

Program 25

October 25, 2022

1 PROGRAM 25

Aim : Program to implement k-NN classification using the datasets (Breastcancer.csv, Telco-Customer-Churn.csv) and find the accuracy of the algorithm

Date : 28/09/2022

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```
[718]: import numpy as np
import pandas as pd
!pip install scikit-learn
```

```
Defaulting to user installation because normal site-packages is not writeable
Requirement already satisfied: scikit-learn in
/opt/anaconda3/lib/python3.9/site-packages (1.0.2)
Requirement already satisfied: threadpoolctl>=2.0.0 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (2.2.0)
Requirement already satisfied: numpy>=1.14.6 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (1.21.5)
Requirement already satisfied: scipy>=1.1.0 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (1.7.3)
Requirement already satisfied: joblib>=0.11 in
/opt/anaconda3/lib/python3.9/site-packages (from scikit-learn) (1.1.0)
```

```
[719]: import pandas as pd
df=pd.read_csv('Telco-Customer-Churn.csv')
#print(df)
```

```
[720]: cols=df.columns
print(cols)
```

```
Index(['customerID', 'gender', 'SeniorCitizen', 'Partner', 'Dependents',
      'tenure', 'PhoneService', 'MultipleLines', 'InternetService',
      'OnlineSecurity', 'OnlineBackup', 'DeviceProtection', 'TechSupport',
      'StreamingTV', 'StreamingMovies', 'Contract', 'PaperlessBilling',
      'PaymentMethod', 'MonthlyCharges', 'TotalCharges', 'Churn'],
      dtype='object')
```

```
[721]: df.describe()
```

```
[721]:
```

	SeniorCitizen	tenure	MonthlyCharges
count	7043.000000	7043.000000	7043.000000
mean	0.162147	32.371149	64.761692
std	0.368612	24.559481	30.090047
min	0.000000	0.000000	18.250000
25%	0.000000	9.000000	35.500000
50%	0.000000	29.000000	70.350000
75%	0.000000	55.000000	89.850000
max	1.000000	72.000000	118.750000

```
[722]: df.isnull().sum()
```

```
[722]:
```

customerID	0
gender	0
SeniorCitizen	0
Partner	0
Dependents	0
tenure	0
PhoneService	0
MultipleLines	0
InternetService	0
OnlineSecurity	0
OnlineBackup	0
DeviceProtection	0
TechSupport	0
StreamingTV	0
StreamingMovies	0
Contract	0
PaperlessBilling	0
PaymentMethod	0
MonthlyCharges	0
TotalCharges	0
Churn	0

dtype: int64

```
[723]: df.value_counts("Churn")
```

```
[723]:
```

Churn	
No	5174
Yes	1869

dtype: int64

```
[724]: df["TotalCharges"] = pd.to_numeric(df["TotalCharges"], errors='coerce')  
df = df.dropna()
```

```
[725]: y=df['Churn'] ##Dependent variable
df.drop('Churn',axis=1,inplace=True)
print(y)
```

```
0      No
1      No
2      Yes
3      No
4      Yes
...
7038   No
7039   No
7040   No
7041   Yes
7042   No
Name: Churn, Length: 7032, dtype: object
```

```
[726]: df.drop('customerID', axis = 1,inplace=True)
cols=df.columns
print(cols)
x=df
print(x)
```

```
Index(['gender', 'SeniorCitizen', 'Partner', 'Dependents', 'tenure',
      'PhoneService', 'MultipleLines', 'InternetService', 'OnlineSecurity',
      'OnlineBackup', 'DeviceProtection', 'TechSupport', 'StreamingTV',
      'StreamingMovies', 'Contract', 'PaperlessBilling', 'PaymentMethod',
      'MonthlyCharges', 'TotalCharges'],
      dtype='object')
gender SeniorCitizen Partner Dependents tenure PhoneService \
0      Female          0     Yes         No        1         No
1       Male          0     No         No       34         Yes
2       Male          0     No         No        2         Yes
3       Male          0     No         No       45         No
4      Female          0     No         No        2         Yes
...      ...          ...     ...         ...      ...
7038   Male          0     Yes         Yes       24         Yes
7039  Female          0     Yes         Yes       72         Yes
7040  Female          0     Yes         Yes       11         No
7041   Male          1     Yes         No        4         Yes
7042   Male          0     No         No       66         Yes

MultipleLines InternetService OnlineSecurity OnlineBackup \
0      No phone service          DSL         No         Yes
1           No          DSL         Yes         No
2           No          DSL         Yes         Yes
3      No phone service          DSL         Yes         No
4           No      Fiber optic         No         No
```

...
7038	Yes	DSL	Yes	No	
7039	Yes	Fiber optic	No	Yes	
7040	No phone service	DSL	Yes	No	
7041	Yes	Fiber optic	No	No	
7042	No	Fiber optic	Yes	No	

	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract \
0	No	No	No	No	Month-to-month
1	Yes	No	No	No	One year
2	No	No	No	No	Month-to-month
3	Yes	Yes	No	No	One year
4	No	No	No	No	Month-to-month

...
7038	Yes	Yes	Yes	Yes	One year
7039	Yes	No	Yes	Yes	One year
7040	No	No	No	No	Month-to-month
7041	No	No	No	No	Month-to-month
7042	Yes	Yes	Yes	Yes	Two year

	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges
0	Yes	Electronic check	29.85	29.85
1	No	Mailed check	56.95	1889.50
2	Yes	Mailed check	53.85	108.15
3	No	Bank transfer (automatic)	42.30	1840.75
4	Yes	Electronic check	70.70	151.65

...
7038	Yes	Mailed check	84.80	1990.50
7039	Yes	Credit card (automatic)	103.20	7362.90
7040	Yes	Electronic check	29.60	346.45
7041	Yes	Mailed check	74.40	306.60
7042	Yes	Bank transfer (automatic)	105.65	6844.50

[7032 rows x 19 columns]

```
[727]: from sklearn.preprocessing import LabelEncoder
label_encoder_y=LabelEncoder()
y=label_encoder_y.fit_transform(y)
print("After labeling :\n",y)
```

After labeling :
[0 0 1 ... 0 1 0]

```
[728]: label_encoder_x=LabelEncoder()
x.iloc[:,0]=label_encoder_x.fit_transform(x.iloc[:,0]) #gender
print("After labeling :\n",x)
```

After labeling :

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	\
0	0	0	Yes	No	1	No	
1	1	0	No	No	34	Yes	
2	1	0	No	No	2	Yes	
3	1	0	No	No	45	No	
4	0	0	No	No	2	Yes	
...	
7038	1	0	Yes	Yes	24	Yes	
7039	0	0	Yes	Yes	72	Yes	
7040	0	0	Yes	Yes	11	No	
7041	1	1	Yes	No	4	Yes	
7042	1	0	No	No	66	Yes	

	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	\
0	No phone service	DSL	No	Yes	
1	No	DSL	Yes	No	
2	No	DSL	Yes	Yes	
3	No phone service	DSL	Yes	No	
4	No	Fiber optic	No	No	
...	
7038	Yes	DSL	Yes	No	
7039	Yes	Fiber optic	No	Yes	
7040	No phone service	DSL	Yes	No	
7041	Yes	Fiber optic	No	No	
7042	No	Fiber optic	Yes	No	

	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	\
0	No	No	No	No	Month-to-month	
1	Yes	No	No	No	One year	
2	No	No	No	No	Month-to-month	
3	Yes	Yes	No	No	One year	
4	No	No	No	No	Month-to-month	
...	
7038	Yes	Yes	Yes	Yes	One year	
7039	Yes	No	Yes	Yes	One year	
7040	No	No	No	No	Month-to-month	
7041	No	No	No	No	Month-to-month	
7042	Yes	Yes	Yes	Yes	Two year	

	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges
0	Yes	Electronic check	29.85	29.85
1	No	Mailed check	56.95	1889.50
2	Yes	Mailed check	53.85	108.15
3	No	Bank transfer (automatic)	42.30	1840.75
4	Yes	Electronic check	70.70	151.65
...
7038	Yes	Mailed check	84.80	1990.50
7039	Yes	Credit card (automatic)	103.20	7362.90

7040	Yes	Electronic check	29.60	346.45
7041	Yes	Mailed check	74.40	306.60
7042	Yes	Bank transfer (automatic)	105.65	6844.50

[7032 rows x 19 columns]

