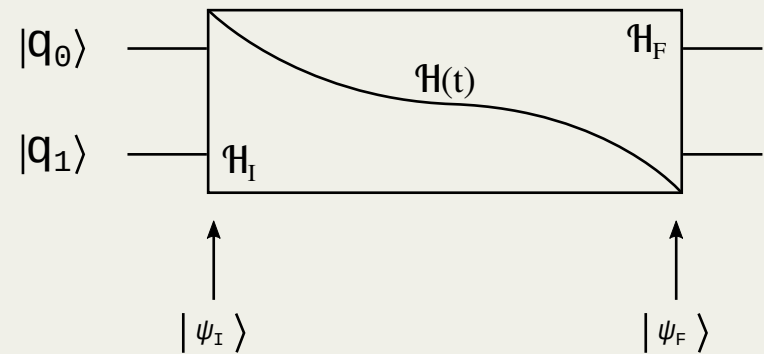
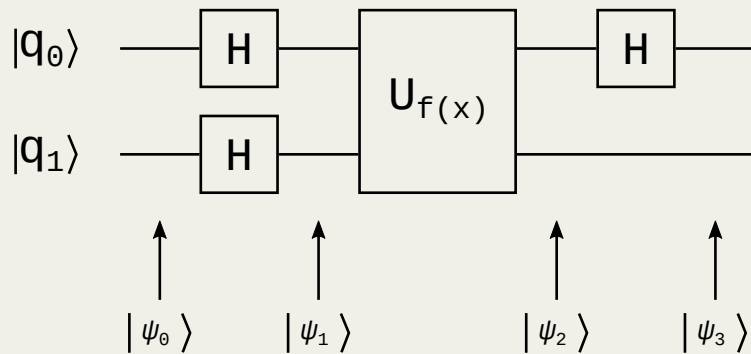


