Screening Task 1

September 27, 2020

0.1 Task instructions

Here, we are asked to build a 4 qubit circuit formed by L layers of even are odd blocks of parametrized gates, where the odd blocks are R_x gates each with a parameter $\theta_{i,n} \in (0, 2\pi)$, where $i \in 1, 2, 3, 4$ and $n \in \text{odd}$; and the even block are formed by R_z gates with different $\theta_{i,n}$ parameters with even n followed by c_z gates between each pair of qubits. The resultant 4 qubit state $|\psi(\theta)\rangle$ is used to calculate the minimum distance

$$\epsilon = \min_{\theta} || |\psi(\theta) > -| \phi(\theta) > ||,$$

where $|\phi(\theta)\rangle$ is a randomly generated 4 qubit vector. Therefore, we need to write a rutine that finds the optimal $\theta_{i,n}$ parameters that results in the minimum ϵ for each L.

```
[1]: from qiskit import QuantumCircuit, execute, Aer from qiskit.tools.jupyter import * from qiskit.visualization import * import numpy as np from qiskit.quantum_info import random_statevector from qiskit_textbook.tools import array_to_latex import random from qiskit.aqua.components.optimizers import COBYLA, AQGD import matplotlib.pyplot as plt
```

0.1.1 Solution

First, we create the random 4 qubit vector $|\phi(\theta)\rangle$ that will be used as the reference vector, and the maximum number of layers L_{max} for the circuit

```
[2]: Phi =np.fromiter(random_statevector(16).to_dict().values(), dtype = "complex_")

→ #phi random vector

L_max = 8 #maximum number of layers
```

And the rutine to get the optimal $\theta_{i,n}$ parameters for each circuit with the number of layers $L \in (1, L_{max})$ is presented here. We use the Constrained Optimization By Linear Approximation (COBYLA) method for the optimization.

```
[3]: Epsilon = []
    for L in range(1, L_max+1): #loop for each L case
        def qc_gates(theta): #function to create the circuit
            theta_d = []
            for i_d in range (0,L):
                theta_r = [theta[8*i_d:8*i_d+4],theta[8*i_d+4:8*(i_d+1)]]
                theta_d.append(theta_r)
            theta = np.asarray(theta_d)
            for i_l in range(0, L):
                for i in range (0, 4):
                    circ.rx(theta[i_1][0][i], i) #rx gates
                for i in range (0, 4):
                    circ.rz(theta[i_1][0][i], i)
                                                   #rz gates
                circ.cz(0, 1)
                                                     #cz gates
                circ.cz(0, 2)
                circ.cz(0, 3)
                circ.cz(1, 2)
                circ.cz(1, 3)
                circ.cz(2, 3)
        def cost(theta_c):
                                       #function to calculate epsilon
            global circ
            circ = QuantumCircuit(4)
                                       #initialization of quantum circuit
            qc_gates(theta_c)
                                       #evaluate gate function
            backend = Aer.get_backend('statevector_simulator')
            job = execute(circ, backend)
            result = job.result()
            st_vec = result.get_statevector()
            return np.linalg.norm(st_vec-Phi)
        theta_0 = np.random.rand(8*L)*2*np.pi #initial random parameters theta
        bound = []
```

```
for i_b in range(0, 8*L):
    bound.append([0, 2*np.pi])

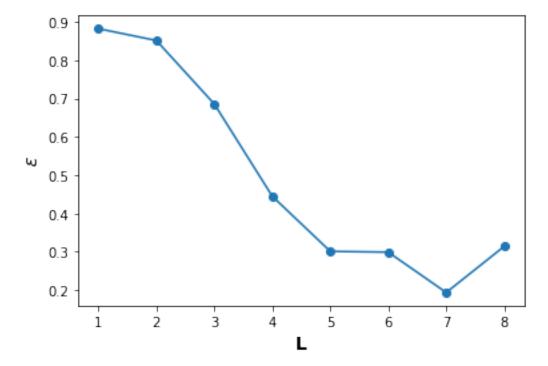
optimizer = COBYLA(maxiter=2000) #optimization

theta_opt, eps, _ = optimizer.optimize(len(theta_0), cost,
→initial_point=theta_0, variable_bounds=bound)

Epsilon.append(eps) #store result
```

Finally, we plot the resultant minimum distance ϵ for each layer L

```
[4]: plt.plot(np.arange(1,L_max+1), Epsilon, marker='o')
    plt.xlabel('L',fontsize=14, fontweight='bold')
    plt.ylabel('$\epsilon$', fontsize=14, fontweight='bold')
    plt.show()
```



As we can see, the minimum distance ϵ decreases as the number of layers L in the circuit increases until it reaches a minimum value $\epsilon \approx 0.2$ for L=7. Further increasing the number L of layers produces an increase of the resultant parameter to $\epsilon \approx 0.3$

We may explore what happens when we use different parametrized gates in our circuit. For instance, if we change the R_x gates in the odd blocks for R_y gates, we get:

```
[5]: Epsilon = []
    for L in range(1, L_max+1): #loop for each L case
        def qc_gates(theta): #function to create the circuit
            theta_d = []
            for i_d in range (0,L):
                theta_r = [theta[8*i_d:8*i_d+4],theta[8*i_d+4:8*(i_d+1)]]
                theta_d.append(theta_r)
            theta = np.asarray(theta_d)
            for i_l in range(0, L):
                for i in range (0, 4):
                    circ.ry(theta[i_1][0][i], i) #ry gates
                for i in range (0, 4):
                    circ.rz(theta[i_1][0][i], i)
                                                   #rz gates
                circ.cz(0, 1)
                                                     #cz gates
                circ.cz(0, 2)
                circ.cz(0, 3)
                circ.cz(1, 2)
                circ.cz(1, 3)
                circ.cz(2, 3)
        def cost(theta_c):
                                       #function to calculate epsilon
            global circ
            circ = QuantumCircuit(4)
                                       #initialization of quantum circuit
            qc_gates(theta_c)
                                       #evaluate gate function
            backend = Aer.get_backend('statevector_simulator')
            job = execute(circ, backend)
            result = job.result()
            st_vec = result.get_statevector()
            return np.linalg.norm(st_vec-Phi)
        theta_0 = np.random.rand(8*L)*2*np.pi #initial random parameters theta
        bound = []
```

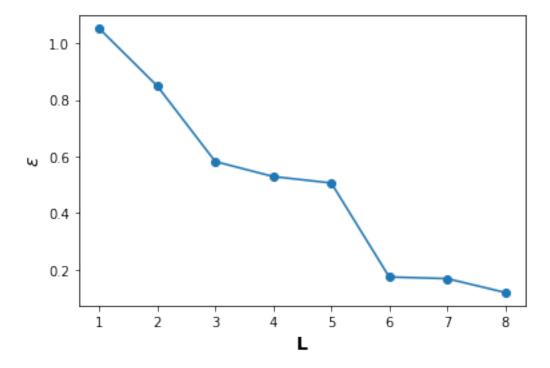
```
for i_b in range(0, 8*L):
    bound.append([0, 2*np.pi])

optimizer = COBYLA(maxiter=2000) #optimization

theta_opt, eps, _ = optimizer.optimize(len(theta_0), cost,
→initial_point=theta_0, variable_bounds=bound)

Epsilon.append(eps) #store result
```

```
[6]: plt.plot(np.arange(1,L_max+1), Epsilon, marker='o')
   plt.xlabel('L',fontsize=14, fontweight='bold')
   plt.ylabel('$\epsilon$', fontsize=14, fontweight='bold')
   plt.show()
```



We can note that the minimum ϵ in this case is lower than the optimal in the previous circuit, reaching a value of $\epsilon = 0.12$ for L = 8.

Now, replacing the R_z gates in the even blocks for R_y gates, we get:

```
[8]: Epsilon = []

for L in range(1, L_max+1): #loop for each L case
```

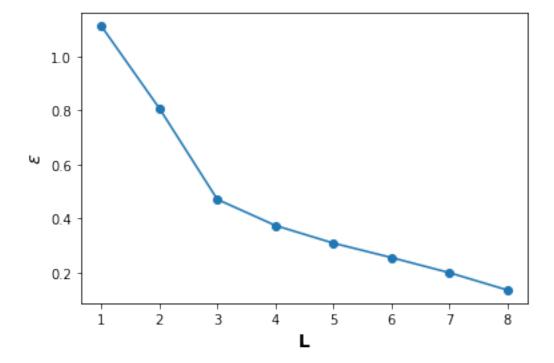
```
def qc_gates(theta): #function to create the circuit
    theta_d = []
   for i_d in range (0,L):
        theta_r = [theta[8*i_d:8*i_d+4],theta[8*i_d+4:8*(i_d+1)]]
       theta_d.append(theta_r)
   theta = np.asarray(theta_d)
   for i_l in range(0, L):
       for i in range (0, 4):
           circ.rx(theta[i_1][0][i], i)
                                          #rx gates
       for i in range (0, 4):
           circ.ry(theta[i_1][0][i], i)
                                          #ry gates
        circ.cz(0, 1)
                                            #cz gates
        circ.cz(0, 2)
        circ.cz(0, 3)
        circ.cz(1, 2)
        circ.cz(1, 3)
        circ.cz(2, 3)
def cost(theta_c):
                               #function to calculate epsilon
   global circ
                               #initialization of quantum circuit
   circ = QuantumCircuit(4)
   qc_gates(theta_c)
                              #evaluate gate function
   backend = Aer.get_backend('statevector_simulator')
   job = execute(circ, backend)
   result = job.result()
   st_vec = result.get_statevector()
   return np.linalg.norm(st_vec-Phi)
theta_0 = np.random.rand(8*L)*2*np.pi #initial random parameters theta
bound = []
for i_b in range(0, 8*L):
   bound.append([0, 2*np.pi])
```

```
optimizer = COBYLA(maxiter=2000) #optimization

theta_opt, eps, _ = optimizer.optimize(len(theta_0), cost, _ optimitial_point=theta_0, variable_bounds=bound)

Epsilon.append(eps) #store result
```

```
[9]: plt.plot(np.arange(1,L_max+1), Epsilon, marker='o')
   plt.xlabel('L',fontsize=14, fontweight='bold')
   plt.ylabel('$\epsilon$', fontsize=14, fontweight='bold')
   plt.show()
```



In this case, we can see that ϵ quickly drops to a value of 0.47 for L=3, followed by a less steep linear decrease for $L \geq 4$ achiving a minimum of $\epsilon = 0.135$ for L=8