



Quantum simulation of one-dimensional system on IBMQ device



Members

Coach

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How to connect the problems and Quantum computer?

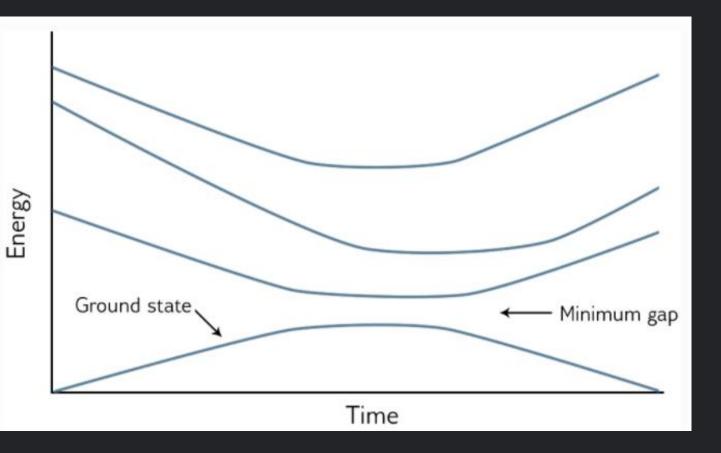
The Hamiltonian of Ising model:

$$H = \frac{-A(s)}{2} \left(\sum_{i} \sigma_{x}^{(i)} \right) + \frac{1 - A(s)}{2} \left(\sum_{i} h_{i} \sigma_{z}^{(i)} + \sum_{i > j} J_{i,j} \sigma_{z}^{(i)} \sigma_{z}^{(j)} \right)$$

Pauli matrices operating on a qubit x_i

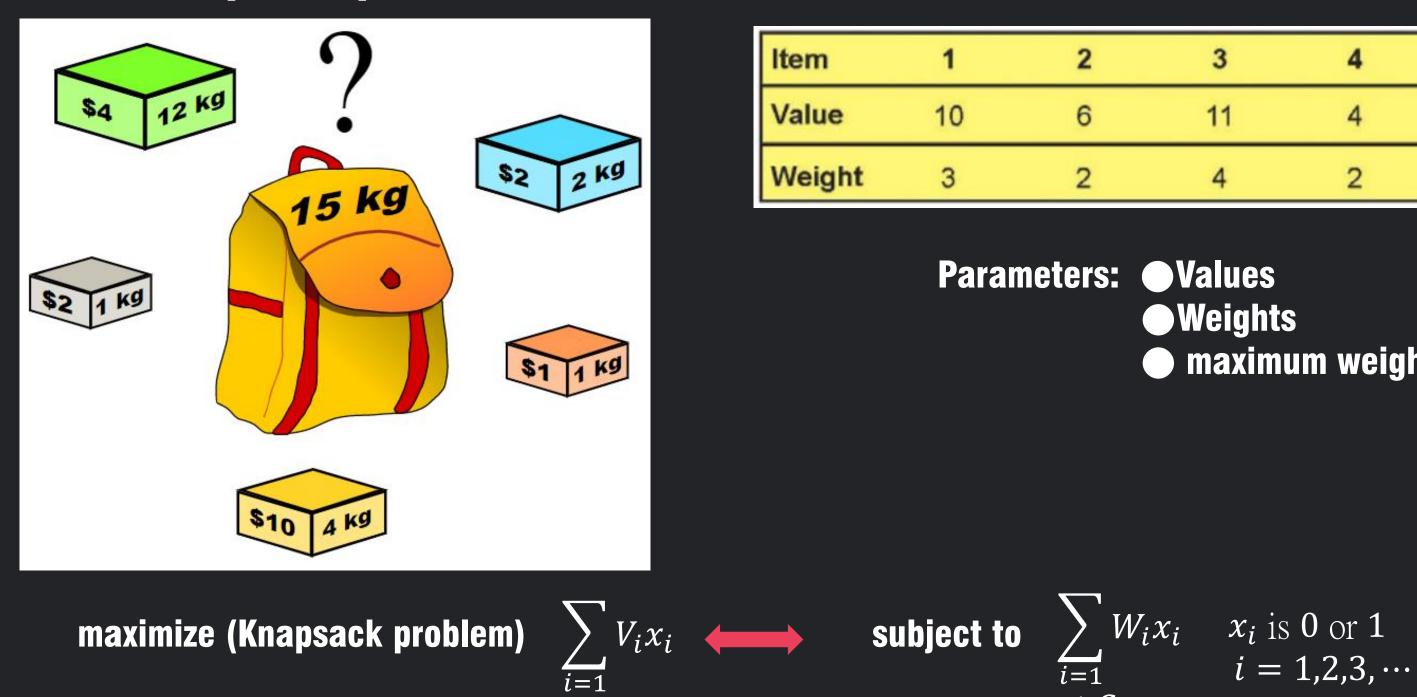
 h_i and interacting parameter $J_{i,i}$ are the qubit biases and coupling strengths

