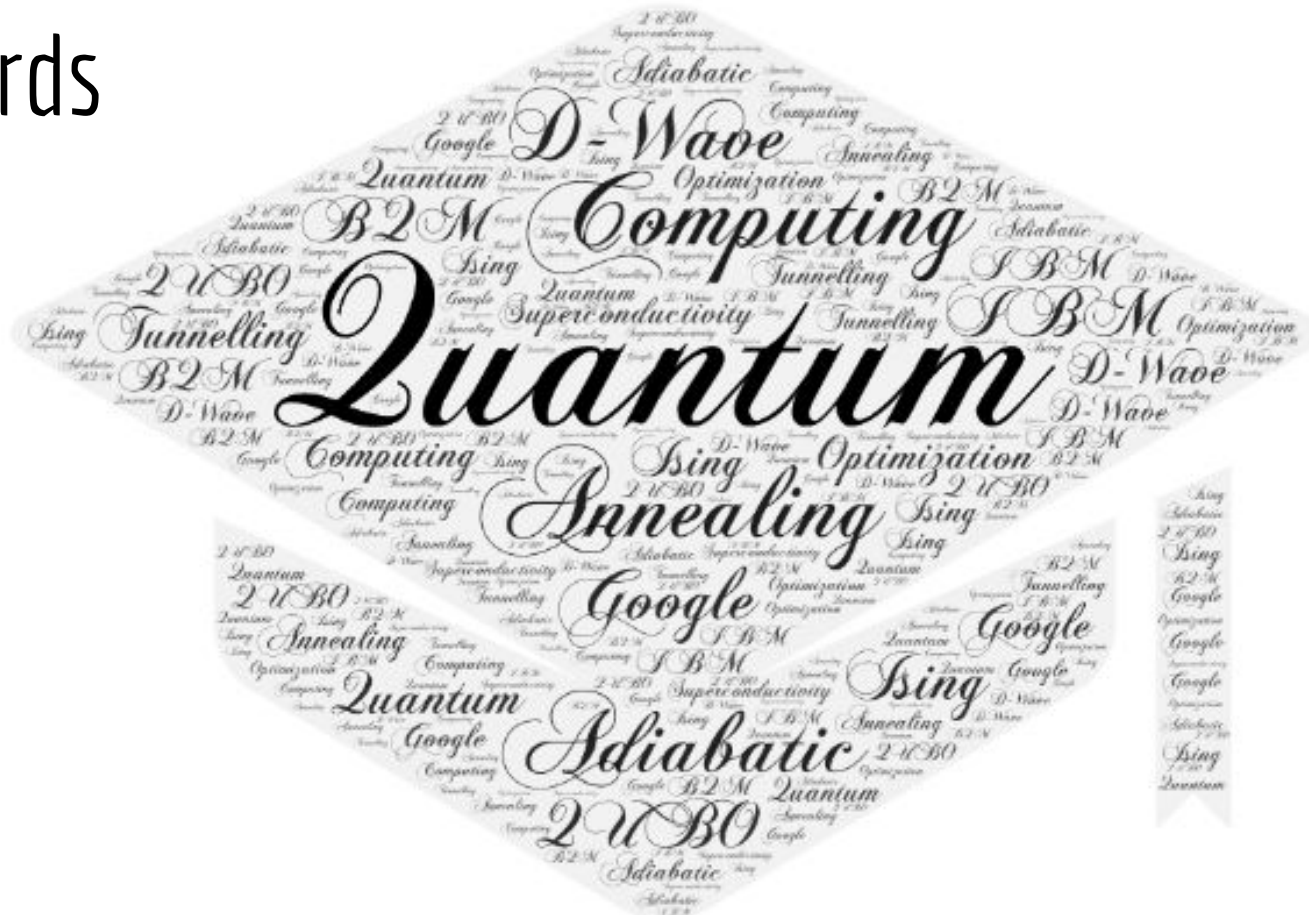


Optimization with Quantum machines

(especially with quantum annealing)

Sevilla, 3rd of December, 2021

Keywords



Outline

1

Quantum Bits

2

D-Wave machines
overview

3

the Quantum Annealing
process (overview)

4

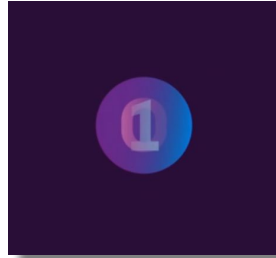
Resolution Process

5

End-to-end example:
the Max-Cut Problem

6

Hybrid Machines

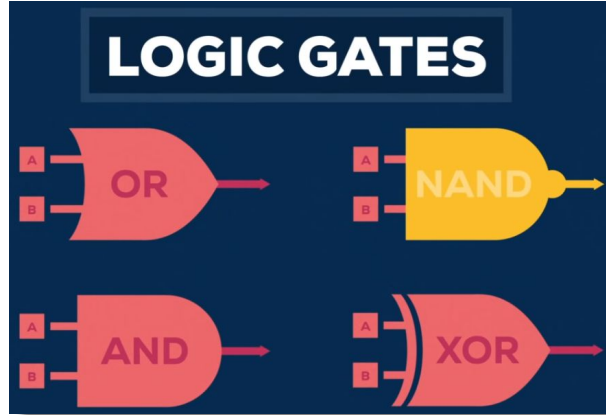
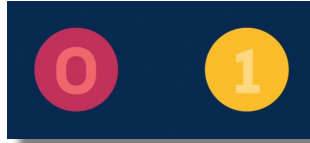


Outline



- Superposition
- Measure
- Entanglement
- Decoherence
- Overview of the 2 quantum computers paradigms

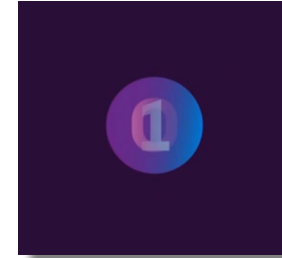
bits



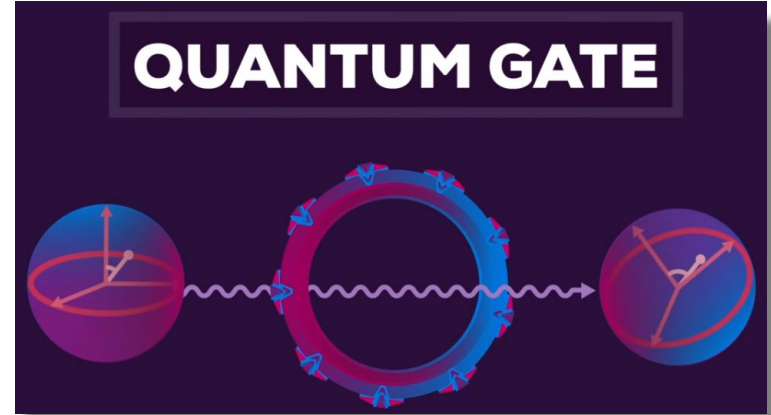
Versus



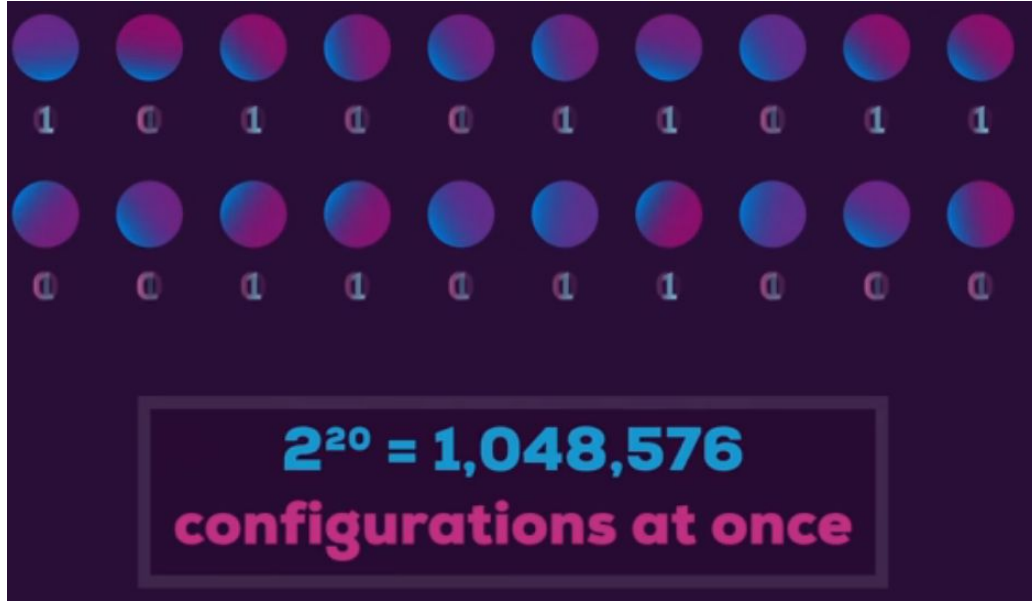
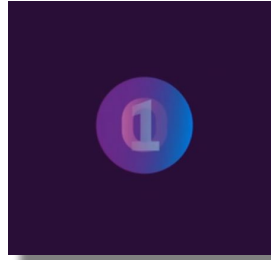
qubits



Superposition 0-1

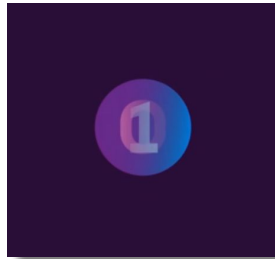


Qubit Superposition

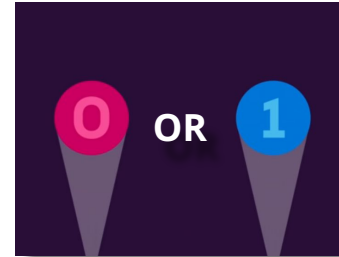
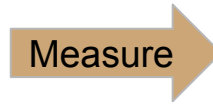


Measuring a qubit

From the superposition of states (0 and 1)
to the classical states (0 or 1)

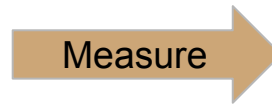
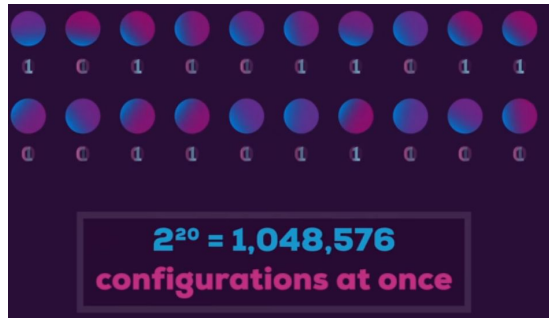


Superposition 0-1



Measure
(before the end of the
coherence time)

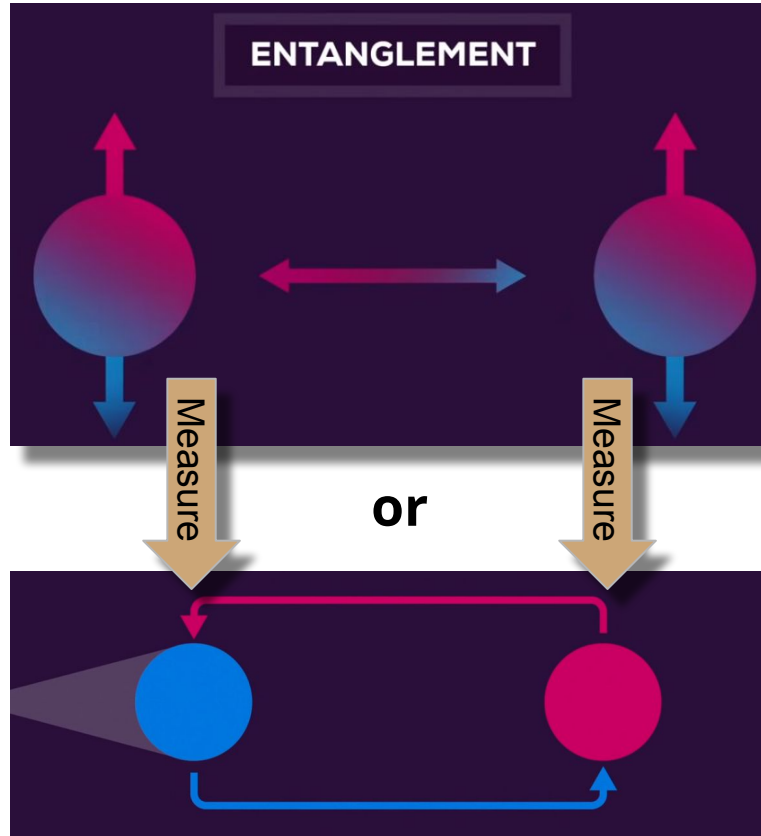
Example:



One solution
(20 qubits became 20 bits
with only one value)

The entanglement

(*sp. entrelazamiento*)



pic: https://www.youtube.com/watch?v=JhHMJCUMq28&ab_channel=Kurzgesagt%E2%80%9393InaNursteil

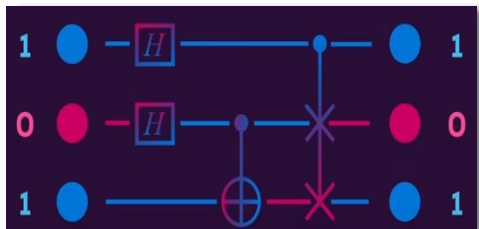
- 9

9



The two paradigms

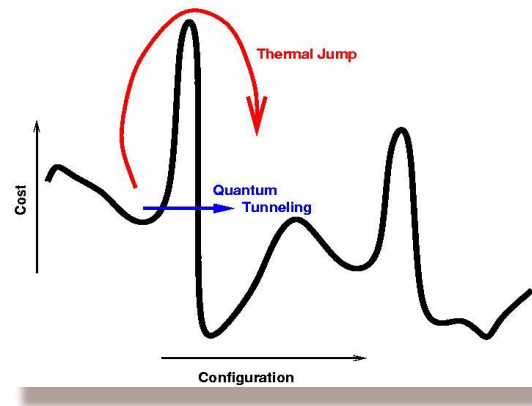
Computer based on Quantum Logic Gates



Main actors:

- IBM
- Google

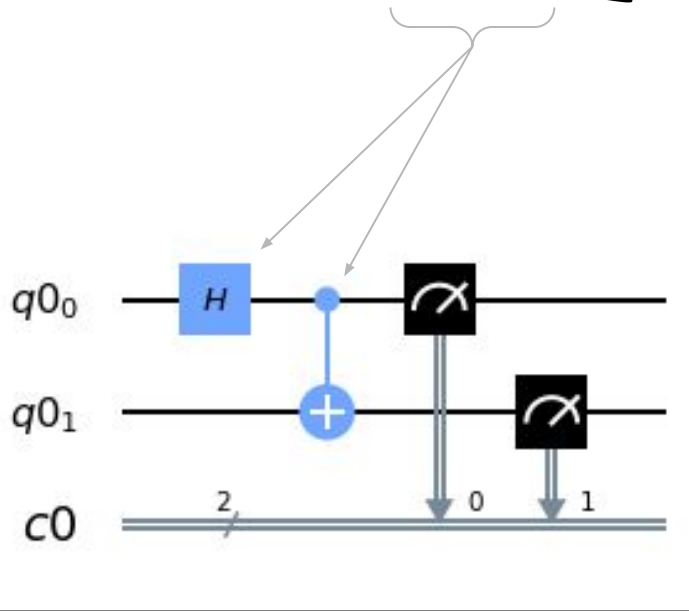
Computer based on Quantum Annealing



Main actors:

- D-Wave
- Qilimanjaro (spain)

in short... Gates Quantum Computers



Main quantum algorithms:

- Shor (factorize an integer N in time $O((\log N)^3)$)
- Grover (quadratic acceleration to find a value)
- QAOA (Quantum Approximate Optimization Algorithm, Nikolaj et al., 2018)

Main APIs:

- Qiskit (IBM)
- Cirq (Google)

But today, even if we can express and solve a very tiny TSP with these gates (QUBO), the result is not good, there is not enough qubits and have a high error rate.

Wait and see...



Outline

2

D-Wave machines overview

- Machines based on quantum annealing
- Hardware
- Software
- Some applications

D-Wave and Quantum Annealing



Hidetoshi Nishimori

- Father of the **Quantum Annealing Metaheuristic**
 - Works on Statistical physics,
 - Paper: Kadowaki, Tadashi, and Hidetoshi Nishimori. "Quantum annealing in the transverse Ising model." *Physical Review E* 58.5 published in **1998**;
- Based on classical computer Metaheuristic **Simulated Annealing**
 - Kirkpatrick, Scott, C. Daniel Gelatt, and Mario P. Vecchi. "Optimization by simulated annealing." *science* 220.4598 (1983): 671-680;
- Prove that the Quantum Annealing gives better results than the Simulated Annealing,
- A good conference from him: <https://www.youtube.com/watch?v=OQ91L96YWck>.



- First (hardware) implementation of the Quantum Annealing,
- Starting the work in **1999**, the 1st computer based on quantum annealing released in **2007**.

Notable Funders

- InQTel (CIA funds),
- Lockheed Martin,
- NSA funds
 - (see Research & Technology Penetrating hard targets' Snowden File)
- and others.

≤ 2013



- (S//SI//REL TO USA, FVEY) Conduct basic research in quantum physics and architecture/engineering studies to determine if, and how, a cryptologically useful quantum computer can be built.

<https://edwardsnowden.com/2014/01/04/penetrating-hard-targets-and-owning-the-net/>

Notable Contracts

Volkswagen, NSA, CIA, NASA, Lockheed Martin, etc.



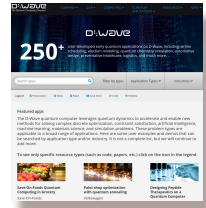
Scope

<https://www.dwavesys.com/applications>

Mainly: IA & **Optimization**.

Examples:

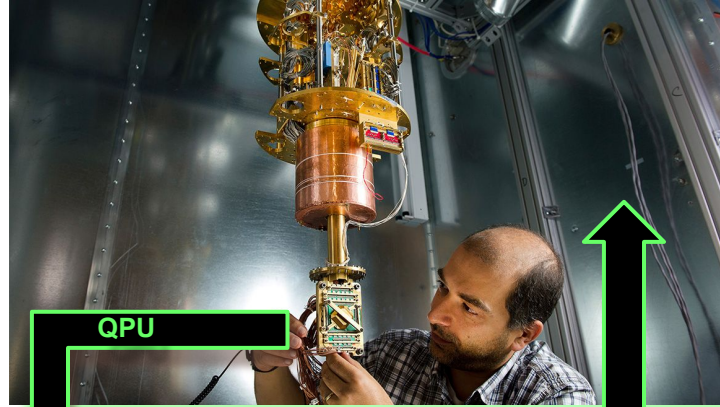
- Logistics,
 - trucks, boats,
 - pick-up and delivery;
- Production planning,
- Design Object,
- Chemistry simulation,
- [Geo]politics.



Hardware: inside a D-Wave quantum machine



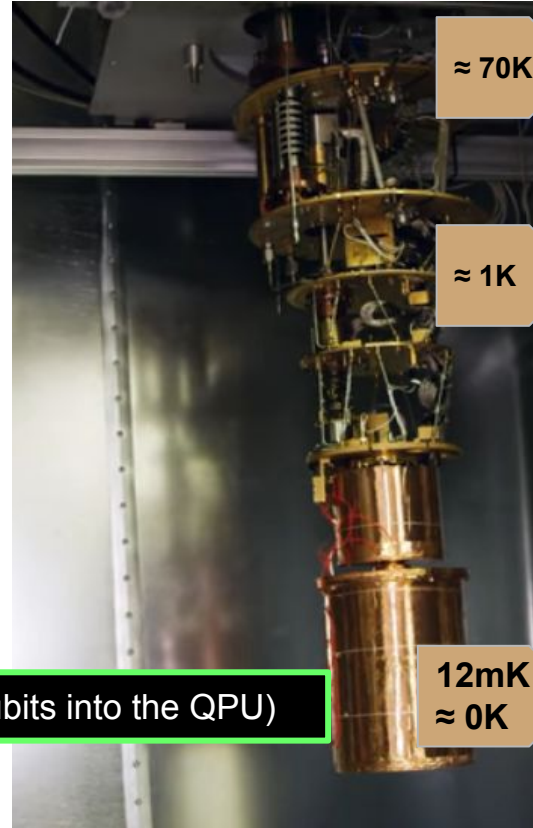
Total Consumption
Example of a 1K qubits machine:
- **25kw** (16kw in cryo').
As a comparison:
26248kw for the Fujitsu A64FX (the most powerful in 02/2021).



QPU

Faraday cage
(for blocking block electromagnetic fields.)

Set of D-Wave Nodium superconductors (qubits into the QPU)



≈ 70K

≈ 1K

12mK
≈ 0K

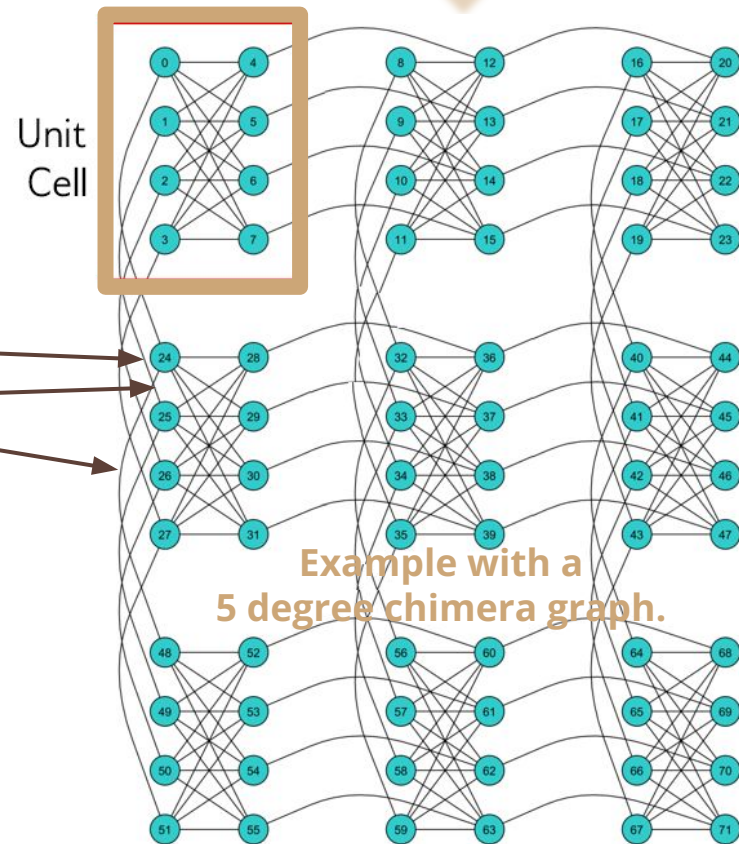
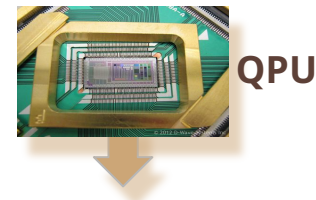
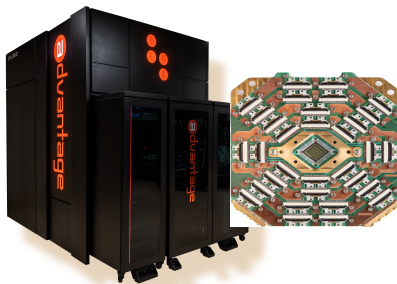
Layers of cooling mechanisms

Chimera Graph

(D-Wave hardware model)

D-Wave's quantum computer hardware can be modeled as a non-oriented graph called chimera graph. This graph has:

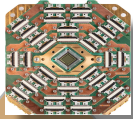
- **qubits** as nodes,
 - quantum **couplers** as edges.
 - since end of 2020 (*Pegasus* topology),
the degree of the qubits is 15 (6 before).
- > 5k qubits
 - 35000 couplers



D-Wave machines: from 2007 to 2020

Advantage QPU:

- 35,000 couplers
- 15 Connectivity



Qubits number in D-Wave machine & Gates based Quantum Computer?



-

Best:
5k qubits

In Quantum Annealing, you let a sort of **natural evolution** of quantum states. **You do not have any control on this evolution:**

- You set-up, launch, wait *ms*, and measure to obtain the solution, **(no program!)**
- Not able to create any possible quantum circuit.



Best:
127 qubits

In Universal Gates Quantum Computers, the aim is way more ambitious, you control the evolution of that quantum state over time:

- **More difficult to build since the quantum system is very delicate to work with, <<< This is why you find a big difference in the number of qubits. >>>;**

To summarize: “way more qubits, but able to solve way less types of problems”.

Frameworks: Background

- **Cirq**
 - Google
 - <https://cirq.readthedocs.io/en/stable/>
- **Qiskit**
 - IBM
 - <https://quantum-computing.ibm.com/>
- **Amazon Braket**
 - <https://aws.amazon.com/fr/braket/>
- and others:
 - Rigetti (Forest/PyQuil),
 - Microsoft (AzureQuantum/LiQui/Q#),
 - From labs: PyTKET, ProjectQ, QuTip etc.)



- **D-Wave Leap**
 - Ocean Library (python)
 - “High Level” programming
 - <https://www.dwavesys.com/take-leap>

<https://ide.dwavesys.io/#https://github.com/SaPlanq/DWAVEQuantum>

How can I
solve a
problem on
a D-Wave
machine?

Your github address

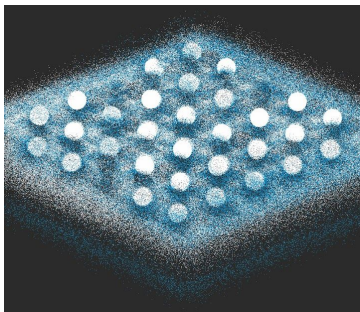
You launch the
resolution directly here,
and you will have an
online terminal showing
the result.

The solution obtained by D-Wave's quantum annealer Advantage_system1.1 is

	0	1	2	3	4	energy	num_oc.	chain_.
0	-1	+1	+1	-1	+1	-4.0	1398	0.0
1	+1	-1	-1	+1	-1	-4.0	1029	0.0
2	-1	+1	+1	-1	-1	-4.0	1487	0.0
3	+1	-1	-1	+1	+1	-4.0	1084	0.0
4	-1	+1	+1	+1	-1	-2.0	1	0.0
5	+1	-1	-1	-1	+1	-2.0	1	0.0

['SPIN', 6 rows, 5000 samples, 5 variables]

Outline



the Quantum Annealing process (overview)

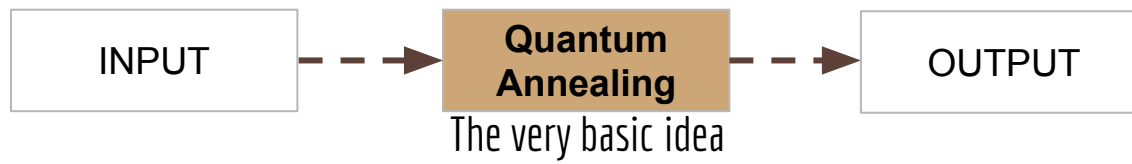
- From superposition to classical states

Quantum annealing (fr : recuit quantique) & Optimization Problems

“Physics can help solve [Optimization Problems] because we can frame them as energy minimization problems. “

No guaranty that quantum annealing will reach the **ground state** (i.e., the global optimum of the problem / Optimal Solution minimizing the cost).
... But, it might **give a better solution than a classical heuristic**.

“The D-Wave system can be viewed as a **hardware heuristic** that **minimizes [..] objective functions** using a physically realized version of quantum annealing.”



First Step

The qubits start in a **superposition** state (**zero and one**).

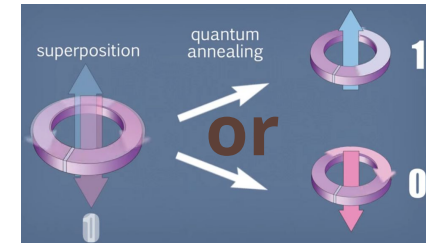


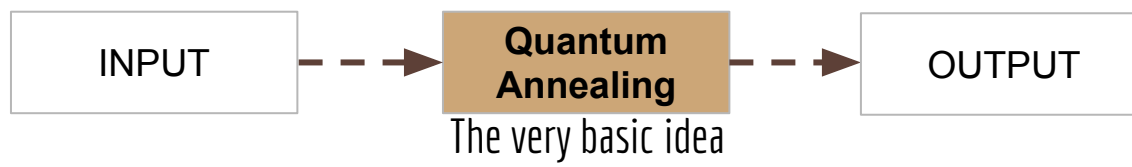
and

Work on the **0 and 1** states probabilities (amplitudes).

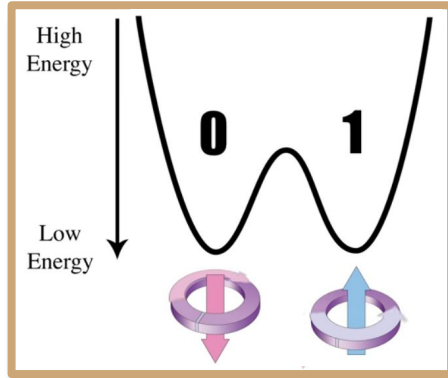
Last step

The measure of each qubit make them going to either the **zero state** **or** the **one state** (according to their probability).





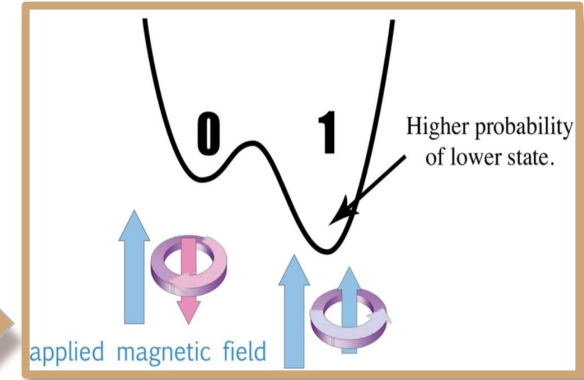
Energy Diagram (Superposition)



Starting Point.
Proba = 50%
for each state.

Control the probability of the state with a programmable external magnetic field (**bias**)

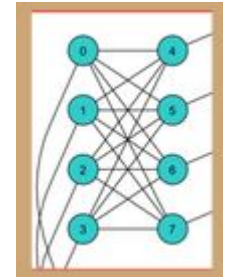
Energy Diagram (After Bias)



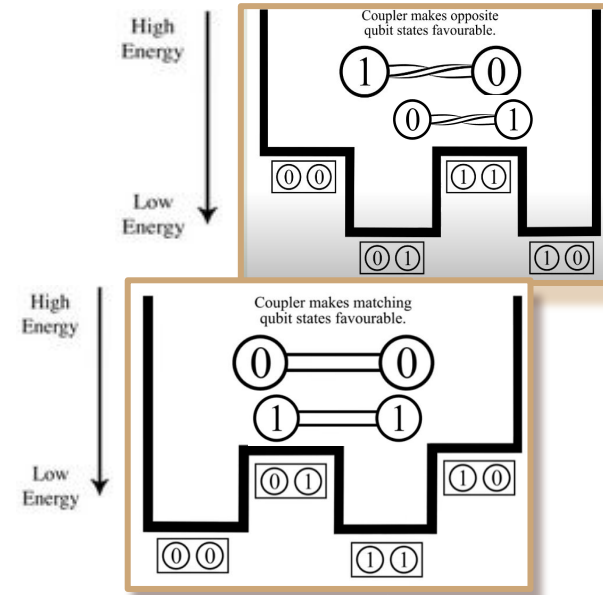
Here: higher proba to
measure the state '1'.

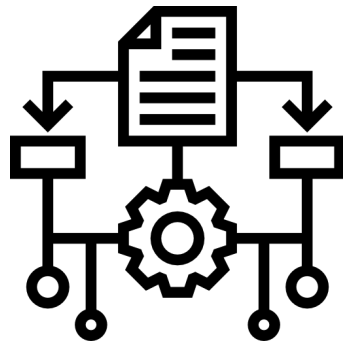


Couplers & Entanglement



- The “real power” of such a system appears once we start to link the qubits together
 - by **couplers** (using **entanglement** behaviour).
 - The couplers define how a pair of qubits ends up, in the same or in opposite state, (0&0, 1&1 or 1&0, 0&1).
- Once 2 qubits are entangled, they can be considered as a **single object**, but with 4 states.
 - When we consider a pair of qubits to be the same (or to be opposite) the machine will “physically” make the equality (or the inequality) “energetically favorable”, i.e., to “low the energy” of those states.





Outline

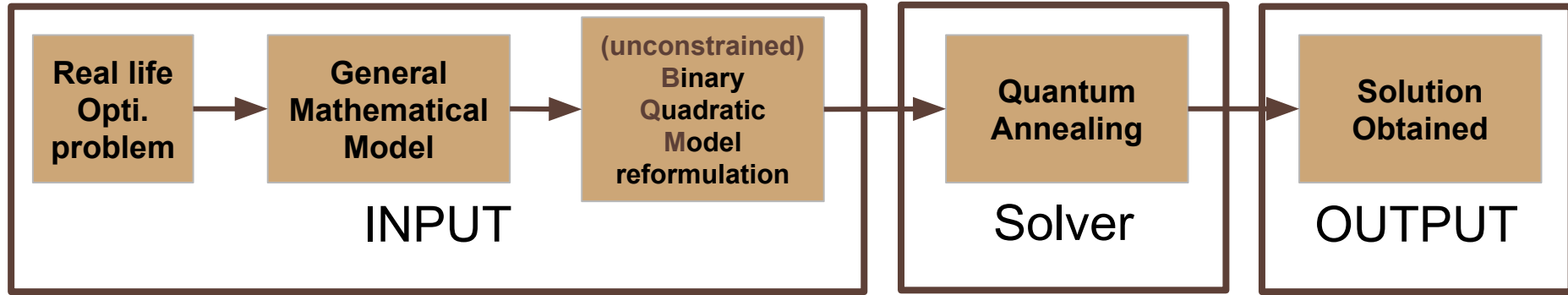


Resolution Process

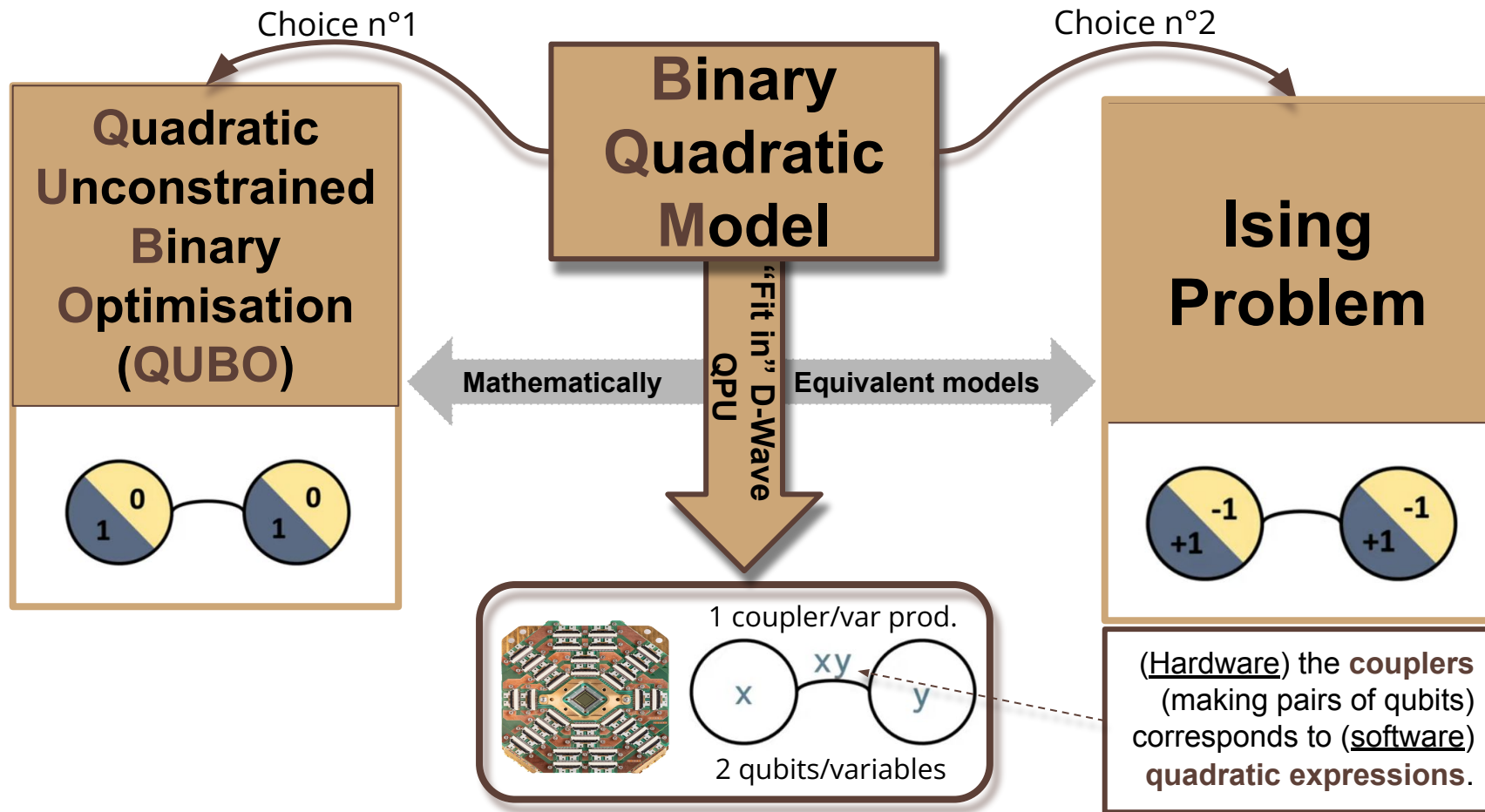
- QUBO & ISING

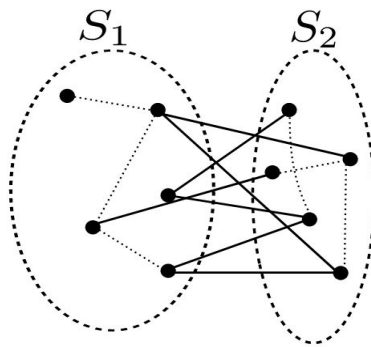
General Process

INPUT > QUANTUM ANNEALING > OUTPUT



Problem formulation: 2 mathematically **equivalent** models



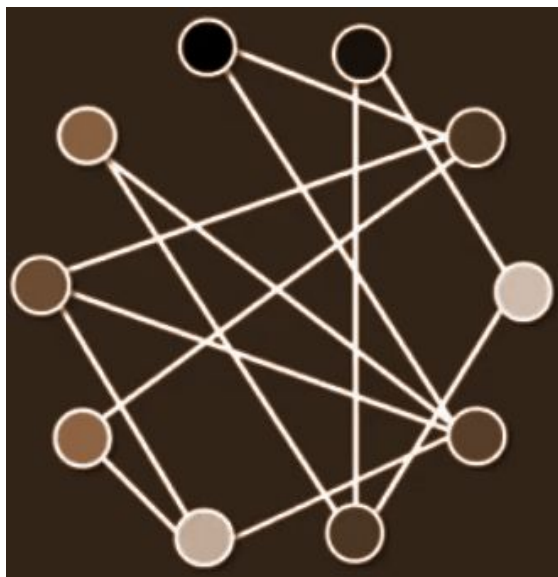


Outline

5

End-to-end example: the Max-Cut Problem

- Problem Definition
- Ising Model Formulation
- Solving one example

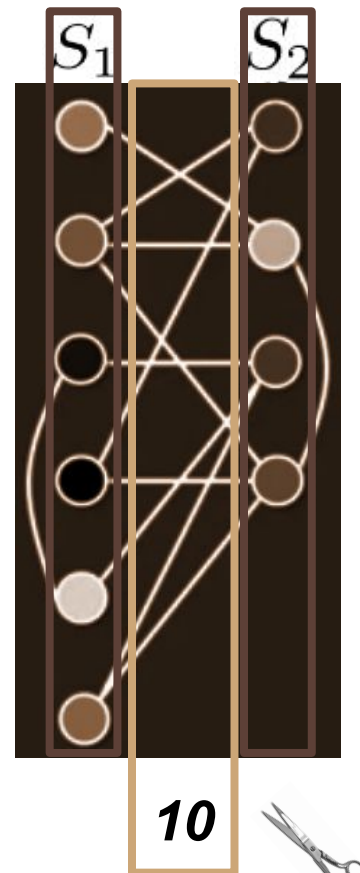


INPUT

The Max-Cut Problem

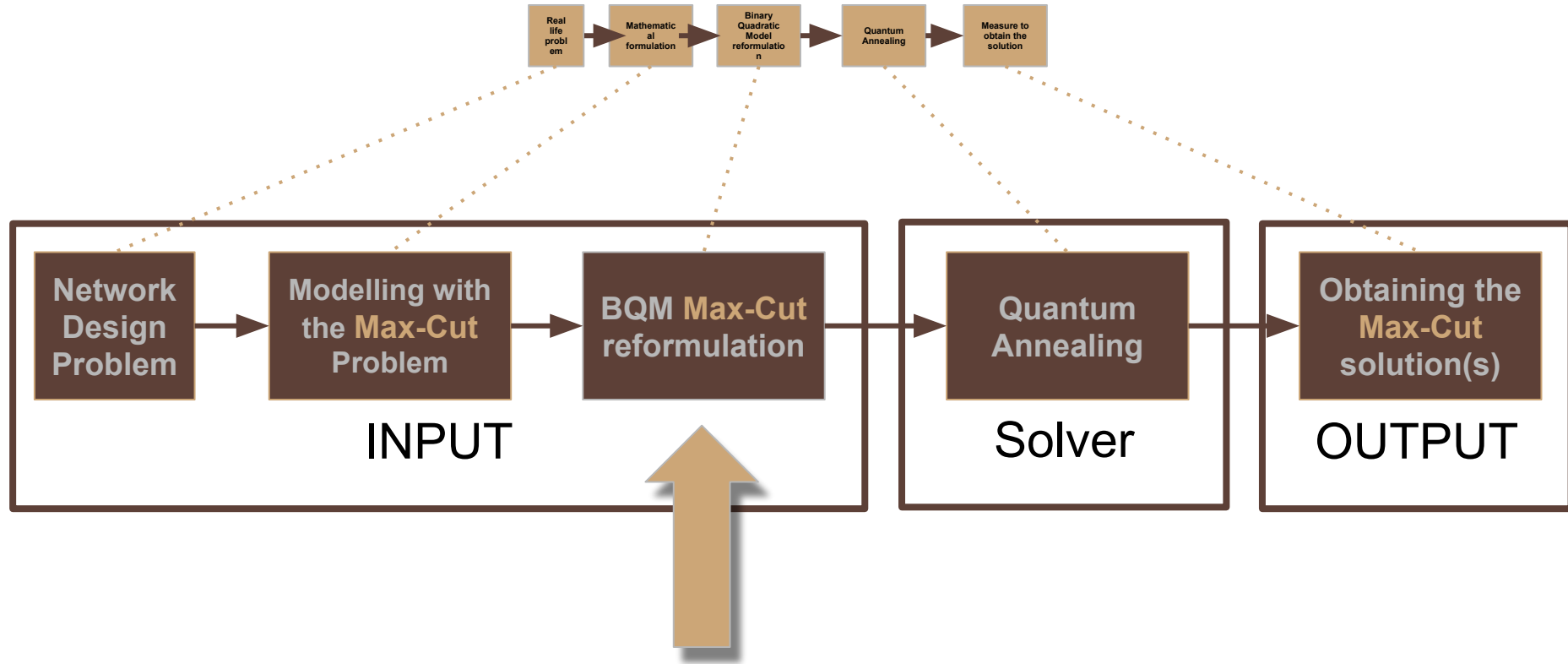
Finding 2 sets of nodes S_1 and S_2 , such that the **number of edges** between the set S_1 and the set S_2 is **as large as possible**.

Complexity: NP-HARD.



OUTPUT

General Process to solve the Max-Cut Problem with a D-Wave machine

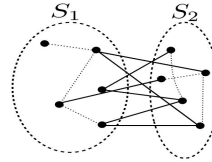
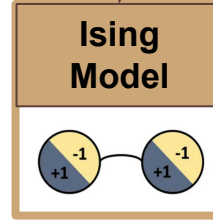


BQM Max-Cut reformulation

- Let's take a Graph $G=(X,E)$, where: X is the set of nodes and E the set of edges.

- We can identify each node i of the graph with a binary variable Z_i such that:

- $Z_i = 1$ if i is in one group S
- $Z_i = -1$ if i is in the other group.



- The Max-Cut Problem consists in finding the vector Z of $|X|$ binaries which **MINIMIZES**:

$$\sum_{(i,j) \in E} Z_i Z_j$$

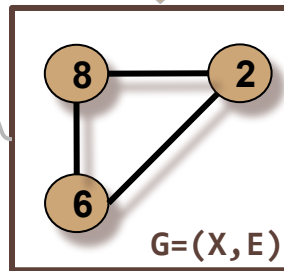
- Exemple. If two nodes (i,j) are in **different** groups:
- $Z_i Z_j = -1$,
- $Z_i Z_j = 1$ otherwise.

Let's take a Graph $G=(X,E)$, where X is the set of nodes and E the set of edges. We can identify each node i of the graph with a binary variable Z_i such that:

- $Z_i = 1$ if i is in one group,
- $Z_i = -1$ otherwise.

Ising Model

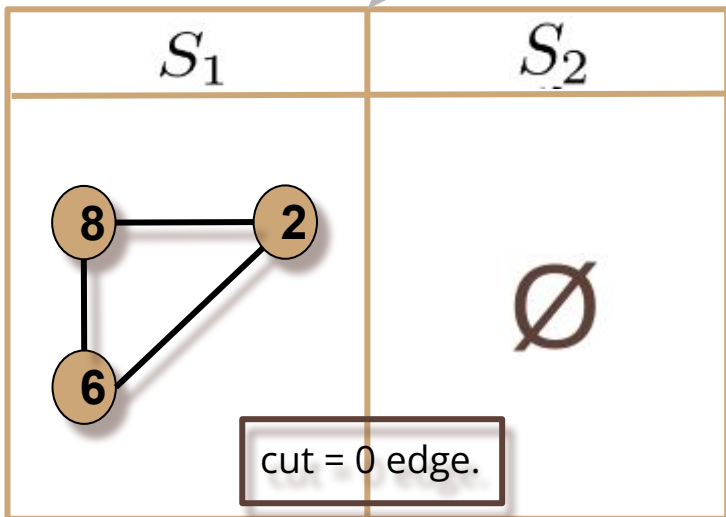
Small Example with a K3 Complete Graph



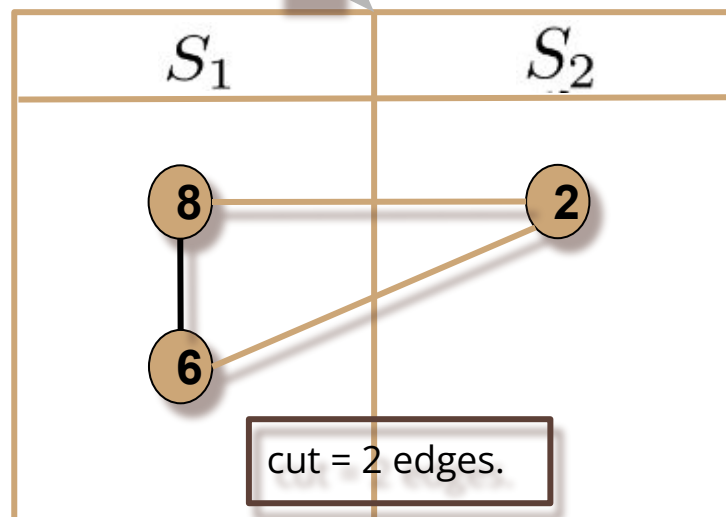
The Max-Cut Problem consists in finding the vector Z of $|X|$ binaries which minimizes:

$$\sum_{(i,j) \in E} Z_i Z_j$$

Solution 1



Solution 2



$$f(\text{Solution 1}) = 3$$

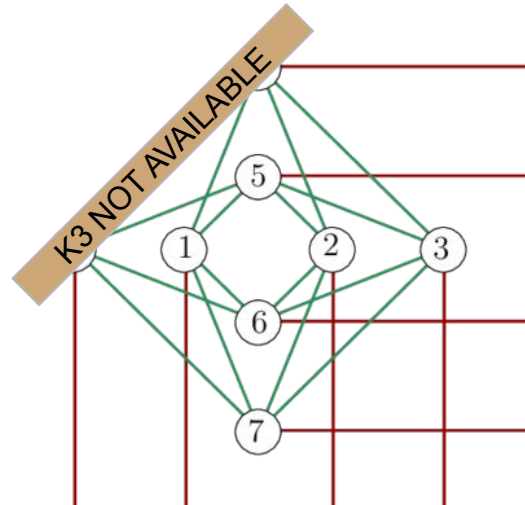
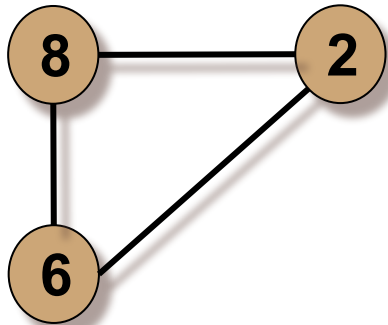
$$f(\text{Solution } i) = Z_2 Z_6 + Z_6 Z_8 + Z_8 Z_2$$

$$f(\text{Solution 2}) = -1$$

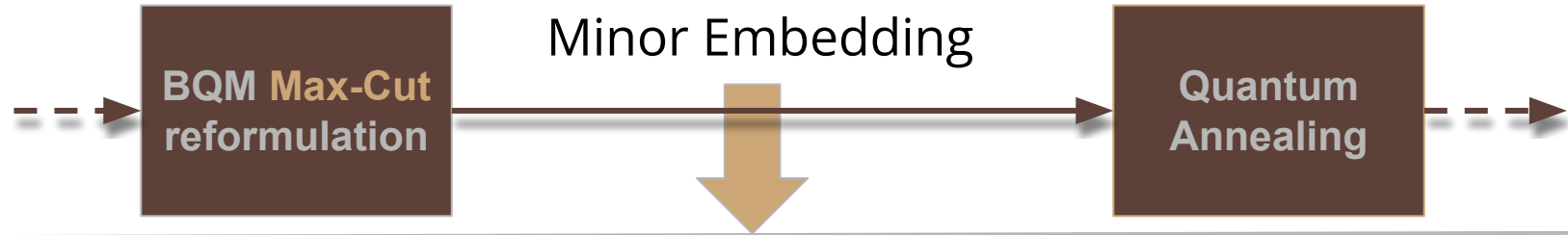
Problem of low connectivity

$$H = Z_2 Z_6 + Z_6 Z_8 + Z_8 Z_2 + \dots \text{blablar} \dots$$

That requires the following K3 complete graph in the hardware:



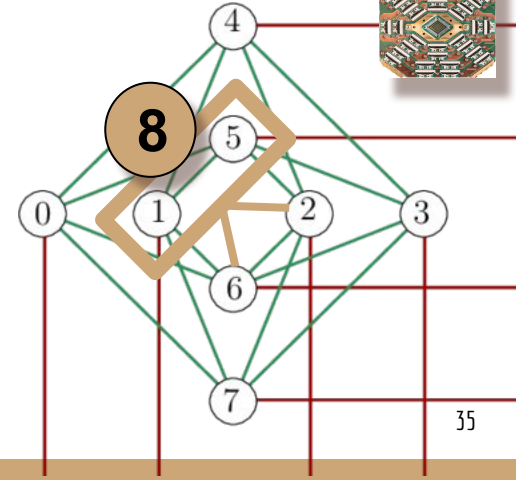
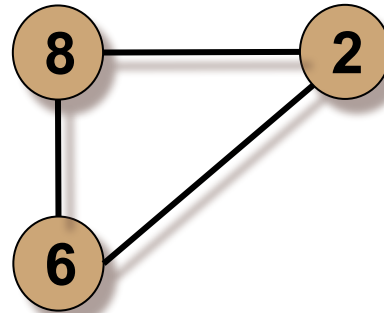
Note: From the Max-Cut Formulation as input To the (hardware) mapping into the QPU



Solution: **1 logical qubit from 2 real qubits**

$$H = Z_2 Z_6 + Z_6 Z_8 + Z_8 Z_2 + \dots \text{blablar} \dots$$

- Generation of **logical qubits**
 - using several physical qubits to "simulate" this connectivity.

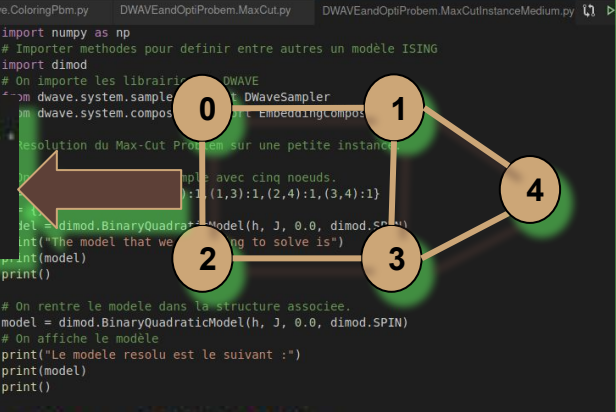


Example of **code** for the Max Cut Problem

```
# We enter the 6 edges of our graph with a weight equals to 1.  
J = {(0,1):1,(0,2):1,(1,3):1,(1,4):1,(2,3):1,(3,4):1}  
h = {} # We do not have external magnetic field in this case
```

No external field
(biais)

Couplers / edges
no weight (':1')



```
model = dimod.BinaryQuadraticModel(h, J, 0.0, dimod.SPIN)
```

```
## Quantum Annealing of D-WAVE Advantage Machine ##  
sampler = EmbeddingComposite(DWaveSampler(solver='Advantage_system1.1'))  
sampler_name = sampler.properties['child_properties']['chip_id']  
response = sampler.sample(model, num_reads=5000)  
print("The solution obtained by D-Wave's quantum annealer", sampler_name, "is")  
print(response)
```



computers)

Number of anneals

```
## Send problem to Sampler  
sampleset = sampler.sample_qubo(Q)
```

Example for a QUBO.

```
53  
54  
55  
56 ## RESOLUTION APPROCHEE ##  
57 Recuit quantique sur Machine D-WAVE  
58 sampler = EmbeddingComposite(DWaveSampler(solver='Advantage_system1.1'))  
59 sampler_name = sampler.properties['child_properties']['chip_id']  
60 response = sampler.sample(model, num_reads=5000)  
61 print("The solution obtained by D-Wave's quantum annealer", sampler_name, "is")  
62 print(response)  
63 print()  
64
```


D-Wave Quantum Annealing results

For sure we did not use all the qubits...

The solution obtained by D-Wave's quantum annealer `Advantage_system1.1` is

	0	1	2	3	4	energy	num oc.	chain
0	-1	+1	+1	-1	+1	-4.0	1398	0.0
1	+1	-1	-1	+1	-1	-4.0	1029	0.0
2	-1	+1	+1	-1	-1	-4.0	1487	0.0
3	+1	-1	-1	+1	+1	-4.0	1084	0.0
4	-1	+1	+1	+1	-1	-2.0	1	0.0
5	+1	-1	-1	-1	+1	-2.0	1	0.0

['SPIN', 6 rows, 5000 samples, 5 variables]

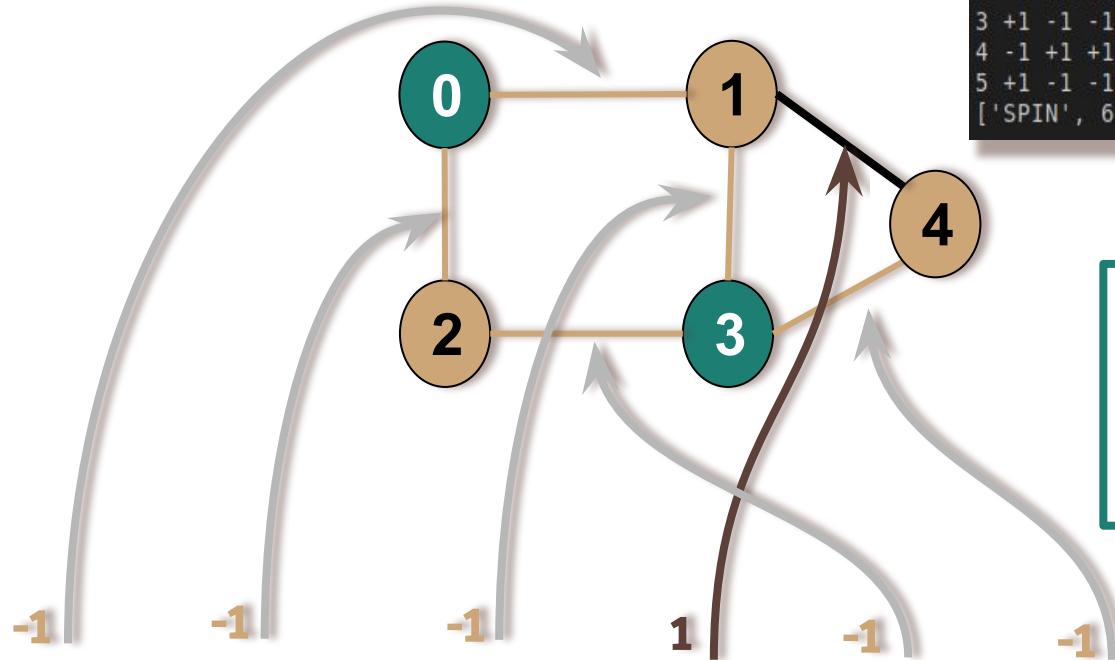
Number of occurrences during the 5000 anneals

$$H = Z_0 Z_1 + Z_0 Z_2 + Z_1 Z_3 + Z_1 Z_4 + Z_2 Z_3 + Z_3 Z_4 = -5 + 1 = -4$$

$$Z_0 = -1 \quad Z_1 = 1 \quad Z_2 = 1 \quad Z_3 = -1 \quad Z_4 = 1$$

Max-cut-Problem: From the solution to the graph.

	0	1	2	3	4	energy	num	oc.	chain
0	-1	+1	+1	-1	+1	-4.0	1398	0.0	
1	+1	-1	-1	+1	-1	-4.0	1029	0.0	
2	-1	+1	-1	-1	-1	-4.0	1487	0.0	
3	+1	-1	-1	+1	+1	-4.0	1084	0.0	
4	-1	+1	+1	+1	-1	-2.0	1	0.0	
5	+1	-1	-1	-1	+1	-2.0	1	0.0	
['SPIN', 6 rows, 5000 samples, 5 variables]									



We deduce
a **5 edges**
cut.

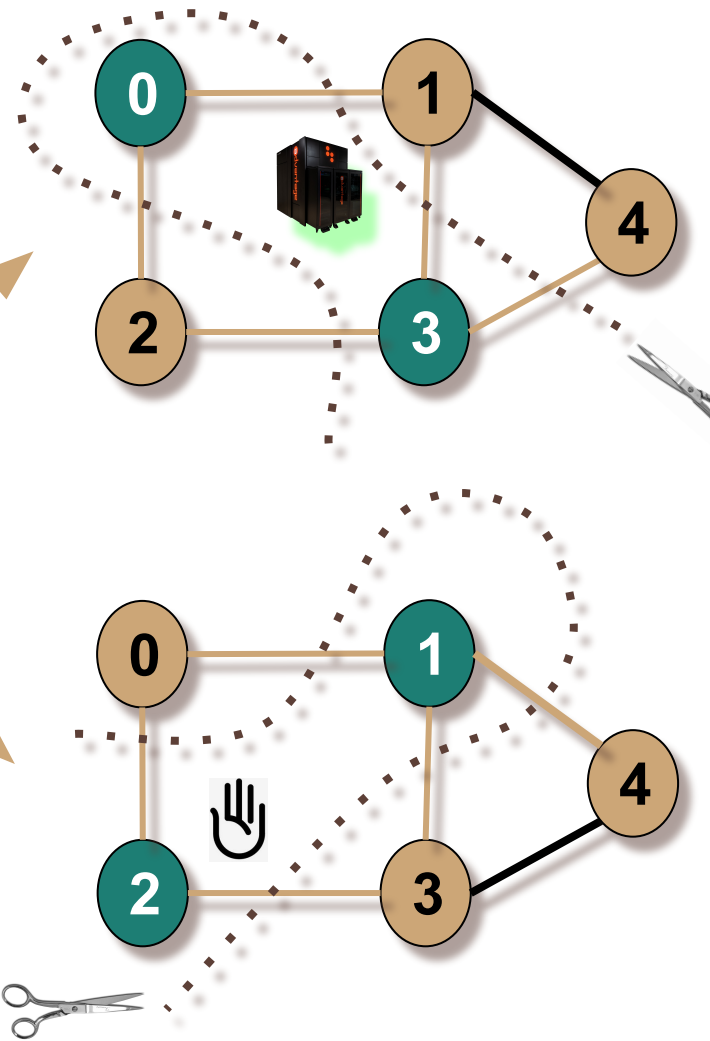
$$H = Z_0 Z_1 + Z_0 Z_2 + Z_1 Z_3 + Z_1 Z_4 + Z_2 Z_3 + Z_3 Z_4 = -5 + 1 = -4$$

$$Z_0 = -1 \quad Z_1 = 1 \quad Z_2 = 1 \quad Z_3 = -1 \quad Z_4 = 1$$

Max-cut-Problem: several optimal solutions

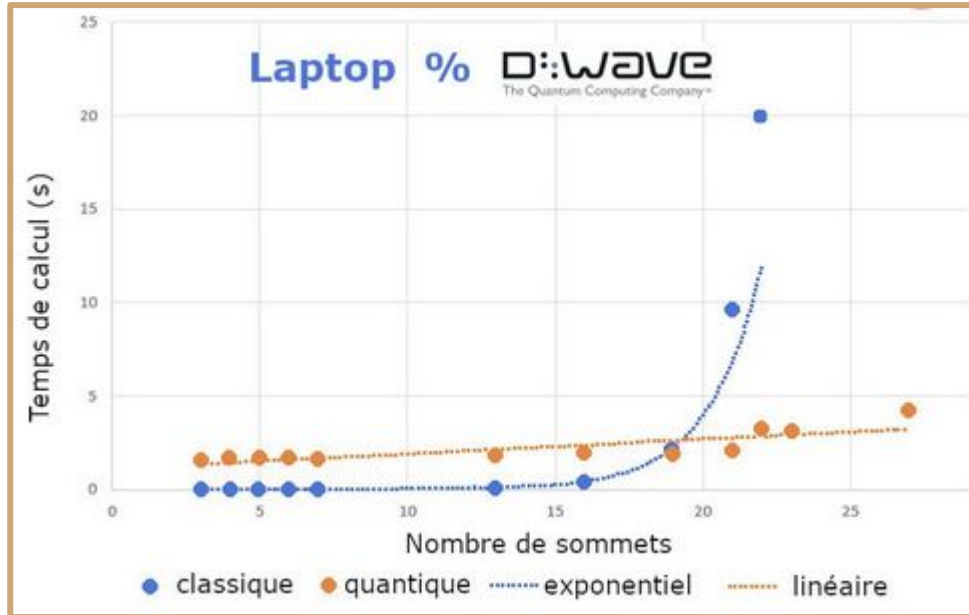
```
The solution obtained by D-Wave's quantum a
  0  1  2  3  4 energy num oc. chain .
0 -1 +1 +1 -1 +1 -4.0 1398 0.0
1 +1 -1 -1 +1 -1 -4.0 1029 0.0
2 -1 +1 +1 -1 -1 -4.0 1487 0.0
3 +1 -1 -1 +1 +1 -4.0 1084 0.0
4 -1 +1 +1 +1 -1 -2.0 1 0.0
5 +1 -1 -1 -1 +1 -2.0 1 0.0
['SPIN', 6 rows, 5000 samples, 5 variables]
```

We obtained ALL the optimal solutions.

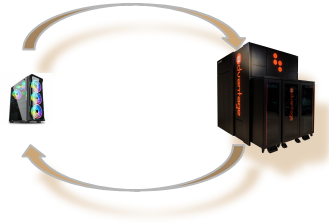


Some results

(Exact versus Quantum Annealing)



(next step: need to be compared with a software heuristic)



Outline



Hybrid Machines

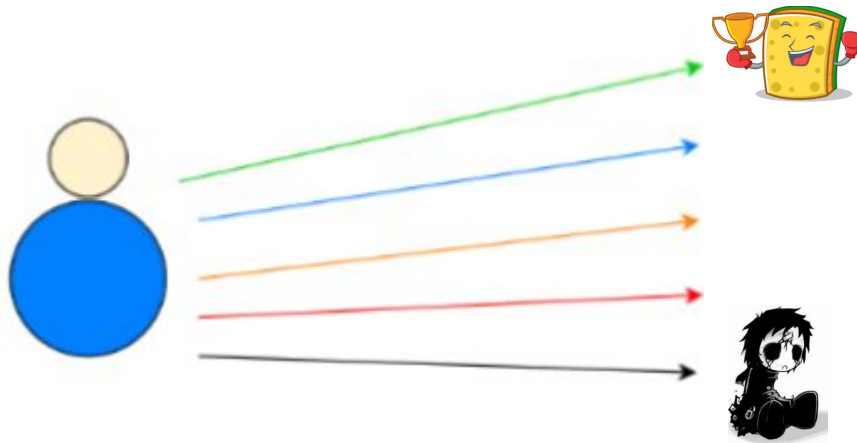
- Focus on few projects using quantum annealing
- *Note: hybrid computers starts to be used for the gate model paradigm.*

Hybrid computers

(for both paradigms)



A Student Project about...Projects



- Each student select **5 projects** and sort **them from the most wanted to the 5th most wanted**.
- Each project has a window to bound the number of students.

Goal: Maximizes general contentment.

$$Obj = \min \sum x_{ij} w_{ij}$$



“Classical” (generalization)

Versus

“Quantum” (example)



$\forall \text{ project } j$

$$\sum_i^{n_students} x_{ij} \leq room_{max}$$

$\forall \text{ student } i$

$$\sum_j^{n_projects} x_{ij} = 1$$

$\forall \text{ project } j$

$$\sum_i^n x_{ij} \leq M \cdot activeProject_j$$

$\forall \text{ project } j$

$$\sum_i^n x_{ij} \geq 3 \cdot activeProject_j$$

$activeProject_j \in \{0, 1\}$ and $M = room_{max\ j}$

($n \Leftrightarrow n_students$)

$$\min \left(\sum_i^9 \sum_j^3 x_{ij} w_{ij} \right) + \gamma_0 \left(\sum_i^9 x_{i0} - 3 \right) + \gamma_1 \left(\sum_i^9 x_{i1} - 4 \right) + \gamma_2 \left(\sum_i^9 x_{i2} - 2 \right) + \gamma_3 \left(\sum_j^3 x_{0j} - 1 \right)^2 + \gamma_4 \left(\sum_j^3 x_{1j} - 1 \right)^2 + \gamma_5 \left(\sum_j^3 x_{2j} - 1 \right)^2 + \dots + \gamma_{11} \left(\sum_j^3 x_{8j} - 1 \right)^2$$

Relaxation of all the constraints to obtain a Quadratic Unconstrained Binary Optimization Model (QUBO).

Pour le projet 0 (3/4)

étudiant : 92 => anthony.monvoisin@isen.yncrea.fr

étudiant : 93 => martin.valet@isen.yncrea.fr

étudiant : 98 => victorien.roussel@isen.yncrea.fr

Pour le projet 1 (5/5)

étudiant : 19 => thibaut.blassele@isen.yncrea.fr

étudiant : 52 => joshua.ofori@student.yncrea.fr

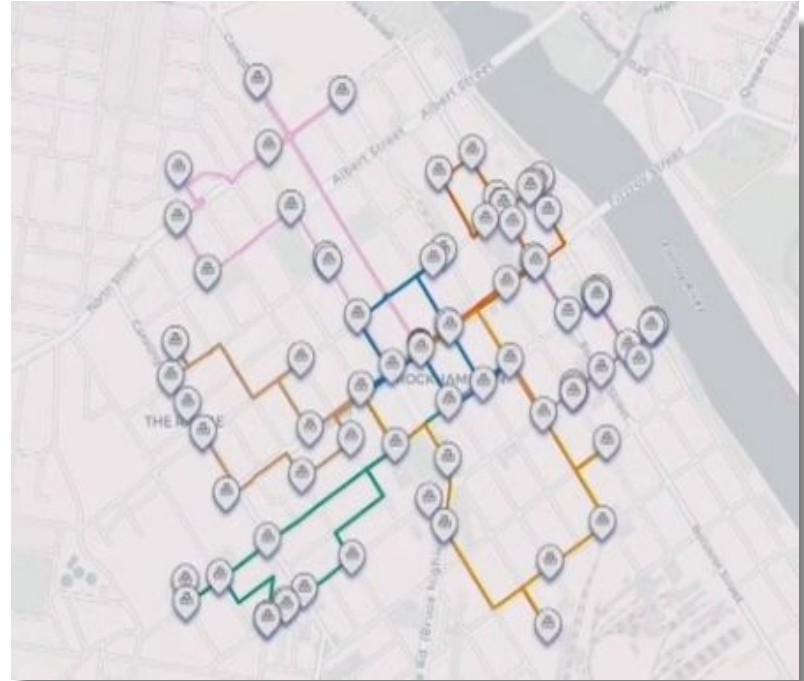
étudiant : 120 => theo.ruchot@isen.yncrea.fr

étudiant : 125 => nicolas.defoort@isen.yncrea.fr

étudiant : 153 => lilian.houplin@isen.yncrea.fr

NEC experiment: CVRP

(David Garvin, Nec Australia at Qubits conf. 2021-10)



Constraints: from one depot node, supply of the other nodes with 1 of the **K vehicles** with a capacity,
Objective: Min total distance.

NEC experiment: CVRP

(David Garvin, Nec Australia at Qubits conf. 2021-10)

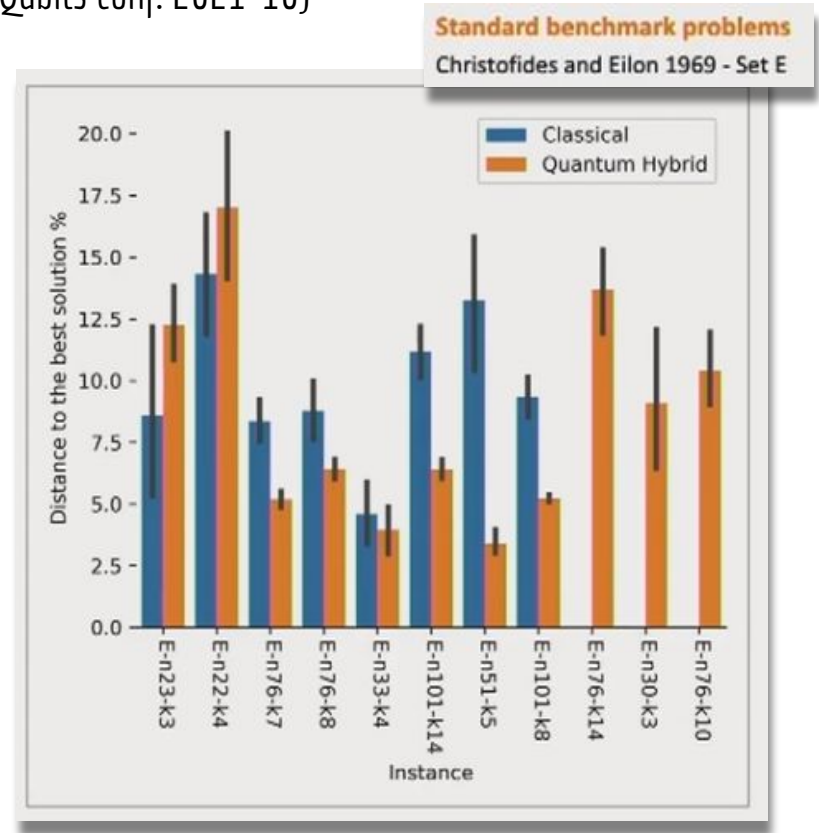
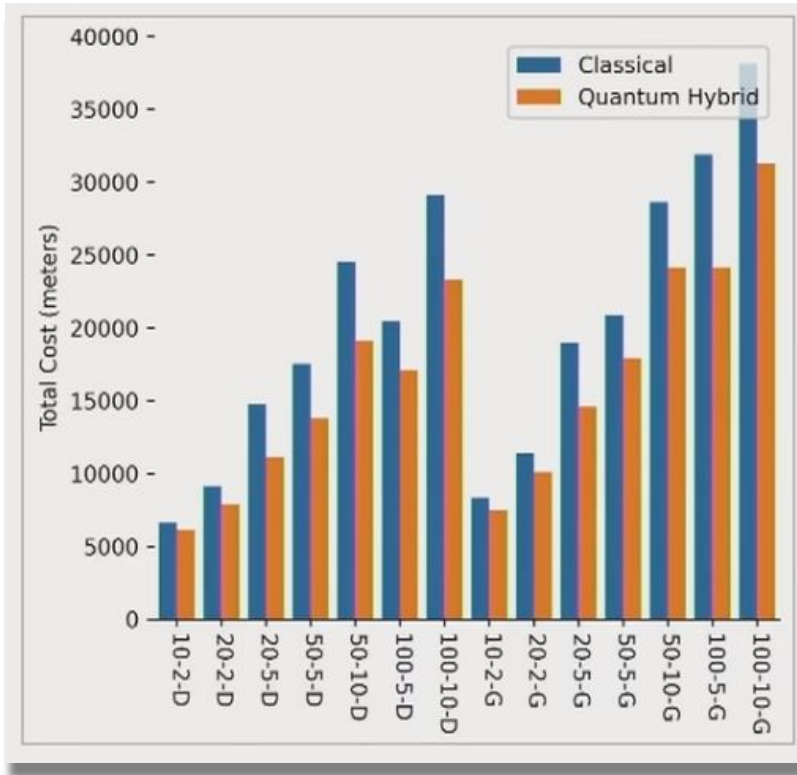
- Resolution Scheme:
 - 1. **Clustering** (for each vehicle)
 - quadratic formulation
 - quantum computer
 - 2. **Routing** (for each vehicle)
 - basic heuristic solving the TSP
 - classical computer

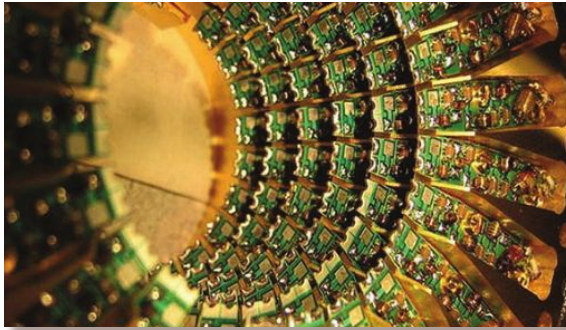
min problem

$$\begin{aligned} f_{\text{clustering}} &= \sum_{i,j} \sum_k c_{ij} x_{ik} x_{jk} \\ \sum_i x_{ik} u_i &\leq C_k \\ \sum_k x_{ik} &= 1 \end{aligned}$$

NEC experiment: CVRP

(David Garvin, Nec Australia at Qubits conf. 2021-10)





Outline

Summary,
General Results, and
Conclusions

Summary

- **Another paradigm:** Method NOT based on universal gates models,
 - But work on smaller number of problems as ones in machine learning and optimization fields;
- **Huge number of qubits, but not controllable as with the universal quantum computers,**
 - Not programmable, no universal gates to apply during the process,
- The Quantum Annealing could be called a “**Hardware Heuristic**” letting the physics solve the problem.
- Last Evolution: MILP API (but how do they map' ?)

Literature: Some Results on a 1K and 2K qubits Machine

- Vert, D., Sirdey, R., & Louise, S. (2021). Benchmarking Quantum Annealing Against “Hard” Instances of the Bipartite Matching Problem. SN Computer Science, 2(2), 1-12.
 - “we argue that unless much denser qubits interconnects are developed, it will be difficult for the approach to compete with classical algorithms on real-world problems” .
 - (But using a 6-connectivity graph with 2K qubits).
- (other document -french thesis)
 - Étude des performances des machines à recuit quantique pour la résolution de problèmes combinatoires. 2021, thèse de Daniel Vert.

Literature: Some Results on a 1K and 2K qubits Machine

- D-Wave 2X again (1K qubits) runs up to **100M times faster** than an optimized implementation of the quantum Monte Carlo algorithm **on a single core**.
 - Denchev, V.S., Boixo, S., Isakov, S.V., Ding, N., Babbush, R., Smelyanskiy, V., Martinis, J. and Neven, H., 2016. What is the computational value of finite-range tunneling?. Physical Review X, 6(3), p.031015.
- D-Wave newsletter of the **18/02/2021** about results on 2K machine:
 - “Today D-Wave marks a major milestone on the [journey to quantum advantage](#) in a new [peer-reviewed paper](#) published in *Nature Communications*. The new research uses a D-Wave lower noise system to demonstrate **3 million times speed-up over classical alternatives in a real-world problem**. [...]”
 - The [Paper](#): King, A.D., Raymond, J., Lanting, T. et al. Scaling advantage over path-integral Monte Carlo in quantum **simulation of geometrically frustrated magnets**. Nature Communication 12, 1113 (2021).
 - **They compared the QA against the best known classical simulation algorithm for this problem (PIMC).**
 - To know which CPU is used in comparison (also the number of threads used), we need to go in the [supplementary materials](#) (they use a light **i7-8650U on one core!** To compare, if it is still possible, and if the algo is fully parallelizable, the Fujitsu A64FX will win since it has more than 7 millions of (ARM) cores)!
- Be careful with all the announcements in the media...and also from the companies!
- But still... waiting for results on **Advantage**_(5k qubits and 15 connectivity)

Maybe not for all optimization problems...

min
s.t.

$$\sum_{r \in R} z_{n_r}$$

(1)

$$x_e^{end} = x_{e+1}^{begin}, \quad e \in RE_r, r \in R : e = 1..n_r - 1,$$

(2)

$$x_e^{end} \geq x_e^{begin} + d_e, \quad e \in E,$$

(3)

$$x_e^{begin} \geq \tau_e^{begin}, \quad e \in E : i_e = 1,$$

(4)

$$x_e^{end} - \tau_e^{end} \leq z_e, \quad e \in E,$$

(5)

$$\sum_{t \in T_s} q_{et} = 1, \quad e \in SE_s, s \in S,$$

(6)

$$q_{et} + q_{\hat{e}t} - 1 \leq \lambda_{e\hat{e}} + \gamma_{e\hat{e}}, \quad e, \hat{e} \in SE_s, t \in T_s, s \in S : e < \hat{e},$$

(7)

$$x_{\hat{e}}^{begin} - x_e^{end} \geq \Delta_s^M \gamma_{e\hat{e}} - M(1 - \gamma_{e\hat{e}}), e < \hat{e} \text{ in } SE_s, s \in S, o_e \neq o_{\hat{e}},$$

(8)

$$x_{\hat{e}}^{begin} - x_e^{end} \geq \Delta_s^F \gamma_{e\hat{e}} - M(1 - \gamma_{e\hat{e}}), e < \hat{e} \text{ in } SE_s, s \in S, o_e = o_{\hat{e}},$$

(9)

$$x_e^{begin} - x_{\hat{e}}^{end} \geq \Delta_s^M \lambda_{e\hat{e}} - M(1 - \lambda_{e\hat{e}}), e < \hat{e} \text{ in } SE_s, s \in S, o_e \neq o_{\hat{e}},$$

(10)

$$x_e^{begin} - x_{\hat{e}}^{end} \geq \Delta_s^F \lambda_{e\hat{e}} - M(1 - \lambda_{e\hat{e}}), e < \hat{e} \text{ in } SE_s, s \in S, o_e = o_{\hat{e}},$$

(11)

$$\lambda_{e\hat{e}} + \gamma_{e\hat{e}} \leq 1, \quad e, \hat{e} \in SE_s, s \in S : e < \hat{e},$$

(12)

$$x_e^{begin} \geq w_t^{end} q_{et} - M\alpha_e^{w_t}, \quad e \in SE_s, t \in T_s, s \in S,$$

(13)

$$x_e^{end} \leq w_t^{begin} q_{et} + M(1 - \alpha_e^{w_t}), \quad e \in SE_s, t \in T_s, s \in S,$$

(14)

$$x_e^{begin}, x_e^{end}, z_e \geq 0, \quad e \in E,$$

(15)

$$\gamma_{e\hat{e}}, \lambda_{e\hat{e}} \in \{0, 1\}, \quad e, \hat{e} \in SE_s, s \in S : e < \hat{e},$$

(16)

$$q_{et}, \alpha_e^{w_t} \in \{0, 1\}, \quad e \in SE_s, t \in T_s, s \in S,$$

(17)

Not a
relaxing
relaxation...



Future of the quantum annealing tech

- Several Projects in Europe:

- **AVaQus**

- the European project developing the first superconducting quantum annealer
 - https://www.quantaneo.com/AVaQus-the-European-project-to-develop-the-first-superconducting-coherent-quantum-annealer_a479.html

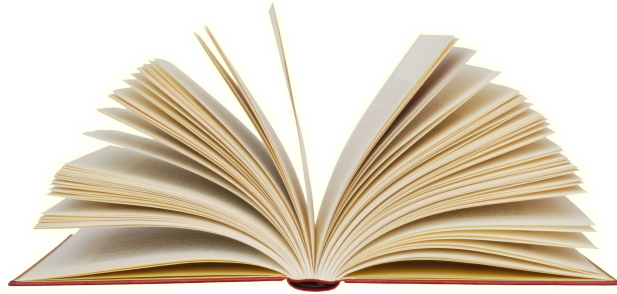
- **ATOS**

- Quantum Annealing **Simulator**
 - Scope: Machine Learning
 - https://atos.net/en/2020/press-release_2020_07_07/atos-opens-up-a-new-path-to-quantum-annealing-simulation

- **Qilimanjaro**



Outline



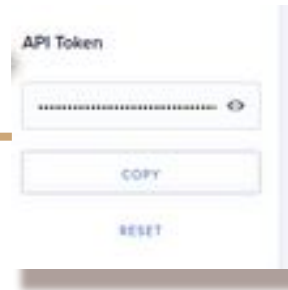
References & Co

- Some Help to install D-Wave Ocean API
- Several types of references:
 - Books, Vidéos, Conf', Papers, Courses,
 - From vulgarisation to theoretical.



Some help to install the D-Wave Ocean API on Debian based machine

- Linux commands :
 - `sudo apt-get install python<version>`
 - `sudo pip install virtualenv`
 - `virtualenv ocean`
 - `pip install dwave-ocean-sdk`
 - `git clone https://github.com/dwavesystems/dwave-ocean-sdk.git`
 - `cd dwave-ocean-sdk` ou `cd ocean`
 - `python setup.py install` ou `./python ./easy_install install`
 - Add to your cod the token available in your dashboard
- More detail: <https://docs.ocean.dwavesys.com/en/stable/overview/install.html>

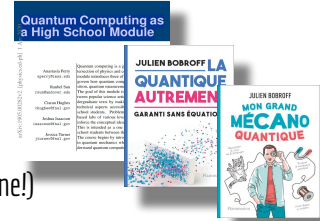


```
(ocean) spydel@spydel-NUC10i5FNH:~/Documents/vracCode/ocean/bin$ ls
activate      activate.ps1  easy_install  easy_install3.8  pip-3.8        python3         wheel3
activate.csh  activate_this.py  easy_install3  pip              pip3.8         python3.8       wheel-3.8
activate.fish activate.xsh     easy_install-3.8  pip3            python         wheel           wheel3.8
(ocean) spydel@spydel-NUC10i5FNH:~/Documents/vracCode/ocean/bin$ dwave config create
Configuration file not found; the default location is: /home/spydel/.config/dwave/dwave.conf
Configuration file path [/home/spydel/.config/dwave/dwave.conf]:
Configuration file path does not exist. Create it? [y/N]: y
Profile (create new) [prod]: spydel
API endpoint URL [skip]:
Authentication token [skip]:
Default client class [skip]:
Default solver [skip]:
Configuration saved.
(ocean) spydel@spydel-NUC10i5FNH:~/Documents/vracCode/ocean/bin$
```

General popularization works

(from the *easiest* to the *hardest*)

- Article (100p): Quantum Computing as a High School Module <https://arxiv.org/abs/1905.00282>
- Book (fr): La quantique autrement. Julien Bobroff.
- Book (fr): Mon grand mécano quantique. Julien Bobroff.
- Video: “Les Ordinateurs Quantiques Expliqués - Limites de la technologie humaine” (so well done!)
https://www.youtube.com/watch?v=JhHMJCUmq28&ab_channel=Kurzgesagt%E2%80%93InaNutshell
- Website article: “You don’t need to be a mathematician to master quantum computing”
<https://towardsdatascience.com/you-dont-need-to-be-a-mathematician-to-master-quantum-computing-161026af8878>
- Slides: US Department of energy: How about quantum computing. Bert de Jong
<https://cs.lbl.gov/assets/CSSSP-Slides/20190624-deJong.pdf>
- Videos: Understanding Quantum Mechanics. Sabine Hossenfelder.
<https://www.youtube.com/watch?v=XJSfgE9LUJw&list=PLwgQsqtH9H5dJlfFhXE6We207beTgUnyL>
- Book: Quantum computation and quantum information. Nielsen & Chuang



Research Papers

- Simulated Annealing: Kirkpatrick, Scott, C. Daniel Gelatt, and Mario P. Vecchi. "Optimization by simulated annealing." *science* 220.4598 (1983): 671-680.
- Vert, D., Sirdey, R., & Louise, S. (2021). Benchmarking Quantum Annealing Against "Hard" Instances of the Bipartite Matching Problem. *SN Computer Science*, 2(2), 1-12.
- Smelyanskiy, Vadim N., et al. "A near-term quantum computing approach for hard computational problems in space exploration." *arXiv preprint arXiv:1204.2821* (2012).
- Hamerly, R., Inagaki, T., McMahon, P.L., Venturelli, D., Marandi, A., Onodera, T., Ng, E., Langrock, C., Inaba, K., Honjo, T. and Enbutsu, K., 2019. Experimental investigation of performance differences between coherent Ising machines and a quantum annealer. *Science advances*, 5(5), p.eaau0823.
 - Here the MaxCut problem is solved on DW2Q has 2048 qubits (lets try with a 5K!)
- Barahona, Francisco; Grötschel, Martin; Jünger, Michael; Reinelt, Gerhard (1988). "An Application of Combinatorial Optimization to Statistical Physics and Circuit Layout Design". *Operations Research*. **36** (3): 493-513.
- Barahona. "On the Computational Complexity of Ising Spin Glass Models." *J. Phys. A* 15 (1982), pp. 3241-3253.
 - About the NP-Hard complexity of Ising and QUBO Problems
- Mengoni, Riccardo, Daniele Ottaviani, and Paolino Iorio. "Breaking RSA Security With A Low Noise D-Wave 2000Q Quantum Annealer: Computational Times, Limitations And Prospects." *arXiv preprint arXiv:2005.02268* (2020).
- King, James, et al. "Benchmarking a quantum annealing processor with the time-to-target metric." *arXiv preprint arXiv:1508.05087* (2015).
- Analytical and numerical evidence suggests that quantum annealing outperforms simulated annealing under certain conditions. Heim, B., Rønnow, T. F., Isakov, S. V., & Troyer, M. (2015). Quantum versus classical annealing of Ising spin glasses. *Science*, 348(6231), 215-217.
- Albash, Tameem, Victor Martin-Mayor, and Itay Hen. "Temperature scaling law for quantum annealing optimizers." *Physical review letters* 119.11 (2017): 110502.

Courses

- “A practical introduction to quantum computing: from qubits to quantum machine learning and beyond (5/7)”
 - 7 courses from the CERN Quantum Technology Initiative.
 - (5/7): “Quantum algorithms for combinatorial optimization. Quantum adiabatic computing and quantum annealing. Introduction to D-Wave Leap. Quantum Approximate Optimization Algorithm.”
 - <https://cds.cern.ch/record/2746545>
- John Preskill lectures:
 - <http://theory.caltech.edu/~preskill/ph229/>
- Serge Haroche (collège de France lectures)
 - https://www.college-de-france.fr/site/serge-haroche/_course.htm

D-Wave ressource

- Leap:
 - <https://cloud.dwavesys.com/leap/>
- Qubo & Ising Models
 - https://docs.dwavesys.com/docs/latest/c_gs_3.html
- D-Wave examples on Github
 - <https://github.com/dwave-examples>
- Tutos (Videos and articles) for developers
 - <https://www.dwavesys.com/practical-quantum-computing-developers>
- Tuto for installing the Ocean API:
 - <https://docs.ocean.dwavesys.com/en/stable/overview/install.html>
- A tons of videos from D-Wave:
 - <https://www.youtube.com/user/dwavesystems/playlists>



**This is
the end**

...

**Thanks
for your
attention!**

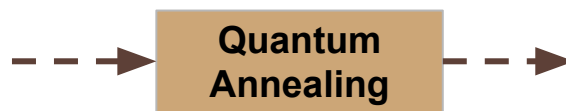
**Cat content?
You're well equipped.**

**Travelling Salesman?
Go quantum!**

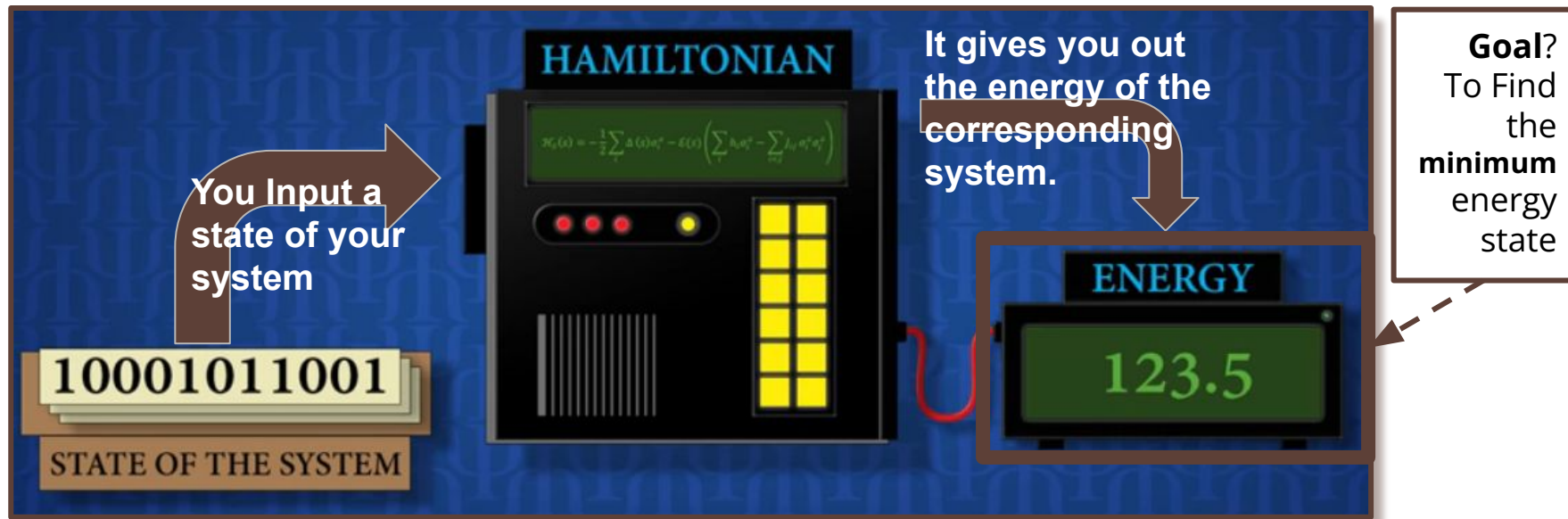
And now...
The rejected slides!

The Hamiltonian

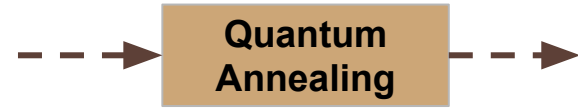
during the quantum annealing....



- The **Hamiltonian (H)** is (here) a mathematical description of the energy of a physical system. (here we focus on the energy inside the QPU).



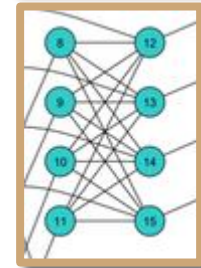
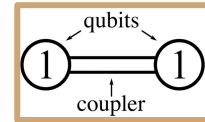
The Hamiltonian during the quantum annealing....



- The **Hamiltonian (H)** includes 2 types of interaction:
 - h_i : the “programmable” external magnetic field called a **bias**,
 - J_{ij} : the interaction between each neighboring qubits i and j (**coupling**).

$$\mathcal{H}_S(s) = -\frac{1}{2} \sum_i \Delta(s) \sigma_i^x + \mathcal{E}(s) \left(- \sum_i \boxed{h_i} \sigma_i^z + \sum_{i < j} \boxed{J_{ij}} \sigma_i^z \sigma_j^z \right)$$

Initial Hamiltonian Final Hamiltonian

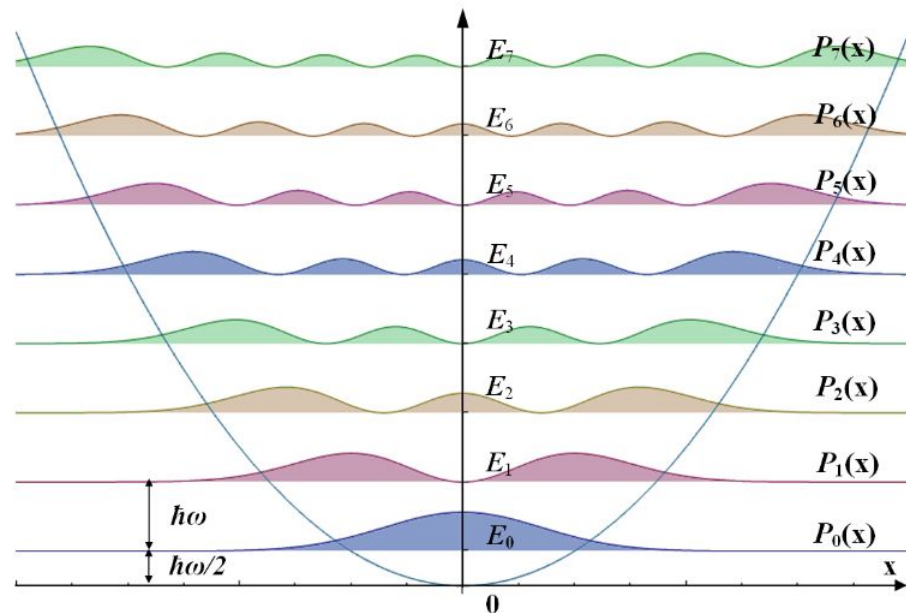


Slow Final Hamiltonian Integration

(Adiabatic Theorem)

Principle: we need to integrate slowly the final Hamiltonian step by step to stay in the ground state ($P_0(x)$ in the figure) and to avoid the excited states ($P_i > 0(x)$ in the figure)

In some way: we map slowly the qubits and couplers from our model.



(very) Few words on Discrete Optimization problems

Goal of an optimization problem

Finding a solution that minimizes the cost of the problem

- an **exact method** is trying to find the best solution(s) with **the lowest cost**.

If it is not possible:

- an **approximate method**, also named **heuristic**, will try to find a good solution with a **low cost**. A heuristic can find the best solution but it is not guaranteed and it will not be proven.

Discrete optimization problems involve discrete variables (**integers and/or binaries**).

Discrete Optimization problems (MILP): Example 1

Traveling Salesman Problem Example

Given a list of cities and the distances between each pair of cities, what is the shortest possible route that visits each city exactly once and returns to the origin city?



$$\min \sum_{i=1}^n \sum_{j \neq i, j=1}^n c_{ij} x_{ij}:$$

Cost

(Objective Function)

$$x_{ij} \in \{0, 1\}$$

$$i, j = 1, \dots, n;$$

$$u_i \in \mathbf{Z}$$

$$i = 2, \dots, n;$$

$$\sum_{i=1, i \neq j}^n x_{ij} = 1$$

$$j = 1, \dots, n;$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1$$

$$i = 1, \dots, n;$$

$$u_i - u_j + nx_{ij} \leq n - 1$$

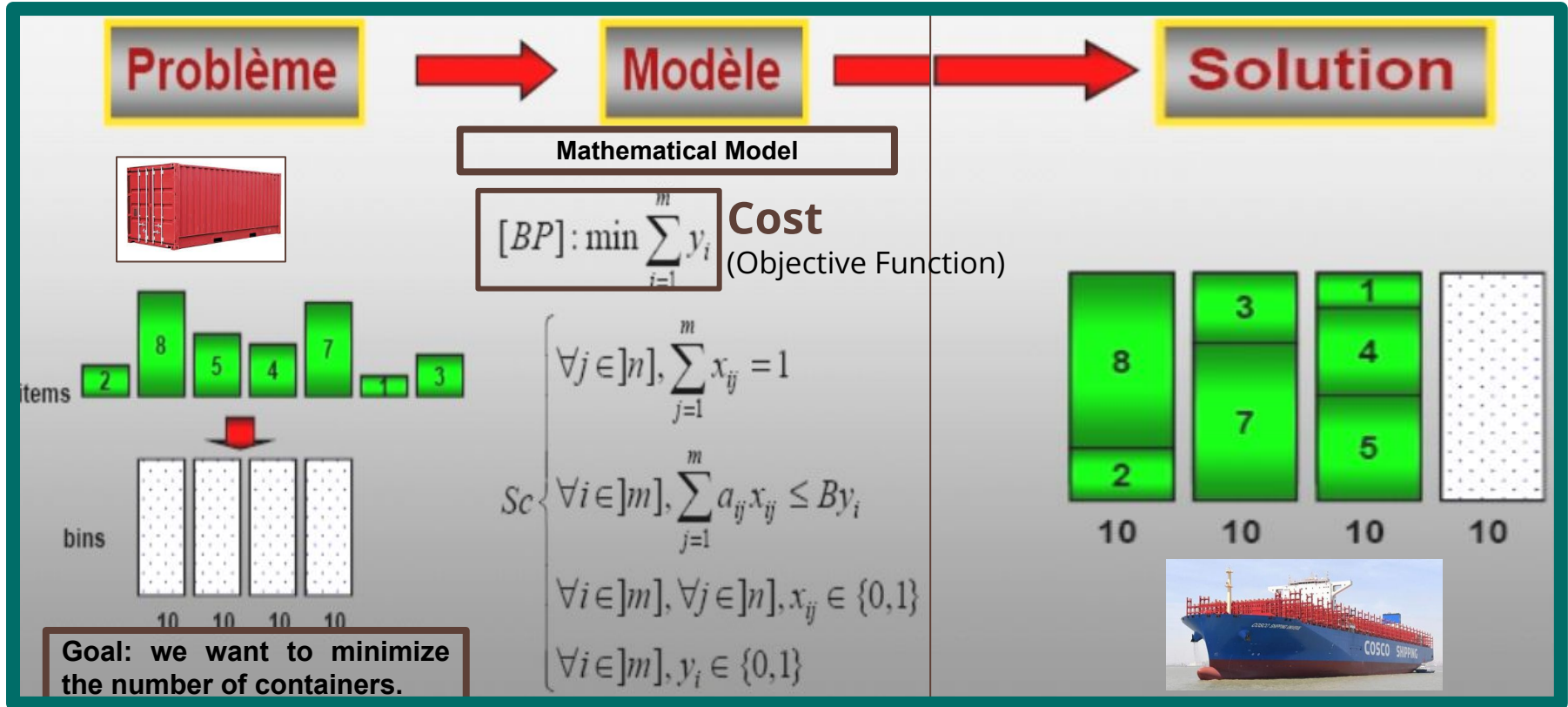
$$2 \leq i \neq j \leq n;$$

$$1 \leq u_i \leq n - 1$$

$$2 \leq i \leq n.$$

Mathematical Model

Discrete Optimization problems (MILP): Example 2



Mathematical Model

INPUT

Classical Computer solvers

Solution

OUTPUT



GUROBI
OPTIMIZATION



FICO
Xpress



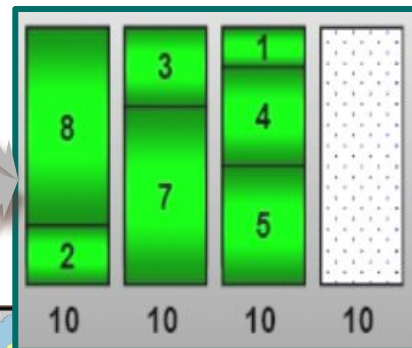
$$[BP]: \min \sum_{i=1}^m y_i$$

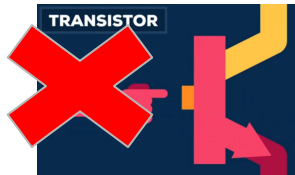
$$Sc \begin{cases} \forall j \in [n], \sum_{i=1}^m x_{ij} = 1 \\ \forall i \in [m], \sum_{j=1}^m a_{ij} x_{ij} \leq B y_i \\ \forall i \in [m], \forall j \in [n], x_{ij} \in \{0,1\} \\ \forall i \in [m], y_i \in \{0,1\} \end{cases}$$

$$\sum_{j=1, j \neq i}^n x_{ij} = 1 \quad i = 1, \dots, n;$$

$$u_i - u_j + n x_{ij} \leq n - 1 \quad 2 \leq i \neq j \leq n;$$

$$1 \leq u_i \leq n - 1 \quad 2 \leq i \leq n.$$





D-Wave **qubits** among the quantum computing world



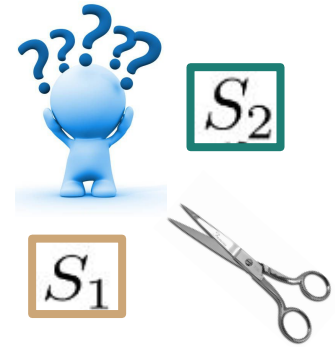
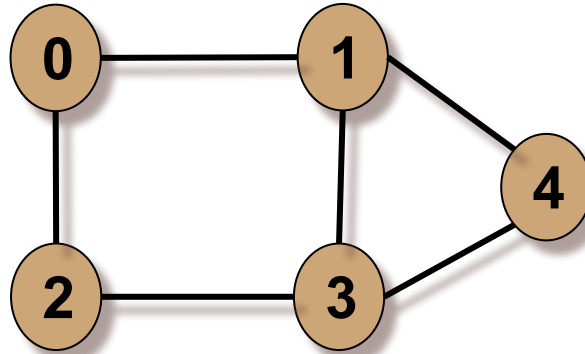
(c) Olivier Ezratty, septembre 2020

Atos

<https://atos.net/wp-content/uploads/2020/07/PR-Atos-opens-up-a-new-path-to-quantum-annealing-simulation.pdf>
 & https://atos.net/fr/2020/communiques-de-presse_2020_07_07/atos-ouvre-la-voie-a-la-simulation-du-recuit-quantique
 original pic: <https://www.oerzatty.net/wordpress/2020/comprendre-informatique-quantique-edition-2020>

Max-cut-Problem

Let's solve another example by hand



Try to find “by hand” the maximal max cut of this simple graph.

Let's take a Graph $G=(X,E)$, where X is the set of nodes and E the set of edges. We can identify each node i of the graph with a binary variable Z_i such that:

- $Z_i = 1$ if i is in one group,
- $Z_i = -1$ otherwise.

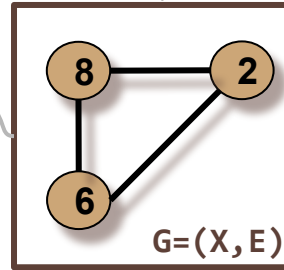
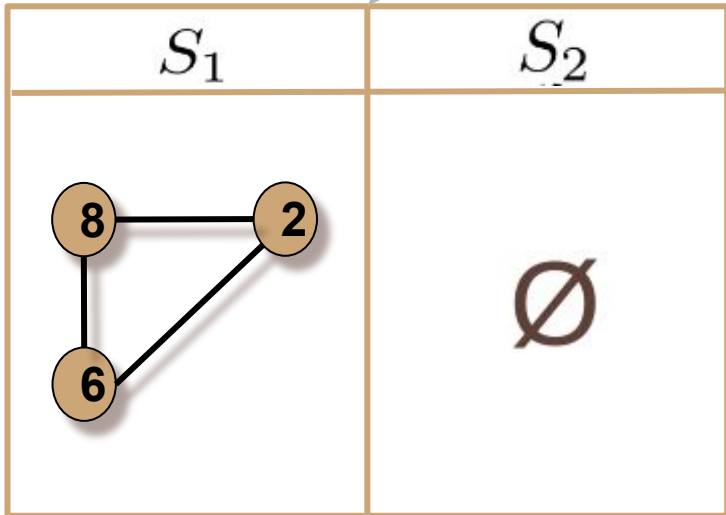
Ising Model

Small Example with a K3 Complete Graph

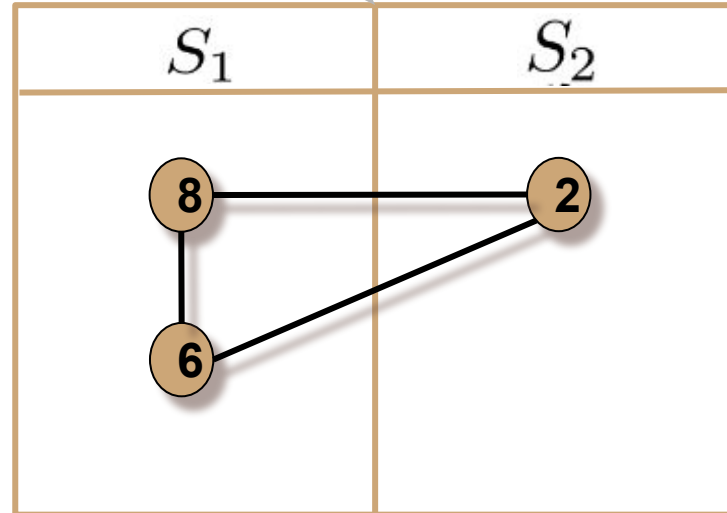
The Max-Cut Problem consists in finding the vector Z of $|X|$ binaries which minimizes:

$$\sum_{(i,j) \in E} Z_i Z_j$$

Solution 1

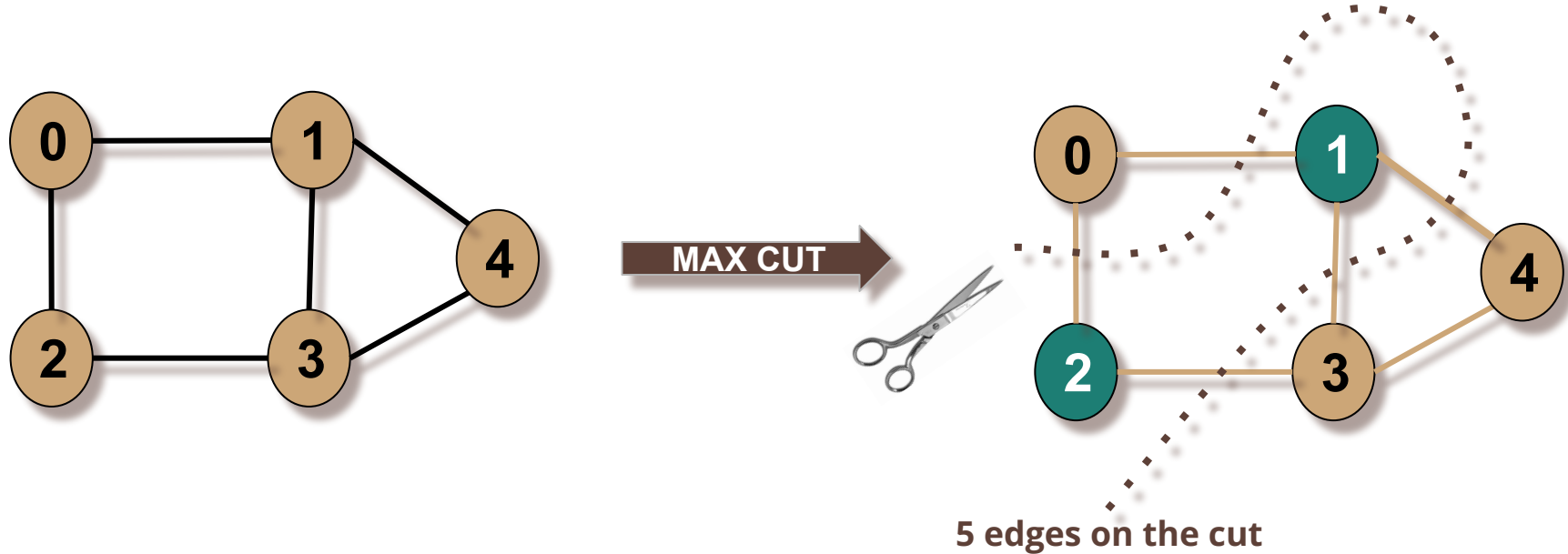


Solution 2



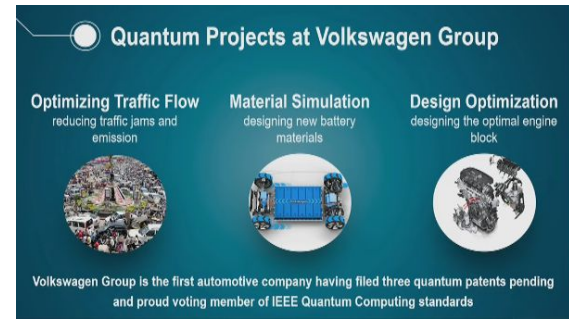
Here you have 2 solutions...You might know which one is the best...
But **try to be sure by calculating the (cost) Objective function $f()$ of each.**

Max-cut-Problem: a first solution

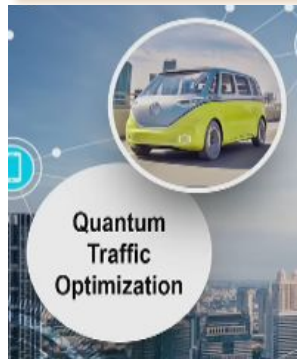


An optimal solution of the Max Cut problem is: {1;2} and {0;3;4} giving 5 edges. The solution can be written as $x = (x_0 \ x_1 \ x_2 \ x_3 \ x_4) = (-1 \ 1 \ 1 \ -1 \ -1)$.

D-Wave & “Volkswagen”



Mobility might be one of the field the most impacted by Quantum Computing in the coming years (with D-Wave machines).



Example (volkswagen/D-wave partnership, 2019):

- **First quantum traffic management system** is Lisbon. 9 buses in Lisbon with a management based on D-Wave quantum computer results (finding almost in real time the fastest route for each of the buses).

D-Wave quantum computers: early applications

Materials Properties

- Atomic Magnetometer,
- Solid state materials simulation,
- Quantum chemistry computation;

Machine Learning

- Image recognition,
- Individual cancer drugs,
- Finding Higgs Boson;

Cyber Security & Fault Detection

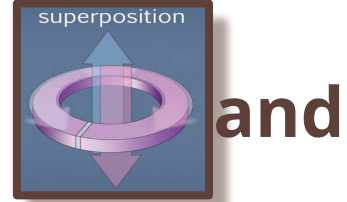
- Formation of Terrorist Networks,
- Facial recognition,
- Fault Detection in circuits;

Optimization

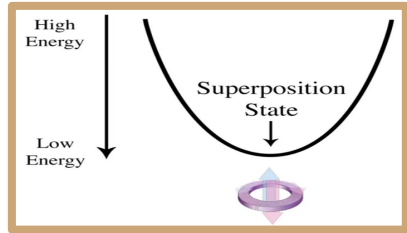
- Location problems
 - Satellites, Network devices, stations;
- Multi-period portfolios (finance),
- Traffic Flow, and so many others!

First
Step

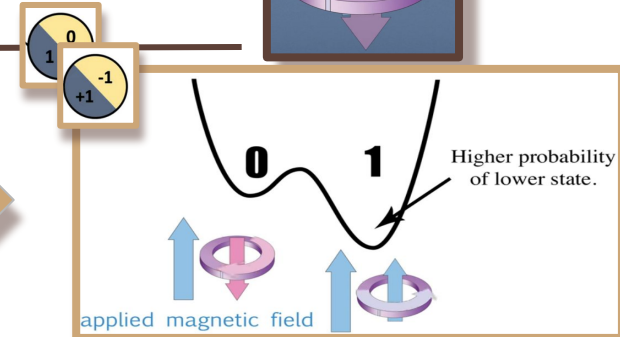
The Qubits start in a superposition state (**zero and one**).



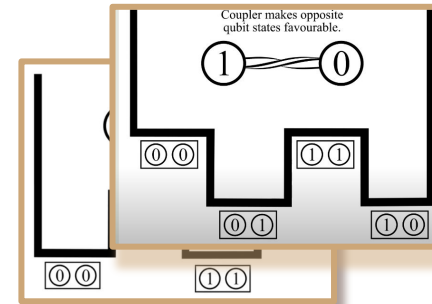
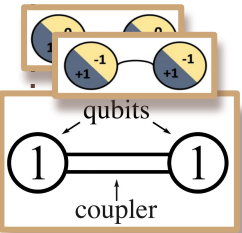
and



Control the probability of the state with a programmable external magnetic field

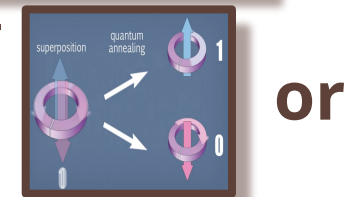


- link the qubits together by **couplers** (entanglement)
- The couplers define how a pair of qubits **ends up**, in the same or in opposite state, resp.: 0 & 0 ; 1 & 1 or, 1 & 0 ; 0 & 1.
- the machine will “physically” make the equality (or the inequality) “**energetically favorable**”, i.e., to “low the energy” of those states.



Last
step

At the end, the measure of each qubit make them going to (with a high probability) either the **zero state** **or** the **one state**.



or

INPUT

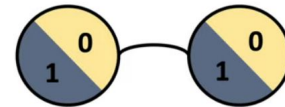
Real life
Opti.
problem

General
Mathematical
Model

Binary
Quadratic
Model
reformulation

Choice
n°2

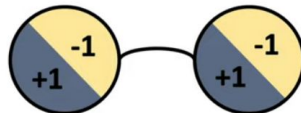
Quadratic Unconstrained
Binary Optimisation
(QUBO)



Choice
n°1

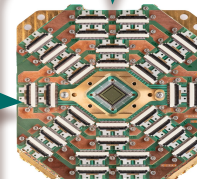
Mathematically Equivalent models

Ising
Problem

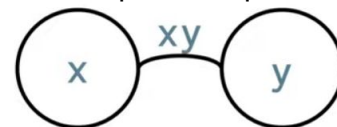


"Fit in"
D-Wave
QPU

"Fit in"
D-Wave
QPU



1 coupler/var prod.



2 qubits/variables

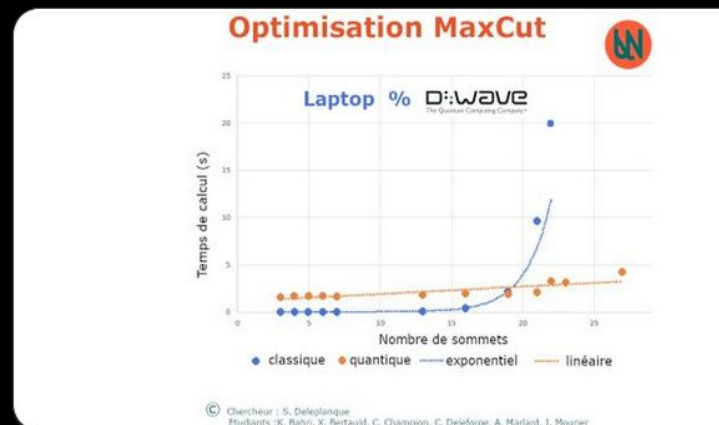


Isabelle LEFEBVRE @ilefebvre_iemn · Mar 24

👤 Projet M1 à @isenlille en informatique quantique: optimisation avec @dwavesys. 1er résultat des étudiants sur la comparaison de temps de calcul. Ca marche!

@DeleplanqueSam2

#QuantumComputing #EnseignementSupérieur #quantique



Isabelle LEFEBVRE @ilefebvre_iemn · Apr 26

👤 Excellente soutenance de projet M1 à @isenlille: optimisation quantique de la répartition des projets selon les préférences des étudiants.

@dwavesys

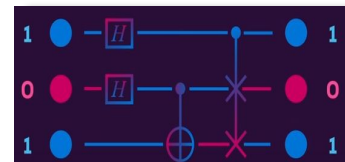
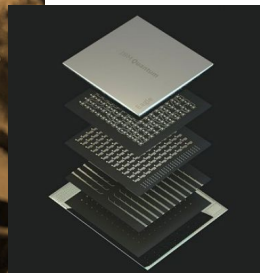
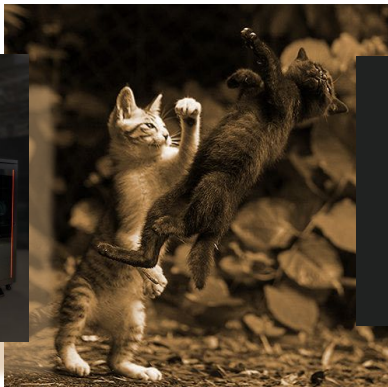
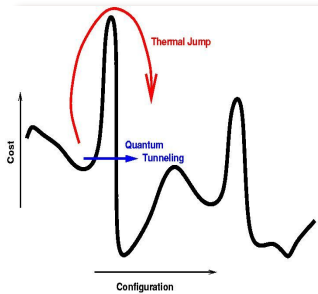
@DeleplanqueSam2

#QuantumComputing #EnseignementSupérieur #quantique



5k qubits... versus 127 ???

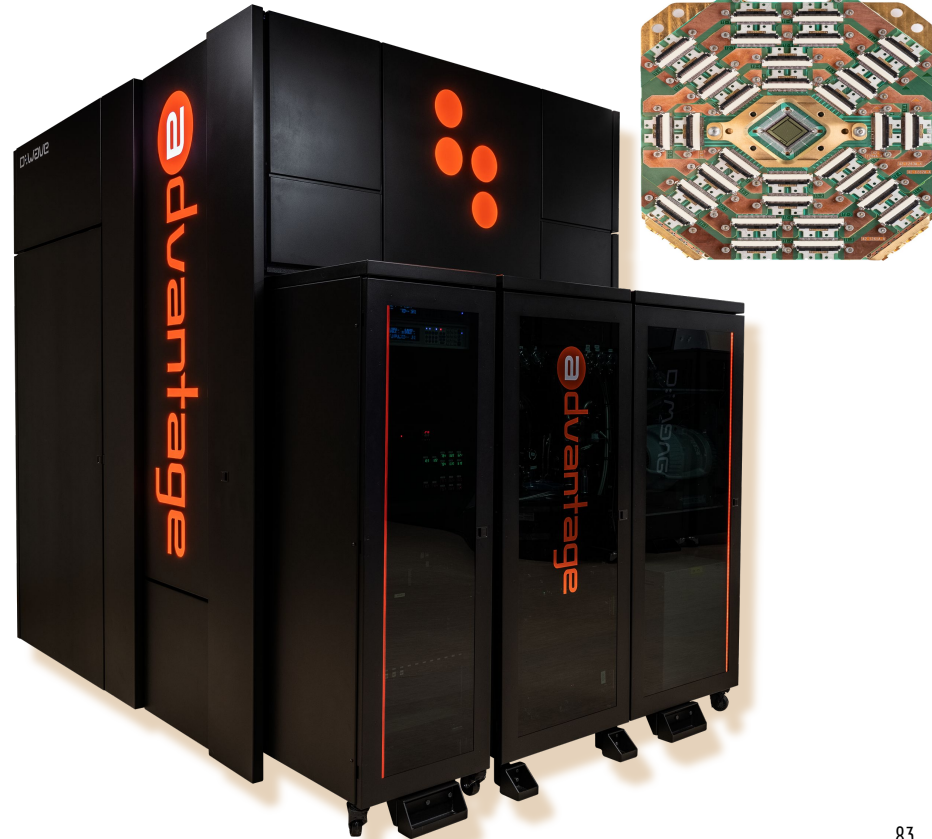
Differences between a D-Wave machine and a gates quantum computer?



There is no match...they are not in the same competition

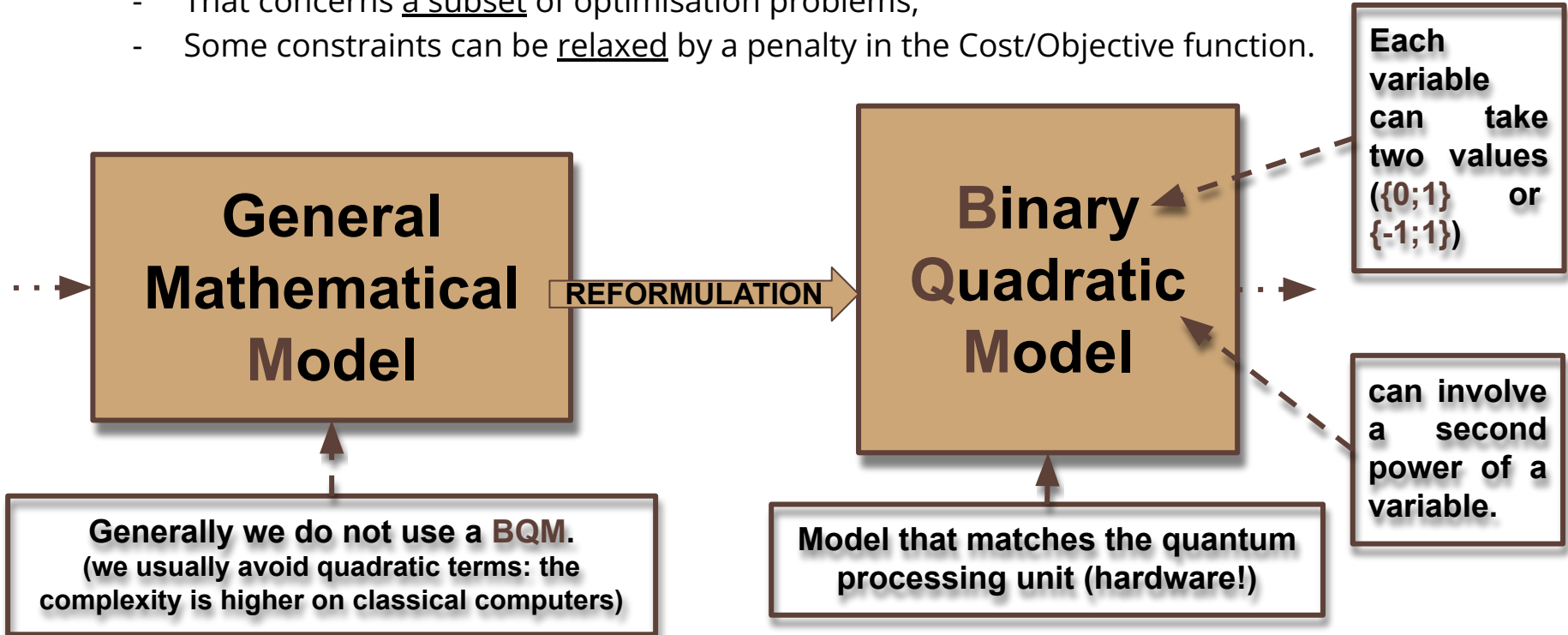
End of 2020, the **Advantage** D-Wave machine is released...

- A new **QPU** with:
 - more than **5,000 qubits**
 - **35,000 couplers**
 - To link pair of qubits
- qubits technology:
 - Superconducting qubits (*SQUID*)
 - Metal used: **niobium**
 - We obtain quantum mechanical effects when the **niobium** becomes a superconductor (once it is cooled down to $\approx 0\text{K}$).

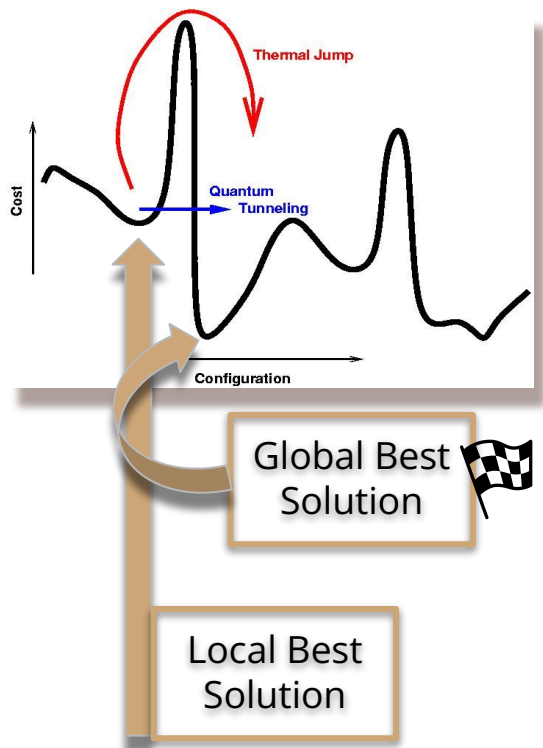


Preparing the input...

- To model an optimisation problem to a Binary Quadratic Model
 - That concerns a subset of optimisation problems,
 - Some constraints can be relaxed by a penalty in the Cost/Objective function.



Simulated Annealing & Quantum annealing: 2 Metaheuristics



Analogy

While the **temperature** is directly related to the probability of moving to a new solution in simulated annealing, in quantum annealing, the strength of **transverse field** determines the quantum-mechanical probability to **change the amplitudes of all states in parallel**.

The advantage of QA compared to SA is due to the quantum mechanics allowing for **an additional escape route from local minima**. While SA must climb over energy barriers to escape traps, **QA can penetrate these barriers without any increase in energy**. This effect is known as **quantum tunneling**.