```
[729]: x.iloc[:,2]=label_encoder_x.fit_transform(x.iloc[:,2]) #partner
print("After labeling :\n",x)
```

After labeling :

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	\
0	0	0	1	No	1	No	
1	1	0	0	No	34	Yes	
2	1	0	0	No	2	Yes	
3	1	0	0	No	45	No	
4	0	0	0	No	2	Yes	
...	
7038	1	0	1	Yes	24	Yes	
7039	0	0	1	Yes	72	Yes	
7040	0	0	1	Yes	11	No	
7041	1	1	1	No	4	Yes	
7042	1	0	0	No	66	Yes	

	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	\
0	No phone service	DSL	No	Yes	
1	No	DSL	Yes	No	
2	No	DSL	Yes	Yes	
3	No phone service	DSL	Yes	No	
4	No	Fiber optic	No	No	
...	
7038	Yes	DSL	Yes	No	
7039	Yes	Fiber optic	No	Yes	
7040	No phone service	DSL	Yes	No	
7041	Yes	Fiber optic	No	No	
7042	No	Fiber optic	Yes	No	

	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	\
0	No	No	No	No	Month-to-month	
1	Yes	No	No	No	One year	
2	No	No	No	No	Month-to-month	
3	Yes	Yes	No	No	One year	
4	No	No	No	No	Month-to-month	
...	
7038	Yes	Yes	Yes	Yes	One year	
7039	Yes	No	Yes	Yes	One year	
7040	No	No	No	No	Month-to-month	
7041	No	No	No	No	Month-to-month	
7042	Yes	Yes	Yes	Yes	Two year	

	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges
0	Yes	Electronic check	29.85	29.85
1	No	Mailed check	56.95	1889.50
2	Yes	Mailed check	53.85	108.15
3	No	Bank transfer (automatic)	42.30	1840.75
4	Yes	Electronic check	70.70	151.65
...
7038	Yes	Mailed check	84.80	1990.50
7039	Yes	Credit card (automatic)	103.20	7362.90
7040	Yes	Electronic check	29.60	346.45
7041	Yes	Mailed check	74.40	306.60
7042	Yes	Bank transfer (automatic)	105.65	6844.50

[7032 rows x 19 columns]

```
[730]: x.iloc[:,3]=label_encoder_x.fit_transform(x.iloc[:,3]) #Dependents
x.iloc[:,5]=label_encoder_x.fit_transform(x.iloc[:,5]) #PhoneService
x.iloc[:,6]=label_encoder_x.fit_transform(x.iloc[:,6]) #multiplelines
x.iloc[:,7]=label_encoder_x.fit_transform(x.iloc[:,7]) #internet service
x.iloc[:,8]=label_encoder_x.fit_transform(x.iloc[:,8]) #Online Security
x.iloc[:,9]=label_encoder_x.fit_transform(x.iloc[:,9]) #Online Backup
x.iloc[:,10]=label_encoder_x.fit_transform(x.iloc[:,10]) #DeviceProtection
x.iloc[:,11]=label_encoder_x.fit_transform(x.iloc[:,11]) #TechSupport
x.iloc[:,12]=label_encoder_x.fit_transform(x.iloc[:,12]) #StreamingTV
x.iloc[:,13]=label_encoder_x.fit_transform(x.iloc[:,13]) #StreamingMovies
x.iloc[:,14]=label_encoder_x.fit_transform(x.iloc[:,14]) #Contract
x.iloc[:,15]=label_encoder_x.fit_transform(x.iloc[:,15]) #PaperlessBilling
x.iloc[:,16]=label_encoder_x.fit_transform(x.iloc[:,16]) #PaymentMethod
print("After labeling :\n",x)
```

After labeling :

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	\
0	0	0	1	0	1	0	
1	1	0	0	0	34	1	
2	1	0	0	0	2	1	
3	1	0	0	0	45	0	
4	0	0	0	0	2	1	
...	
7038	1	0	1	1	24	1	
7039	0	0	1	1	72	1	
7040	0	0	1	1	11	0	
7041	1	1	1	0	4	1	
7042	1	0	0	0	66	1	

	MultipleLines	InternetService	OnlineSecurity	OnlineBackup	\
0	1	0	0	2	
1	0	0	2	0	

2	0	0	2	2
3	1	0	2	0
4	0	1	0	0
...
7038	2	0	2	0
7039	2	1	0	2
7040	1	0	2	0
7041	2	1	0	0
7042	0	1	2	0

	DeviceProtection	TechSupport	StreamingTV	StreamingMovies	Contract	\
0	0	0	0	0	0	
1	2	0	0	0	1	
2	0	0	0	0	0	
3	2	2	0	0	1	
4	0	0	0	0	0	
...	
7038	2	2	2	2	1	
7039	2	0	2	2	1	
7040	0	0	0	0	0	
7041	0	0	0	0	0	
7042	2	2	2	2	2	

	PaperlessBilling	PaymentMethod	MonthlyCharges	TotalCharges
0	1	2	29.85	29.85
1	0	3	56.95	1889.50
2	1	3	53.85	108.15
3	0	0	42.30	1840.75
4	1	2	70.70	151.65
...
7038	1	3	84.80	1990.50
7039	1	1	103.20	7362.90
7040	1	2	29.60	346.45
7041	1	3	74.40	306.60
7042	1	0	105.65	6844.50

[7032 rows x 19 columns]