Quantum Annealing

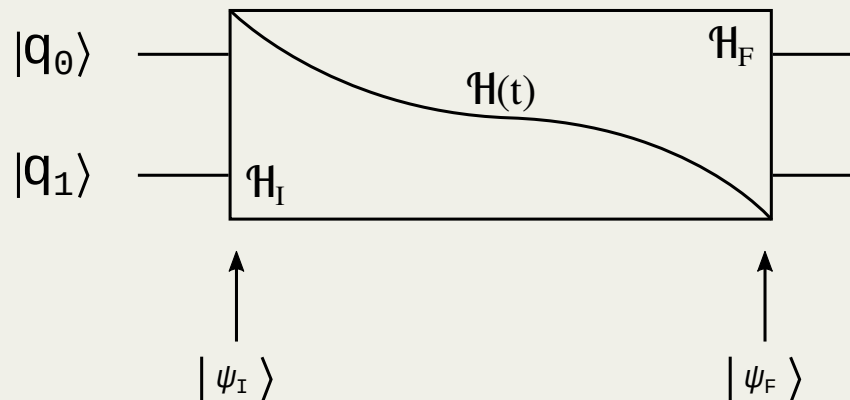
Quantum Capita Selecta

Bernardo Villalba Frías, PhD

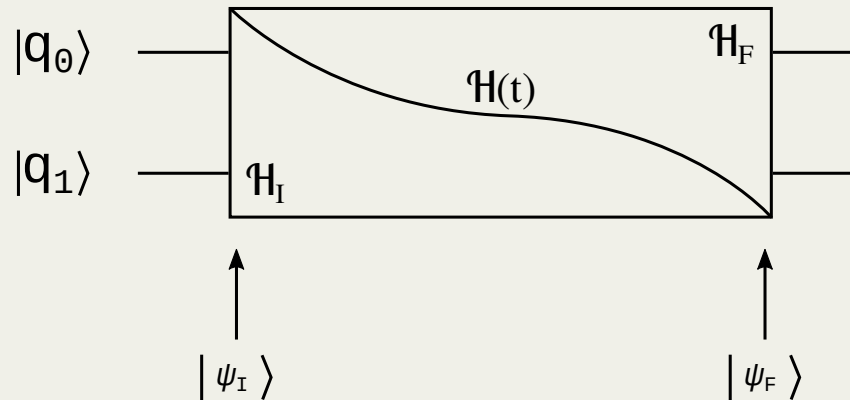
`b.r.villalba.frias@hva.nl`



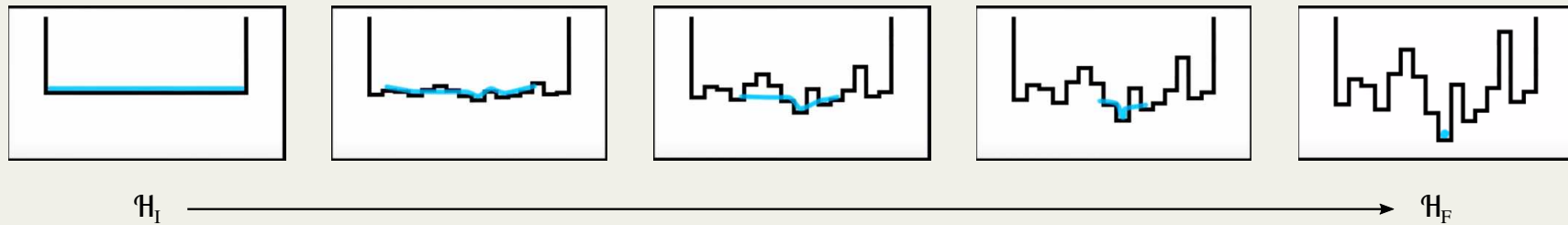
- Alternative to the gate model
 - Polynomially equivalent
 - Discrete versus analog nature
- Not full CPUs
 - Smart memory accelerators
 - Solve NP–hard optimization problems
 - Using Quantum Annealing



- States gradually evolve according to certain forces
 - Represented by Hamiltonians
- Transition from initial to final Hamiltonian
 - $\mathcal{H}_I \rightarrow \mathcal{H}_F$
- State is read at the end of the transition
- Each observable state has an associated energy
- Ground state \rightarrow lowest energy over all states

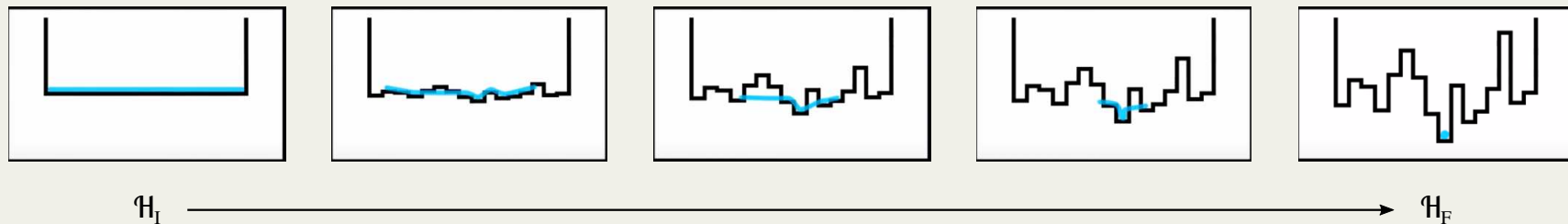


- Adiabatic Theorem:
 - If the qubits start in ground state, and
 - the transition is slow enough; then,
 - the system, with high probability, will finish in ground state



- Key components:
 - Initial Hamiltonian (\mathcal{H}_I): ground state
 - Final Hamiltonian (\mathcal{H}_F): optimal solution
 - Adiabatic path ($s(t)$): transition from \mathcal{H}_I to \mathcal{H}_F

$$s(t) = 1 - t$$

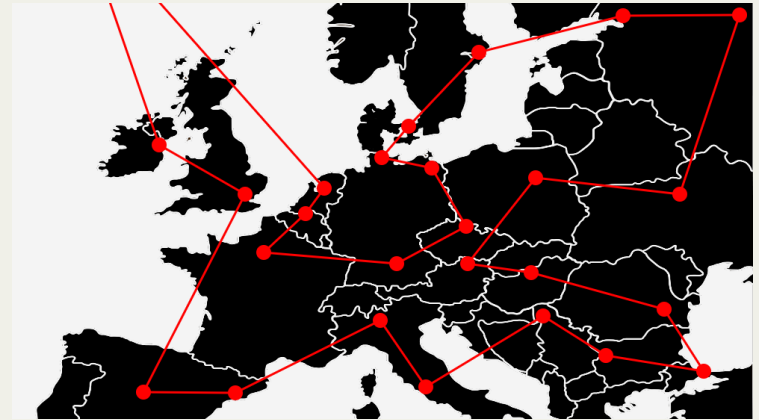
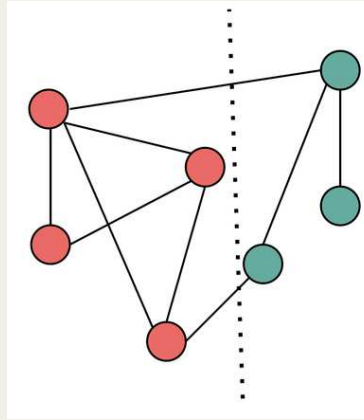
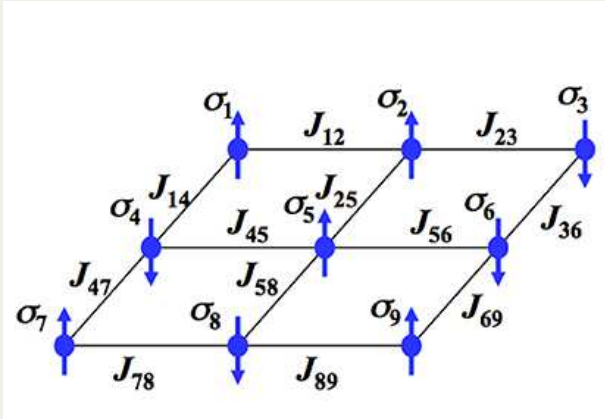


