

```
import pandas as pd
import numpy as np
XXX = pd.read_csv("tmp_hms_sav_ass.csv", sep=';')
XXX
```



	id	idResource	idProject	refType	refId	rate	assignedWork	realWork	left
0	1	5	1	Activity	1	100	NaN	21.00	
1	4	15	4	Activity	4	100	NaN	131.00	
2	6	15	4	Activity	6	100	NaN	21.00	
3	8	15	4	Activity	9	100	NaN	4.00	
4	9	15	4	Activity	10	100	NaN	2.00	
...	
7196	7601	9	34	Activity	594	100	NaN	0.00	
7197	7602	32	34	Activity	594	100	NaN	NaN	
7198	7603	9	34	Activity	597	100	NaN	0.25	
7199	7604	7	12	Activity	1524	100	NaN	0.50	
7200	7605	7	50	Activity	1525	100	NaN	2.50	

7201 rows × 25 columns

```
XX = XXX[['idResource','idProject','refId','realCost','realWork']]
dataset = XX.replace(np.nan,0)
dataset = XX.replace(np.nan,0)
dataset['realWork'] = dataset['realWork'].astype(int)
dataset['realCost'] = dataset['realCost'].astype(int)
dataset
```

	idResource	idProject	refId	realCost	realWork
0	5	1	1	0	21
1	15	1	1	0	131

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

```
dataset = pd.read_csv('dataset.csv')
X = dataset.iloc[:, :-1].values
y = dataset.iloc[:, 4].values
```

7196	9	34	594	0	0
------	---	----	-----	---	---

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20)
```

7196	7	12	1524	0	0
------	---	----	------	---	---

```
from sklearn.preprocessing import StandardScaler
scaler = StandardScaler()
scaler.fit(X_train)
```

```
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)
```

```
from sklearn.neighbors import KNeighborsClassifier
classifier = KNeighborsClassifier(n_neighbors=1)
classifier.fit(X_train, y_train)
```

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

```
y_pred = classifier.predict(X_test)
```

```
from sklearn.metrics import classification_report, confusion_matrix
print(confusion_matrix(y_test, y_pred))
print(classification_report(y_test, y_pred))
```

```
[[1308  15   6 ...   0   0   0]
 [  9   8   1 ...   0   0   0]
 [  6   4   1 ...   0   0   0]
 ...
 [  1   0   0 ...   0   0   0]
 [  1   0   0 ...   0   0   0]
 [  0   0   0 ...   0   0   0]]
```

	precision	recall	f1-score	support
0	0.97	0.97	0.97	1351
1	0.24	0.38	0.29	21
2	0.08	0.08	0.08	13
3	0.00	0.00	0.00	7
4	0.00	0.00	0.00	5

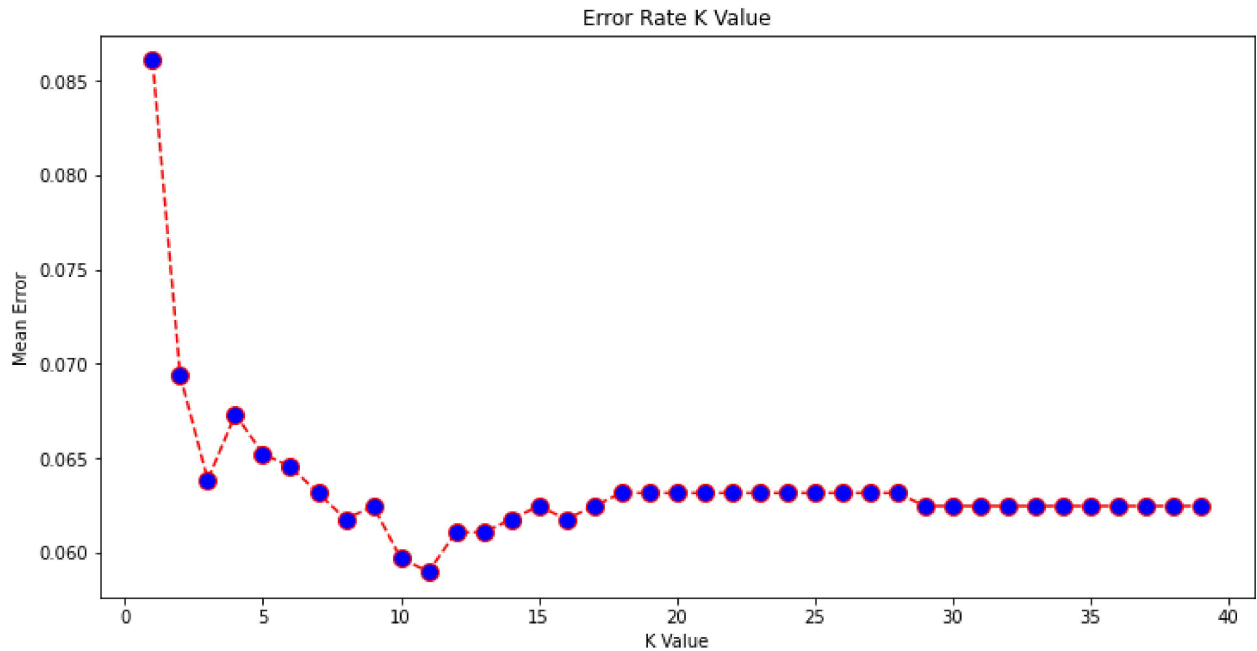
5	0.00	0.00	0.00	5
6	0.00	0.00	0.00	4
7	0.00	0.00	0.00	3
8	0.00	0.00	0.00	6
9	0.00	0.00	0.00	2
10	0.00	0.00	0.00	1
11	0.00	0.00	0.00	0
12	0.00	0.00	0.00	1
13	0.00	0.00	0.00	1
14	0.00	0.00	0.00	2
15	0.00	0.00	0.00	2
18	0.00	0.00	0.00	3
19	0.00	0.00	0.00	1
20	0.00	0.00	0.00	0
21	0.00	0.00	0.00	2
22	0.00	0.00	0.00	1
23	0.00	0.00	0.00	1
27	0.00	0.00	0.00	0
32	0.00	0.00	0.00	0
35	0.00	0.00	0.00	0
37	0.00	0.00	0.00	1
38	0.00	0.00	0.00	1
39	0.00	0.00	0.00	1
54	0.00	0.00	0.00	2
62	0.00	0.00	0.00	1
81	0.00	0.00	0.00	1
163	0.00	0.00	0.00	1
214	0.00	0.00	0.00	1
235	0.00	0.00	0.00	0
accuracy				0.91
macro avg				0.04
weighted avg				0.92

```
C:\Users\saifn\anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1221: l
_warn_prf(average, modifier, msg_start, len(result))
C:\Users\saifn\anaconda3\lib\site-packages\sklearn\metrics\_classification.py:1221: l
_warn_prf(average, modifier, msg_start, len(result))
```

```
error = []
# Calculating error for K values between 1 and 40
for i in range(1, 40):
    knn = KNeighborsClassifier(n_neighbors=i)
    knn.fit(X_train, y_train)
    pred_i = knn.predict(X_test)
    error.append(np.mean(pred_i != y_test))
```

```
plt.figure(figsize=(12, 6))
plt.plot(range(1, 40), error, color='red', linestyle='dashed', marker='o',
         markerfacecolor='blue', markersize=10)
plt.title('Error Rate K Value')
plt.xlabel('K Value')
plt.ylabel('Mean Error')
```

Text(0, 0.5, 'Mean Error')



```

from sklearn.neighbors import KNeighborsClassifier

#Setup arrays to store training and test accuracies
neighbors = np.arange(1,40)
train_accuracy = np.empty(len(neighbors))
test_accuracy = np.empty(len(neighbors))
#Compute accuracy on the training set
for i,k in enumerate(neighbors):
    #Setup a knn classifier with k neighbors
    knn = KNeighborsClassifier(n_neighbors=k)

    #Fit the model
    knn.fit(X_train, y_train)

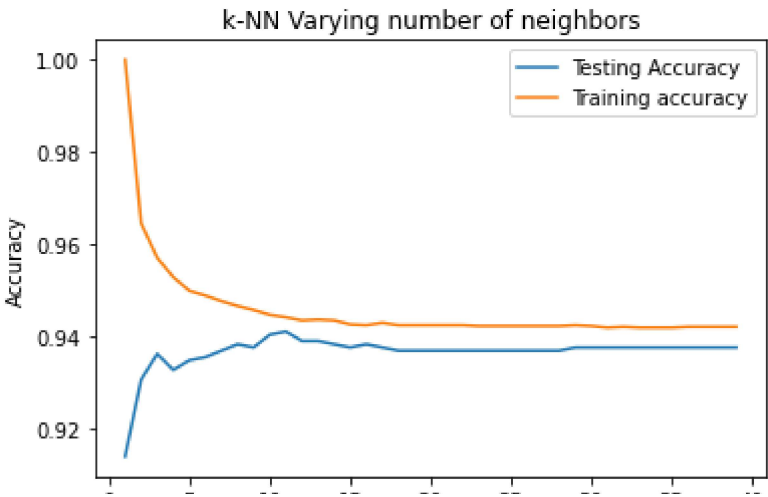
    train_accuracy[i] = knn.score(X_train, y_train)

#Compute accuracy on the test set

    test_accuracy[i] = knn.score(X_test, y_test)

#Generate plot
plt.title('k-NN Varying number of neighbors')
plt.plot(neighbors, test_accuracy, label='Testing Accuracy')
plt.plot(neighbors, train_accuracy, label='Training accuracy')
plt.legend()
plt.xlabel('Number of neighbors')
plt.ylabel('Accuracy')
plt.show()

```



```
knn.score(X_test, y_test)
```

0.9375433726578765