

```

import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
%matplotlib inline

df =
pd.read_csv(r"C:\Users\GLAU\Desktop\K-NEarest-Neighbor-master/Classifi
ed Data",index_col=0)
df.head()

```

	WTT	PTI	EQW	SBI	LQE	QWG
FDJ \						
0	0.913917	1.162073	0.567946	0.755464	0.780862	0.352608
0.759697						
1	0.635632	1.003722	0.535342	0.825645	0.924109	0.648450
0.675334						
2	0.721360	1.201493	0.921990	0.855595	1.526629	0.720781
1.626351						
3	1.234204	1.386726	0.653046	0.825624	1.142504	0.875128
1.409708						
4	1.279491	0.949750	0.627280	0.668976	1.232537	0.703727
1.115596						

	PJF	HQE	NXJ	TARGET CLASS
0	0.643798	0.879422	1.231409	1
1	1.013546	0.621552	1.492702	0
2	1.154483	0.957877	1.285597	0
3	1.380003	1.522692	1.153093	1
4	0.646691	1.463812	1.419167	1

```

from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
scaler.fit(df.drop('TARGET CLASS',axis=1))

```

```
StandardScaler()
```

```
scaled_features = scaler.transform(df.drop('TARGET CLASS',axis=1))
```

```
df_feat = pd.DataFrame(scaled_features,columns=df.columns[:-1])
df_feat.head()
```

	WTT	PTI	EQW	SBI	LQE	QWG
FDJ \						
0	-0.123542	0.185907	-0.913431	0.319629	-1.033637	-2.308375
0.798951						
1	-1.084836	-0.430348	-1.025313	0.625388	-0.444847	-1.152706
1.129797						
2	-0.788702	0.339318	0.301511	0.755873	2.031693	-0.870156
2.599818						
3	0.982841	1.060193	-0.621399	0.625299	0.452820	-0.267220
1.750208						

```
4  1.139275 -0.640392 -0.709819 -0.057175  0.822886 -0.936773
0.596782
```

```
      PJF      HQE      NXJ
0 -1.482368 -0.949719 -0.643314
1 -0.202240 -1.828051  0.636759
2  0.285707 -0.682494 -0.377850
3  1.066491  1.241325 -1.026987
4 -1.472352  1.040772  0.276510
```

```
## Train Test split
```

```
from sklearn.model_selection import train_test_split
```

```
X_train, X_test, y_train, y_test =
train_test_split(scaled_features,df['TARGET CLASS'],
                  test_size=0.30)
```

```
## USING KNN
```

```
## Remember that we are trying to come up with a model to predict
whether someone will TARGET CLASS or not. We'll start with k=1.
```

```
from sklearn.neighbors import KNeighborsClassifier
```

```
knn = KNeighborsClassifier(n_neighbors=1)
```

```
knn.fit(X_train,y_train)
```

```
KNeighborsClassifier(n_neighbors=1)
```

```
pred = knn.predict(X_test)
```

```
## Prediction and Evaluation
```

```
from sklearn.metrics import classification_report,confusion_matrix
```

```
print(confusion_matrix(y_test,pred))
```

```
[[139  17]
 [ 14 130]]
```

```
print(classification_report(y_test,pred))
```

	precision	recall	f1-score	support
0	0.91	0.89	0.90	156
1	0.88	0.90	0.89	144
accuracy			0.90	300
macro avg	0.90	0.90	0.90	300
weighted avg	0.90	0.90	0.90	300

```
## choosing a K value
```

```
error_rate = []
```

```

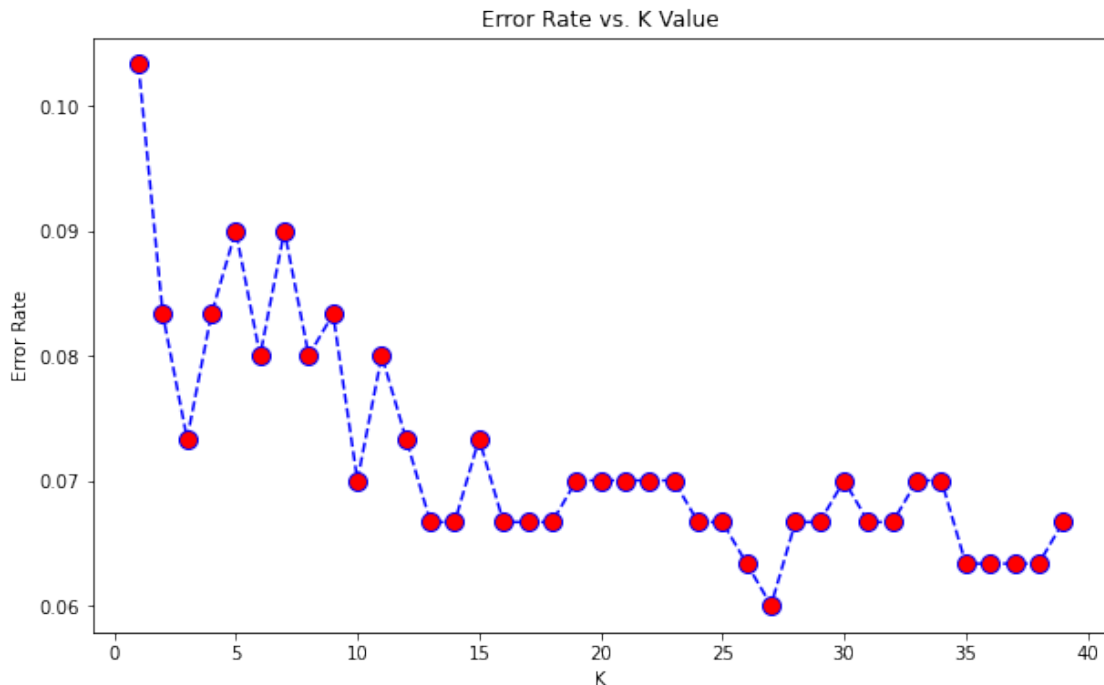
# Will take some time
for i in range(1,40):

    knn = KNeighborsClassifier(n_neighbors=i)
    knn.fit(X_train,y_train)
    pred_i = knn.predict(X_test)
    error_rate.append(np.mean(pred_i != y_test))

plt.figure(figsize=(10,6))
plt.plot(range(1,40),error_rate,color='blue', linestyle='dashed',
marker='o',
        markerfacecolor='red', markersize=10)
plt.title('Error Rate vs. K Value')
plt.xlabel('K')
plt.ylabel('Error Rate')

Text(0, 0.5, 'Error Rate')

```



```

# FIRST A QUICK COMPARISON TO OUR ORIGINAL K=1
knn = KNeighborsClassifier(n_neighbors=1)

knn.fit(X_train,y_train)
pred = knn.predict(X_test)

print('WITH K=1')
print('\n')
print(confusion_matrix(y_test,pred))

```

```
print('\n')
print(classification_report(y_test,pred))
```

WITH K=1

```
[[139  17]
 [ 14 130]]
```

	precision	recall	f1-score	support
0	0.91	0.89	0.90	156
1	0.88	0.90	0.89	144
accuracy			0.90	300
macro avg	0.90	0.90	0.90	300
weighted avg	0.90	0.90	0.90	300

NOW WITH K=23

```
knn = KNeighborsClassifier(n_neighbors=23)
```

```
knn.fit(X_train,y_train)
pred = knn.predict(X_test)
```

```
print('WITH K=23')
print('\n')
print(confusion_matrix(y_test,pred))
print('\n')
print(classification_report(y_test,pred))
```

WITH K=23

```
[[143  13]
 [  8 136]]
```

	precision	recall	f1-score	support
0	0.95	0.92	0.93	156
1	0.91	0.94	0.93	144
accuracy			0.93	300
macro avg	0.93	0.93	0.93	300
weighted avg	0.93	0.93	0.93	300

