```
In [1]:
import pandas as pd
import numpy as np
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
from sklearn.neighbors import KNeighborsClassifier
                                                                                                                     In [2]:
glass = pd.read csv('C:/Users/acer/Sandesh Pal/Data Science Assgn/KNN/glass.csv')
                                                                                                                     In [3]:
glass.head()
                                                                                                                    Out[3]:
        RI
                              Si
                                   Κ
                                      Ca Ba Fe Type
             Na
                  Mg
                        Αl
0 1.52101 13.64 4.49 1.10 71.78 0.06 8.75 0.0 0.0
                                                     1
1 1.51761 13.89 3.60 1.36 72.73 0.48
                                     783 00 00
                                                     1
2 1.51618 13.53 3.55 1.54 72.99 0.39
                                     7.78 0.0 0.0
                                                     1
3 1.51766 13.21 3.69 1.29 72.61 0.57 8.22 0.0 0.0
                                                     1
4 1.51742 13.27 3.62 1.24 73.08 0.55 8.07 0.0 0.0
                                                     1
                                                                                                                     In [4]:
 # checking for null values
glass.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 214 entries, 0 to 213
Data columns (total 10 columns):
    Column Non-Null Count Dtype
              -----
---
 0
     RI
              214 non-null
                                 float64
 1
     Na
              214 non-null
                                 float64
 2
              214 non-null
                                 float64
     Mg
 3
              214 non-null
     Al
                                 float64
 4
     Si
              214 non-null
                                 float64
 5
     K
              214 non-null
                                 float64
              214 non-null
 6
     Ca
                                 float64
 7
     Ва
              214 non-null
                                 float64
 8
     Fe
               214 non-null
                                 float64
 9
     Type
               214 non-null
                                 int64
dtypes: float64(9), int64(1)
memory usage: 16.8 KB
                                                                                                                     In [5]:
glass.describe()
                                                                                                                    Out[5]:
              RI
                        Na
                                  Mg
                                              Αl
                                                         Si
                                                                    Κ
                                                                              Ca
                                                                                        Ba
                                                                                                   Fe
                                                                                                            Type
count 214.000000 214.000000 214.000000 214.000000 214.000000 214.000000 214.000000 214.000000 214.000000 214.000000
 mean
        1.518365
                  13.407850
                              2.684533
                                         1.444907
                                                  72.650935
                                                              0.497056
                                                                         8.956963
                                                                                   0.175047
                                                                                              0.057009
                                                                                                         2.780374
        0.003037
                   0.816604
                              1.442408
                                         0.499270
                                                   0.774546
                                                              0.652192
                                                                        1.423153
                                                                                   0.497219
                                                                                              0.097439
                                                                                                         2.103739
  std
  min
        1.511150
                  10.730000
                              0.000000
                                         0.290000
                                                  69.810000
                                                              0.000000
                                                                         5.430000
                                                                                   0.000000
                                                                                              0.000000
                                                                                                         1.000000
        1.516523
  25%
                  12.907500
                              2.115000
                                         1.190000
                                                  72.280000
                                                              0.122500
                                                                         8.240000
                                                                                   0.000000
                                                                                              0.000000
                                                                                                         1.000000
                                                                         8.600000
  50%
        1.517680
                  13.300000
                              3.480000
                                         1.360000
                                                  72.790000
                                                              0.555000
                                                                                   0.000000
                                                                                              0.000000
                                                                                                         2.000000
        1.519157
                                                              0.610000
                                                                                                         3.000000
  75%
                  13.825000
                              3.600000
                                         1.630000
                                                  73.087500
                                                                         9.172500
                                                                                   0.000000
                                                                                              0.100000
        1.533930
                  17.380000
                              4 490000
                                                  75 410000
                                                              6.210000
                                                                                   3 150000
                                                                                              0.510000
                                                                                                         7 000000
  max
                                         3.500000
                                                                        16.190000
```

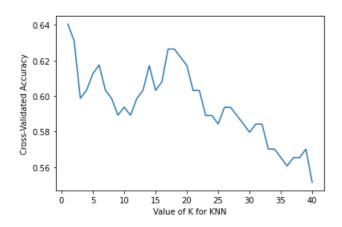
X = glass.drop(['Type'], axis =1)

X.head()

In [6]:

```
Out[6]:
                Mg
                                   Ca Ba
0 1.52101 13.64 4.49 1.10 71.78 0.06 8.75 0.0 0.0
1 1.51761 13.89 3.60 1.36 72.73 0.48 7.83 0.0 0.0
2 1.51618 13.53 3.55 1.54 72.99 0.39 7.78 0.0 0.0
3 1.51766 13.21 3.69 1.29 72.61 0.57 8.22 0.0 0.0
4 1.51742 13.27 3.62 1.24 73.08 0.55 8.07 0.0 0.0
                                                                                                           In [8]:
Y = glass['Type']
Y.head()
                                                                                                          Out[8]:
0
     1
2
     1
3
     1
Name: Type, dtype: int64
                                                                                                           In [9]:
# Let's use Grid search CV to find out best value for K
                                                                                                          In [10]:
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model selection import GridSearchCV
                                                                                                         In [11]:
n neighbors = np.array(range(1,40))
param_grid = dict(n_neighbors=n_neighbors)
                                                                                                         In [12]:
model = KNeighborsClassifier()
grid = GridSearchCV(estimator=model, param grid=param grid)
grid.fit(X, Y)
                                                                                                         Out[12]:
GridSearchCV(estimator=KNeighborsClassifier(),
             param grid={'n neighbors': array([ 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15
 16. 17.
       18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34,
       35, 36, 37, 38, 39])})
                                                                                                            P
                                                                                                          In [13]:
print(grid.best_score_)
print(grid.best_params_)
0.640531561461794
{ 'n_neighbors': 1}
                                                                                                         In [14]:
# Visualizing the CV results
                                                                                                         In [15]:
import matplotlib.pyplot as plt
%matplotlib inline
# choose k between 1 to 41
k range = range(1, 41)
k_scores = []
\# use iteration to caclulator different k in models, then return the average accuracy based on the cross
for k in k range:
 knn = KNeighborsClassifier(n_neighbors=k)
 scores = cross val score(knn, X, Y, cv=5)
 k_scores.append(scores.mean())
# plot to see clearly
plt.plot(k range, k scores)
plt.xlabel('Value of K for KNN')
plt.ylabel('Cross-Validated Accuracy')
```

plt.show()



#KNN Classification
num\_folds = 30
kfold = KFold(n\_splits=30)
model = KNeighborsClassifier(n\_neighbors=1)
results = cross\_val\_score(model, X, Y, cv=kfold)
print(results.mean())
0.6642857142857143

In [16]:

In [ ]: