## Task1

## September 27, 2020

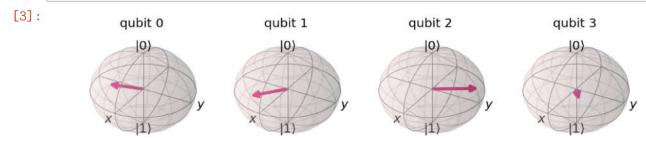
```
[1]: %matplotlib inline
    # Importing standard Qiskit libraries and configuring account
    from qiskit import QuantumCircuit, execute, Aer, IBMQ
    from qiskit.compiler import transpile, assemble
    from qiskit.tools.jupyter import *
    from qiskit.visualization import *
    # Loading your IBM Q account(s)
    provider = IBMQ.load_account()
```

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:192: UserWarning: Timestamps in IBMQ backend properties, jobs, and job results are all now in local time instead of UTC.

warnings.warn('Timestamps in IBMQ backend properties, jobs, and job results '

```
[2]: import numpy as np
from qiskit import BasicAer
from qiskit.circuit import Parameter
from qiskit_textbook.tools import random_state, array_to_latex
from qiskit.aqua.components.optimizers import COBYLA
import matplotlib.pyplot as plt
```

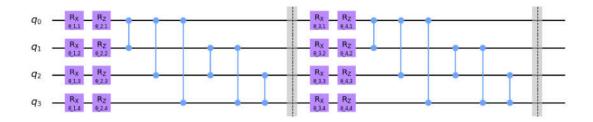
```
[3]: np.random.seed(123)
phi = random_state(4) # target
plot_bloch_multivector(phi*2) # *2 just to see things a bit better
```



```
[4]: class ParamQuantumCircuit:
         def __init__(self,n,L):
             self.blocks = 0
             self.layers = 0
             self.thetas = []
             self.circuit = QuantumCircuit(n)
             for i in range(L):
                 self.add_layer(n)
         # wrapper for drawing circuits
         def draw(self):
             return self.circuit.draw()
         # add 1 more layer to the main circuit
         def add_layer(self,n):
             i = self.blocks
             self.thetas += [Parameter(f'_{i+1}, \{j+1\}') for j in range(n)] + \setminus
                             [Parameter(f'_{i+2},{j+1}') for j in range(n)] # starts_
      \hookrightarrow from 1
             # U odd
             for j in range(n): # Rx gates
                 self.circuit.rx(self.thetas[j-2*n], j)
             # U_even
             for j in range(n): # Rz gates
                 self.circuit.rz(self.thetas[j-n], j)
             for j in range(n): # CZ gates
                 for k in range(j+1,n):
                     self.circuit.cz(j,k)
             # end of block
             self.circuit.barrier()
             self.blocks += 2
             self.layers += 1
         # run circuit
         def run(self, thetas):
             job = execute(self.circuit,
                              BasicAer.get_backend('statevector_simulator'),
                              parameter_binds = [{bind : theta for bind, theta in_
      ⇒zip(self.thetas,thetas)}])
             return job.result().get_statevector()
```

```
[5]: # example with 2 layers
qc = ParamQuantumCircuit(4,2)
qc.draw()
```

[5]:

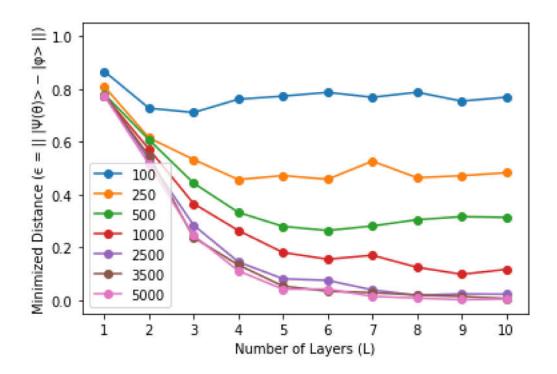


```
[6]: # = // /Ψ()> - /> //
def cost_func(thetas):
    psi = qc.run(thetas.tolist())
    return np.linalg.norm(psi-phi)
```

```
[10]: max_layers = 10
      \max iters = [100, 250, 500, 1000, 2500, 3500, 5000]
      sample_counts = [30,20,15, 5, 5, 5] # just to show average trend
      # preallocate results
      # results = { max iter: [ []
                                   for L in range(max_layers) ]
                                       for max_iter in max_iters }
      epsilons = { max_iter: [ 0 for layer in range(max_layers) ]
                                  for max_iter in max_iters }
      # optimize for various numbers of layers and iteration
      for L in range(max_layers):
          # create circuit with L+1 layers
          qc = ParamQuantumCircuit(4,L+1)
          # for every circuit try out various iterations
          for max_iter,sample_count in zip(max_iters, sample_counts):
              # Initialize the COBYLA optimizer
              optimizer = COBYLA(maxiter=max iter, tol=1e-4)
              # take more samples for showing overall average results
              for sample in range(sample count):
                  print(f'optimizing layer [{L+1}] with iteration [{max_iter}] sample_
      → [{sample+1}/{sample_count}]')
                  # optimize from random parameters
                  thetas = np.random.rand(len(qc.thetas)) * 2 * np.pi
                  result = optimizer.optimize(len(thetas), cost_func,
       →variable_bounds=[(0, 2 * np.pi)] * len(thetas),
       →initial_point=thetas)
                  epsilons[max_iter][L] += result[1]
              # average
              epsilons[max_iter][L] /= sample_count
```

```
optimizing layer [10] with iteration [500] sample [3/15]
     optimizing layer [10] with iteration [500] sample [4/15]
     optimizing layer [10] with iteration [500] sample [5/15]
     optimizing layer [10] with iteration [500] sample [6/15]
     optimizing layer [10] with iteration [500] sample [7/15]
     optimizing layer [10] with iteration [500] sample [8/15]
     optimizing layer [10] with iteration [500] sample [9/15]
     optimizing layer [10] with iteration [500] sample [10/15]
     optimizing layer [10] with iteration [500] sample [11/15]
     optimizing layer [10] with iteration [500] sample [12/15]
     optimizing layer [10] with iteration [500] sample [13/15]
     optimizing layer [10] with iteration [500] sample [14/15]
     optimizing layer [10] with iteration [500] sample [15/15]
     optimizing layer [10] with iteration [1000] sample [1/5]
     optimizing layer [10] with iteration [1000] sample [2/5]
     optimizing layer [10] with iteration [1000] sample [3/5]
     optimizing layer [10] with iteration [1000] sample [4/5]
     optimizing layer [10] with iteration [1000] sample [5/5]
     optimizing layer [10] with iteration [2500] sample [1/5]
     optimizing layer [10] with iteration [2500] sample [2/5]
     optimizing layer [10] with iteration [2500] sample [3/5]
     optimizing layer [10] with iteration [2500] sample [4/5]
     optimizing layer [10] with iteration [2500] sample [5/5]
     optimizing layer [10] with iteration [3500] sample [1/5]
     optimizing layer [10] with iteration [3500] sample [2/5]
     optimizing layer [10] with iteration [3500] sample [3/5]
     optimizing layer [10] with iteration [3500] sample [4/5]
     optimizing layer [10] with iteration [3500] sample [5/5]
     optimizing layer [10] with iteration [5000] sample [1/5]
     optimizing layer [10] with iteration [5000] sample [2/5]
     optimizing layer [10] with iteration [5000] sample [3/5]
     optimizing layer [10] with iteration [5000] sample [4/5]
     optimizing layer [10] with iteration [5000] sample [5/5]
[11]: x = np.arange(max_layers) + 1
      for max iter in max iters:
          y = np.array(epsilons[max_iter])
          plt.plot(x,y, marker='o', label=max_iter)
      plt.ylim(-0.05, 1.05)
      plt.xlim(0.5,max_layers+0.5)
      plt.xticks(x)
      plt.legend(loc='lower left')
      plt.xlabel('Number of Layers (L)')
      plt.ylabel('Minimized Distance ( = || |\Psi() > - | > ||)')
      plt.show()
```

optimizing layer [10] with iteration [500] sample [2/15]



```
[15]: # 6 layers seems good enough. Comparison with 5000 iteration
     qc = ParamQuantumCircuit(4,6)
      # optimize from random parameters
     thetas = np.random.rand(len(qc.thetas)) * 2 * np.pi
     result = COBYLA(maxiter=3500, tol=1e-4).optimize(len(thetas), cost_func,
                                                       variable bounds=[(0, 2 * np.
      →pi)] * len(thetas),
                                                       initial_point=thetas)
     print(f'Error: {result[1]*100}%')
     print(f'Parameters: {result[0]}')
     psi = qc.run(result[0])
     Error: 0.8144699128586688%
     Parameters: [ 4.37217648 5.48012572 4.60231892
                                                      5.64928367
                                                                  0.64252781
     3.66225745
       1.47285971 3.70697221 5.95741961 1.43753137
                                                      6.74144137
                                                                  6.19559652
      -0.01040421 2.65233409 3.76955369 1.82782718
                                                      1.62749458
                                                                  6.15699765
       4.59894412 0.60289469 1.74684644 5.36904271
                                                      3.63472187
                                                                  6.88126512
       0.1940274
                   4.11019317
                              4.39500115 4.29212799
                                                      3.7467245
                                                                  3.63640277
       1.82585482 0.58147699 2.20050707
                                          1.275086
                                                       4.29498194
                                                                  2.52044257
       2.40916174 6.40480019 3.79007609
                                          2.34552582
                                                      2.6356579
                                                                  3.93596913
       5.86164765 1.91420503 3.91394106 0.90867697
                                                      3.70370503 1.06699614]
[16]: plot_bloch_multivector(phi*2) # target
```

[16]:

