KNN Alogrithm Classification Example

```
#Objective is to build classifier and predict class of unknown classes using K-nearest n
import itertools
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import matplotlib.ticker as ticker
from matplotlib.ticker import NullFormatter
from sklearn import preprocessing
from matplotlib import rc,font_manager
ticks_font = font_manager.FontProperties(family='Times New Roman', style='normal',
    size=12, weight='normal', stretch='normal')
ax=plt.gca()
## Loading Data ##
df=pd.read_csv('D:\Python\edx\Machine Learning\Classification\elecom_customer_data.csv')
with open('KNN.txt','a') as f:
    print(df.head(),file=f)
    print(df.describe(),file=f)
    print('Classes are: ', df['custcat'].value_counts(),file=f)
#1- Basic Service 2- E-Service 3- Plus Service 4- Total Service#
df.hist(column='income',bins=50)
##Feature Set X,Y##
print(df.columns)
X=df[['region', 'tenure', 'age', 'marital', 'address', 'income', 'ed',
       'employ', 'retire', 'gender', 'reside',]].values
X[0:5]
y=df['custcat'].values
y [0:5]
# Normalize Data #
X=preprocessing.StandardScaler().fit(X).transform(X.astype(float))
X[0:5]
```

```
# Classification - K nearest neighbour#
from sklearn.neighbors import KNeighborsClassifier
k=9 # Train model and predict with k=9 (best value for high accuracy, check KNN_k_values
neigh=KNeighborsClassifier(n_neighbors=k).fit(X_train,y_train)
yhat=neigh.predict(X_test)
with open('KNN.txt', 'a') as f:
    print(yhat[0:5],file=f)
#Predict using accuracy_score which is similar to jaccard_similarity_score function#
from sklearn import metrics
with open('KNN.txt', 'a') as f:
    print('Train set accuracy: ', metrics.accuracy_score(y_train,neigh.predict(X_train))
    print('Test set accuracy: ',metrics.accuracy_score(y_test,yhat),file=f)
plt.show()
Solution:
   region tenure
                    age marital
                                  address
                                            income
                                                    ed
                                                        employ
                                                                retire
                                                                         gender
custcat
0
        2
               13
                     44
                               1
                                        9
                                              64.0
                                                     4
                                                             5
                                                                    0.0
                                                                              0
2
         1
                               1
                                        7
                                                             5
1
        3
               11
                     33
                                             136.0
                                                     5
                                                                    0.0
                                                                              0
         4
6
2
        3
               68
                     52
                               1
                                       24
                                             116.0
                                                            29
                                                                    0.0
2
         3
3
        2
               33
                     33
                               0
                                       12
                                              33.0
                                                                    0.0
1
         1
        2
               23
                     30
                                              30.0
                                                                    0.0
4
                                                     1
         3
4
          region
                        tenure
                                                  marital
                                                                address
                                                                              income
                                        age
                                                  reside
ed
         employ
                       retire
                                    gender
                                                               custcat
       1000.0000 1000.00000 1000.000000 1000.000000 1000.000000 1000.000000
1000.000000 1000.000000 1000.000000 1000.000000 1000.000000 1000.000000
```

X_train, X_test, y_train, y_test=train_test_split(X,y,test_size=0.2,random_state=4)

Train Test Split#

with open('KNN.txt', 'a') as f:

from sklearn.model_selection import train_test_split

print('Train set: ', X_train.shape,y_train.shape,file=f)
print('Test set: ', X_test.shape,y_test.shape,file=f)

```
2.0220
                     35.526000
                                   41.684000
                                                 0.495000
mean
                                                              11.551000
                                                                            77.535000
2.671000
            10.987000
                           0.047000
                                         0.517000
                                                       2.331000
                                                                     2.487000
std
          0.8162
                     21.359812
                                   12.558816
                                                 0.500225
                                                              10.086681
                                                                           107.044165
1.222397
            10.082087
                           0.211745
                                         0.499961
                                                       1.435793
                                                                     1.120306
min
          1.0000
                      1.000000
                                   18.000000
                                                 0.000000
                                                               0.000000
                                                                             9.000000
1.000000
              0.000000
                           0.000000
                                         0.000000
                                                       1.000000
                                                                     1.000000
25%
          1.0000
                     17.000000
                                   32.000000
                                                 0.00000
                                                               3.000000
                                                                            29.000000
2.000000
             3.000000
                           0.000000
                                         0.00000
                                                       1.000000
                                                                     1.000000
50%
          2.0000
                     34.000000
                                   40.000000
                                                 0.000000
                                                               9.000000
                                                                            47.000000
3.000000
              8.000000
                           0.000000
                                         1.000000
                                                       2.000000
                                                                     3.000000
          3.0000
                     54.000000
                                   51.000000
                                                 1.000000
                                                              18.000000
                                                                            83.000000
75%
            17.000000
4.000000
                           0.000000
                                         1.000000
                                                       3.000000
                                                                     3.000000
          3.0000
                     72.000000
                                  77.000000
                                                 1.000000
                                                              55.000000 1668.000000
max
5.000000
            47.000000
                           1.000000
                                         1.000000
                                                       8.000000
                                                                     4.000000
Classes are: 3
                    281
     266
1
     236
4
     217
Name: custcat, dtype: int64
Train set: (800, 11) (800,)
Test set: (200, 11) (200,)
[3 1 3 2 4]
Train set accuracy: 0.5025
Test set accuracy: 0.34
```

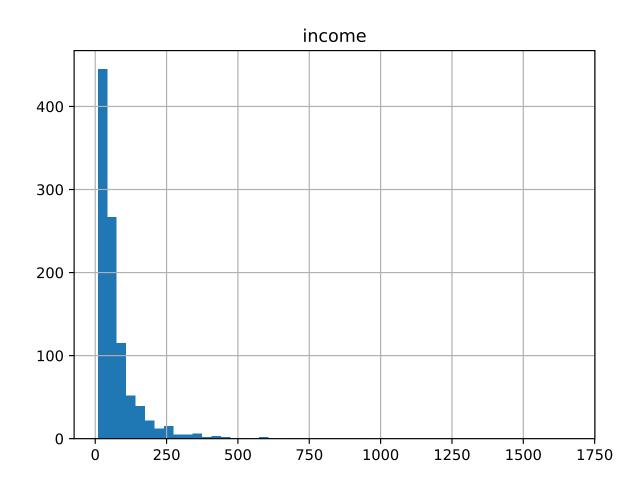
KNN Alogrithm k Value program

Objective is to find best value of K in KNN classifier to achieve maximum accuracy

```
import itertools
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import matplotlib.ticker as ticker

from matplotlib.ticker import NullFormatter
from sklearn import preprocessing
from matplotlib import rc,font_manager

ticks_font = font_manager.FontProperties(family='Times New Roman', style='normal', size=12, weight='normal', stretch='normal')
```



```
ax=plt.gca()
## Loading Data ##
\tt df=pd.read\_csv('D:\Python\edx\Machine\ Learning\Classification\elecom\_customer\_data.csv')
#1- Basic Service 2- E-Service 3- Plus Service 4- Total Service#
##Feature Set X,Y##
print(df.columns)
X=df[['region', 'tenure', 'age', 'marital', 'address', 'income', 'ed',
       'employ', 'retire', 'gender', 'reside',]].values
X[0:5]
y=df['custcat'].values
y [0:5]
# Normalize Data #
X=preprocessing.StandardScaler().fit(X).transform(X.astype(float))
X[0:5]
# Train Test Split#
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test=train_test_split(X, y, test_size=0.2, random_state=4)
#Finding the value of k to achieve maximum accuracy via Confusion matrix and then use th
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
Ks = 10
acc_mean=np.zeros((Ks-1))
acc_std=np.zeros((Ks-1))
Cmat = []
for n in range(1,Ks):
    neigh=KNeighborsClassifier(n_neighbors=n).fit(X_train,y_train)
    yhat=neigh.predict(X_test)
    acc_mean[n-1] = metrics.accuracy_score(y_test,yhat)
    acc_std[n-1]=np.std(yhat==y_test)/np.sqrt(yhat.shape[0])
with open('KNN_kvalue.txt', 'a') as f:
    print(acc_mean,file=f)
    print('The best accuracy was with: ', acc_mean.max(), 'with k= ',acc_mean.argmax()+1
#Plot of k values vs accuracy #
plt.figure()
plt.plot(range(1,Ks),acc_mean,'g')
plt.fill_between(range(1,Ks),acc_mean - 1 * acc_std,acc_mean + 1 * acc_std, alpha=0.10)
plt.legend(('Accuracy ', '+/- 3xstd'))
plt.ylabel('Accuracy ',fontname='Times New Roman',fontsize=12)
```

```
plt.xlabel('Number of Nabors (K)',fontname='Times New Roman',fontsize=12)
plt.tight_layout()
plt.show()
```

Solution:

[0.3 0.29 0.315 0.32 0.315 0.31 0.335 0.325 0.34] The best accuracy was with: 0.34 with $k=\ 9$

