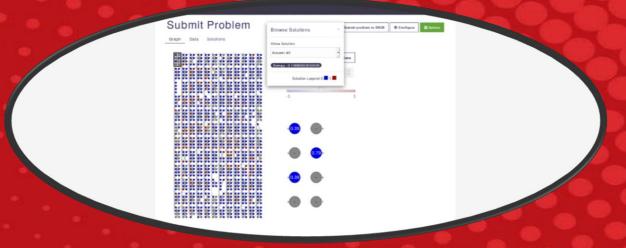
JOURNEE QUANTIQUE organisée par le GT RO Quantique du GDR RO

12 mai 2022 au vendredi 13 mai 2022





DE LA RECHERCHE À L'INDUSTRIE

TP: Solving the maximum matching problem with quantum annealing

VERT Daniel



Déroulement

Introduction

- > D-Wave
- ➤ Limitation de l'architecture / Les instances QUBO
- Famille des graphes G_n

Partie 1 : Résolution du problème de maximum matching

- ➢ De G_n à QUBO
- > De QUBO à D-Wave

Partie 2 : Résultats

- Interpretation
- > Conclusion?

Partie 3 : Pour aller plus loin

- Question 1
- Question 2

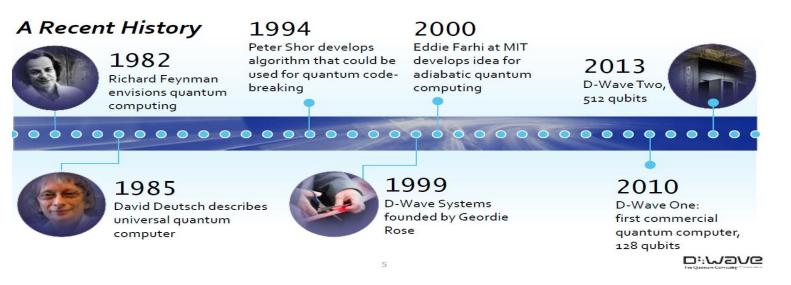
"In my opinion, the crucial experiment (which has not yet been done) would be to compare the adiabatic algorithm head-on against simulated annealing and other classical heuristics. The evidence for the adiabatic algorithm's performance would be much more convincing if the known classical algorithms took exponential time on the same random instances."

Scott Aaronson



L'ORDINATEUR QUANTIQUE ANALOGIQUE





15 millions de dollars pour l'ordinateur quantique D-Wave 2000Q

TOUTE L'ACTUALITÉ / DATACENTER / MAINFRAME ET SUPERCALCULATEUR

Sharon Gaudin / IDG News Service (adapté par Jean Elyan), publié le 25 Janvier 2017

D-Wave Systems a annoncé hier la disponibilité générale du 2000Q, un ordinateur présenté comme quantique à 2000 qubits, soit deux fois plus que son système précédent. Une entreprise de cybersécurité a déjà passé commande. Prix du nouveau système: 15 millions de dollars.



Le 20000 de D-Wave se présente comme un ordinateur quantique d'une puissance de 2000 qubits. Son prix: 15 millons de doite

Anviès un noemier confinateur muantique D.Wave 2Y à 1000 quibits actuellement tecté noir Conselle la

Machine quantique analogique

- Machine spécialisée dans la résolution d'un difficile problème NP-difficiles.
- Utilise un algorithme similaire au recuit simulé → (méta)heuristique pour résoudre rapidement des classes de problèmes complexes (problèmes d'optimisation, d'apprentissage machine ou de recherche opérationnelle) inspirés de la physique statistique.
- D-Wave est donc une sorte d'oracle d'optimisation pour le problème typique de la machine (verre de spin) utilisant les phénomènes quantiques.

L'idée du TP :

Comportement/Performance sur un problème connu pour être difficile (existe-t-il un facteur d'accélération entre QA et SA)



LE RECUIT QUANTIQUE

- Principe [FARHI 2000] : Interpolation entre un Hamiltonien difficile à déterminer (HP) et un Hamiltonien facile à décrire (H0)
 - Configuration de l'Hamiltonien initial: $\mathcal{H}(t) = A(t)\mathcal{H}_0 + B(t)\mathcal{H}_P$
 - Avec un état initial facile : $\mathcal{H}_0 = \sum_i \sigma_i^x$
 - Evolution "lente" vers l'Hamiltonien décrit comme un modèle d'Ising : $\mathcal{H}_P = \sum_i h_i \sigma_i^z + \sum_{(ij)} J_{i,j} \sigma_i^z \sigma_j^z$
 - Initialement B(t=0)=0 et $A(t=\tau)=0$ \rightarrow $\mathcal{H}(t=0)=A(t=0)\mathcal{H}_0$
 - Evolution de l'état t = 0 à $t = \tau \rightarrow \mathcal{H}(t = \tau) = B(t = \tau)\mathcal{H}_P$ H(t) est "recuit "sous une forme purement classique (l'état fondamental H(0) =H0 évolue vers un état H(τ) = HP)

Théorème adiabatique [MESSIAH 1979]: Si l'évolution temporelle est assez lente (τ est assez grand) \rightarrow La solution optimale $\epsilon(\sigma)$ est obtenue avec une forte probabilité (le système reste dans l'état fondamental et son état final décrit une solution au problème d'Ising).



PROBLEMES ADRESSABLES?

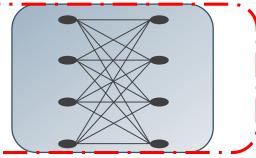
Problèmes combinatoires : Problèmes discrets avec un nombre extrêmement grand de solutions mais ... fini!

Voyageur de commerce (NP-hard) : Chercher un chemin entre N villes → N! possibilités Aucun algorithme (classique et/ou quantique) ne peut trouver une solution exacte rapidement!



The traveling salesperson problem is an optimization problem that can be solved using exactly-one-true constraints. © 2009), Google.

Bipartie (polynomial): problem de matching: Problème de couplage couvrant autant de sommets que possible -> N! solutions possibles sur un graphe complet biparti



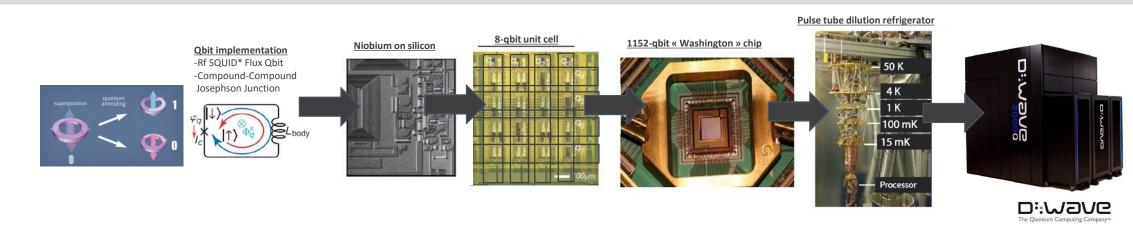
Optimisation des flux de trafic : Minimiser le temps de parcours d'un ensemble de voitures → Algorithmes classiques inefficaces pour ~ milliers de véhicules



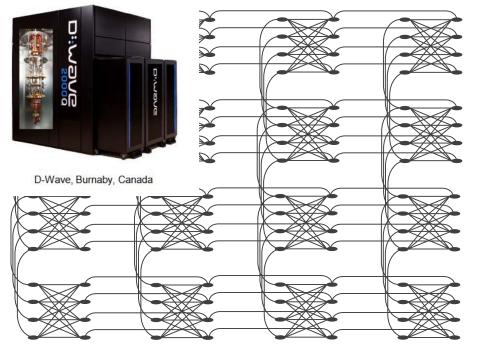


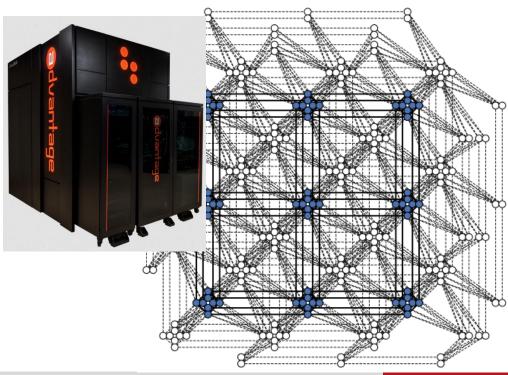
ARCHITECTURE D-Wave QPU: TOPOLOGIES

QPU:



■ D-Wave 2000Q/Advantage → plus de 2000/5600 qubits







RESOLUTION DES PROBLEMES AVEC D-WAVE

- Pour le QPU : Ising ou QUBO
- Modèle d'Ising: Traditionnellement utilisé en mécanique statistique → Les variables sont de type "spin up" (↑: +1)

et "spin down" (\downarrow : -1).

Fonction objectif :

$$E_{ising}(\mathbf{s}) = \sum_{i=1}^{N} h_i \sigma_i + \sum_{i< j}^{N} J_{ij} \sigma_i \sigma_j$$



hi : coefficients linéaires correspondant aux biais des gubits

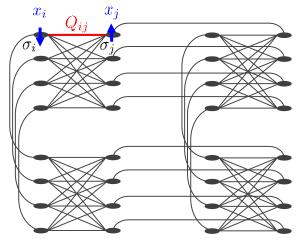
Ji,j les coefficients quadratiques correspondant aux forces de couplage

- Modèle QUBO : Traditionnellement utilisé en informatique (variable binaire) → matrice triangulaire supérieure Q de taille N x N
 - Fonction objectif:

$$O(\mathbf{Q}, \mathbf{x}) = \sum_{i} Q_{ii} x_i + \sum_{i < j} Q_{ij} x_i x_j$$

■ ISING QUBO :

$$\forall i, Q_{ii} = h_i, \forall i, j/i \neq j, Q_{ij} = J_{ij}, \forall i, s_i = 2x_i - 1$$



Qij est la force de couplage entre les variables

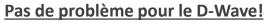
Qii est le biais pour la variable xi.

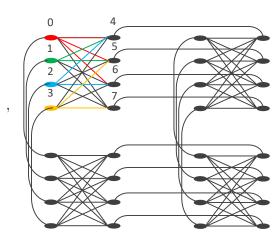


INTEGRATION DU QUBO SUR D-WAVE ET LIMITATION

- Intégration de la matrice QUBO dans le graphe interne des qubits physiques
 - Approche directement intégrable (exemple fictif) :

1						4	5		7
1	0	17	0	0	0	-5	2	-21	0
ı	1	0	17	0	0	-16	-1	0	9
ı	2	0	0	17	0	-1	0		0
ı	3	0	0	0		0	16	0	8
ı	4	0	0	0	0	17	0	0	0
ı	5	0	0	0	0	0	17	0	0
ı	6	0	0	0	0	0	0	17	0
	7	0	0	0	0	0	0	0	17 /



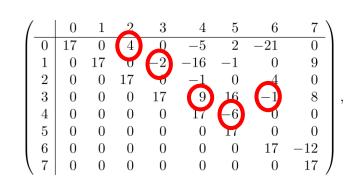


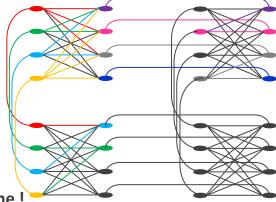


1 variable → N qubits

Pas assez de connexions entre les qubits (hi) → maximum 6!

- Variable 1,2,5,7 dupliqué 3 fois
- Variable 0,3,4,6 dupliqué 2 fois
- Nombre total de couplages: 15
- Nombre total de qubits utilisés: 20





Plusieurs qubits physiques sont UNE variable du problème!

Si le graphe induit par les couplages non nuls de la matrice n'est pas isomorphe au graphe de Chimera

→ il est nécessaire de dupliquer la ou les variables sur plusieurs qubits physiques !



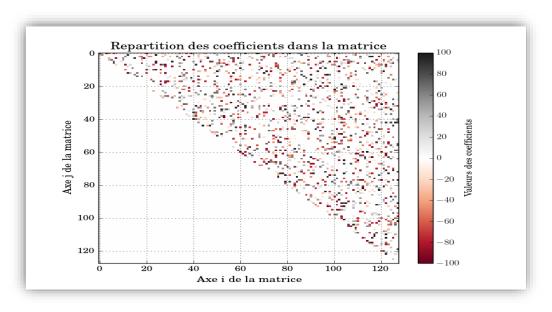
LIMITATION (SUITE)

Intégration de la matrice QUBO dans le graphe interne des qubits physiques

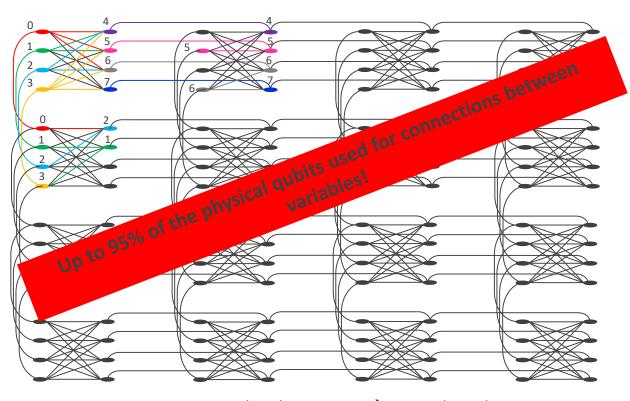


1 variable → 1 qubit

Nombre total de couplages : 384 arêtes!



128 h_i (qubits) + 8128 J_{ij} (termes de couplages)



Graphe Chimera $4x4 \rightarrow 128$ qubits physiques

La topologie ne représente que 4% du nombre total de coupleurs nécessaires pour intégrer la matrice dans le pire des cas.

Définition du problème de couplage

Benchmarking Quantum Annealing Against "Hard" Instances of the Bipartite Matching Problem



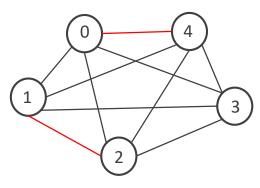
DEFINITION DU PROBLEME



Problème de Matching dans le cas général :

But : sélectionner un ensemble d'arêtes qui donne un **couplage** et qui maximise le nombre de sommets couverts.

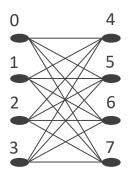
Problème Polynomial → Algorithme polynomial (Edmonds) permet de le résoudre!



Exemple sur un graphe complet

Problème de Matching sur un graphe bipartie :

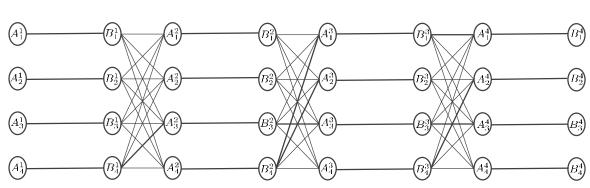
Problème Polynomial Polonais / algorithme de flow le résout



Problème de Maximum matching :

[sasaki-1988] → Cas particuliers nécessitant un nombre exponentiel d'itérations pour obtenir la solution optimale.

- → (trivial à résoudre algorithmiquement)
- → Et même l'oeil



Sommets A non connectés entre eux et idem pour les sommets B



DEFINITION DU PROBLEME DE MATCHING

Sélection des arêtes au bords :

→ Valide <u>et</u> solution optimale

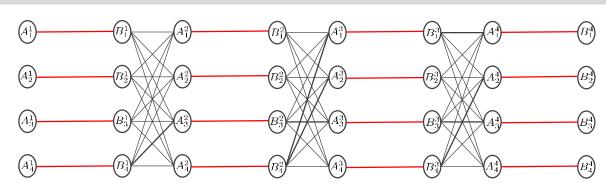
Sélection d'une arêtes dans un bipartie :

→ suppression de deux sommets

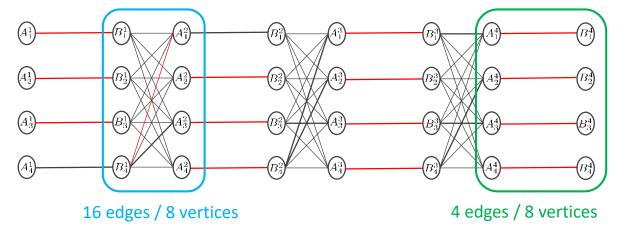
Nombre de sommets : O(n³) Nombre d'arêtes : O(n+1)³

Nombre d'arêtes pour la solution optimal : O(n²)

→ <u>Diffcile pour le recuit simulé</u>



Example all the red edges have a value of 1



[sasaki-1988]: Le temps de calcul croît **exponentiellement** avec le nombre de sommets du graphe (la plupart des variables se trouvent dans des sous-graphes denses).

Objectif: Comparer QA avec SA

► Résolution du problème de matching avec l'ordinateur quantique analogique

Benchmarking Quantum Annealing Against "Hard" Instances of the Bipartite Matching Problem



PROBLEME DE COUPLAGE VERS QUBO

- Problème de couplage s'écrit naturellement sous la forme d'un <u>programme linéaire en nombres entiers</u> (des contraintes) vs QUBO (pas de contraintes)
 - \rightarrow 1 variable pour chaque arrête du graphe $x_e \in \{0,1\}$ 1 si e \in C (couplage) 0 sinon
- Fonction économique : Maximiser le nombre d'arrêtes dans le couplage $\sum_{e \in E} x_e$
- Contraintes :

Pour chaque sommets v : $\sum_{e \in \Gamma(v)} x_e \le 1$ Chaque sommet est couverts au plus une fois

→ Parmi toutes les arrêtes incidentes en un sommet : Une seule fait parti du couplage (contrainte linéaire)

Transformation du couplage en QUBO

- → QUBO sans contraintes : Rajouter une pénalité sur la violation des contraintes de couplages !
- Technique de pénalité → Mettre ces contraintes dans la fonction économique :

Maximiser $\sum_{e \in E} x_e - \lambda \sum_{v \in V} \left(1 - \sum_{e \in \Gamma(v)} x_e\right)^2$ (QUBO) Avec λ grand de manière à pénaliser les solutions qui ne satisfont pas les contraintes

<u>Contraintes « molles » → Solution qui accepte les non couplages (</u>toutes les solutions même celles qui violent les contraintes)



STEP 1 : FORMULATION DU PROBLEME

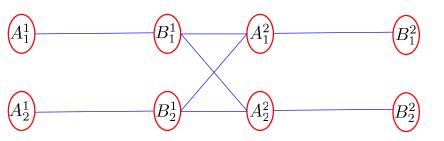
Pénalité sur la violation des contraintes de couplage → Maximise

$$\sum_{e \in E} x_e - \lambda \sum_{v \in V} \left(1 - \sum_{e \in \Gamma(v)} x_e \right)^2$$

Construction de la matrice QUBO:

Termes diagonaux: $Q_{ee} = -(1+2\lambda)$

Termes Non-diagonaux: $Q_{ee'} = \begin{cases} 2\lambda & \text{if } e \cap e' \neq \emptyset, \\ 0 & \text{otherwise.} \end{cases}$



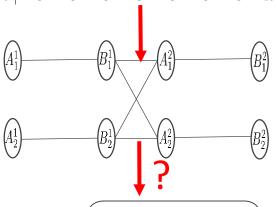
(0	1	2	3	4	5	6	7 \	١
0	17	0	16	16	0	0	0	0	
1	0	-17	0	0	16	16	0	0	
2	0	0	-17	16	16	0	16	0	
3	0	0	0	-17	0	16	0	16	Ι,
4	0	0	0	0	-17	16	16	0	
5	0	0	0	0	0	-17	0	16	
6	0	0	0	0	0	0	-17	0	
\setminus 7	0	0	0	0	0	0	0	-17	

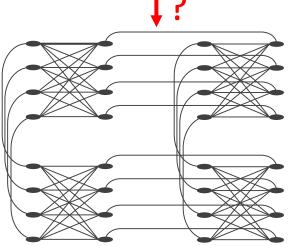
```
danielvm@danielvm-VirtualBox: ~/Desktop/tp_dwave
nielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$ python Gn2QUBO.py
  normalisation: 17.0
5 16
7 16
nielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$
```



STEP 2: INTEGRATION DES VARIABLES

	0	1	2	3	4	5	6	7
0	-17	0	16	16	0	0	0	0
1	0	-17	0	0	16	16	0	0
2	0	0	-17	16	16	0	16	0
3	0	0	0	-17	0	16	0	16
4	0	0	0	0	-17	16	16	0
5	0	0	0	0	0	-17	0	16
6	0	0	0	0	0	0	-17	0
7	0	0	0	0	0	0	0	-17
`								,





nb qbits = 31g node = read qbits(nb qbits, "noeuds 2.txt") #print("nb_qubits :", g_node) q dw2x = read coupleurs(nb qbits, "coupleurs dwaves.txt") #print("nb_coupleurs :", g_dw2x) #print() g nodep = read qbitsp(nb qbits, "noeuds p.txt") #print("nb qubits p :", g nodep) g_dw2xp = read_coupleursp(nb_qbits, "coupleurs_dwavep.txt") #print("nb coupleurs p :", g dw2xp) 01,norma=gntoqubo(n) print 01 H = nx.Graph()H.add_nodes_from(g_nodep) H.add edges from(g dw2xp) embedding = find embedding(01, g dw2xp, random seed=10000000000)

print("transformation : ", embedding),"\n'

General Embedding

General embedding refers to embedding that may be useful for any type of graph.

The primary utility function, find_embedding(), is an implementation of the heuristic algorithm described in [1]. It accepts various optional parameters used to tune the algorithm's execution or constrain the given problem.

This implementation performs on par with tuned, non-configurable implementations while providing users with hooks to easily use the code as a basic building block in research.

[1] https://arxiv.org/abs/1406.2741

find_embedding() [source]

Heuristically attempt to find a minor-embedding of source graph S into a target graph T.

Parameters

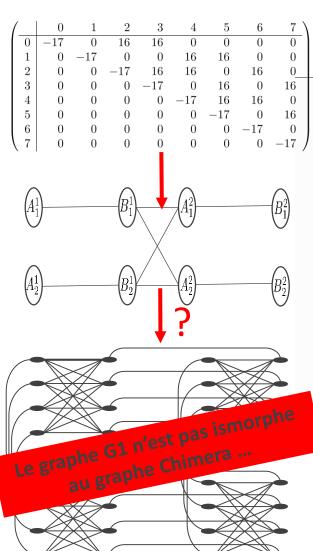
- \$ (iterable/NetworkX Graph) The source graph as an iterable of label pairs representing the edges, or a NetworkX Graph.
- T (iterable/NetworkX Graph) The target graph as an iterable of label pairs representing the edges, or a NetworkX Graph.
- **params (optional) See below

Returns

When the optional parameter return overlap is False (the default), the function returns a



STEP 2 : INTEGRATION DES VARIABLES



General Embedding

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Return

When the optional parameter return overlap is False (the default), the function returns a

```
danielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$ python umbedQ.py
('QUBO init :', {(6, 7): 16, (4, 6): 16, (0, 0): -17, (3, 7): 16, (3, 3): -17, (6, 6): -17, (4, 5): 16, (5, 5): -17, (1, 7): 16, (2, 2): -17, (2, 4): 16, (2, 6): 16, (0, 5): 16, (0, 4): 16, (7, 7): -17, (1, 6): 16, (5, 7): 16, (1, 1): -17, (4, 4): -17, (3, 5): 16})

('coupleur :', [(0, 4), (1, 4), (2, 4), (3, 4), (0, 5), (1, 5), (2, 5), (3, 5), (0, 6), (1, 6), (2, 6), (3, 6), (0, 7), (1, 7), (2, 7), (3, 7), (4, 12), (8, 12), (9, 12), (10, 12), (11, 12), (5, 13), (8, 13), (9, 13), (10, 13), (11, 13), (6, 14), (8, 14), (9, 14), (10, 14), (11, 14), (7, 15), (8, 15), (9, 15), (10, 15), (11, 15), (12, 20), (16, 20), (17, 20), (18, 20), (19, 20), (13, 21), (16, 21), (17, 21), (18, 21), (19, 21), (16, 22), (17, 22), (18, 22), (19, 22), (15, 23), (16, 23), (17, 23), (18, 23), (19, 23), (20, 28), (24, 28), (25, 28), (26, 28), (27, 28), (21, 29), (24, 29), (25, 29), (26, 29), (27, 29), (22, 30), (24, 30), (25, 30), (26, 30), (27, 30), (23, 31), (24, 31), (25, 31), (26, 31), (27, 31)]

('transformation : ', {0: [26], 1: [18], 2: [25], 3: [16], 4: [29, 27], 5: [23, 31], 6: [20, 28], 7: [21, 17]})

danielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$
```

Solution : dupliquer les qubits physiques !



STEP 2 : INTEGRATION DES VARIABLES

```
danielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$ python umbedQ.py
('QUBO init :', {(6, 7): 16, (4, 6): 16, (0, 0): -17, (3, 7): 16, (3, 3): -17, (6, 6): -17, (4, 5): 16, (5, 5): -17, (1, 7): 16, (2, 2): -17, (2, 4): 16, (2, 6): 16, (0, 5): 16, (0, 4): 16, (7, 7): -17, (1, 6): 16, (5, 7): 16, (1, 1): -17, (4, 4): -17, (3, 5): 16})

('coupleur :', [(0, 4), (1, 4), (2, 4), (3, 4), (0, 5), (1, 5), (2, 5), (3, 5), (0, 6), (1, 6), (2, 6), (3, 6), (0, 7), (1, 7), (2, 7), (3, 7), (4, 12), (8, 12), (9, 12), (10, 12), (11, 12), (5, 13), (8, 13), (9, 13), (10, 13), (11, 13), (6, 14), (8, 14), (9, 14), (10, 14), (11, 14), (7, 15), (8, 15), (9, 15), (10, 15), (11, 15), (12, 20), (16, 20), (17, 20), (18, 20), (19, 20), (13, 21), (16, 21), (17, 21), (18, 21), (19, 21), (16, 22), (17, 22), (18, 22), (19, 22), (15, 23), (16, 23), (17, 23), (18, 23), (19, 23), (20, 28), (24, 28), (25, 28), (26, 28), (27, 28), (21, 29), (24, 29), (25, 29), (26, 29), (27, 29), (22, 30), (24, 30), (25, 30), (26, 30), (27, 30), (23, 31), (24, 31), (25, 31), (26, 31), (27, 31)])

('transformation : ', {0: [26], 1: [18], 2: [25], 3: [16], 4: [29, 27], 5: [23, 31], 6: [20, 28], 7: [21, 17]})

danielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$
```

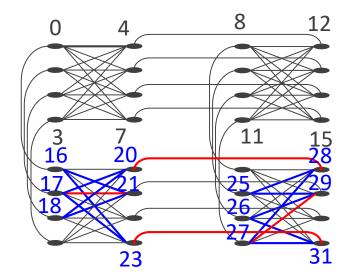
Solution:

 $V0 \rightarrow \text{qubits } 26$ V1 \rightarrow \text{qubits } 18

 $V4 \rightarrow qubits 29 et 27$

Total:

8 Variables pour 12 qubits (1.5 fois plus de qubits)



<u>Duplication de Qubit → Plus de variables et de coefficients non nuls dans le QUBO</u>

- → Contraintes supplémentaires (contrainte forte sur les qubits dupliqués qui doivent être à la même valeur)
- → Augmentation de l'amplitude des coefficients pour représenter mon problème



INTEGRATION DE G1 SUR D-WAVE (D2000Q)

STEP 2: INTEGRATION DES VARIABLES

```
Optional Parameters:
max_no_improvement (int, optional, default=10):
                                                                                                     Maximum number of failed iterations to improve the current solution, where each
iteration attempts to find an embedding for each variable of S such that it is adjacent to
                                                                                                     all its neighbours.
nb_qbits = 31
                                                                                                   random_seed (int, optional, default=None):
q node = read qbits(nb qbits, "noeuds 2.txt")
                                                                                                     Seed for the random number generator. If None, seed is set by os.urandom().
#print("nb qubits :", g node)
g_dw2x = read_coupleurs(nb_qbits, "coupleurs_dwaves.txt")
#print("nb coupleurs :", q dw2x)
                                                                                                   timeout (int, optional, default=1000):
#print()
                                                                                                     Algorithm gives up after timeout seconds.
q nodep = read qbitsp(nb qbits, "noeuds p.txt")
#print("nb qubits p :", g nodep)
                                                                                                   max beta (double, optional, max beta=None):
g dw2xp = read coupleursp(nb qbits, "coupleurs dwavep.txt")
                                                                                                     Qubits are assigned weight according to a formula (beta^n) where n is the number of
#print("nb_coupleurs_p :", g_dw2xp)
                                                                                                     chains containing that qubit. This value should never be less than or equal to 1. If None,
                                                                                                     max beta is effectively infinite.
01.norma=qntoqubo(n)
print Q1
                                                                                                   tries (int. optional, default=10):
                                                                                                     Number of restart attempts before the algorithm stops. On D-WAVE 2000Q, a typical
H = nx.Graph()
                                                                                                     restart takes between 1 and 60 seconds.
H.add nodes from(q nodep)
H.add_edges_from(g_dw2xp)
                                                                                                   inner rounds (int, optional, default=None):
                                                                                                     The algorithm takes at most this many iterations between restart attempts; restart
############################## recherche du mappage de G_n s le graphe #############################
                                                                                                     attempts are typically terminated due to max no improvement . If None, inner rounds is
embedding = find embedding(01, g dw2xp, random seed=10000000000)
                                                                                                     effectively infinite.
print("transformation : ", embedding),"\n'
```

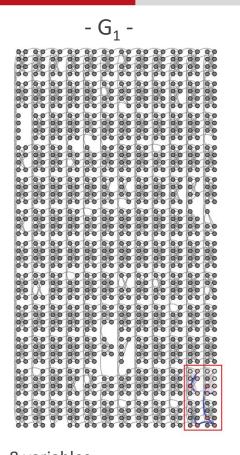
Changer la seed (donc changer l'endroit d'intégration du problème sur la puce) change la qualité de la solution ?

('transformation : ', {0: [201], 1: [205], 2: [210, 212], 3: [331], 4: [215, 207], 5: [206, 203], 6: [202, 204], 7: [200, 328, 332]})

119, 115], 6: [125, 123], 7: [126, 118]})

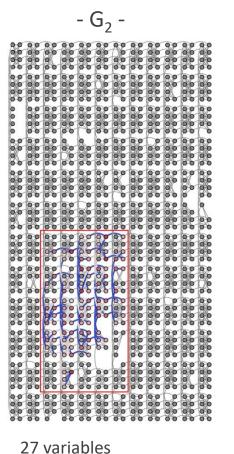


INSTANCES SUR DWAVE (DW2X → 1095 qubits)



8 variables 16 qubits physiques → ~2% Moy. dup. : 2.0

12 coefficients non 0



100 qubits physiques → ~10% Moy. dup. : 3.7

72 coefficients non 0

- G₃ -

64 variables
431 qubits physiques → ~40%
Moy. dup.: 6.7
240 coefficients non 0

125 variables
951 qubits physiques → ~87%
Moy. dup.: 7.6
600 coefficients non 0

Daniel Vert. Étude des performances des machines à recuit quantique pour la résolution de problèmes combinatoires. Université Paris-Saclay, 2021. (tel-03208838v2)



STEP 3 : QUBO SUR DWAVE

1		0	1	2	3	4	5	6	7	\	
1	0	-17	0	16	16	0	0	0	0	١	
	1	0	-17	0	0	16	16	0	0	ı	
	2	0	0	-17	16	16	0	16	0	ı	
	3	0	0	0	-17	0	16	0	16		
	4	0	0	0	0	-17	16	16	0	ı	
	5	0	0	0	0	0	-17	0	16	ı	
l	6	0	0	0	0	0	0	-17	0		
	7	0	0	0	0	0	0	0	-17)	
QUBO init											

QUBO norm

```
danielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$ python umbed().py
 QUBO init___', {(6, 7): 16, (4, 6): 16, (0, 0): -17, (3, 7): 16, (3, 3): -17,
             , 5): 16, (5, 5): -17, (1, 7): 16, (2, 2): -17, (2, 4): 16, (2, 6)
         7: 16, (0, 4): 16, (7, 7): -17, (1, 6): 16, (5, 7): 16, (1, 1): -17,
   upleur :', [(0, 4), (1, 4), (2, 4), (3, 4), (0, 5), (1, 5), (2, 5), (3, 5)
 (11, 15), (12, 20), (16, 20), (17, 20), (18, 20), (19, 20), (13, 21), (16, 21
 (17, 21), (18, 21), (19, 21), (16, 22), (17, 22), (18, 22), (19, 22), (15, 23
 (16, 23), (17, 23), (18, 23), (19, 23), (20, 28), (24, 28), (25, 28), (26, 28
 (27, 28), (21, 29), (24, 29), (25, 29), (26, 29), (27, 29), (22, 30), (24, 30
 (25, 30), (26, 30), (27, 30), (23, 31), (24, 31), (25, 31), (26, 31), (27, 31
  ransformation : ', {0: [26], 1: [18], 2: [25], 3: [16], 4: [29, 27], 5: [23,
   6: [20, 28], 7: [21, 17]})
      norm :', {(6, 7): 0.9411764705882353, (4, 6): 0.9411764705882353, (0, 0)
          Y): 0.9411764705882353, (3, 3): -1.0, (6, 6): -1.0, (4, 5): 0.9411764
             5): -1.0, (1, 7): 0.9411764705882353, (2, 2): -1.0, (2, 4): 0.941
             (2, 6): 0.9411764705882353, (0, 5): 0.9411764705882353, (0, 4):
9411764705882353, (7, 7): -1.0, (1, 6): 0.9411764705882353, (5, 7): 0.941176470
 82353, (1, 1): -1.0, (4, 4): -1.0, (3, 5): 0.9411764705882353})
```

h_range

```
Supported Leap's Hybrid QPU Other
Solvers:
```

Range of values possible for the qubit biases (linear coefficients), h, for this solver.

The auto_scale parameter, which rescales h and J values in the problem to use as much of the range of h (h_range) and the range of J (i_range) as possible, enables you to submit problems with values outside these ranges and have the system automatically scale them to fit.

Example

For the solver configured in the example setup above,

```
>>> qpu_advantage.properties["h_range"]
[-2.0, 2.0]
```

j_range



Range of values possible for the coupling strengths (quadratic coefficients), J, for this solver.

The auto_scale parameter, which rescales h and J values in the problem to use as much of the range of h (h_range) and the range of J (j_range) as possible, enables you to submit problems with values outside these ranges and have the system automatically scale them to fit.

See also extended_j_range.

Example

For the solver configured in the example setup above,

```
>>> qpu_advantage.properties["j_range"]
[-1.0. 1.0]
```

Normalisation dans les paramètres de DWave



STEP 3 : QUBO SUR DWAVE

```
-17
                                   0
                    16
                                   0
 0
               16
                                   0
                         16
                                  16
                         16
                              16
                                   0
                                  16
                            -17
                                 -17
                              0
          QUBO init
```

QUBO norm

```
# Embed the QUBO
target_Q = dimod.embed_qubo(Q1, embedding, H)
#print np.asarray(target_Q.items())
```

```
Ud= np.asarray(target_Q.items())[:,1]/4 # normalisation de dwave pour la duplication
QUBO=dict(zip(list(target_Q.keys()),Ud))
print("QUBO dwave:", QUBO), "\n"
```

```
danielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$ python umbedQ.py
 QUBO init:', {(6, 7): 16, (4, 6): 16, (0, 0): -17, (3, 7): 16, (3, 3): -17,
 16, (0, 5): 16, (0, 4): 16, (7, 7): -17, (1, 6): 16, (5, 7): 16, (1, 1): -17,
 , 4): -17, (3, 5): 16})
 coupleur :', [(0, 4), (1, 4), (2, 4), (3, 4), (0, 5), (1, 5), (2, 5), (3, 5)
 (9, 12), (10, 12), (11, 12), (5, 13), (8, 13), (9, 13), (10, 13), (11, 13), (
 14), (8, 14), (9, 14), (10, 14), (11, 14), (7, 15), (8, 15), (9, 15), (10, 15)
 (11, 15), (12, 20), (16, 20), (17, 20), (18, 20), (19, 20), (13, 21), (16, 21 (17, 21), (18, 21), (19, 21), (16, 22), (17, 22), (18, 22), (19, 22), (15, 23
 (16, 23), (17, 23), (18, 23), (19, 23), (20, 28), (24, 28), (25, 28), (26, 28)
(27, 28), (21, 29), (24, 29), (25, 29), (26, 29), (27, 29), (22, 30), (24, 30)
(25, 30), (26, 30), (27, 30), (23, 31), (24, 31), (25, 31), (26, 31), (27, 31)
 transformation: ', {0: [26], 1: [18], 2: [25], 3: [16], 4: [29, 27], 5: [23,
31], 6: [20, 28], 7: [21, 17]})
 QUBO norm :', {(6, 7): 0.9411764705882353, (4, 6): 0.9411764705882353, (0, 0)
 1.0, (3, 7): 0.9411764705882353, (3, 3): -1.0, (6, 6): -1.0, (4, 5): 0.9411764
 05882353, (5, 5): -1.0, (1, 7): 0.9411764705882353, (2, 2): -1.0, (2, 4): 0.941
 764705882353, (2, 6): 0.9411764705882353, (0, 5): 0.9411764705882353, (0, 4):
.9411764705882353, (7, 7): -1.0, (1, 6): 0.9411764705882353, (5, 7): 0.941176470
882353, (1, 1): -1.0, (4, 4): -1.0, (3, 5): 0.9411764705882353})
 QUBO dwave:', {(16, 16): -0.25, (23, 31): -1.0) (20, 20): 0.375, (16, 21): 0.
 29411764705882, (20, 17): 0.235294<del>11764705882, (29, 27): -1.0, (23, 17): 0.23</del>
 9411764705882, (26, 29): 0.23529411764705882, (27, <del>28): 0.23</del>529411764705882, (
  20): 0.23529411764705882, (25, 29): 0.23529411764705882, (26, 26): -0.25, (2
 25): -0.25, (16, 23): 0.23529411764705882, (21, 21): 0.375, (27, 31): 0.23529
 764705882, (18, 21): 0.23529411764705882, (25, 28): 0.23529411764705882, (27
 ): 0.375, (29, 29): 0.375, (28, <del>28): 0.3</del>75, (23, 23): 0.375, (31, <del>31): 0.3</del>75
    31): 0.23529411764705882, (20, 28): -1.0, (18, 18): -0.25, (21, 17): -1.0
anielvm@danielvm-VirtualBox:~/Desktop/tp_dwave$
```

Matrice QUBO intégrable sur le graphe de D-Wave



DWAVE LEAP



Leap In

EMAIL ADDRESS

jane@gmail.com

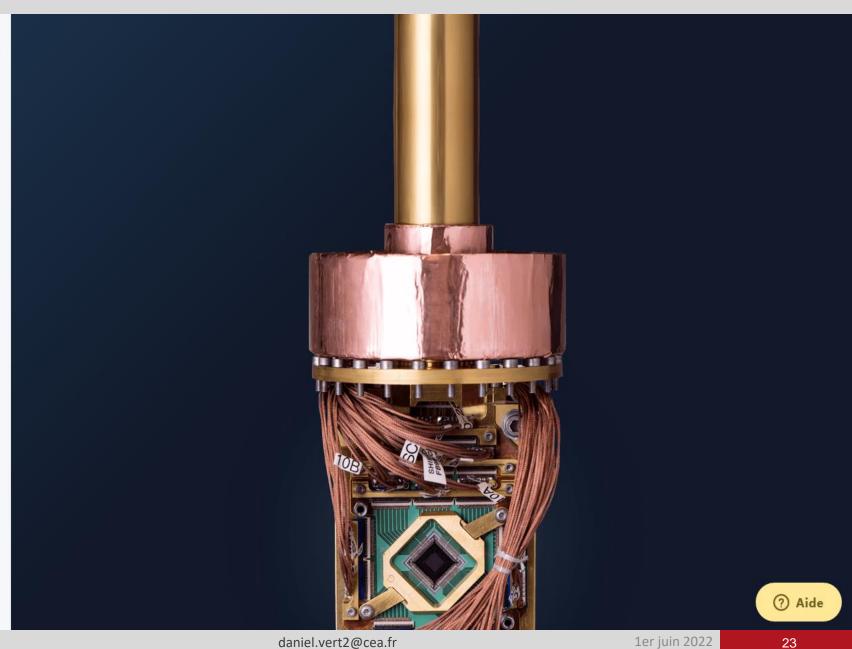
PASSWORD

Forgot password? Having trouble logging in?

LOG IN

Don't have an account? Sign up

Lost activation link? Resend link



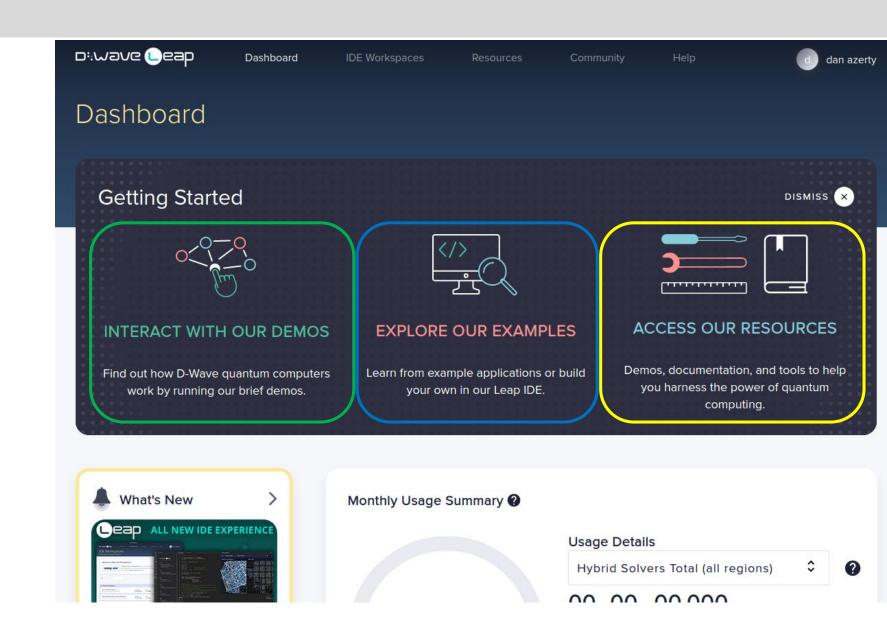
DWAVE LEAP

STEP 4 : Paramètres DWave

Interact with our demos

Explore our examples

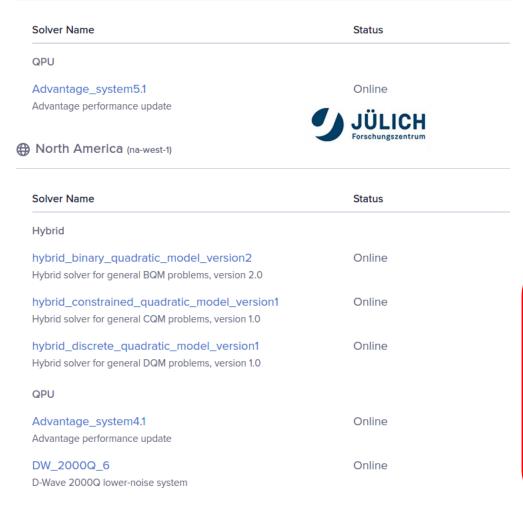
Access our resources

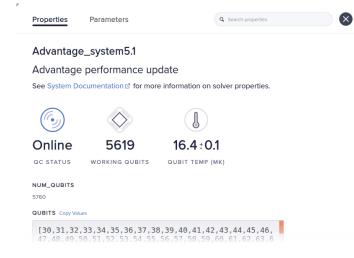


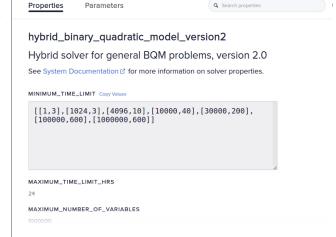
DWAVE LEAP

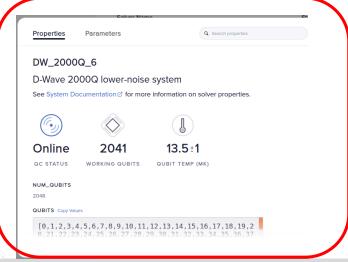
STEP 4: PARAMETRES DE D-WAVE

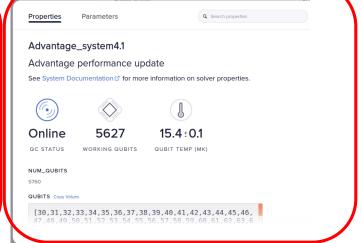
⊕ Europe (eu-central-1)



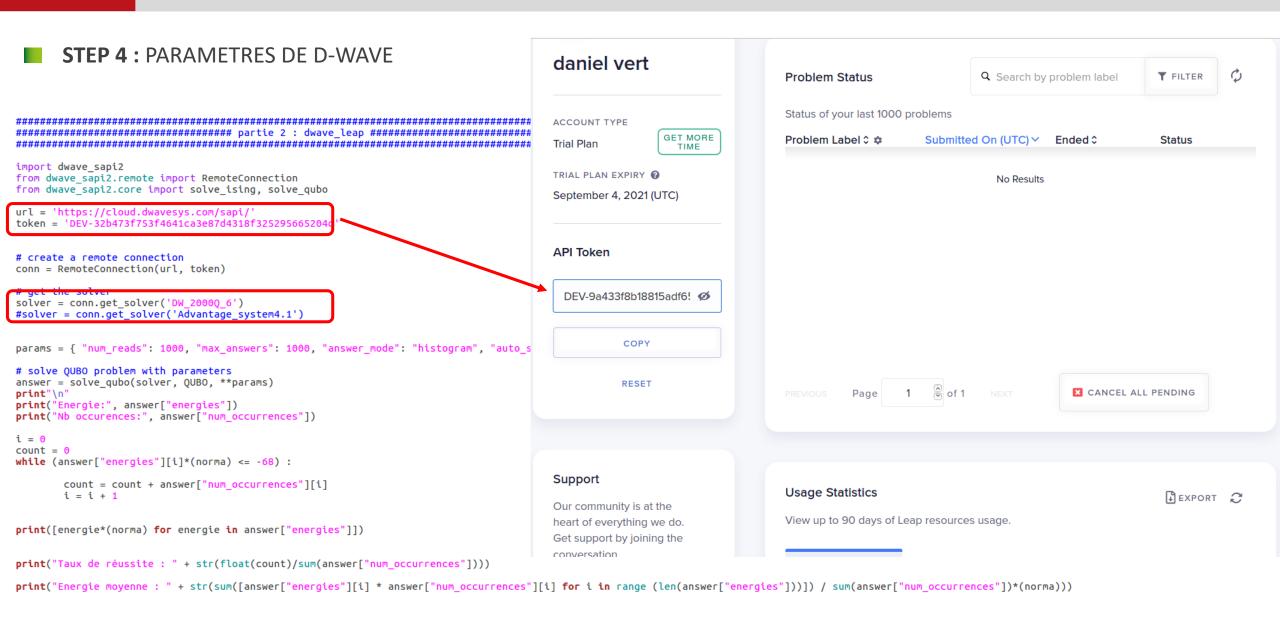














STEP 4 : PARAMETRES DE D-WAVE

```
anneal_offset_ranges
anneal_offset_step
anneal_offset_step_phi0
import dwave sapi2
                                                                                                                                                   · annealing_time_range
from dwave_sapi2.remote import RemoteConnection

    beta range

from dwave sapi2.core import solve ising, solve qubo
                                                                                                                                                   · category
                                                                                                                                                   chip_id
url = 'https://cloud.dwavesys.com/sapi/'

    couplers

token = 'DEV-32b473f753f4641ca3e87d4318f325295665204

    default_annealing_time

    default beta

# create a remote connection
conn = RemoteConnection(url, token)
                                                                                                                                                   · default readout thermalization

    extended i range

    h_gain_schedule_range

solver = conn.get solver('DW 20000 6')

    h_range

#solver = conn.get_solver('Advantage_system4.1')

    j_range

    max_anneal_schedule_points

                                                                                                                         "num_spin_reversal_trans • max_h_gain_schedule_points
params = { "num_reads": 1000, "max_answers": 1000, "answer_mode": "histogram", "auto_scale": True, "annealing_time": 20}

    maximum_number_of_biases

# solve QUBO problem with parameters
                                                                                                                                                   · maximum number of constraints
answer = solve_qubo(solver, QUBO, **params)
print"\n'

    maximum_number_of_variables

print("Energie:", answer["energies"])
                                                                                                                                                   · maximum time limit hrs
print("Nb occurences:", answer["num_occurrences"])

    minimum_time_limit

    minimum_time_limit_s

i = 0
                                                                                                                                                   · num biases multiplier
count = 0
while (answer["energies"][i]*(norma) <= -68) :</pre>

    num_constraints_multiplier

                                                                                                                                                   num_qubits
        count = count + answer["num occurrences"][i]

    num_reads_range

        i = i + 1

    num_variables_multiplier

    parameters

                                                                                                                                                   · per_qubit_coupling_range
print([energie*(norma) for energie in answer["energies"]])

    problem_run_duration_range

                                                                                                                                                   · programming_thermalization_range
print("Taux de réussite : " + str(float(count)/sum(answer["num_occurrences"])))
print("Energie moyenne : " + str(sum([answer["energies"][i] * answer["num occurrences"][i] for i in range (len(answer["energies"]))]) / sum(answer["num occurrences"])*(norma)))
```

The examples below use the following setup: >>> from dwave.system import DWaveSampler, LeapHybridSampler, LeapHybridCQMSampler >>> qpu_advantage = DWaveSampler(solver={'topology__type': 'pegasus'}) >>> qpu_2000q = DWaveSampler(solver={'topology__type': 'chimera'}) >>> hybrid_bqm_sampler = LeapHybridSampler() >>> hybrid_cqm_sampler = LeapHybridCQMSampler() · default_programming_thermalization · maximum number of quadratic variables



RESULTATS DE G1 SUR DWAVE

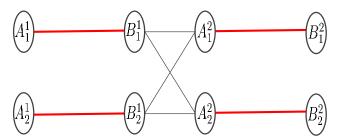
STEP 4 : RESULTATS

Sortie 2000Q

#q_p (va)	26(0)	18(1)	25(2)	16(3)	29(4)	27(4)	23(5)	31(5)	20(6)	28(6)	21(7)	17(7)
Ising	1	1	1	1	1	1	-1	-1	-1	-1	-1	-1
QUBO	1	1	1	1	1	1	0	0	0	0	0	0

Solution :11110000 Energie: -1x17x4 = -68!

Solution optimale : $4 \times (-17) = -68 (11110000)$





RESULTATS DE G1 SUR DWAVE

- **STEP 4 :** RESULTATS
 - Sortie 2000Q

Sortie Advantage4.1

```
transformation : ', {0: [600], 1: [571], 2: [3123], 3: [3107], 4: [3138], 5: [615], 6
  [586], 7: [3092]})

    Pas de duplication

 QUBO norm: ', {(6, 7): 0.9411764705882353, (4, 6): 0.9411764705882353, (0, 0): -1.0,
  , 7): 0.9411764705882353, (3, 3): -1.0, (6, 6): -1.0, (4, 5): 0.9411764705882353, (5,
5): -1.0, (1, 7): 0.9411764705882353, (2, 2): -1.0, (2, 4): 0.9411764705882353, (2, 6)
0.9411764705882353, (0, 5): 0.9411764705882353, (0, 4): 0.9411764705882353, (7, 7): -
.0, (1, 6): 0.9411764705882353, (5, 7): 0.9411764705882353, (1, 1): -1.0, (4, 4): -1.0
 (3, 5): 0.9411764705882353})
 QUBO dwave:', {(3138, 615): 0.23529411764705882, (615, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.23529411764705882, (515, 3092): 0.235294117647058411764705882, (515, 3092): 0.25529411764705841176470584117647058411764705841765841765841765841765
      3092): 0.23529411764705882, (3123, 3123): -0.25, (600, 615): 0.23529411764705882,
 36, 586): -0.25, (3092, 3092): -0.25, (3107, 3092): 0.23529411764705882, (615, 615):
.25, (3138, 586): 0.23529411764705882, (571, 586): 0.23529411764705882, (600, 600): -0
25, (600, 3138): 0.23529411764705882, (3138, 3138): -0.25, (3107, 3107): -0.25, (3107
615): 0.23529411764705882, (3123, 586): 0.23529411764705882, (3123, 3138): 0.235294117
4705882, (571, 571): -0.25, (586, 3092): 0.23529411764705882})
  Energie:', [-0.99999999999999])
 'Nb occurences:', [1000])
67.9999999999999]
                                                                                                                                                                                                                                                                     -Wahou 100% de réussite 🙂
 aux de réussite : 100.
nergie moyenne : -68.0
antelvm@dantelvm-VirtualBox:~/Desktop/tp_dwave$
```



RESULTATS DE G1 SUR SOLVEUR CLASSIQUE

- **STEP 4**: RESULTATS (suite)
 - Solver exact

```
# State the problem that we want to solve is very easy
# We will start with a simple case

model = dimod.BinaryQuadraticModel.from_qubo(Q1, offset = 0.0)
model.linear
print("The model that we are going to solve is")
print(model)
print()

# We can solve it exactly

from dimod.reference.samplers import ExactSolver
sampleset = dimod.ExactSolver().sample_qubo(Q1)
#solution = sampler.sample(model)
print("The exact solution is")
print(sampleset)
print()
```

```
ne model that we are going to solve is
inaryQuadraticModel(\{0: -17, 1: -17, 2: -17, 3: -17, 4: -17, 5: -17, 6: -17, 7: -17\}, \{(2, 6): 16, (2, 6): 16, (2, 6): 16\}
16, (2, 4): 16, (3, 5): 16}, 0.0, Vartype.BINARY)
he exact solution is
esponse(rec.array([([0, 0, 0, 0, 0, 0, 0, 0], 0., 1),
          ([1, 0, 0, 0, 0, 0, 0, 0], -17., 1),
          ([1, 1, 0, 0, 0, 0, 0, 0], -34., 1),
          ([0, 1, 0, 0, 0, 0, 0, 0], -17., 1),
           [0, 1, 1, 0, 0, 0, 0, 0], -34., 1),
          ([0, 0, 1, 0, 0, 0, 0, 0], -17., 1),
          ([0, 0, 1, 1, 0, 0, 0, 0], -34., 1),
          ([1, 1, 1, 1, 0, 0, 0, 0], -68,, 1),
          ([0, 1, 1, 1, 0, 0, 0, 0], -51., 1),
           ([1, 1, 0, 1, 0, 0, 0, 0], -51., 1),
           ([1, 0, 0, 1, 0, 0, 0, 0], -34., 1),
           [[1, 1, 1, 0, 1, 1, 0, 0], -21., 1),
          ([0, 0, 1, 1, 1, 1, 0, 0], -20., 1),
          ([1, 0, 1, 1, 1, 1, 0, 0], -5., 1),
          ([1, 1, 1, 1, 1, 1, 0, 0], -22., 1),
```



print(sampleset)

print()

RESULTATS DE G1 SUR SOLVEUR CLASSIQUE

- **STEP 4**: RESULTATS (suite)
 - Recuit simulé (~100% de réussites)

```
# State the problem that we want to solve is very easy
# We will start with a simple case

model = dimod.BinaryQuadraticModel.from_qubo(Q1, offset = 0.0)
model.linear
print("The model that we are going to solve is")
print(model)
print()

# We can solve it exactly

from dimod.reference.samplers import ExactSolver
sampleset = dimod.ExactSolver().sample_qubo(Q1)
#solution = sampler.sample(model)
print("The exact solution is")
```

```
he solution with simulated annealing is
          ([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
           ([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
           ([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
           ([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
            [1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
             1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
            [1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
           ([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
           ([1, 1, 1, 1, 0, 0, 0, 0], -68., 1),
```



RESULTATS (suite)

STEP 4 : G_max?

```
608, 224, 352, 229, 237], 4: [381, 373, 365], 5: [505, 377, 249], 6: [114], 7:
108], 8: [372], 9: [614, 630, 617, 622], 10: [360, 616, 492, 488], 11: [508, 6
 , 498, 500], 12: [613, 625, 629, 621], 13: [618, 362, 490], 14: [382, 366, 374
624, 368, 496], 15: [363, 619, 491], 16: [497, 369, 502, 494], 17: [509, 499,
93, 501], 18: [245, 117, 240, 112], 19: [107, 104, 109], 20: [232, 238, 370, 24
. 242], 21: [378, 250, 122, 127, 119], 22: [367, 111, 361, 233, 105], 23: [380
376, 383, 375], 24: [118, 121, 126, 115], 25: [255, 239, 234, 247, 110, 106], 2
: [252, 244, 371, 243]})
Nb occurences:', [23, 2, 1, 3, 1, 4, 3, 2, 3, 1, 6, 1, 2, 3, 3, 3, 2, 3, 2, 5
7, 3, 5, 2, 3, 8, 1, 1, 1, 2, 7, 1, 1, 2, 1, 4, 2, 1, 1, 2, 1, 2, 1, 1, 2, 1, 1
1, 4, 1, 1, 2, 2, 1, 3, 2, 2, 1, 1, 2, 2, 1, 1, 2, 1, 2, 3, 5, 1, 1, 2, 7, 5,
 1, 2, 1, 4, 1, 2, 1, 2, 1, 3, 1, 8, 2, 2, 1, 3, 1, 1, 1, 2, 2, 1, 1, 4, 1, 1
495.0, -442.00000000000021, -442.000000000013, -442.000000000013, -442.00000
00013, -442.0000000000005, -442.000000000005, -442.000000000005, -442.00000
)00005, -442.0000000000005, -442.0000000000005, -442.000000000005, -442.00000
000005, -441.99999999998, -441.99999999999, -441.99999999998, -441.99999
 99998. -441.99999999998. -441.99999999998. -441.99999999998. -441.99999
999998, -441.999999999998, -441.999999999998, -441.99999999998, -441.99999
999998, -441.999999999998, -441.99999999999, -441.9999999999, -441.999999
999, -441.9999999999, -441.99999999999, -441.999999999, -441.999999999
-441.9999999999, -441.9999999999, -441.000000000022, -441.000000000002
                                     -441.0000000000007
440.9999999999999, -440.999999999999,
440.99999999999, -440.99999999999, -440.999999999999,
440.99999999999, -440.99999999999, -440.99999999999
440.99999999991, -440.99999999991, -440.999999999991,
440.99999999991, -440.999999999991, -440.99999999991, -440.999999999983
440.000000000016. -440.000000000016. -440.00000000016. -440.00000000000
440.000000000008, -440.0000000000008, -440.00000000008, -440.0, -440.0, -440
0, -440.0]
aux de réussite : 0.0934959349593
nergie movenne : -446.361788618
```

G2

~100 qubits pour 27 varaibles ...

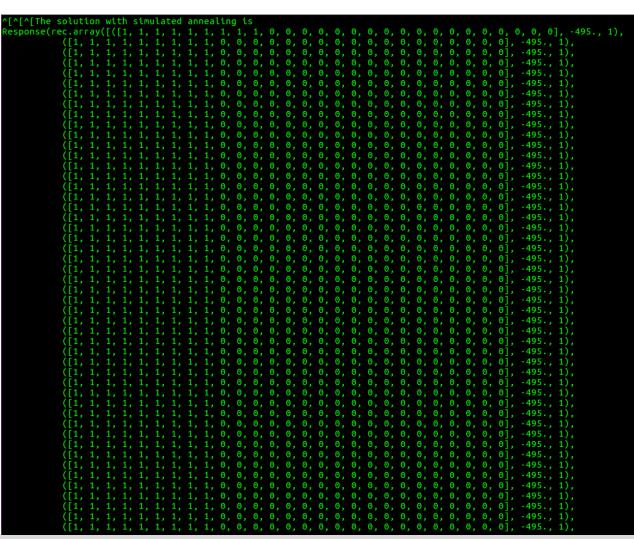
Trouve la solution optimale mais → 23 fois sur 1000 ...



RESULTATS (suite)

STEP 4 : G_max?

G2



Trouve la solution optimale ~100% sur 1000

RESULTATS (suite)

STEP 4 : G_max?

G3

219, 222], 5: [476, 468, 484, 594, 466, 338, 210, 207, 199, 215], 6: [327, 343, 212, 340, 335, 211, 339], 7: [221, 218, 48] 463, 495, 346, 474, 487, 479, 471], 8: [482, 610, 354, 738, 866], 9: [363, 491, 1131, 748, 1003, 875, 747, 619], 10: [752 985, 857, 729, 734], 14: [745, 749, 757, 765, 761, 1017, 895, 889], 15: [997], 16: [606, 598, 582, 576, 448, 48, 176, 590 304, 432, 437, 453, 445], 17: [205, 197, 191, 440, 312, 189, 184], 18: [187, 319, 443, 315], 19: [442, 446, 450, 454], 20: 55, 67, 64, 95, 63, 87, 79, 71], 21: [190, 192, 206, 198], 22: [320, 326, 57, 318, 185, 313], 23: [195, 317, 325, 451, 3 . 24: [92, 84, 52, 60, 68, 201, 73, 76], 25: [78, 72, 200], 26: [188, 196, 203, 204], 27: [69, 70, 449, 321, 65, 193], 28: 214, 461, 80, 208, 464, 336, 333, 329, 457, 470, 462], 29: [458, 334, 74, 330, 202], 30: [316, 324, 331, 332], 31: [452, 45] 460], 32: [253, 377, 505, 633, 617, 638, 630, 249, 614, 622], 33: [241, 365, 369, 381, 373], 34: [632, 504, 248, 376, 508 00], 35: [225, 497, 502, 609, 353, 481, 494, 478, 486], 36: [621, 605, 597, 613], 37: [620, 618, 490, 362, 492], 38: [593, , 485, 499, 477, 501, 493], 39: [473, 601, 602, 345, 596, 612, 604], 40: [350, 342, 611, 483, 227, 358, 355], 41: [3-372, 356, 364], 42: [220, 224, 228, 236, 244, 240, 496, 368], 43: [341, 349, 357, 608, 480, 352], 44: [242, 233, 616, 226 60, 488, 232, 229, 245, 237], 45: [374, 489, 366, 361], 46: [375, 370, 503, 498], 47: [351, 367, 359], 48: [872, 870, 878] 49: [884, 860, 876, 868], 50: [869, 877, 1001, 873], 51: [871, 993, 1121, 865], 52: [1116, 1002, 1140, 1124, 1130, 874, 1132 53: [737, 1117, 740, 732, 730, 858, 1125, 1114, 986], 54: [888, 1016, 1144, 1149, 1141, 1129, 1133], 55: [994, 1122, 1118] 1126, 1134], 56: [1006, 1143, 754, 1138, 1014, 1010, 886, 882], 57: [883, 742, 750, 755, 758], 58: [893, 880, 885], 59: [100 1137, 879, 887, 1009, 1013, 753, 1142, 881], 60: [1120, 998, 864, 992], 61: [1011, 1012, 1004, 731, 996, 859, 987, 988], 6 [1008, 1023, 1015, 1007, 991, 999], 63: [1123, 759, 751, 743, 739, 867, 995]}) 1935.99999999777, -1934.999999999486, -1810.00000000000211, -1810.0000000000064, -1809.999999999845, -1809.9999999999 000000214, -1809.0000000000214, -1809.0000000000066. -1809.00000 -1808.99999999992, -1808.999999999773, -1808.999999999627, -1808.99999999627, -1808.999999999627, -1808.99999999948 1808.9999999948, -1808.0000000000362, -1808.000000000014, -1807.9999999995, -1807.999999999923, -1807.9999999999 .807.99999999923, -1807.999999999923, -1807.99999999775, -1807.99999999775, -1807.999999999702, -1807.999999999 1807.99999999702, -1807.99999999554, -1807.99999999554, -1807.999999999482, -1807.999999999336, -1807.999999999 -1806.99999999704, -1806.999999999632, -1805.999999999927, -1683.0000000005, -1683.0000000000007, -1683.00000000000132 00006. -1683.00000000006. -1683.000000000006, -1682.999999999914, -1682.99999999914, -1682.999999991 82.99999999693, -1682.999999999545, -1682.000000000575, -1682.0000000000355, -1682.000000000282, 00021, -1682.0000000000136, -1682.000000000136, -1682.000000000061, -1682.000000000061, .682.0000000000061, -1682.000000000000061, -1682.000000000000001, -1681.99999999999, -1681.9999999999, -1681.999999999991681.99999999999, -1681.99999999999, -1681.99999999999, -1681.99999999999, -1681.9999999999, -1681.999999999916, -1681.99999999916, -1681.999999999916, -1681.99999999916, -1681.999999999916, -1681.99999999998 -1681.9999999984, -1681.99999999984, -1681.9999999984, -1681.9999999984, -1681.99999999984, -1681.999999999984, .999999999768, -1681.99999999768, -1681.999999999768, -1681.99999999768, -1681.99999999768, -1681.99999999768 31.999999999695, -1681.999999999695, -1681.999999999623, -1681.999999999623, -1681.9999999955, -1681.99999999955, -16 [.999999999475, -1681.999999999475, -1681.9999999999402, -1681.999999999402] aux de réussite : 0.0 rgie moyenne : -1733.97

~430 qubits pour 64 varaibles ...

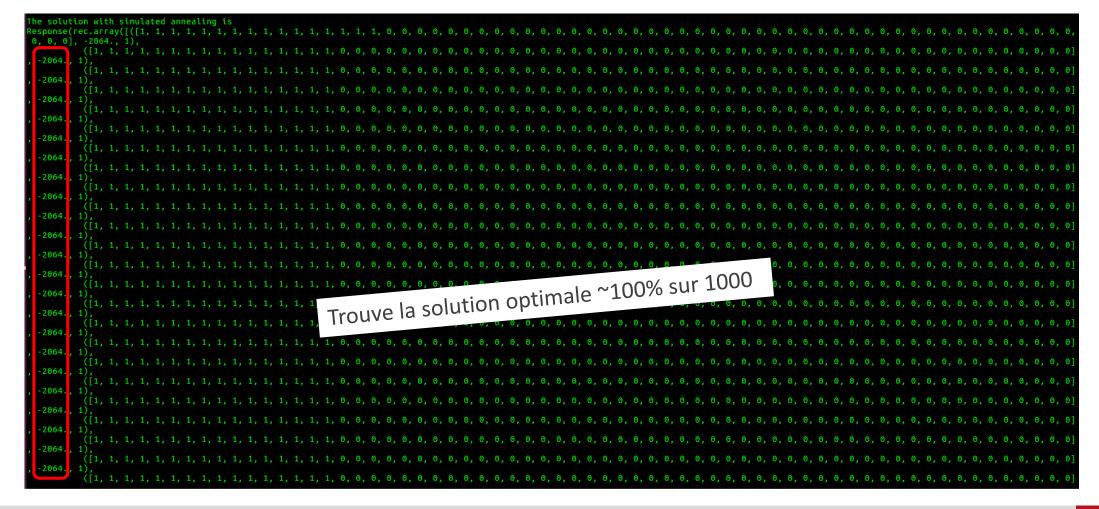
~ 6% d'écart dans le meilleur des cas (ici on a eu de la chance)

Solution optimale 0 ...



STEP 4 : G_max?

G3



daniel.vert2@cea.fr



RESULTATS (suite et fin)

STEP 4 : G_max?

G4





PREMIERE CONCLUSION

Solutions sans topologie

		opt.	best	worst	mean	median	stdev
G_1	n	-68	-68	-68	-68	-68	0
	$n^{1.5}$	-68	-68	-68	-68	-68	0
	n^2	-68	-68	-68	-68	-68	0
G ₂	n	-495	-495	-495	-495	-495	0
	$n^{1.5}$	-495	-495	-495	-495	-495	0
	n^2	-495	-495	-495	-495	-495	0
G_3	n	-2064	-2064	-1810	-2004.7	-2064	79.9
	$n^{1.5}$	-2064	-2064	-2064	-2064	-2064	0
	n^2	-2064	-2064	-2064	-2064	-2064	0
G_4	n	-6275	-6275	-5528	-5785.3	-5777	178.9
	$n^{1.5}$	-6275	-6275	-6026	-6241.8	-6275	86.1
	n^2	-6275	-6275	-6275	-6275	-6275	0

Paramètres Annealing: - Standard cooling time of the form T_{k+1}=0,95T_k

- Stop annealing at T<10^-3.

- Number of iterations of the Metropolis algorithm

→ For each k to T = cst and n: number of variables in QUBO.

For **n iterations** per temperature plate: Lower quality results → standard solutions.

For **n^2 iterations** per plateau: Results of much better quality, but with a much longer computing time.

- Better performance simulated annealing
- Worst solutions over 30 cycles: Almost always better than D-Wave over 10,000 cycles.
- Asymptotic regime of the exponential number of iterations of Sasaki & Hajek's theorem not reached with annealing (sol. Opt. reached for G4)
- Instances small enough to remain easy to anneal in the conventional way

Unduplicated resolved instances for comparison with D-Wave



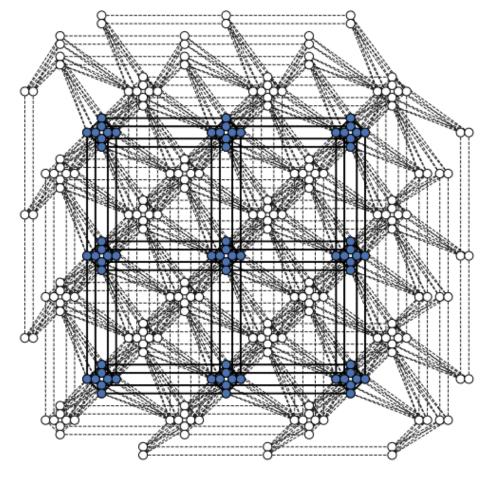
ETUDE DU BIAIS DE LA TOPOLOGIE (DEUXIÈME CONCLUSION)

Solutions avec topologie (Chim. & Peg.)

		opt.	best	worst	mean	median
G_4	n	-6275	-2213	3662	1453.9	1401.0
(Chim.)	$n^{1.5}$	-6275	-4526	-2654	-3585.6	-3699.8
	n^2	-6275	-5028	-4027	-4473.1	-4527.0
D-Wave		-6275	-5025	-3551	-4447.7	-4525
G_4	n	-6275	-3930	-785	-2609.3	-2708.5
(Peg.)	$n^{1.5}$	-6275	-5028	-3580	-4305.5	-4281.0
	n^2	-6275	-5278	-4530	-5035.9	-5028.0

Topologie d'interconnexions entre qubits de 15

	#var.	#qubits (Chim.)	#qubits (Peg.)
G_1	8	16	8
G_1 G_2 G_3 G_4	27	100	46
G_3	64	431	164
G_4	125	958	513



- Résultats sur D-Wave **compétitif** avec ceux du SA.
- Les instances dupliquées sont beaucoup plus difficiles à résoudre que les instances non dupliquées sur SA.
- La topologie plus dense de Pegasus conduit à des QUBO dupliqués plus petits et donne de meilleurs résultats (mais pas optimaux).



DEUXIEME CONCLUSION

Need for duplication → Limits the size of the instances to be mapped.

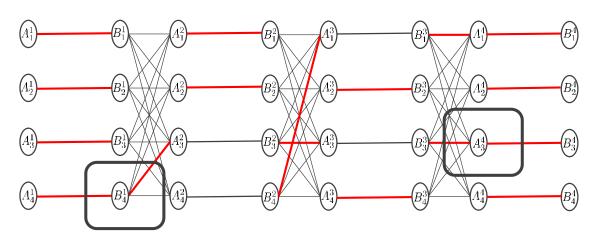
Max G4 size: 125 variables for 958/1098 available qubits (~ 87%) sur DW2X

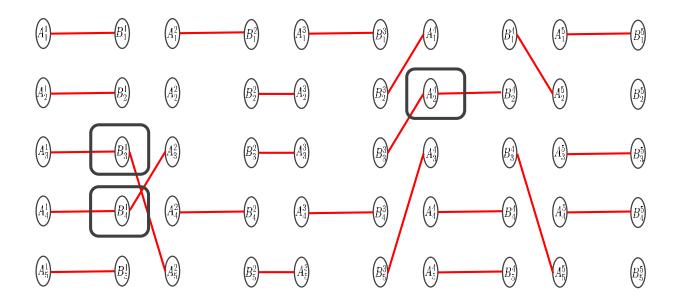
125 variables for 1031/2041 avaibles qubits (~ 50%) sur D2000Q

	#var.	#qubits	average dup.	max. dup.
G_1	8	16	2.0	6
G_2	27	100	3.7	6
G_3	64	431	6.7	18
G_4	125	951	7.6	18

Observation of results :

- Best solution found for G3 / G4 not a matching
- D-Wave does not get the optimal solution







DEUXIEME CONCLUSION AVEC ADVANTAGE

Advantage4.1 (Peg.)

5627 qubits

	#var.	#qubits	#Opt.	Opt.	Best	Worst	Mean	Median	Stdev
G1	8	8	1000	-68	-68	-68	-68	-68	0
G2	27	49	80	-495	-495	-225	-389.2	-387	34.5
G3	64	174	0	-2064	-1937 (x3)	-916	-1565.5	-1553	137.8
G4	125	471	0	-6275	-5278	-2782	-4268.9	-4278	370.4
G5	216	1033	0	-15588	-12134	-6527	-9629.3	-9549	912.3
G6	343	2372	0	-33663	-23383	-8301	-16299.9	-16528	2286.4
G7	512	4186	0	-65600	-40019	-11361	-27278.3	-27734	4645.5

Résultats sur Advantage Identique à D2000Q ?

On passe de 6 connexions à 15!

Question1: besoin de post-processing? Si oui → Constrainte(s): rapide (garde l'avantage quantique) et donne une solution <u>valide</u>!

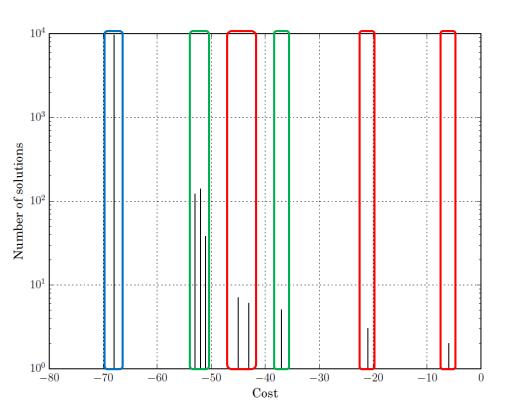
Question 2 : Possible d'améliorer les résultats ?

40





- Résultat sur 10000 runs de G_1 :
 - Solution optimale trouvée (9673 times)
 - Solution non couplage
 - Solution avec mauvaise duplication

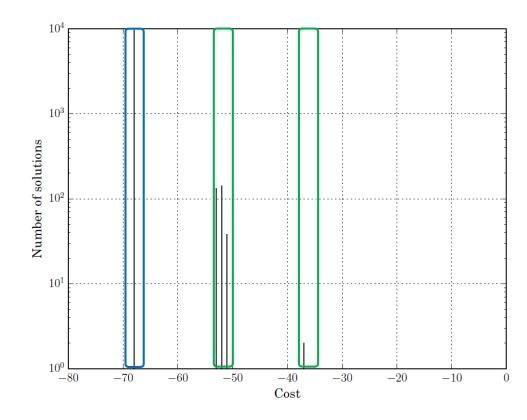


Mauvaise duplication sur le qubit « 6 »

→ Solution non valide

qubits	variable	best	worst
1040	0	1	1
1041	4	0	0
1042	5	0	0
1044	5	0	0
1045	4	0	0
1047	5	0	0
1048	1	1	1
1050	6	0	0
1051	3	1	1
1052	7	0	0
1053	2	1	1
1054	6	0	0
1055	3	1	1
1137	6	0	0
1143	6	0	1
1146	6	0	0
1151	6	0	1

- Post-traitement → vote majoritaire :
 - Solution optimale trouvée (9686 times)
 - Solution non couplage

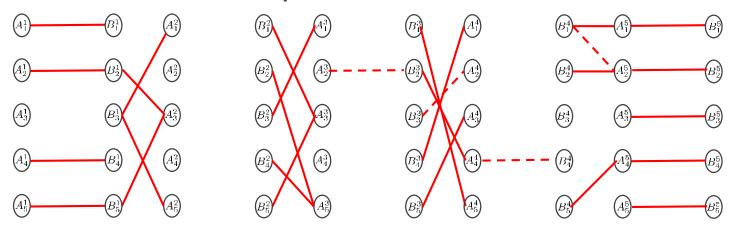


D Vert, R Sirdey, S Louise - arXiv preprint arXiv:1910.05129, 2019



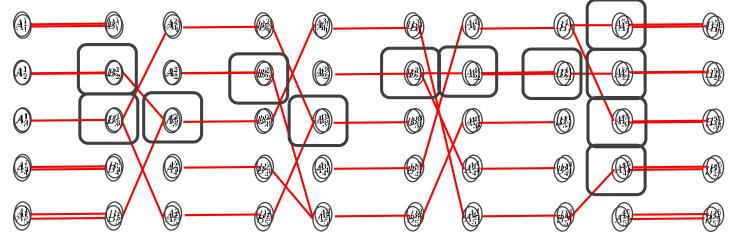
INCONSISTENT DUPLICATIONS

■ Résult → Inconsistent duplications



Need for post-processing Constraint(s): quick and have a valid solution!

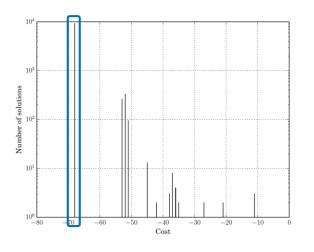
Repul Soliubiomajority vote

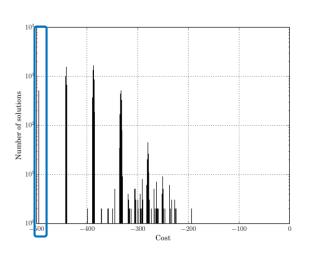


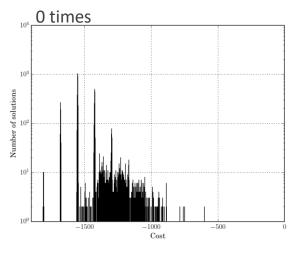


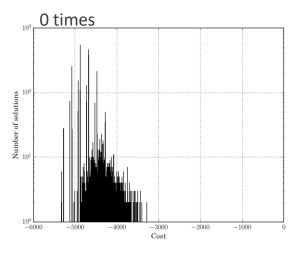
QUESTION 1

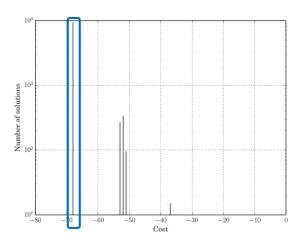
					median						
G_1	-68	-68	-9	-66.8	-68	4.6	-68	-37	-66.8	-68	4.2
G_2	-495	-495	-29	-398.2	-388	48.1	-495	-277	-400.4	-388	44.6
G_3	-2064	-1810	-505	-1454.8	-1548	157.7	-1810	-911	-1496.5	-1550	111.8
G_4	-6275	-5527	-2507	-4609.9	-68 -388 -1548 -4675	346.5	-5527	-3030	-4579.2	-4527	314.1

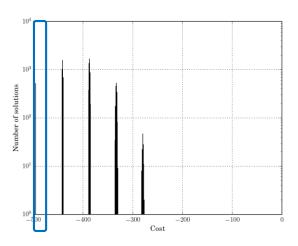


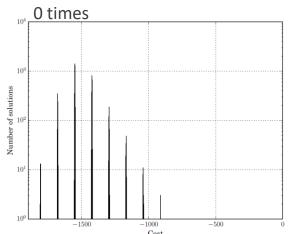


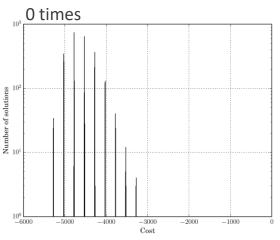














Améliorer les résultats?

Théorème adiabatique : Si l'évolution temporelle est assez lente (τ est assez grand) \rightarrow La solution optimale $\varepsilon(\sigma)$ est

obtenue avec une forte probabilité.

```
Q Search properties
                                                                                                                   Properties
                                                                                                                            Parameters
ANNEALING_TIME_RANGE
                                                                                                                   [1, 2000]
import dwave sapi2
                                                                                                                   CHIP_ID
from dwave sapi2.remote import RemoteConnection
from dwave sapi2.core import solve ising, solve qubo
                                                                                                                   DW 2000Q 6
                                                                                                                   DEFAULT_ANNEALING_TIME
url = 'https://cloud.dwavesys.com/sapi/'
token = 'DEV-32b473f753f4641ca3e87d4318f325295665204d'
                                                                                                                   DEFAULT_PROGRAMMING_THERMALIZATION
# create a remote connection
conn = RemoteConnection(url, token)
                                                                                                                   DEFAULT_READOUT_THERMALIZATION
# get the solver
solver = conn.get solver('DW 20000 6')
                                                                                                                   EXTENDED_J_RANGE
#solver = conn.get_solver('Advantage_system4.1')
                                                                                                                   [-2, 1]
                                                                                                                   H_GAIN_SCHEDULE_RANGE
params = { "num_reads": 1000, "max_answers": 1000, "answer_mode": "histogram", "auto_scale": True, (annealing_time": 2
                                                                                                     "num_spin_reve
                                                                                                                   [-5, 5]
# solve OUBO problem with parameters
                                                                                                                   MAX ANNEAL SCHEDULE POINTS
answer = solve_qubo(solver, QUBO, **params)
print"\n'
print("Energie:", answer["energies"])
print("Nb occurences:", answer["num_occurrences"])
i = 0
count = 0
while (answer["energies"][i]*(norma) <= -68) :</pre>
      count = count + answer["num occurrences"][i]
      i = i + 1
```

print("Energie moyenne : " + str(sum([answer["energies"][i] * answer["num occurrences"][i] for i in range (len(answer["energies"]))]) / sum(answer["num occurrences"])*(norma)))

print([energie*(norma) for energie in answer["energies"]])

print("Taux de réussite : " + str(float(count)/sum(answer["num occurrences"])))



■ Améliorer les résultats?

	AT	#var.	#qubits	#Opt.	Opt.	Best	Worst	Mean	Median	Stdev	
	20	64	360	0	-2064	-1808	-1105	-1482.15	-1486.7	86.8	Résultats « de base »
G3	100	64	360	0	-2064	-1937	-1680	-1714.3	-1785.9	77.7	Résultats sensiblement similaire → AT
ds	1000	64	360	0	-2064	-1810	-1680	-1700.2	-1713.9	82.5	permet « d'écraser » les pires solutions
	2000	64	360	0	-2064	-1810	-1426	-1565.3	-1573.6	89.9	vers les meilleures
	20	125	1031	0	-6275	-5029	-4776	-4808.5	-4996.4	197.3	Résultats « de base »
G4	100	125	1031	0	-6275	-5275	-4778	-4754.8	-4859.8	178.6	Améliore la qualité de la solution
04	1000	125	1031	0	-6275	-5276	-4280	-4554.8	-4598.9	221.4	optimale
	2000	125	1031	0	-6275	-5277	-4280	-4562.7	-4603.3	219.1	

Sur D2000Q → **Non** ...



Et sur Advantadge4.1?

```
N = 3, RCS:0.496
                           (best,
                                   worst,
                                            mean,
                                                     median,
                                                              stdev)
AT:40.89, NSRT:20, Energy: [(-1937., -1170., -1628.17, -1616.5, 164.46063693)]
AT:404.44, NSRT:0, Energy: [(-1936., -1043., -1659.94, -1680., 153.89514742)]
AT:1232.52, NSRT:0, Energy: [(-2064., -1170., -1690.67, -1681., 165.02612248)]
                                                                                                 Solution optimale!
AT:1616.26, NSRT:0, Energy: [(-2064., -1171., -1735.756, -1683., 140.86804628, b'9/1000')]
N = 4, RCS:0.667
                           (best, worst, mean,
                                                    median,
                                                              stdev)
AT:182.27, NSRT:5, Energy: [(-5275., -3775., -4541.33, -4527., 341.89059815)]
AT:626.61, NSRT:0, Energy: [(-5527., -3526., -4766.153, -4776., 308.55241628, b'1/1000')]
                                                                                                Mieux que DW2000Q!
AT:868.97, NSRT:20, Energy: [(-5028., -3280., -4516.4, -4527., 381.59651466)]
AT:1434.48, NSRT:0, Energy: [(-5028., -3773., -4618.57, -4528.5, 288.32891825)]
                                                                                                (11.92% best vs opt.)
AT:1697.05, NSRT:18, Energy: [(-5027., -3777., -4551.25, -4527., 283.53131661)]
N = 5, RCS:0.569
                           (best, worst, mean,
                                                    median,
AT:384.24, NSRT:25, Energy: [(-12134., -8245., -10835.38, -10834., 772.65500425)]
AT:626.61, NSRT:50, Energy: [(-12564., -8246., -10696.91, -10835., 863.21299915)]
AT:990.15, NSRT:0, Energy: [(-12562., -7385., -10666.95, -10834.5, 885.29777335)]
AT:1575.86, NSRT:5, Energy: [(-12562., -8678., -10792.02, -10834., 746.78836333)]
AT:1919.21, NSRT:0, Energy: [(-12994., -8676., -11122.638, -11266., 733.42026898, b'1/1000')]
N = 6, RCS:0.888
                                                     median,
                           (best, worst, mean,
AT:364.05, NSRT:0, Energy: [(-26805., -16521., -22291.68, -22683., 1796.02446464)]
                                                                                                                 G5 \rightarrow ^{\sim}16.64\%
AT:687.2, NSRT:18, Energy: [(-25433., -17205., -21824.25, -22002., 1571.67414164)]
                                                                                                                 G6 → ~18.35%
AT:929.56, NSRT:20, Energy: [(-25430., -15835., -21947.48, -22002., 1851.31472462)]
AT:1414.29, NSRT:5, Energy: [(-25429., -17885., -22468.23, -22685.5, 1548.27271406)]
                                                                                                                 G7 → ~21.85%
AT:1656.65, NSRT:5, Energy: [(-27487., -18574., -23302.174, -23374., 1463.85336073, b'1/1000')]
N = 7, RCS:1.157
                           (best, worst, mean,
                                                     median.
AT:61.09, NSRT:15, Energy: [(-43075., -29769., -36523.03, -36930.5, 3067.31622255)]
AT:444.83, NSRT:10, Energy: [(-44098., -20554., -36778.89, -36931., 3393.44609474)]
AT:1091.14, NSRT:0, Energy: [(-46148., -28746., -39074.07, -38982.5, 3295.13821032)]
AT:1535.47, NSRT:5, Energy: [(-47167., -32837., -40003.79, -40004.5, 2973.82983472)]
AT:1899.02, NSRT:5, Energy: [(-51266., -31819., -41678.632, -42048., 2807.29091556, b'1/1000')
```

CONCLUSION

Difficult problem for simulated and quantum annealing

First benchmark on a D-Wave With topology → Similar results between D-Wave and SA.

The constraints imposed by the graph show that SA and QA cannot solve the problem!

Low quality results

- The need to duplicate qubits strongly limits the size of accessible problems.
- Necessity of post-treatments **Duplication errors** ⊗

 not representing a valid solution

Sparse topology

Need more connections between qubits \rightarrow mapping denser and larger problems!

► MERCI

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[SN] « Benchmarking quantum annealing against "hard" instances of the bipartite matching problem », in Special issue for Springer journals (SN Computer Science).