- Time-varying Hamiltonian:
 - Creates a gradual transition: $\mathcal{H}_I \rightarrow \mathcal{H}_F$

$$\begin{aligned}\mathcal{H}(t) &= s(t)\mathcal{H}_I + (1 - s(t))\mathcal{H}_F \\ &= (1 - t)\mathcal{H}_I + t\mathcal{H}_F\end{aligned}$$

- This Hamiltonian is an AQC algorithm for solving the problem

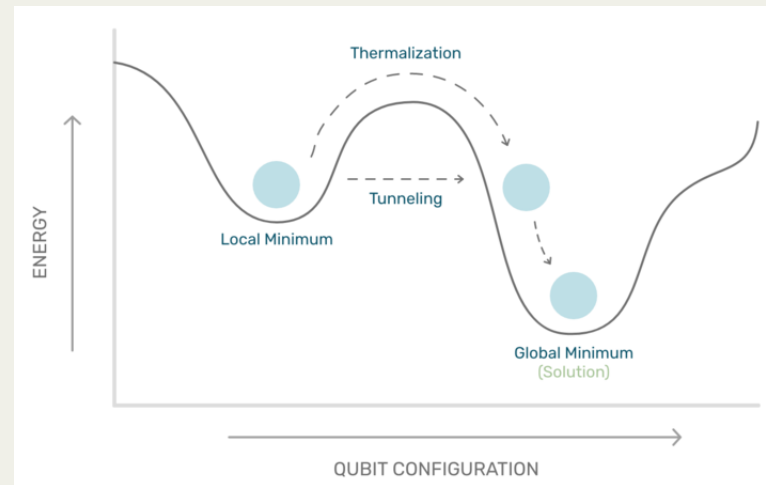
- Robust against decoherence:
 - Computation takes place in ground state
- Relaxed computational goal:
 - Designed to solve one NP–hard optimization problem (the Ising Model)
 - Find an assignment to N variables that minimizes a given objective function
 - Feasible solutions can be verified
 - Optimal solutions cannot
 - Using a specific solution approach (Quantum Annealing)



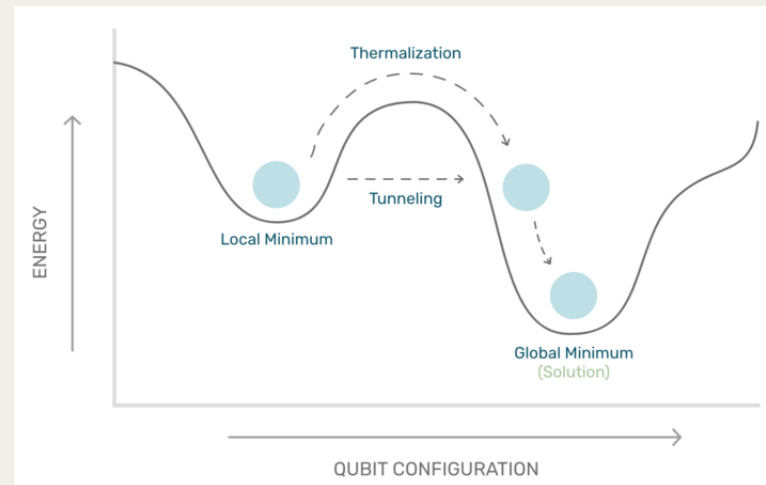
- A spin system: up and down
- Graph-like arrangement
 - Allowing spin interaction
- The energy of the system is given by:

$$\mathcal{H}(\sigma) = - \sum_{\langle ij \rangle} J_{ij} \sigma_i \sigma_j - \sum_j h_j \sigma_j$$

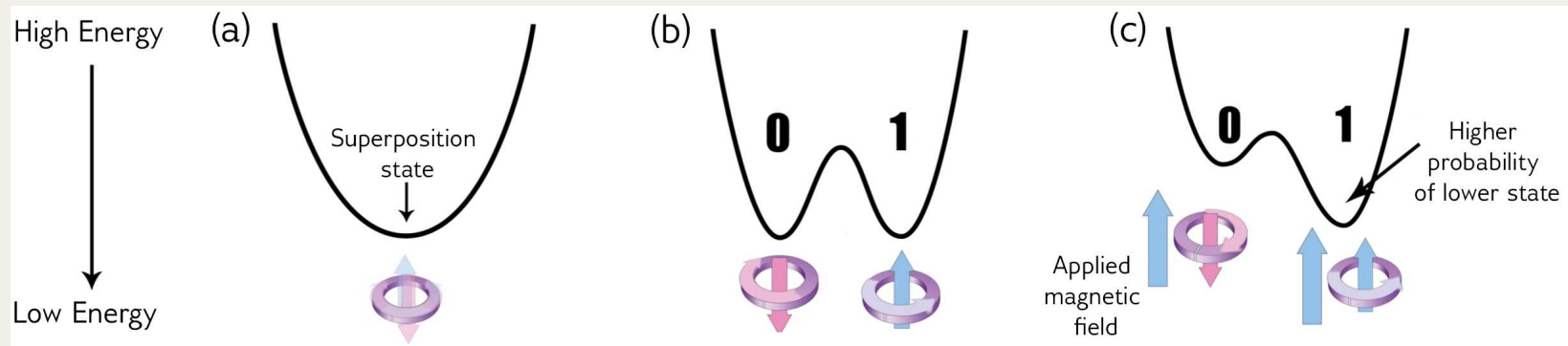
- Reformulated to many interesting NP-hard problems



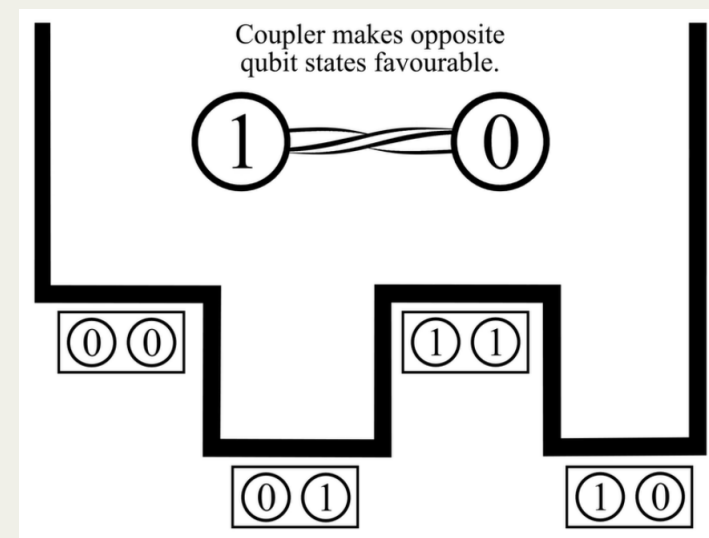
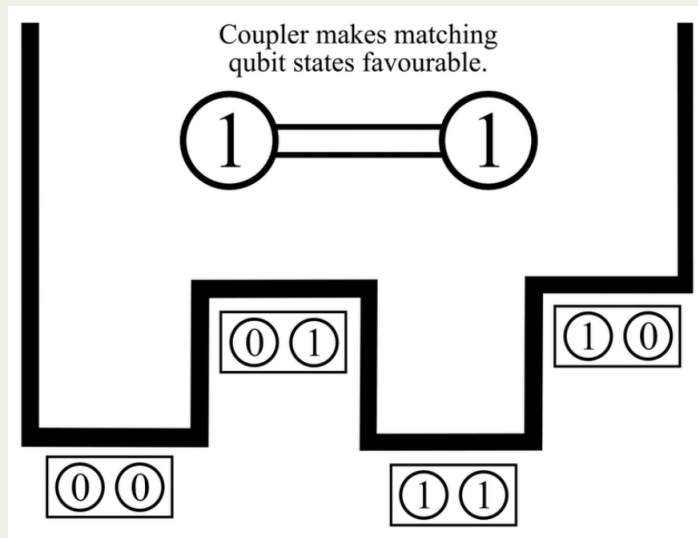
- Each spin configuration produces an energy value
 - Energy landscape
- Goal: find the global minimum
 - What about local minimum?
 - Simulated annealing
 - Quantum annealing



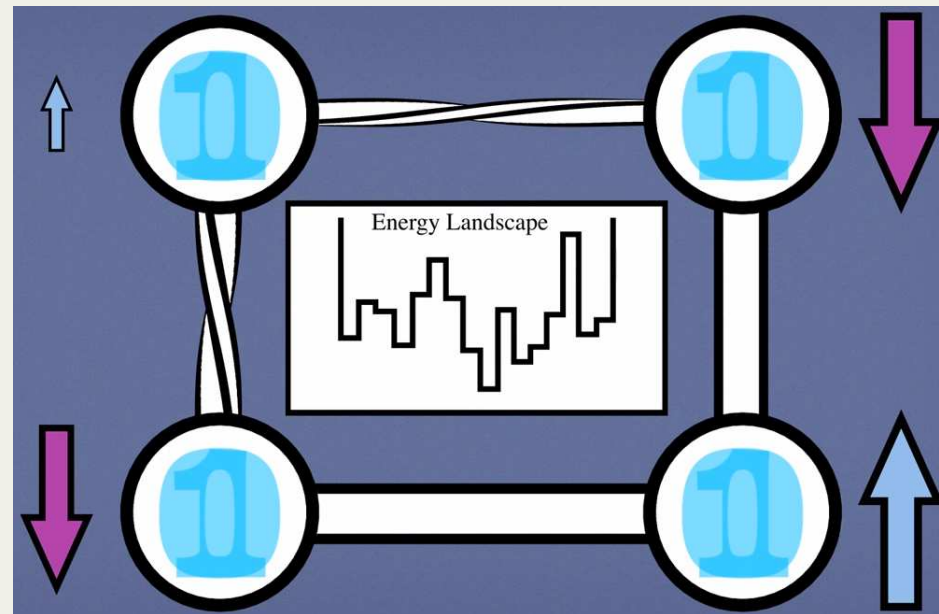
- Heuristic approach to solving combinatorial optimization problems
 - Restriction on final Hamiltonian \mathcal{H}_F
- Similar to simulated (thermal) annealing
- Can “tunnel” through hills
- Can run on classical or adiabatic platforms



- The system is initially in superposition
 - Lowest point in the valley
- Quantum annealing runs, a barrier is raised
 - Double-well potential
- Apply an external magnetic field (bias)
 - Tilting the double-well potential

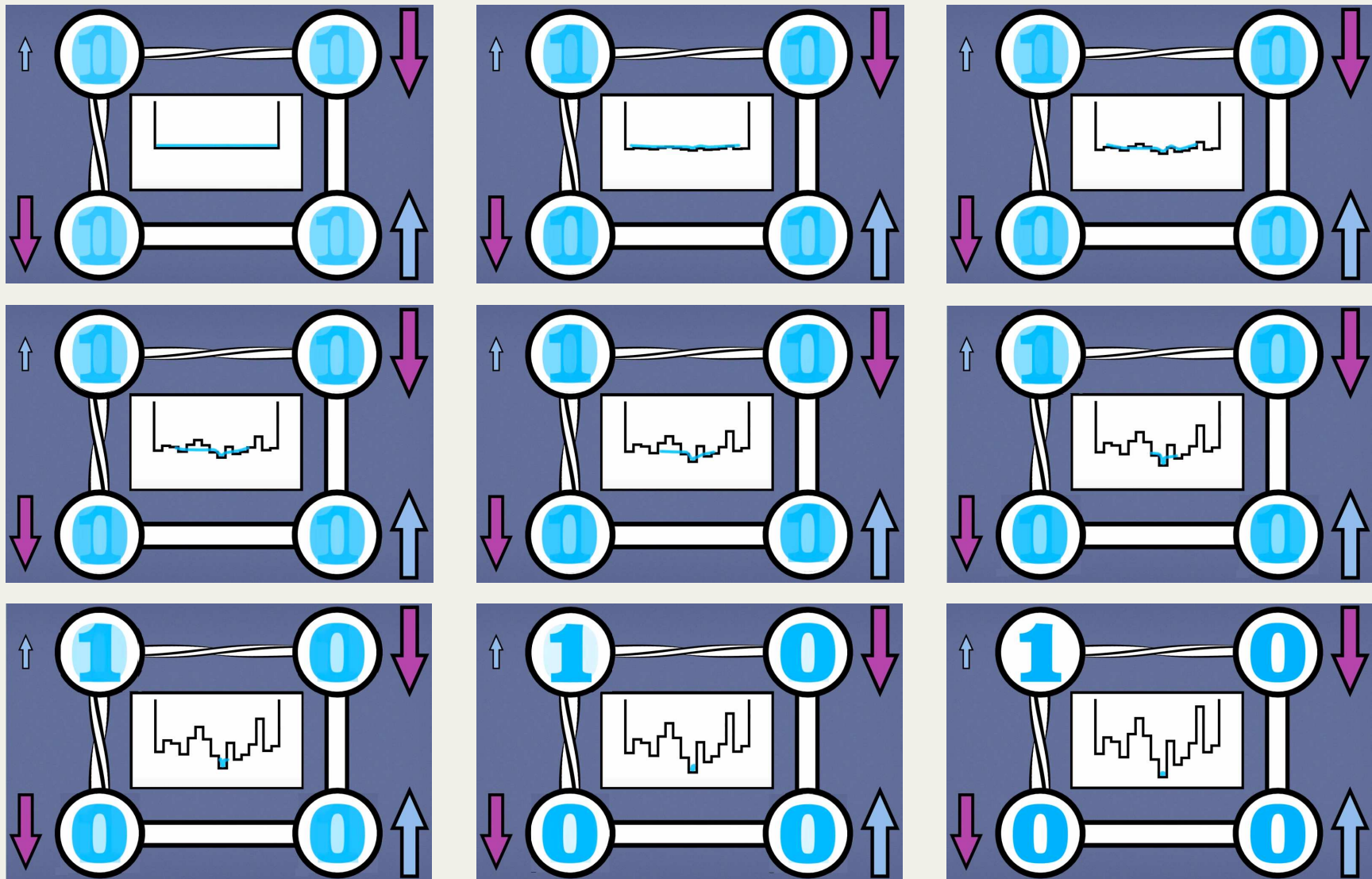


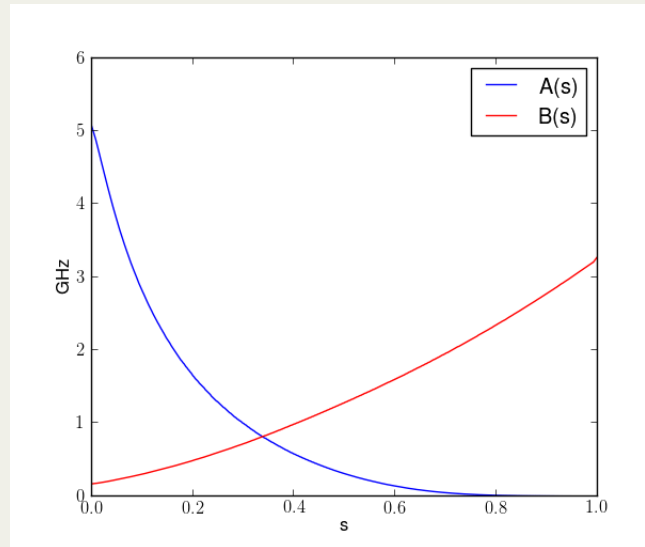
- Defines how qubits influence each other
- Sets the correlation between two interacting qubits
 - Qubits end up in the same state
 - Qubits end up in opposite states
- Effectively enabling entanglement



- Choose the values for biases and couplers
 - Direction and strength
- Defines an energy landscape
- The quantum annealing solves and finds the minimum energy

Programming in Quantum Annealing





- The D-Wave's quantum annealing system incorporates:

$$\mathcal{H}_I = \sum_i \sigma_i^x$$

$$\mathcal{H}_F = \sum_i h_i \sigma_i^z + \sum_{ij} J_{ij} \sigma_i^z \sigma_j^z$$

$$\mathcal{H}(s) = A(s)\mathcal{H}_I + B(s)\mathcal{H}_F$$



- Advantage2
 - Announced in 2021 (Released ???)
 - 7000+ qubits
 - Zephyr graph topology

