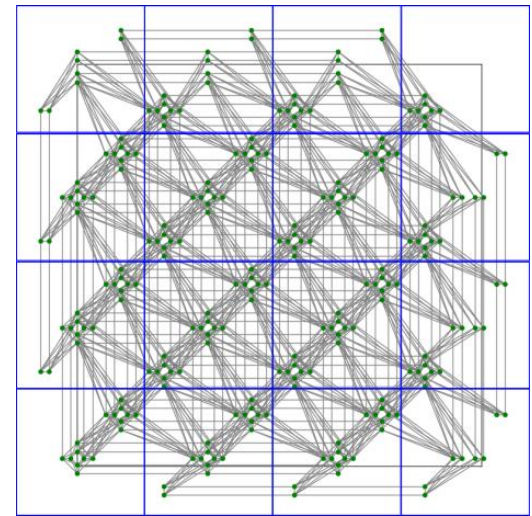


- **Major paradigms**
 - gate based
 - quantum gates – produce qubits from input qubits
 - adiabatic quantum computing
 - quantum annealing
 - and more...

[illegible]

ADIABATIC QUANTUM COMPUTERS

- **Limited number of qubits** (a few 1000s)
 - many more than gate-based model (a few 100s)
- **Limited connectivity**
 - hardware does not implement a fully connected graph
 - chained qubits (use extra qubits)
 - 5000 qubits and current topology can implement a fully connected graph with 177 qubits
- **But simple: one sample** – annealing, readout, cooling

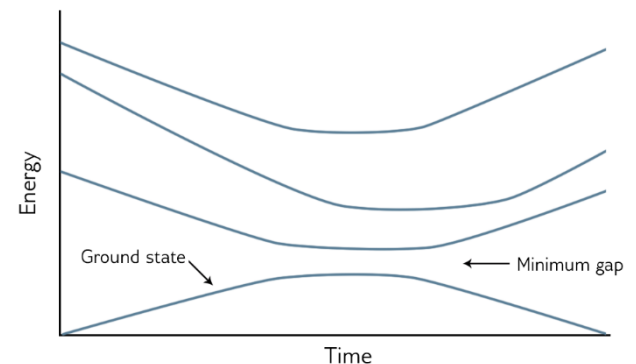


ADIABATIC QUNTUM COMPUTATION

- **Adiabatic computation**

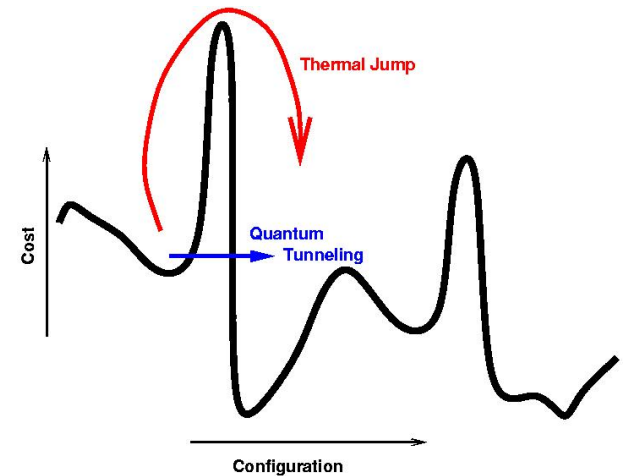
$$\mathcal{H}_{ising} = \underbrace{-\frac{A(s)}{2} \left(\sum_i \hat{\sigma}_x^{(i)} \right)}_{\text{Initial Hamiltonian}} + \underbrace{\frac{B(s)}{2} \left(\sum_i h_i \hat{\sigma}_z^{(i)} + \sum_{i>j} J_{i,j} \hat{\sigma}_z^{(i)} \hat{\sigma}_z^{(j)} \right)}_{\text{Final Hamiltonian}}$$

- By varying $A(s)$ and $B(s)$ over time
 - $0 \nearrow 1, 1 \searrow 0$
- The system moves from initial Hamiltonian and stays in the ground state (minimal energy)



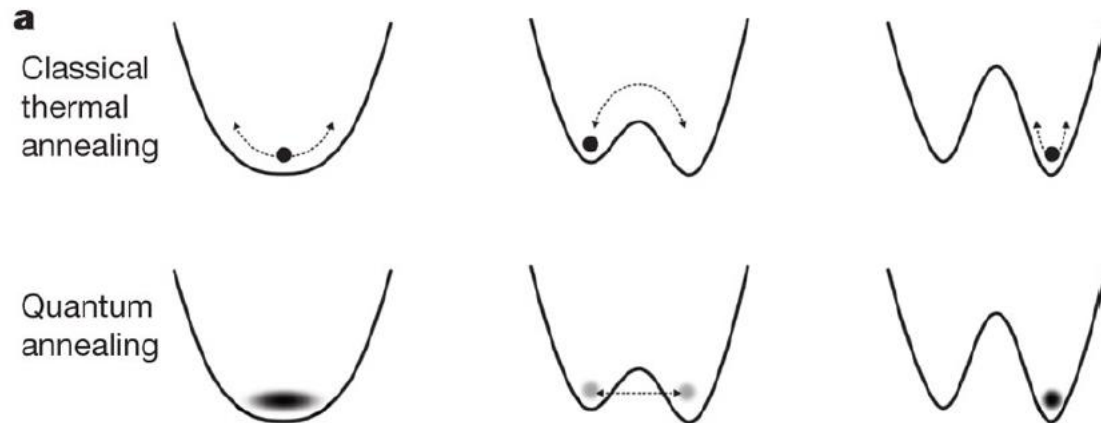
QUANTUM ANNEALING

- **Adiabatic computation** can be interpreted as quadratic optimization problem
 - n Boolean variables x_1, x_2, \dots, x_n
 - decision variables (we want to determine their values)
 - minimize
$$\sum_{i=1}^n a_i x_i + \sum_{i=1}^n a_i' x_i^2 + \sum_{i,j} a_{ij} x_i x_j$$
 - a_i, a_i', a_{ij} are real-valued coefficients
- **Equivalent to:**
 - minimize $\sum_{i \leq j} q_{ij} x_i x_j$ (QUBO)
 - using matrices $\min x^T Q x$
 - can be solved by classical simulated annealing but adiabatic computation can utilize quantum tunneling effect
 - $q_{ij} \neq 0$ requires hardware connection between qubits i and j



QUANTUM EFFECTS

- Temperature corresponds to the strength of tunneling effect
 - any height can be crossed
 - but the tunnel can be only of limited length depending on the temperature
 - the temperature can change the landscape
- If a certain cooling scheme is followed, then the process can reach the global optimum thanks to tunneling effect



PROBLEM COMPILATION TO QUBO

- **QUBO:** Quadratic Unconstrained Binary Optimization
 - minimize $\sum_{i \leq j} q_{ij} x_i x_j$
- How to express constraints in an unconstrained formalism?
- Use **penalties**.
 - Examples:
 - (pseudo-Boolean) constraint $x+y \leq 1$, penalty: xy
 - constraint $x = y$, penalty: $x+y - 2xy$
 - implication $x \Rightarrow y$, equivalent to $x \leq y$, penalty: $x - xy$
 - Penalties are compositional – addition, penalty coefficients
- More complex compilation, **one-hot encoding**
 - n Boolean variables x_1, x_2, \dots, x_n , exactly one has value 1
 - (pseudo-Boolean) constraint $\sum_{i=1}^n x_i = 1 \Leftrightarrow \sum_{i=1}^n x_i - 1 = 0$,
penalty: $(\sum_{i=1}^n x_i - 1)^2 \Leftrightarrow -\sum_{i=1}^n x_i + 2 \sum_{i < j} x_i x_j$