https://docs.dwavesys.com/docs/latest/c_gs_2.html?fbclid=lwAR1zq5W45Lr0SHCGS5KrKc NZBoSBCKcTpBqrd2-aGaNUW-D_LhZ56NQDLPc



There's an interesting topic!

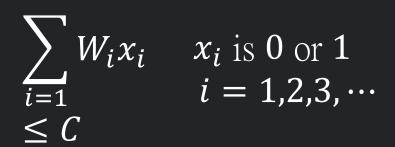
Knapsack problem



https://en.wikipedia.org/wiki/Knapsack problem

2	3	4	5
6	11	4	5
2	4	2	3

Parameters: •Values Weights maximum weight



Max-Cut problem

Model :

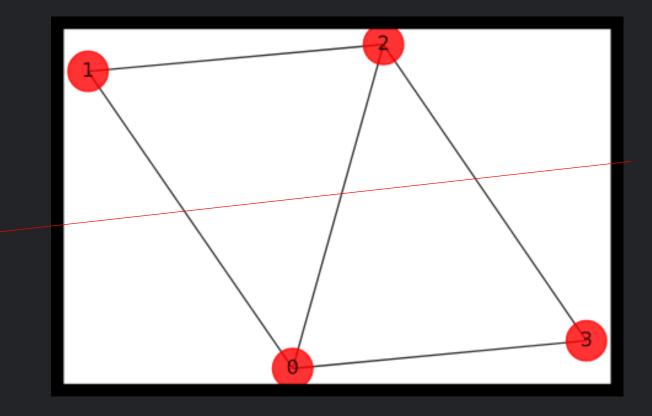
$$C(\mathbf{x}) = \sum_{i,j} w_{ij} x_i (1 - x_j) + \sum_i w_i x_i$$

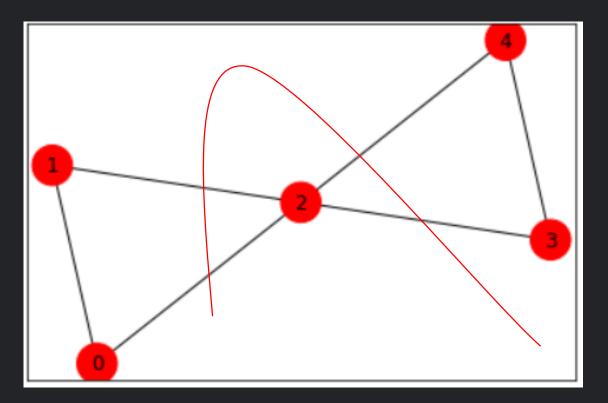
$$x_i \to (1 - Z_i)/2, \text{ where } Z_i \text{ is the Pauli Z operator, } \sigma_z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

But we can solve it only if we mapping this model to the lsing model :

$$H = \sum_{i} w_i Z_i + \sum_{i < j} w_{ij} Z_i Z_j$$

After mapping, we use quantum adiabatic algorithm to do optimization.



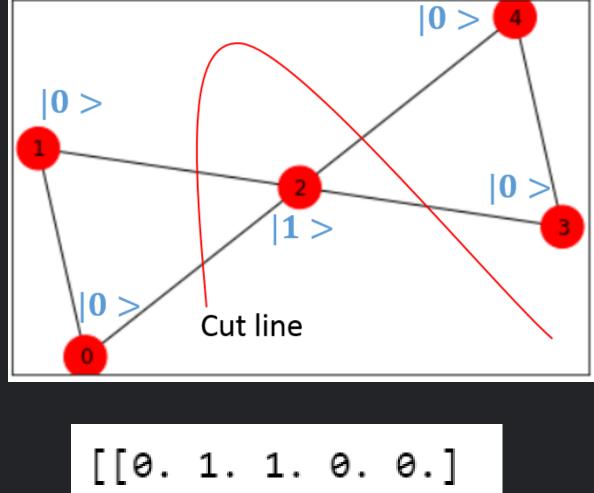


What we want to solve, with n=5.

Compared By Process

Coding process: Set up qubit dots connection

```
w=np.zeros([n,n])
for i in range(n):
    for j in range(n):
        temp=G.get_edge_data(i,j,default=0)
        if temp != 0:
            w[i,j]=temp['weight']
print(w)
```



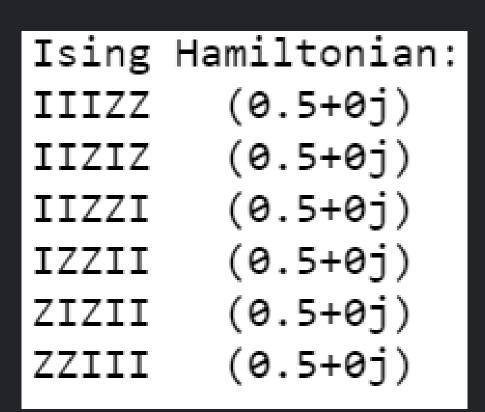
$$\begin{bmatrix} [0. 1. 1. 0. 0.] \\ [1. 0. 1. 0. 0.] \\ [1. 1. 0. 1. 1.] \\ [0. 0. 1. 0. 1.] \\ [0. 0. 1. 0. 1.] \end{bmatrix}$$

Compared By Process

Coding process: Transform into Ising model

from qiskit.optimization.applications.ising import max_cut
qubitOp, offset = max_cut.get_operator(w)
print('Offset:', offset)
print('Ising Hamiltonian:')
print(qubitOp.print_details())

 $H = \sum_{i} w_i Z_i + \sum_{i \le i} w_{ij} Z_i Z_j.$

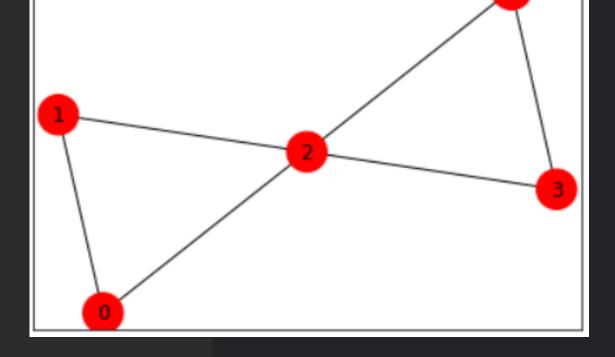


S 9 Process

Coding process: Set up 5-qubit dots

```
# Generating the butterfly graph with 5 nodes
     = 5
n
     = np.arange(0,n,1)
V
     = [(0,1,1.0), (0,2,1.0), (1,2,1.0), (3,2,1.0), (3,4,1.0), (4,2,1.0)]
E
     = nx.Graph()
G
G.add_nodes_from(V)
G.add_weighted_edges_from(E)
# Generate plot of the Graph
colors
       = ['r' for node in G.nodes()]
default_axes = plt.axes(frameon=True)
    = nx.spring_layout(G)
pos
```

nx.draw_networkx(G, node_color=colors, node_size=600, alpha=1, ax=default_axes, pos=pos)



Hamiltonian in Quantum Annealing

$$H = \underbrace{\frac{-A(s)}{2} \left(\sum_{i} \sigma_{x}^{(i)}\right)}_{2} + \underbrace{\frac{1 - A(s)}{2} \left(\sum_{i} h_{i} \sigma_{z}^{(i)} + \sum_{i>j} J_{i} + \sum_{i>j} J_{i}\right)}_{HO, \text{ initial state}}$$

HD, initial state

- Use 5 qubits with initial state evolving to final state
- Final state is from the Max-cut problems