```
[731]: x.shape
```

```
[731]: (7032, 19)
```

```
[732]: x=df
one_hot_encoding_columns=["MultipleLines","InternetService","OnlineSecurity","OnlineBackup","D
x=pd.get_dummies(x,columns=one_hot_encoding_columns)
print(x)
```

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	\
--	--------	---------------	---------	------------	--------	--------------	---

0	0	0	1	0	1	0
1	1	0	0	0	34	1
2	1	0	0	0	2	1
3	1	0	0	0	45	0
4	0	0	0	0	2	1
...
7038	1	0	1	1	24	1
7039	0	0	1	1	72	1
7040	0	0	1	1	11	0
7041	1	1	1	0	4	1
7042	1	0	0	0	66	1

	PaperlessBilling	MonthlyCharges	TotalCharges	MultipleLines_0	...	\
0	1	29.85	29.85	0	...	
1	0	56.95	1889.50	1	...	
2	1	53.85	108.15	1	...	
3	0	42.30	1840.75	0	...	
4	1	70.70	151.65	1	...	
...	
7038	1	84.80	1990.50	0	...	
7039	1	103.20	7362.90	0	...	
7040	1	29.60	346.45	0	...	
7041	1	74.40	306.60	0	...	
7042	1	105.65	6844.50	1	...	

	StreamingMovies_0	StreamingMovies_1	StreamingMovies_2	Contract_0	\
0	1	0	0	1	
1	1	0	0	0	
2	1	0	0	1	
3	1	0	0	0	
4	1	0	0	1	
...	
7038	0	0	1	0	
7039	0	0	1	0	
7040	1	0	0	1	
7041	1	0	0	1	
7042	0	0	1	0	

	Contract_1	Contract_2	PaymentMethod_0	PaymentMethod_1	\
0	0	0	0	0	
1	1	0	0	0	
2	0	0	0	0	
3	1	0	1	0	
4	0	0	0	0	
...	
7038	1	0	0	0	
7039	1	0	0	1	
7040	0	0	0	0	

7041	0	0	0	0
7042	0	1	1	0

	PaymentMethod_2	PaymentMethod_3
0	1	0
1	0	1
2	0	1
3	0	0
4	1	0
...
7038	0	1
7039	0	0
7040	1	0
7041	0	1
7042	0	0

[7032 rows x 40 columns]

```
[734]: x.isnull().sum()
```

```
[734]: gender                0
SeniorCitizen              0
Partner                    0
Dependents                  0
tenure                      0
PhoneService                0
PaperlessBilling            0
MonthlyCharges              0
TotalCharges                0
MultipleLines_0             0
MultipleLines_1             0
MultipleLines_2             0
InternetService_0           0
InternetService_1           0
InternetService_2           0
OnlineSecurity_0            0
OnlineSecurity_1            0
OnlineSecurity_2            0
OnlineBackup_0              0
OnlineBackup_1              0
OnlineBackup_2              0
DeviceProtection_0          0
DeviceProtection_1          0
DeviceProtection_2          0
TechSupport_0               0
TechSupport_1               0
TechSupport_2               0
```

```

StreamingTV_0      0
StreamingTV_1      0
StreamingTV_2      0
StreamingMovies_0  0
StreamingMovies_1  0
StreamingMovies_2  0
Contract_0         0
Contract_1         0
Contract_2         0
PaymentMethod_0    0
PaymentMethod_1    0
PaymentMethod_2    0
PaymentMethod_3    0
dtype: int64

```

```

[735]: #convert total charge into numeric
x["TotalCharges"]=pd.to_numeric(x["TotalCharges"],errors='coerce')
print(x)

```

	gender	SeniorCitizen	Partner	Dependents	tenure	PhoneService	\
0	0	0	1	0	1	0	
1	1	0	0	0	34	1	
2	1	0	0	0	2	1	
3	1	0	0	0	45	0	
4	0	0	0	0	2	1	
...	
7038	1	0	1	1	24	1	
7039	0	0	1	1	72	1	
7040	0	0	1	1	11	0	
7041	1	1	1	0	4	1	
7042	1	0	0	0	66	1	

	PaperlessBilling	MonthlyCharges	TotalCharges	MultipleLines_0	...	\
0	1	29.85	29.85	0	...	
1	0	56.95	1889.50	1	...	
2	1	53.85	108.15	1	...	
3	0	42.30	1840.75	0	...	
4	1	70.70	151.65	1	...	
...	
7038	1	84.80	1990.50	0	...	
7039	1	103.20	7362.90	0	...	
7040	1	29.60	346.45	0	...	
7041	1	74.40	306.60	0	...	
7042	1	105.65	6844.50	1	...	

	StreamingMovies_0	StreamingMovies_1	StreamingMovies_2	Contract_0	\
0	1	0	0	1	

1	1	0	0	0
2	1	0	0	1
3	1	0	0	0
4	1	0	0	1
...
7038	0	0	1	0
7039	0	0	1	0
7040	1	0	0	1
7041	1	0	0	1
7042	0	0	1	0

	Contract_1	Contract_2	PaymentMethod_0	PaymentMethod_1	\
0	0	0	0	0	
1	1	0	0	0	
2	0	0	0	0	
3	1	0	1	0	
4	0	0	0	0	
...	
7038	1	0	0	0	
7039	1	0	0	1	
7040	0	0	0	0	
7041	0	0	0	0	
7042	0	1	1	0	

	PaymentMethod_2	PaymentMethod_3
0	1	0
1	0	1
2	0	1
3	0	0
4	1	0
...
7038	0	1
7039	0	0
7040	1	0
7041	0	1
7042	0	0

[7032 rows x 40 columns]

```
[736]: x.dtypes
```

```
[736]: gender          int64
SeniorCitizen        int64
Partner              int64
Dependents           int64
tenure               int64
PhoneService         int64
```

```

PaperlessBilling      int64
MonthlyCharges        float64
TotalCharges          float64
MultipleLines_0       uint8
MultipleLines_1       uint8
MultipleLines_2       uint8
InternetService_0     uint8
InternetService_1     uint8
InternetService_2     uint8
OnlineSecurity_0      uint8
OnlineSecurity_1      uint8
OnlineSecurity_2      uint8
OnlineBackup_0        uint8
OnlineBackup_1        uint8
OnlineBackup_2        uint8
DeviceProtection_0    uint8
DeviceProtection_1    uint8
DeviceProtection_2    uint8
TechSupport_0         uint8
TechSupport_1         uint8
TechSupport_2         uint8
StreamingTV_0         uint8
StreamingTV_1         uint8
StreamingTV_2         uint8
StreamingMovies_0     uint8
StreamingMovies_1     uint8
StreamingMovies_2     uint8
Contract_0            uint8
Contract_1            uint8
Contract_2            uint8
PaymentMethod_0       uint8
PaymentMethod_1       uint8
PaymentMethod_2       uint8
PaymentMethod_3       uint8
dtype: object

```

```

from sklearn.preprocessing import MinMaxScaler st_x=MinMaxScaler() x=st_x.fit_transform(x)
print(x)

```

```
[737]: x.isnull().sum()
```

```

[737]: gender      0
      SeniorCitizen  0
      Partner      0
      Dependents    0
      tenure       0
      PhoneService  0

```

PaperlessBilling	0
MonthlyCharges	0
TotalCharges	0
MultipleLines_0	0
MultipleLines_1	0
MultipleLines_2	0
InternetService_0	0
InternetService_1	0
InternetService_2	0
OnlineSecurity_0	0
OnlineSecurity_1	0
OnlineSecurity_2	0
OnlineBackup_0	0
OnlineBackup_1	0
OnlineBackup_2	0
DeviceProtection_0	0
DeviceProtection_1	0
DeviceProtection_2	0
TechSupport_0	0
TechSupport_1	0
TechSupport_2	0
StreamingTV_0	0
StreamingTV_1	0
StreamingTV_2	0
StreamingMovies_0	0
StreamingMovies_1	0
StreamingMovies_2	0
Contract_0	0
Contract_1	0
Contract_2	0
PaymentMethod_0	0
PaymentMethod_1	0
PaymentMethod_2	0
PaymentMethod_3	0

dtype: int64

```
[738]: x=x.dropna()
```

```
[739]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x, y,test_size= 0.
↳25,random_state=0)
```

```
[740]: from sklearn.preprocessing import MinMaxScaler
st_x=MinMaxScaler()
x_train=st_x.fit_transform(x_train)
x_test=st_x.fit_transform(x_test)
print("Taining data \n",x_train)
```

```
print("Testing data \n",x_test)
```

Taining data

```
[[0. 0. 1. ... 0. 0. 0.]  
[1. 0. 1. ... 0. 0. 0.]  
[0. 0. 0. ... 0. 1. 0.]  
...  
[1. 0. 1. ... 1. 0. 0.]  
[0. 0. 1. ... 1. 0. 0.]  
[1. 0. 0. ... 1. 0. 0.]]
```

Testing data

```
[[1. 0. 0. ... 0. 1. 0.]  
[1. 0. 0. ... 1. 0. 0.]  
[0. 0. 0. ... 1. 0. 0.]  
...  
[1. 0. 0. ... 0. 1. 0.]  
[1. 1. 0. ... 0. 0. 0.]  
[0. 0. 1. ... 0. 0. 1.]]
```

```
[741]: from sklearn.neighbors import KNeighborsClassifier  
classifier=KNeighborsClassifier(n_neighbors=5)  
classifier.fit(x_train,y_train)
```

```
[741]: KNeighborsClassifier()
```

```
[742]: y_pred=classifier.predict(x_test)  
print(y_pred)
```

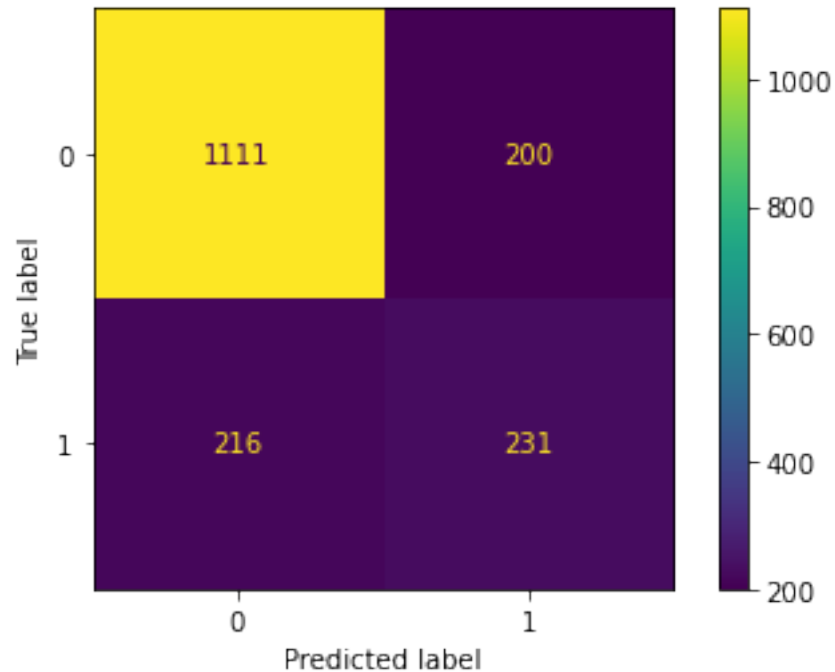
```
[0 0 0 ... 1 0 0]
```

```
[743]: from sklearn.metrics import confusion_matrix  
cm=confusion_matrix (y_test,y_pred,labels=classifier.classes_)  
print(cm)
```

```
[[1111  200]  
 [ 216  231]]
```

```
[744]: from sklearn.metrics import ConfusionMatrixDisplay  
disp=ConfusionMatrixDisplay(confusion_matrix=cm,display_labels=classifier.  
    ↪classes_)  
disp.plot()
```

```
[744]: <sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at  
0x7f15f1363eb0>
```



```
[745]: training_score=classifier.score(x_train, y_train)
test_score=classifier.score(x_test, y_test)
print("Training Accuracy :",training_score)
print("Testing Accuracy :",test_score)
```

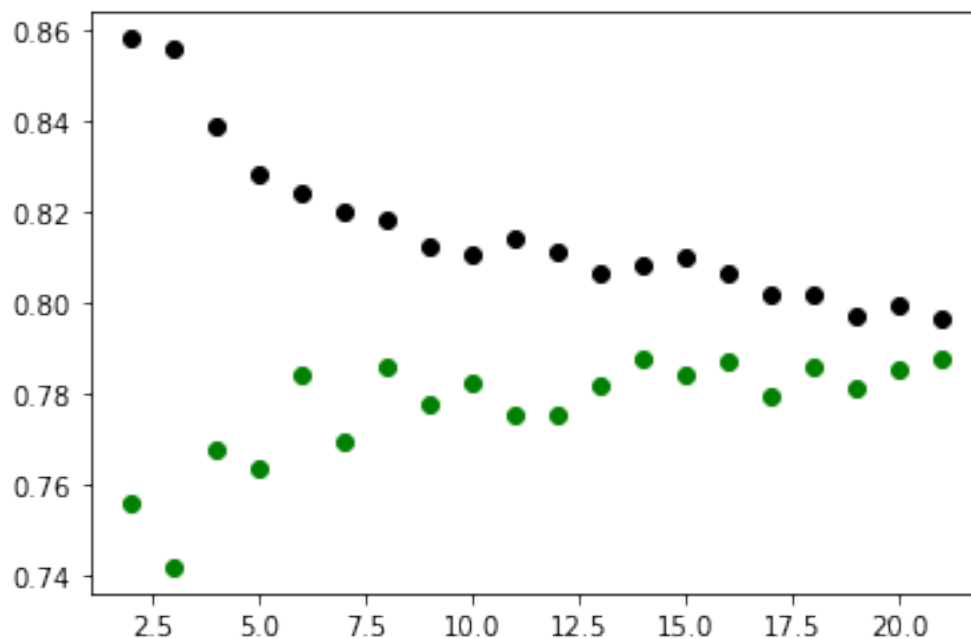
Training Accuracy : 0.8282138794084186
Testing Accuracy : 0.7633674630261661

```
[748]: K=[]
training=[]
test=[]
scores={}
for k in range(2, 22):
    clf=KNeighborsClassifier(n_neighbors=k)
    clf.fit(x_train,y_train)
    training_score=clf.score(x_train, y_train)
    test_score=clf.score(x_test, y_test)
    K.append(k)
    training.append(training_score)
    test.append(test_score)
    scores[k]=[training_score, test_score]
for keys, values in scores.items():
    print(keys,':',values)
# visualization
import matplotlib.pyplot as plt
```



```
plt.scatter(K,training,color='k')
plt.scatter(K,test,color='g')
plt.show()
```

```
2 : [0.8581721653394009, 0.7559726962457338]
3 : [0.8558968524838833, 0.7417519908987485]
4 : [0.8386423966628744, 0.767349260523322]
5 : [0.8282138794084186, 0.7633674630261661]
6 : [0.8244216913158893, 0.7838452787258248]
7 : [0.820250284414107, 0.7696245733788396]
8 : [0.8179749715585893, 0.7861205915813424]
9 : [0.8122866894197952, 0.7775881683731513]
10 : [0.810580204778157, 0.7821387940841866]
11 : [0.8141827834660599, 0.7753128555176336]
12 : [0.8109594235874099, 0.7753128555176336]
13 : [0.8064087978763747, 0.7815699658703071]
14 : [0.8081152825180129, 0.7878270762229806]
15 : [0.8102009859689041, 0.7838452787258248]
16 : [0.8064087978763747, 0.7872582480091013]
17 : [0.8020477815699659, 0.7792946530147895]
18 : [0.801668562760713, 0.7861205915813424]
19 : [0.7973075464543041, 0.7810011376564278]
20 : [0.7992036405005688, 0.785551763367463]
21 : [0.7963594994311718, 0.7878270762229806]
```



[]: