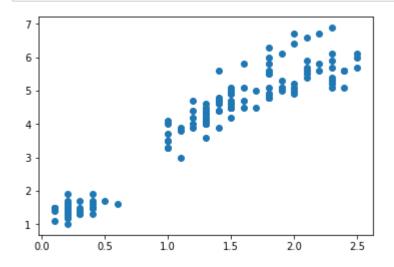
## **Chapter 5: Demo KNN Regression**

```
In [0]: # from google.colab import drive
         # drive.mount("/content/gdrive", force_remount=True)
In [0]: # %cd '/content/qdrive/My Drive/LDS6 MachineLearning/practice/Chapter5 KNN/'
In [0]:
        import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         from sklearn.model selection import train test split
In [0]: # import some data to play with
         iris = pd.read excel("Iris.xls")
         iris.info()
         <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 150 entries, 0 to 149
        Data columns (total 5 columns):
        sepallength
                        150 non-null float64
                        150 non-null float64
        sepalwidth
                        150 non-null float64
        petallength
        petalwidth
                        150 non-null float64
                        150 non-null object
        iris
        dtypes: float64(4), object(1)
        memory usage: 5.9+ KB
In [0]: | iris.head()
Out[3]:
            sepallength sepalwidth petallength petalwidth
                                                           iris
         0
                   5.1
                             3.5
                                        1.4
                                                  0.2 Iris-setosa
                   4.9
                             3.0
                                        1.4
                                                  0.2 Iris-setosa
                   4.7
                             3.2
                                        1.3
                                                  0.2 Iris-setosa
                   4.6
                             3.1
                                        1.5
                                                  0.2 Iris-setosa
                                                  0.2 Iris-setosa
                   5.0
                             3.6
                                        1.4
In [0]:
        # Cho pentalwidth dư đoán pentallenath
         petalwidth = iris[['petalwidth']] # x hay input
         pentallength = iris['petallength'] # y hay output
```

## In [0]: plt.scatter(petalwidth, pentallength) plt.show()



Type *Markdown* and LaTeX:  $\alpha^2$ 

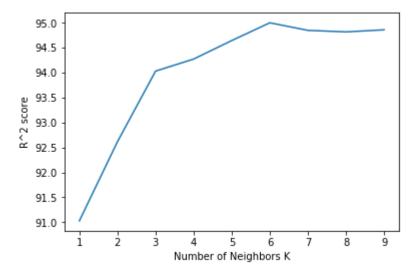
In [0]: from sklearn.neighbors import KNeighborsRegressor

```
In [0]: # neu nhu co mot so k co score = nhau => tim mse cua cac k bang nhau =>
# mse << => chon k
list_k = []
list_r_score = []
for K_value in range(2,int(y_train.shape[0]**0.5)):
    list_k.append(K_value)
    neigh = KNeighborsRegressor(n_neighbors = K_value)
    neigh.fit(X_train, y_train)
    score = neigh.score(X_test,y_test)*100
    list_r_score.append(score)
    print("The R^2 score is ", score, " for K-Value:",K_value)
vi_tri = list_r_score.index(max(list_r_score))
k = list_k[vi_tri]
print("\nThe optimal number of neighbors is:", k, "with R^2 score is:",
    list_r_score[vi_tri])
```

```
The R^2 score is 91.02944417134236 for K-Value: 1
The R^2 score is 92.61607309341787 for K-Value: 2
The R^2 score is 94.03093022335983 for K-Value: 3
The R^2 score is 94.27262556574509 for K-Value: 4
The R^2 score is 94.6437034902446 for K-Value: 5
The R^2 score is 94.99968922866555 for K-Value: 6
The R^2 score is 94.84781487391967 for K-Value: 7
The R^2 score is 94.8173935890699 for K-Value: 8
The R^2 score is 94.85852370431896 for K-Value: 9
```

The optimal number of neighbors is: 6 with R^2 score is: 94.99968922866555

```
In [0]: plt.plot(list_k, list_r_score)
    plt.xlabel('Number of Neighbors K')
    plt.ylabel('R^2 score')
    plt.show()
```



```
In [0]: # seleck k => model
         knn = KNeighborsRegressor(n_neighbors=6)
         knn.fit(X train, y train)
Out[11]: KNeighborsRegressor(algorithm='auto', leaf_size=30, metric='minkowski',
                   metric params=None, n jobs=None, n neighbors=6, p=2,
                   weights='uniform')
         y pred = knn.predict(X test)
 In [0]:
         y pred
Out[12]: array([3.9]
                          , 1.36666667, 5.71666667, 4.63333333, 4.6
                           , 4.16666667, 5.71666667, 4.63333333, 3.9
                                      , 1.38333333, 1.4
                5.53333333, 1.4
                                                           , 1.36666667,
                                      , 3.91666667, 4.16666667, 5.85
                       , 5.85
                4.85
                1.38333333, 5.36666667, 1.6 , 5.66666667, 5.53333333,
                5.71666667, 5.36666667, 5.71666667, 1.36666667, 1.38333333])
 In [0]: print("The Training R^2 score is: ", knn.score(X_train,y_train))
         print("The Testing R^2 score is: ", knn.score(X_test,y_test))
         The Training R^2 score is: 0.9581011061128362
         The Testing R^2 score is: 0.9499968922866555
 In [0]:
         from sklearn import metrics
         print('Mean Squared Error:', metrics.mean_squared_error(y_test, y_pred))
         Mean Squared Error: 0.16387962962962974
 In [0]: | df = pd.DataFrame({'Actual': pd.DataFrame(y_test.values)[0].values,
                             'Prediction': pd.DataFrame(y_pred)[0].values})
         df.head()
Out[15]:
            Actual Prediction
               4.7
                    3.900000
          0
               1.7
          1
                    1.366667
                    5.716667
          2
               6.9
          3
               4.5
                    4.633333
               4.8
                    4.600000
 In [0]: x_now = [[0.25]]
         y_now = knn.predict(x_now)
         y_now
Out[16]: array([1.38333333])
```