 $I_{i,j}\sigma_z^{(i)}\sigma_Z^{(j)}$

Coding process: HamiltonianGate \rightarrow Set up Hamiltonian

-A(s)/2*(x.tensor(i).tensor(i).tensor(i).tensor(i)+ HO i.tensor(x).tensor(i).tensor(i).tensor(i)+ i.tensor(i).tensor(x).tensor(i).tensor(i)+ i.tensor(i).tensor(i).tensor(x).tensor(i)+ i.tensor(i).tensor(i).tensor(i).tensor(x)) $H_p = (1-A(s))/2*(qubit0p.to_opflow())$ circ.append(HamiltonianGate(H0, 1), qr) circ.append(HamiltonianGate(H_p, 1), qr)

Coding process: Measuring backend with simulator

sim_backend = provider.get_backend('ibmq_qasm_simulator')

s = round((s-0.01), 6)

circ.measure(qr,cr)

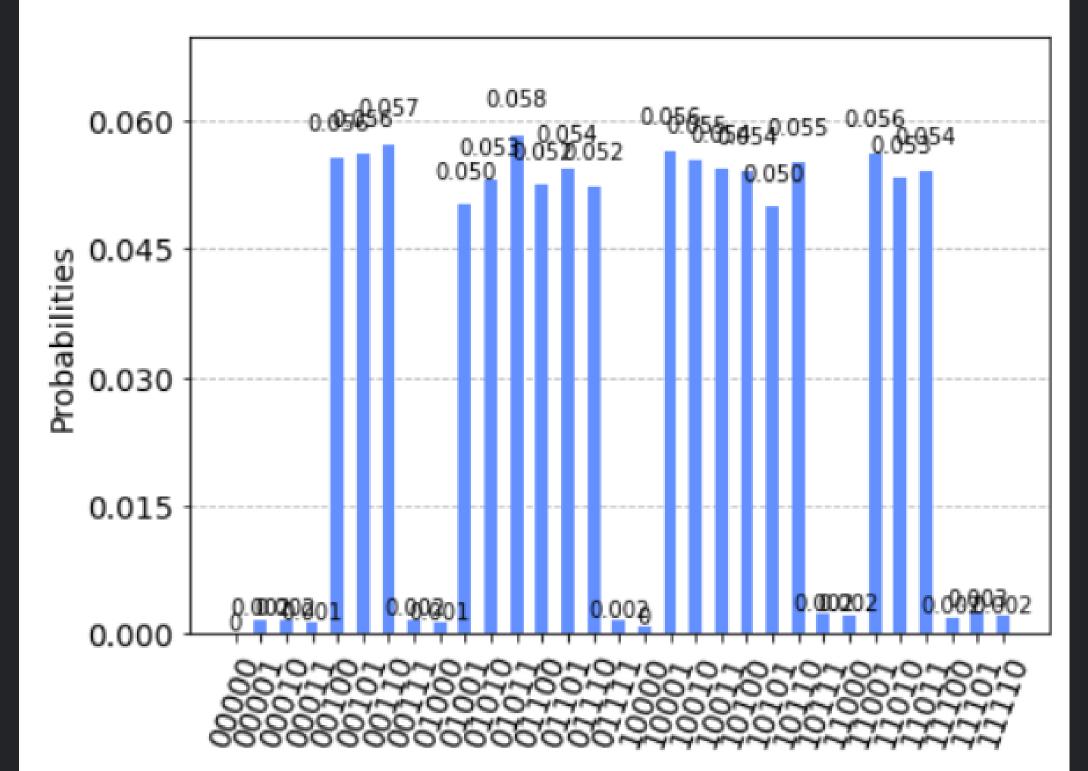
job = execute(circ,backend = sim_backend, shots = 8192)

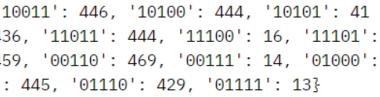
output = job.result()

answer = output.get_counts(circ)

J Result

{'00000': 1, '00001': 14, '10000': 7, '10001': 462, '10010': 453, '10011': 446, '10100': 444, '10101': 41 0, '10110': 451, '10111': 19, '11000': 18, '11001': 459, '11010': 436, '11011': 444, '11100': 16, '11101': 22, '11110': 17, '00010': 14, '00011': 11, '00100': 455, '00101': 459, '00110': 469, '00111': 14, '01000': 11, '01001': 412, '01010': 434, '01011': 477, '01100': 430, '01101': 445, '01110': 429, '01111': 13}





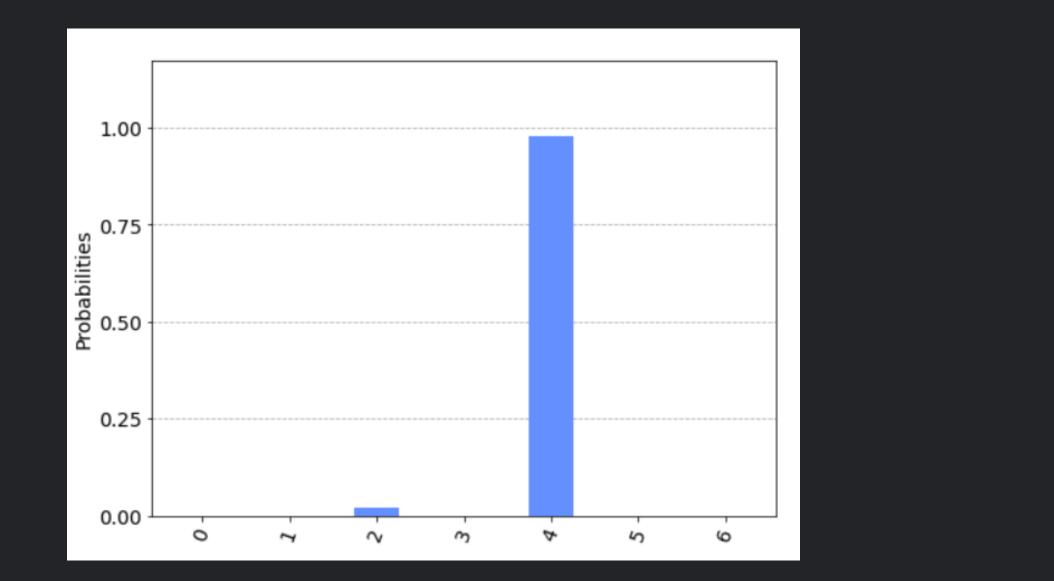
S 9 Process

Coding process: Counts for eigenvalue which means 'cut'

```
p3=[]
n=5
best_cost_brute = 0
for b in range(2**n):
    x = [int(t) for t in reversed(list(bin(b)[2:].zfill(n)))]
    cost = 0
    for i in range(n):
        for j in range(n):
            cost = cost + w[i,j]*x[i]*(1-x[j])
    if best_cost_brute < cost:</pre>
        best_cost_brute = cost
        xbest_brute = x
    p3.append(best_cost_brute)
    print('case = ' + str(x)+ ' cost = ' + str(cost))
colors = ['r' if xbest_brute[i] == 0 else 'b' for i in range(n)]
nx.draw_networkx(G, node_color=colors, node_size=600, alpha=.8, pos=pos)
```

Solution Process

Coding process: Calculate the eigenvalue value vary with steps



Coding process: Calculate the expectation value vary with steps

[111.1111111111111, 125.0, 142.85714285714286, 166.666666666666666666, 200.0, 250.0, 333.3333333333333, 500. 0, 1000.0] [3.68701171875, 3.689697265625, 3.685546875, 3.6865234375, 3.704345703125, 3.686767578125, 3.69 0673828125, 3.692138671875, 3.685302734375]

The expectation value is near the maximum value 4 !



Lucas, Andrew. "Ising formulations of many NP problems." Frontiers in Physics 2 (2014): 5.

Coffey, Mark W. "Adiabatic quantum computing solution of the knapsack problem." arXiv preprint arXiv:1701.05584 (2017).

https://giskit.org/documentation/stubs/giskit.optimization.applications.ising.knapsack.html?fbclid=lwAR2GWY01Qjtzlai9NhudxMN42q Zzq8d9510a9IRBgBy0Zb80gxFo uL8M

Murawski, Carsten, and Peter Bossaerts. "How humans solve complex problems: The case of the Knapsack problem." Scientific reports 6 (2016): 34851.

THANK YOU FOR LISTENING!

ANY QUESTIONS?

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