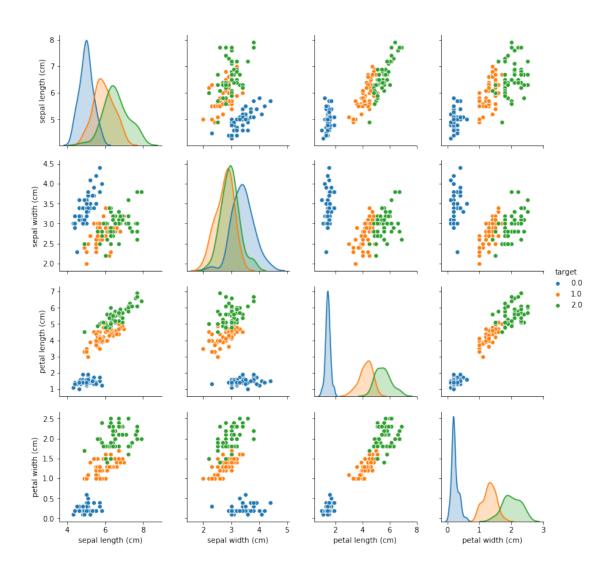
IRIS

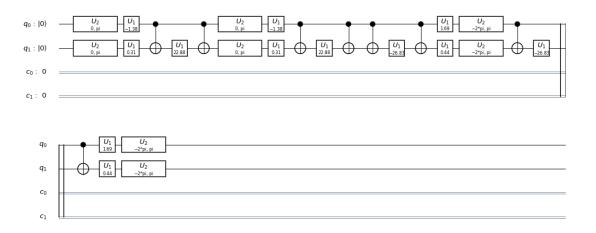
August 27, 2019

```
In [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()
In [2]: import numpy as np
        import pandas as pd
        import seaborn as sns
        import matplotlib.pyplot as plt
        from sklearn import datasets
        from sklearn.model_selection import train_test_split
        from sklearn.preprocessing import StandardScaler
        from sklearn.preprocessing import MinMaxScaler
       from sklearn.decomposition import PCA
       from qiskit import Aer
        from qiskit.aqua.utils import split_dataset_to_data_and_labels, map_label_to_class_nam-
        from qiskit.aqua import run_algorithm, QuantumInstance
        from qiskit.aqua.input import ClassificationInput
        from qiskit.aqua.components.feature_maps import SecondOrderExpansion
        from qiskit.aqua.components.multiclass_extensions.one_against_rest import OneAgainstRe
        from qiskit.aqua.algorithms.many_sample.qsvm import _QSVM_Estimator
        from qiskit.aqua.algorithms import QSVM
        from sklearn.metrics import classification_report
        from sklearn.metrics import confusion_matrix
In [3]: def Data(dataset, n, PLOT_DATA):
            class_labels = [r'Setosa', r'Versicolor', r'Virginica']
            data, target = datasets.load_iris(True)
            sample_train, sample_test, label_train, label_test = train_test_split(data, target
            # Now we standarize for gaussian around 0 with unit variance
            std_scale = StandardScaler().fit(sample_train)
            sample_train = std_scale.transform(sample_train)
            sample_test = std_scale.transform(sample_test)
```

```
# Now reduce number of features to number of qubits
            pca = PCA(n_components=n).fit(sample_train)
            sample_train = pca.transform(sample_train)
            sample_test = pca.transform(sample_test)
            # Scale to the range (-1,+1)
            samples = np.append(sample_train, sample_test, axis=0)
            minmax_scale = MinMaxScaler((-1, 1)).fit(samples)
            sample_train = minmax_scale.transform(sample_train)
            sample_test = minmax_scale.transform(sample_test)
            training_input = {key: (sample_train[label_train == k, :]) for k, key in enumerate
            test_input = {key: (sample_test[label_test == k, :]) for k, key in enumerate(class)
            if PLOT_DATA:
                for k in range(0, 3):
                    plt.scatter(sample_train[label_train == k, 0],
                                sample_train[label_train == k, 1])
                plt.title("Iris dataset")
                plt.show()
            return sample_train, training_input, test_input, class_labels
In [4]: feature_dim = 2 # dimension of each data point
        dataset=datasets.load_iris()
        sample_Total, training_input, test_input, class_labels = Data(datasets.load_iris(True)
                                                                       n=feature_dim, PLOT_DATA
        datapoints, class_to_label = split_dataset_to_data_and_labels(test_input)
        print(class_to_label)
{'Setosa': 0, 'Versicolor': 1, 'Virginica': 2}
In [5]: df = pd.DataFrame( np.c_[dataset['data'], dataset['target']],
                                 columns = np.append(dataset['feature_names'], ['target']))
        sns.pairplot(df, hue = 'target',
                     vars = ['sepal length (cm)', 'sepal width (cm)',
                             'petal length (cm)', 'petal width (cm)'])
Out[5]: <seaborn.axisgrid.PairGrid at 0x7f41d0ae5828>
```



Out[7]:



In [8]: %time result = run_algorithm(params, algo_input, backend=backend)

Entry point 'HartreeFock = qiskit.chemistry.aqua_extensions.components.initial_states:HartreeFock

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar
DeprecationWarning)
/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:547: DeprecationWar
DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar
DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar
DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:547: DeprecationWar
DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar DeprecationWarning)

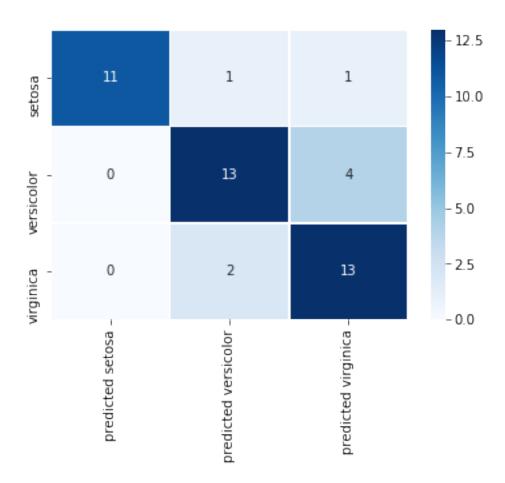
/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWar
DeprecationWarning)

/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWa

```
DeprecationWarning)
/opt/conda/lib/python3.7/site-packages/qiskit/providers/ibmq/ibmqfactory.py:594: DeprecationWa
  DeprecationWarning)
CPU times: user 9min 28s, sys: 1min 5s, total: 10min 34s
Wall time: 24min 12s
In [9]: print("ground truth:
                              {}".format(datapoints[1]))
       print("prediction:
                              {}".format(result['predicted_labels']))
       print("predicted class: {}".format(result['predicted_classes']))
       print("accuracy:
                              {}".format(result['testing_accuracy']))
                ground truth:
[0\ 0\ 1\ 2\ 0\ 0\ 0\ 0\ 0\ 0\ 0\ 1\ 1\ 1\ 2\ 1\ 1\ 2\ 1\ 1\ 1\ 2\ 1\ 1\ 2\ 1\ 1\ 2\ 2\ 2\ 2\ 2\ 2\ 2
prediction:
 2 1 2 2 2 1 2 2]
predicted class: ['Setosa', 'Setosa', 'Versicolor', 'Virginica', 'Setosa', 'Setosa', 'Setosa',
                0.8222222222222
accuracy:
In [10]: print(classification_report(datapoints[1], result['predicted_labels'],
                                   target_names=dataset.target_names))
             precision
                          recall f1-score
                                            support
     setosa
                  1.00
                            0.85
                                     0.92
                                                 13
                  0.81
                            0.76
                                     0.79
 versicolor
                                                 17
  virginica
                  0.72
                            0.87
                                     0.79
                                                 15
                                     0.82
                                                 45
   accuracy
  macro avg
                  0.84
                            0.83
                                     0.83
                                                 45
weighted avg
                  0.84
                            0.82
                                     0.83
                                                 45
In [11]: cm = np.array(confusion_matrix(datapoints[1], result['predicted_labels']))
        confusion = pd.DataFrame(cm, index=dataset.target_names,
                                columns= ['predicted ' + s for s in dataset.target_names])
        sns.heatmap(confusion, annot=True, fmt="d", linewidths=.5, cmap="Blues")
        plt.title("Heatmap\n", fontdict = {'fontsize': 18})
```

plt.show()

Heatmap



In []: