

# Adiabatic quantum computing II

Alternative to the standard "gate model" of quantum computing

## Adiabatic quantum computation (AQC)

- Engineer a Hamiltonian that changes in time so that  $\mathcal{H}(t=0)$  has a known product state as its ground state, e.g.

$$|E_0(0)\rangle = |+\rangle |+\rangle \dots |+\rangle$$

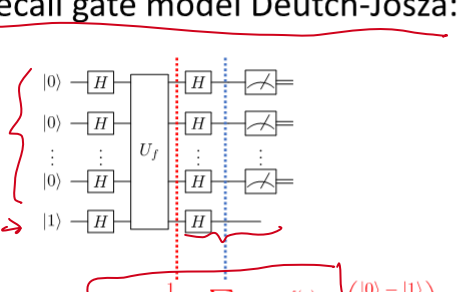
- And for  $t=t_f$ ,  $\mathcal{H}(t_f)$  has a ground state that encodes the solution to the problem you want to solve. E.g. for adiabatic Deutsch-Josza:

$$|E_0(t_f)\rangle = \frac{1}{\sqrt{2^n}} \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle$$

- So that performing a measurement (in a particular basis) will give the desired answer to the problem.

## Example: Adiabatic version of Deutsch-Josza algorithm

- Recall gate model Deutsch-Josza:



- Adiabatic version: Want to find Hamiltonian whose ground state is

$$|E_0(t=t_f)\rangle = |E_0(1)\rangle = \frac{1}{\sqrt{2^n}} \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle$$

- a) Starting Hamiltonian:

$$\mathcal{H}(t=0) = \mathcal{H}(0) = \omega \sum_{i=1}^n \left( \frac{I - X_i}{2} \right)$$

$$|E_0(0)\rangle = |+\rangle |+\rangle \dots |+\rangle$$

$$\mathcal{H}(0) |E_0(0)\rangle = 0 \cdot |E_0(0)\rangle = 0$$

$$|E_1(0)\rangle = |-\rangle |+\rangle \dots |+\rangle, \quad E_1(0) = \omega$$

$$|E_2(0)\rangle = |+\rangle |-\rangle |+\rangle \dots |+\rangle, \quad E_2(0) = \omega$$

$$|E_3(0)\rangle = |+\rangle |+\rangle |-\rangle |+\rangle \dots |+\rangle, \quad E_3(0) = \omega$$

- b) We have  $|E_0(1)\rangle = U |E_0(0)\rangle$  for  $U = \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle\langle x|$

$$U |E_0(0)\rangle = \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle\langle x| \left( \frac{1}{\sqrt{2^n}} \sum_{x \in \{0,1\}^n} |x\rangle \right) = \frac{1}{\sqrt{2^n}} \sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle = |E_0(1)\rangle$$

- b) The Hamiltonian  $\mathcal{H}(1) = U \mathcal{H}(0) U^\dagger$  has  $|E_0(1)\rangle$  as its ground state:

$$\mathcal{H}(1) |E_0(1)\rangle = U \mathcal{H}(0) U^\dagger U |E_0(0)\rangle = U \mathcal{H}(0) |E_0(0)\rangle = U \cdot 0 = 0 = \mathcal{H}(1) |E_0(1)\rangle$$

- b) The interpolation between  $\mathcal{H}(0)$  and  $\mathcal{H}(1)$  can be achieved by:

$$\mathcal{H}(s) = \bar{U}(s) \mathcal{H}(0) \bar{U}^\dagger(s), \quad \text{where } \bar{U}(s) = e^{i \int_0^s \mathcal{H}(t) dt}$$

BECAUSE  $U = \bar{U}(1) \Rightarrow \bar{U}(s) = e^{i \int_0^s \mathcal{H}(t) dt} = e^{i \int_0^s \omega \sum_{i=1}^n \left( \frac{I - X_i}{2} \right) dt} = e^{i \omega s \sum_{i=1}^n \left( \frac{I - X_i}{2} \right)} = U$

- c) The unitary transf. preserves the energy levels, so  $E_1(1) - E_0(1) = \omega$ . Hence runtime is set by:

$$t_f \gg \max_{x \in \{0,1\}^n} \frac{\hbar |E_1(s) - E_0(s)|}{|E_1(s) - E_0(s)|^2} \sim \frac{\hbar}{\omega^2} = \frac{1}{\omega} \Rightarrow \text{INDEPENDENT OF } m! \text{ ADIABATIC RUNTIME IS } O(1), \text{ JUST LIKE THE GATE MODEL VERSION!}$$

MORE GENERALLY EIGENVECTORS OF  $(U \mathcal{H}(0) U^\dagger)$  ARE  $U |E_m(0)\rangle$  AND EIGENVALUES ARE  $E_m$  (SAME!)  $\Rightarrow$  EIGENVALUES OF  $U \mathcal{H}(0) U^\dagger$  ARE ALSO  $E_m(0)$ !

## AQC = gate model up to polynomial overhead!

- One can construct a n-qubit circuit that efficiently simulates the AQC time evolution of n qubits with the number of gates scaling as  $t_f^2 \text{ poly}(n)$ .

- It is also possible to explicitly construct a final Hamiltonian  $\mathcal{H}(1)$  whose ground state is the result of one and two-qubit gates (circuit with depth L). It was proven that the result of L one and two-qubit gates to evolve adiabatically into  $|E_0(1)\rangle$ !

- Which one is better? Depends on the problem. AQC seems really good for optimization problems such as travelling salesman.

- Gate model and AQC are not the only "models" for QC. There are a total of 5 different models:

- Measurement-based QC (done by 1-qubit measurements on a highly entangled initial state)
- Topological QC (done by braiding anyons on a 2d lattice)
- Quantum Turing machine

## Combinatorial problems: Reduction to 3-satisfiability (3-SAT)

- Example combinatorial problem: Exact cover: Tiling a region of 60 squares using each of 12 pentominoes only once.



- This problem can be formulated as a decision problem for choices and constraints: The choices are the different locations and orientations for each pentomino, and the constraints are (1) Each square must be covered exactly once and (2) each pentomino appears once.

- Answer: 65 different tilings, e.g.



- See more at <https://garethrees.org/2015/11/09/exact-cover/>

- Exact cover is one of the so called NP-complete problems. There are hundreds of them, e.g. travelling salesman for integer distances. All of them map into each other with polynomial overhead, so if you solve one you solve all! Another example is the 3-satisfiability (3-SAT) problem that we now show can be solved with AQC.

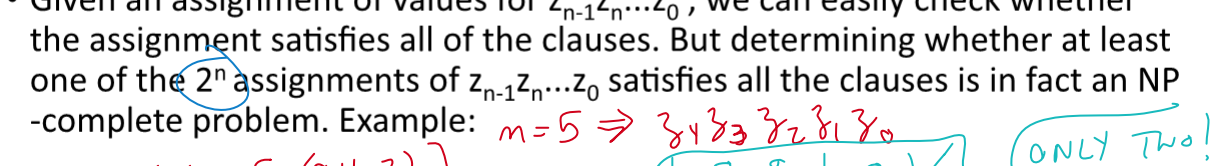
## The 3-satisfiability (3-SAT) problem

- 3-SAT: Find the n-bit string  $z_1, z_2, \dots, z_n$  that satisfies a set of 3-bit constraint clauses. Each clause involves 3 bits  $i, j, k$  and the constraint is that one of the 3 bits must have value 1 and the other two must be 0:

$$c(i, j, k) = z_i + z_j + z_k - 1 = 0$$

- An n-bit instance of 3-SAT is a list of triples  $(i, j, k)$  indicating which groups of three bits are involved in clauses. The problem is to determine whether there is some assignment of the n-bit values that satisfies all of the clauses.

- Given an assignment of values for  $z_1, z_2, \dots, z_n$ , we can easily check whether the assignment satisfies all of the clauses. But determining whether at least one of the  $2^n$  assignments of  $z_1, z_2, \dots, z_n$  satisfies all the clauses is in fact an NP-complete problem. Example:  $m=5 \Rightarrow z_1 z_2 z_3 z_4 z_5$  (ONLY TWO!)



## AQC algorithm for 3-SAT E. Farhi et al, Science 292, 472 (2001)

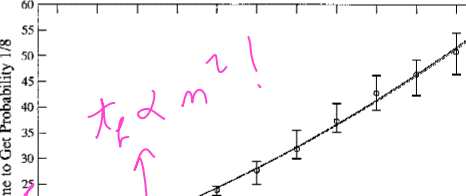
- a) Starting Hamiltonian:  $\mathcal{H}(t=0) = \mathcal{H}(0) = \omega \sum_{i=1}^n \left( \frac{I - X_i}{2} \right)$  (USUAL ONE,  $|E_0(0)\rangle = |+\rangle |+\rangle \dots |+\rangle = \frac{1}{\sqrt{2^n}} \sum_{x \in \{0,1\}^n} |x\rangle$ )

- b) Final Hamiltonian:  $\mathcal{H}(t=t_f) = \mathcal{H}(1) = \sum_{\text{all } c(i,j,k)} \left[ \left( \frac{I - Z_i}{2} \right) + \left( \frac{I - Z_j}{2} \right) + \left( \frac{I - Z_k}{2} \right) - 1 \right]^2$  ( $E_0(1) = 0$  WHEN  $(z_1, z_2, \dots, z_n)$  SATISFIES ALL CLAUSES!)

- c) Picture shows median  $t_f$  required to get the right answer with prob. 1/8 using exact integration of Schrödinger eqn for linear  $s(t)$ :

$$\mathcal{H}(t) = \left( 1 - \frac{t}{t_f} \right) \mathcal{H}(0) + \frac{t}{t_f} \mathcal{H}(1)$$

$s(t) = \frac{t}{t_f}$  (A FIN CLAUSE OTHER ONES!)



## Annealing method: A heuristic algorithm

An algorithm based on the "annealing" metal processing phenomenon

**Annealing Phenomenon**

When brought up to high temperature then gradually cooled, the structure of metal becomes stable (low energy).

High Temperature: High energy - Unstable atoms

Low Temperature: Low energy - Stable atoms

**Annealing Method**

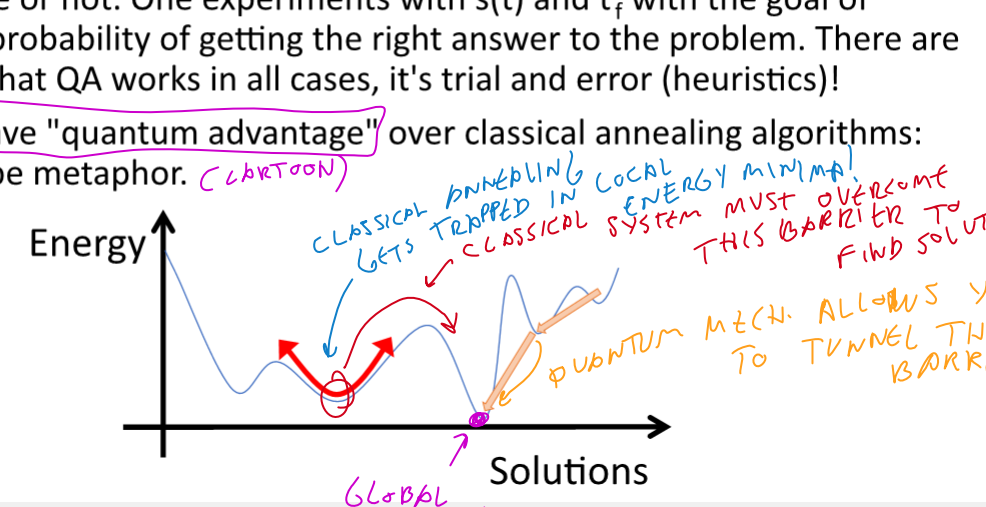
Find a way to quickly fit all the pieces by first shaking the whole system, then gradually reducing the shaking

When exploring optimal solutions, first search all solutions including those far from optimal, and then gradually close in to an optimal solution.

## Quantum annealing (QA) as heuristic AQC

- In Quantum annealing, one performs a "loose run" of the 3-SAT AQC for particular choices of  $s(t)$  and  $t_f$ , without worrying whether the system stays in the ground state or not. One experiments with  $s(t)$  and  $t_f$  with the goal of optimizing the probability of getting the right answer to the problem. There are no guarantees that QA works in all cases, it's trial and error (heuristics)!

- Why QA may have "quantum advantage" over classical annealing algorithms: Energy landscape metaphor. (CARICATURE)



## 3-SAT is a particular case of QUBO

- Let's expand the 3-SAT  $\mathcal{H}(1)$ :

$$\mathcal{H}(1) = \sum_{\text{all } c(i,j,k)} \left[ \left( \frac{I - Z_i}{2} \right) + \left( \frac{I - Z_j}{2} \right) + \left( \frac{I - Z_k}{2} \right) - 1 \right]^2$$

$$= \sum_{\text{all } c} \left\{ \frac{1}{4} (z_i^2 + z_j^2 + z_k^2 - 2(z_i z_j + z_j z_k + z_k z_i)) + 2(z_i z_j + z_j z_k + z_k z_i) \right\}$$

- This is a binary quadratic form. So 3-SAT is a particular case of

### Quadratic Unconstrained Binary Optimization (QUBO)

$$\mathcal{H}_{\text{QUBO}}(1) = - \sum_i b_i z_i + \sum_{i < j} c_{ij} z_i z_j$$

CHOOSE YOUR  $b_i$  AND  $c_{ij}$  AS YOU WISH.

## D-Wave Leap: Cloud access to QA hardware

- D-Wave from Burnaby, BC is the leading developer of QA hardware and software. Next class we will access their devices using cloud-based D-Wave's platform, D-Wave Leap.



HOW COME THEY ALREADY HAVE 2000 QUBITS?  $\rightarrow$  BY FOCUSING ON A SPECIFIC ALGORITHM (QUBO) MUCH LESS CONTROL IS REQUIRED! AND ALSO, QA IS MORE ROBUST TO NOISE. SO D-WAVE IS ABLE TO SCALE UP FASTER THAN THEIR COMPETITION!