

QOSF - Problem 4

September 25, 2020

1 QOSF Problem-4: The VQE Algorithm (from scratch)

The following code can be used to find the minimum eigenvalue of **any** 2-qubit Hamiltonian using the Variational Quantum Eigensolver (VQE) algorithm. We will be coding using IBM's open-source Quantum SDK QISKit.

```
[1]: import numpy as np
      np.random.seed(20)
```

1.1 Hamiltonian

```
[2]: # building the Hamiltonian
      # this code will work for any 2-qubit (4x4) Hamiltonian

      H = np.zeros((4, 4))
      H[0][0], H[1][2], H[2][1], H[3][3] = 1, -1, -1, 1
      print(H)
```

```
[[ 1.  0.  0.  0.]
 [ 0.  0. -1.  0.]
 [ 0. -1.  0.  0.]
 [ 0.  0.  0.  1.]]
```

Now, we will decompose the Hamiltonian into Pauli terms. Any 2-qubit Hamiltonian H can be written as a sum of Pauli-strings as follows -

$$H = \sum_{i,j=0}^3 a_{ij} P_i \otimes P_j$$

where, $(P_0, P_1, P_2, P_3) = (I, X, Y, Z)$.

Then, using the Hilbert-Schmidt Inner Product (or, Trace Inner Product), we can write -

$$a_{ij} = \frac{1}{4} \text{Tr} [(P_i \otimes P_j) H]$$

```
[3]: # creating the list of Pauli Matrices
      I = np.array([[1, 0], [0, 1]])
```

```

X = np.array([[0, 1], [1, 0]])
Y = np.array([[0, -1j], [1j, 0]])
Z = np.array([[1, 0], [0, -1]])

PAULI = [I, X, Y, Z]

print(I, '\n', X, '\n', Y, '\n', Z)

```

```

[[1 0]
 [0 1]]
[[0 1]
 [1 0]]
[[ 0.+0.j -0.-1.j]
 [ 0.+1.j  0.+0.j]]
[[ 1  0]
 [ 0 -1]]

```

[4]: *# finding the Pauli-coefficients using the Trace Inner Product*

```

dim = (4, 4)

# coefficients matrix
coeffs = np.zeros(dim)*1j

for i in range(dim[0]):
    for j in range(dim[1]):

        # trace inner product
        coeffs[i][j] = np.trace(np.kron(PAULI[i], PAULI[j])@H)/4

print(coeffs)

```

```

[[ 0.5+0.j  0. +0.j  0. +0.j  0. +0.j]
 [ 0. +0.j -0.5+0.j  0. +0.j  0. +0.j]
 [ 0. +0.j  0. +0.j -0.5+0.j  0. +0.j]
 [ 0. +0.j  0. +0.j  0. +0.j  0.5+0.j]]

```

We will use Numpy's linear algebra module to find the actual eigenvalues. This will serve as a reference to find the error in our computed value.

[5]:

```

eigs = np.linalg.eigvals(H)
min_eig = np.amin(eigs)
print('All Eigenvalues', eigs)
print('Minimum Eigenvalue:', min_eig)

```

```

All Eigenvalues [ 1. -1.  1.  1.]
Minimum Eigenvalue: -1.0

```

```
[6]: np.linalg.eig(H)
```

```
[6]: (array([ 1., -1.,  1.,  1.]),
      array([[ 0.,          0.,          1.,          0.],
             [-0.70710678,  0.70710678,  0.,          0.],
             [ 0.70710678,  0.70710678,  0.,          0.],
             [ 0.,          0.,          0.,          1.]])
```

For the given Hamiltonian, the lowest eigenvalue is (-1) , which occurs for the state -

$$\begin{bmatrix} 0 \\ 1/\sqrt{2} \\ 1/\sqrt{2} \\ 0 \end{bmatrix} = \frac{1}{\sqrt{2}}|10\rangle + \frac{1}{\sqrt{2}}|01\rangle$$

1.2 Creating Ansatz

```
[7]: # we will be building and running circuits in Qiskit
      from qiskit import QuantumCircuit, ClassicalRegister, QuantumRegister
```

```
[8]: # creating ansatz circuit
      def get_ansatz(param):
          circ = QuantumCircuit(2)

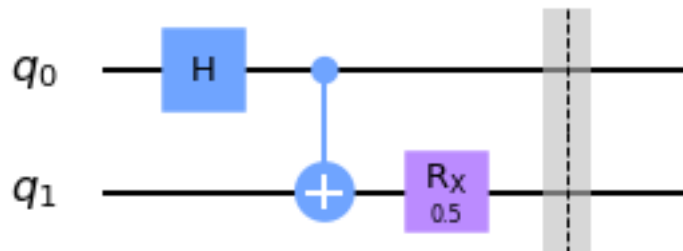
          circ.h(0)
          circ.cx(0, 1)
          circ.rx(param[0], 1)

          circ.barrier()

          return circ

      # checking circuit
      get_ansatz([0.5]).draw(output='mpl')
```

[8]:



```
[9]: # generating the VQE Measurement Circuit for a Pauli String
def get_measurement_circuit(string):
    """Generates a Pauli-Measurement Circuit.
       :string: is a 2-element list or tuple, indicating the measurement-basis_
       ↪ of the corresponding qubit
       input the following numbers for the corresponding basis -
       0 - I, 1 - X, 2 - Y, 3 - Z"""

    meas = QuantumCircuit(2, 2)

    for i in range(len(string)):
        if string[i]==1:
            # measuring in X-basis
            meas.h(i)
        elif string[i]==2:
            # measuring in Z-basis
            meas.u2(0, (np.pi)/2, i)

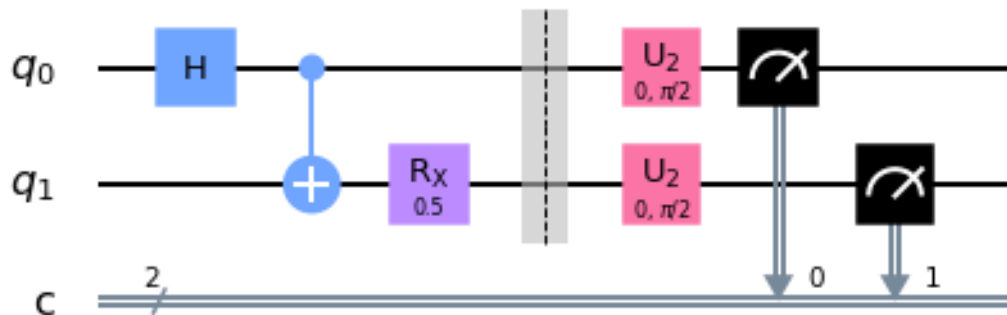
    # default measurement is in Z-basis
    # so, don't have to do anything for string[i]=3

    meas.measure([0, 1], [0, 1])

    return meas
```

```
[10]: # generating a measurement circuit for YY measurement to check
(get_ansatz([0.5]) + get_measurement_circuit([2, 2])).draw(output='mpl')
```

[10]:



2 Optimization

2.1 Configuring the Simulator

```
[11]: # Import Aer - qiskit's simulation module
      from qiskit import Aer, execute
      # Run the quantum circuit on a simulator backend
      backend = Aer.get_backend('qasm_simulator')
```

2.2 Cost Function

```
[12]: # making a 4-element probability vector from qiskit's dictionary of simulation_
      ↪ results

      def get_probability_distribution(counts):
          output_distr = np.zeros(4)

          for i in range(4):
              # generate reversed bit-string of i to access the counts dictionary
              output_string = format(i, '02b')[::-1]
              if output_string in counts:
                  output_distr[i] = counts[output_string]
                  # print(output_distr)

          # normalizing the probability distribution
          output_distr = output_distr/np.sum(output_distr)
          return output_distr

      # checking
      count = {'01': 500, '11':500}
      get_probability_distribution(count)
```

```
[12]: array([0. , 0. , 0.5, 0.5])
```

```
[13]: # list of eigen-values of each Pauli-Matrix and I
      # we will need them to add the outcome probabilities with the required sign
      pauli_eigen = [np.array([1, 1]), np.array([1, -1]), np.array([1, -1]), np.
      ↪ array([1, -1])]
      print(pauli_eigen)
```

```
[array([1, 1]), array([ 1, -1]), array([ 1, -1]), array([ 1, -1])]
```

```
[14]: # this is the cost-function
      def objective_function(params):
          """Outputs the expectation value of the Hamiltonian for the state
```

```

    corresponding to the parameters (params)"""

    # Obtain a quantum circuit instance from the paramters
    circ = get_ansatz(params)

    cost = 0
    for i in range(dim[0]):
        for j in range(dim[1]):
            # no-need to simulate if the Pauli-Weight is 0
            if coeffs[i][j]==0:
                continue

            # Obtain a measurement circuit for the Pauli-string
            test = circ + get_measurement_circuit([i, j])

            # Create a Quantum Program for execution
            result = execute(test, backend, shots=NUM_SHOTS).result()

            # Obtain the counts for each measured state, and convert those
            ↪ counts into a probability vector
            output_distr = get_probability_distribution(result.get_counts())
            # print('Output Distribution', output_distr)

            # computing the sign vector to add the probabilities
            sign_vector = np.kron(pauli_eigen[i], pauli_eigen[j])
            # print('Sign Vector', sign_vector)

            cost += coeffs[i][j]*(np.dot(sign_vector, output_distr))
            # print('Present cost is', cost, '\n')

    return np.real(cost.real)

```

```

[15]: # checking if the cost-function works
      NUM_SHOTS = 1000
      objective_function([np.pi])

```

```

[15]: -1.0

```

2.3 Configuring Optimizer

```

[16]: from qiskit.aqua.components.optimizers import COBYLA, SPSA
      import time

      # Initialize the Constrained Optimization BY Linear Approximation (COBYLA)
      ↪ optimizer

```

```
optimizer = COBYLA(maxiter=1000)

# Create the initial parameters
INITIAL = np.random.rand(1)

# bounds of the parameters
BOUNDS = [(0, 2*np.pi)]*len(INITIAL)
```

2.4 Optimizing Circuit

```
[17]: NUM_SHOTS = 500
print("Sampling Circuit", NUM_SHOTS, "times.")

# timing simulation
start = time.time()

# optimize circuit
ret = optimizer.optimize(num_vars=len(INITIAL),
    ↳objective_function=objective_function,
    ↳initial_point=INITIAL)

print("Circuit Optimized!", round(time.time()-start), "seconds elapsed.\n")

print('Value of parameter', (ret[0]/np.pi)%2, '\b*pi\n', 'Minimum eigenvalue',
    ↳ret[1], '\n')
print('Error = ', np.abs((ret[1]-min_eig)/min_eig), '%')
```

Sampling Circuit 500 times.
Circuit Optimized! 1 seconds elapsed.

Value of parameter [1.00287693]*pi
Minimum eigenvalue -1.0

Error = 0.0 %

2.5 References

- [1] Quantum Computation and Quantum Information, M. A. Nielsen & I. L. Chuang
- [2] Simulating Molecules using VQE, Qiskit Textbook
- [3] Variational Quantum Eigensolver, Davit Khachatryan

[]: