## **Muhammad Arham Adeel**

## **Support Vector Machine**

Practice and task given in the video.

```
In [8]:
          ## import sckit learn data set library
          from sklearn import datasets
          #Load Data set
          cancer = datasets.load breast cancer()
 In [9]:
          ## Print the name of the features
          print("Features: ",cancer.feature names)
          #print the label type of cancer (malignant, benign)
          print("Labels : ", cancer.target_names)
         Features: ['mean radius' 'mean texture' 'mean perimeter' 'mean area'
          'mean smoothness' 'mean compactness' 'mean concavity'
          'mean concave points' 'mean symmetry' 'mean fractal dimension'
          'radius error' 'texture error' 'perimeter error' 'area error'
          'smoothness error' 'compactness error' 'concavity error'
          'concave points error' 'symmetry error' 'fractal dimension error'
          'worst radius' 'worst texture' 'worst perimeter' 'worst area'
          'worst smoothness' 'worst compactness' 'worst concavity'
          'worst concave points' 'worst symmetry' 'worst fractal dimension']
         Labels : ['malignant' 'benign']
In [10]:
          #print data feature shape
          cancer.data.shape
         (569, 30)
Out[10]:
In [11]:
          # print the cancer data features (top 5 records)
          print(cancer.data[0:5])
         [[1.799e+01 1.038e+01 1.228e+02 1.001e+03 1.184e-01 2.776e-01 3.001e-01
           1.471e-01 2.419e-01 7.871e-02 1.095e+00 9.053e-01 8.589e+00 1.534e+02
           6.399e-03 4.904e-02 5.373e-02 1.587e-02 3.003e-02 6.193e-03 2.538e+01
           1.733e+01 1.846e+02 2.019e+03 1.622e-01 6.656e-01 7.119e-01 2.654e-01
           4.601e-01 1.189e-01]
          [2.057e+01 1.777e+01 1.329e+02 1.326e+03 8.474e-02 7.864e-02 8.690e-02
           7.017e-02 1.812e-01 5.667e-02 5.435e-01 7.339e-01 3.398e+00 7.408e+01
           5.225e-03 1.308e-02 1.860e-02 1.340e-02 1.389e-02 3.532e-03 2.499e+01
           2.341e+01 1.588e+02 1.956e+03 1.238e-01 1.866e-01 2.416e-01 1.860e-01
           2.750e-01 8.902e-02]
          [1.969e+01 2.125e+01 1.300e+02 1.203e+03 1.096e-01 1.599e-01 1.974e-01
           1.279e-01 2.069e-01 5.999e-02 7.456e-01 7.869e-01 4.585e+00 9.403e+01
```

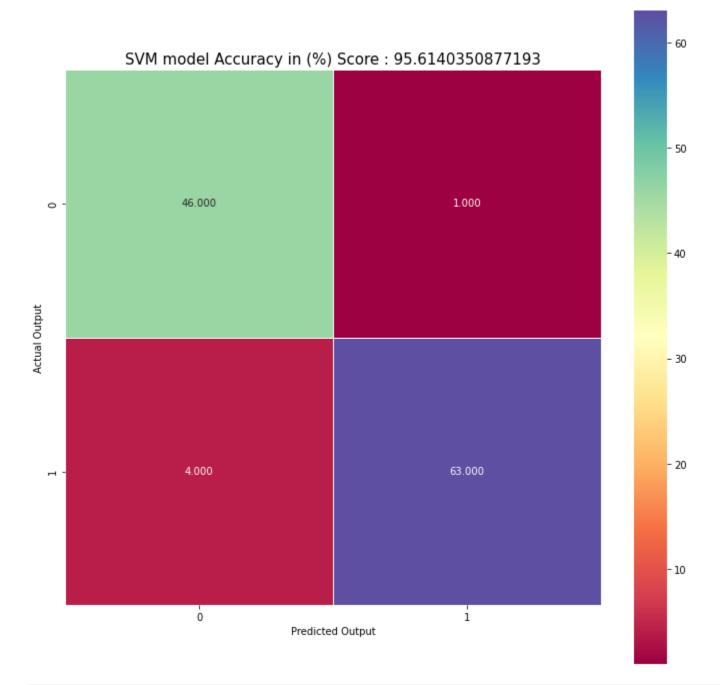
```
2.553e+01 1.525e+02 1.709e+03 1.444e-01 4.245e-01 4.504e-01 2.430e-01
       3.613e-01 8.758e-02]
      [1.142e+01 2.038e+01 7.758e+01 3.861e+02 1.425e-01 2.839e-01 2.414e-01
       1.052e-01 2.597e-01 9.744e-02 4.956e-01 1.156e+00 3.445e+00 2.723e+01
       9.110e-03 7.458e-02 5.661e-02 1.867e-02 5.963e-02 9.208e-03 1.491e+01
       2.650e+01 9.887e+01 5.677e+02 2.098e-01 8.663e-01 6.869e-01 2.575e-01
       6.638e-01 1.730e-01]
      [2.029e+01 1.434e+01 1.351e+02 1.297e+03 1.003e-01 1.328e-01 1.980e-01
       1.043e-01 1.809e-01 5.883e-02 7.572e-01 7.813e-01 5.438e+00 9.444e+01
       1.149e-02 2.461e-02 5.688e-02 1.885e-02 1.756e-02 5.115e-03 2.254e+01
       1.667e+01 1.522e+02 1.575e+03 1.374e-01 2.050e-01 4.000e-01 1.625e-01
       2.364e-01 7.678e-02]]
In [12]:
      # print the cancer labels (0 : malignant , 1 : benign)
      print(cancer.target)
      1\ 1\ 1\ 1\ 1\ 1\ 1\ 1\ 0\ 1\ 1\ 1\ 0\ 0\ 1\ 1\ 0\ 0\ 1\ 1\ 1\ 1\ 0\ 0\ 1\ 0
      1 0 1 1 1 0 1 1 0 0 1 0 0 0 0 1 0 0 0 1 0 1 0 1 1 0 1 0 0 0 0 1 1 0 0 1 1
      1 1 1 1 1 1 1 0 0 0 0 0 0 1
In [13]:
      from sklearn.model selection import train test split
      X_train,X_test,y_train,y_test = train_test_split(cancer.data,cancer.target, test_size=0.2, random_
In [17]:
      # import svm library
      from sklearn import svm
      #Create a SVM classifier
      clf = svm.SVC(kernel="linear").fit(X train,y train)
      # making prediction on data set
      y_pred = clf.predict(X_test)
      y pred
      array([0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 0, 0, 0,
Out[17]:
          1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1,
          0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 1, 0, 1,
          0, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1,
          0, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 0, 1, 0, 0,
          1, 0, 0, 1])
In [19]:
      from sklearn import metrics
      score = metrics.accuracy_score(y_test,y_pred)
```

6.150e-03 4.006e-02 3.832e-02 2.058e-02 2.250e-02 4.571e-03 2.357e+01

```
# Model Accuracy: how often is the classifier correct?
          print("Accuracy: ",metrics.accuracy_score(y_test,y_pred))
         Accuracy: 0.956140350877193
In [21]:
          #Model Precision: What percentage of positive tuoles are labeled as such?
          print("Precision: ",metrics.precision_score(y_test,y_pred))
          #Model Recall: What percentage of positive tuoles are labeled as such?
          print("Recall: ",metrics.recall_score(y_test,y_pred))
         Precision: 0.984375
         Recall: 0.9402985074626866
In [22]:
          from sklearn import metrics
          cm = metrics.confusion_matrix(y_test,y_pred)
         array([[46, 1],
Out[22]:
                [ 4, 63]], dtype=int64)
In [23]:
          import seaborn as sns
          import matplotlib.pyplot as plt
          plt.figure(figsize=(12,12))
          sns.heatmap(cm, annot = True, fmt=".3f",linewidths=".5",square=True,cmap = "Spectral")
          plt.xlabel("Predicted Output")
          plt.ylabel("Actual Output")
          all_sample_title = "SVM model Accuracy in (%) Score : {0}".format(score*100)
          plt.title(all_sample_title,size=15)
```

Text(0.5, 1.0, 'SVM model Accuracy in (%) Score : 95.6140350877193')

Out[23]:



In [ ]: