MT18052_Quiz_3

November 8, 2019

```
In [135]: import numpy as np
          import pandas as pd
          from sklearn.model_selection import KFold
          from sklearn.model_selection import cross_val_score
          from sklearn.metrics import confusion_matrix, accuracy_score
          import warnings
          warnings.filterwarnings('ignore')
          import matplotlib.pyplot as plt
In [52]: data = pd.read_csv("./Telco_Data.csv")
In [3]: data.head()
Out[3]:
           {\tt customerID}
                        gender
                                SeniorCitizen Partner Dependents
                                                                    tenure PhoneService
        0
           7590-VHVEG
                        Female
                                             0
                                                    Yes
                                                                          1
                                                                No
                                                                                       No
           5575-GNVDE
                                             0
                                                                         34
        1
                          Male
                                                     No
                                                                No
                                                                                      Yes
                                                    No
           3668-QPYBK
                                             0
                                                                          2
                          Male
                                                                No
                                                                                      Yes
                                             0
          7795-CFOCW
                          Male
                                                     No
                                                                No
                                                                         45
                                                                                       No
           9237-HQITU Female
                                                     No
                                                                No
                                                                                      Yes
              MultipleLines InternetService OnlineSecurity
                                                               ... DeviceProtection
        0
           No phone service
                                          DSL
                                                           No
                                                                                  No
        1
                          No
                                          DSL
                                                          Yes
                                                                                 Yes
                                          DSL
                                                          Yes ...
                                                                                  No
                          No
        3
                                          DSL
                                                          Yes
                                                                                 Yes
           No phone service
        4
                                  Fiber optic
                                                           No
                                                                                  No
          TechSupport StreamingTV StreamingMovies
                                                            Contract PaperlessBilling
        0
                    No
                                                      Month-to-month
                                                                                    Yes
                                No
                                                 No
        1
                    No
                                                                                    No
                                No
                                                 No
                                                            One year
        2
                    No
                                 No
                                                 No
                                                      Month-to-month
                                                                                    Yes
        3
                   Yes
                                 No
                                                  No
                                                            One year
                                                                                    No
        4
                    No
                                 No
                                                      Month-to-month
                                                                                    Yes
                                                  No
                        PaymentMethod MonthlyCharges
                                                        TotalCharges Churn
        0
                     Electronic check
                                                29.85
                                                               29.85
                                                                         No
        1
                         Mailed check
                                                56.95
                                                              1889.5
                                                                         Nο
        2
                         Mailed check
                                                53.85
                                                              108.15
                                                                        Yes
```

```
3 Bank transfer (automatic)
                                            42.30
                                                        1840.75
                   Electronic check
                                            70.70
        [5 rows x 21 columns]
In [54]: # data = data.drop(columns=['customerID'])
In [55]: label = data.Churn
        data = data.drop(columns=['Churn'])
In [56]: for i in data.keys():
            print (i , " : ",len(set(data[i])))
gender: 2
SeniorCitizen : 2
Partner: 2
Dependents : 2
tenure: 73
PhoneService : 2
MultipleLines : 3
InternetService : 3
OnlineSecurity : 3
OnlineBackup : 3
DeviceProtection : 3
TechSupport : 3
StreamingTV : 3
StreamingMovies : 3
Contract : 3
PaperlessBilling : 2
PaymentMethod : 4
MonthlyCharges : 1585
TotalCharges : 6531
In [6]: # y
In [7]: totalcharges = np.array(data.TotalCharges)
In [13]:
In []:
In [11]: missing = []
        summ = 0.0
        for i in range(len(totalcharges)):
            try:
                totalcharges[i] = float(totalcharges[i])
                summ+=float(totalcharges[i])
            except:
                missing.append(i)
```

No

Yes

151.65

```
In [12]: for i in missing:
             totalcharges[i] = summ/(len(totalcharges)-len(missing))
In [57]: totalcharges = list(map(float,totalcharges))
In [58]: data['TotalCharges'] = totalcharges
In [59]: data['MonthlyCharges'] = list(map(float,data['MonthlyCharges']))
In [17]: # data.to_csv('Telco_Data_2.csv', index=False)
In [30]: data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 7043 entries, 0 to 7042
Data columns (total 20 columns):
                    7043 non-null object
gender
                    7043 non-null int64
SeniorCitizen
                    7043 non-null object
Partner
Dependents
                    7043 non-null object
tenure
                    7043 non-null int64
                    7043 non-null object
PhoneService
                    7043 non-null object
MultipleLines
InternetService
                    7043 non-null object
                    7043 non-null object
OnlineSecurity
OnlineBackup
                    7043 non-null object
DeviceProtection
                    7043 non-null object
TechSupport
                    7043 non-null object
StreamingTV
                    7043 non-null object
StreamingMovies
                    7043 non-null object
                    7043 non-null object
Contract
                    7043 non-null object
PaperlessBilling
                    7043 non-null object
PaymentMethod
MonthlyCharges
                    7043 non-null float64
TotalCharges
                    7043 non-null float64
                    7043 non-null object
dtypes: float64(2), int64(2), object(16)
memory usage: 1.1+ MB
In [28]: # datatypes[0]
In [61]: keys = list(data.keys())
In [62]: encodingcols = []
In [63]: for i in keys:
             if (type(data[i][0]) == str):
                 encodingcols.append(i)
```

```
In [64]: keys
Out [64]: ['gender',
          'SeniorCitizen',
          'Partner',
          'Dependents',
          'tenure',
          'PhoneService',
          'MultipleLines',
          'InternetService',
          'OnlineSecurity',
          'OnlineBackup',
          'DeviceProtection',
          'TechSupport',
          'StreamingTV',
          'StreamingMovies',
          'Contract',
          'PaperlessBilling',
          'PaymentMethod',
          'MonthlyCharges',
          'TotalCharges']
In [65]: encodingcols
Out[65]: ['gender',
          'Partner',
          'Dependents',
          'PhoneService',
          'MultipleLines',
          'InternetService',
          'OnlineSecurity',
          'OnlineBackup',
          'DeviceProtection',
          'TechSupport',
          'StreamingTV',
          'StreamingMovies',
          'Contract',
          'PaperlessBilling',
          'PaymentMethod']
In [66]: from sklearn.preprocessing import LabelEncoder
         labelencoder_X = LabelEncoder()
         for i in encodingcols:
                         = labelencoder_X.fit_transform(data[i])
In [67]: data.head(15)
Out [67]:
             gender SeniorCitizen Partner Dependents tenure PhoneService \
         0
                                  0
                                           1
                                                        0
                                                                1
```

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	
5	0	0	0	0		8		1	
6	1	0	0	1		22		1	
7	0	0	0	0		10		0	
8	0	0	1	0		28		1	
9	1	0	0	1		62		1	
10	1	0	1	1		13		1	
11	1	0	0	0		16		1	
12	1	0	1	0		58		1	
13	1	0	0	0		49		1	
14	1	0	0	0		25		1	
	MultipleLines	Interne	tService	OnlineSecur	ity	Online	Backup	\	
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		
5	2		1	0			0		
6	2		1	0			2		
7	1		0		2		0		
8	2		1		0		0		
9	0		0		2		2		
10	0		0	2			0		
11	0		2				1		
12	2		1				0		
13	2			1 0			2		
14	0		1		2		0		
	DeviceProtection			StreamingTV	Str	eamingMo		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	
5	2		0				2	0	
6	0			0 2			0	0	
7	0		0		0		0	0	
8		2	2	2			2	0	
9	0		0	0			0	1	
10		0	0	0			0	0	
11		1	1	1			1	2	
12		2	0	2 2			2 2	1	
13 14		2	0 2	2			2	0	
14		2	2	2			2	0	

```
PaperlessBilling
                               PaymentMethod MonthlyCharges
                                                               TotalCharges
         0
                                                        29.85
                                                                       29.85
                            1
                                            2
         1
                            0
                                                        56.95
                                                                     1889.50
                                            3
         2
                            1
                                            3
                                                        53.85
                                                                      108.15
         3
                                            0
                                                        42.30
                            0
                                                                     1840.75
         4
                            1
                                            2
                                                        70.70
                                                                      151.65
         5
                            1
                                            2
                                                        99.65
                                                                      820.50
         6
                                                                     1949.40
                            1
                                            1
                                                        89.10
         7
                            0
                                            3
                                                        29.75
                                                                      301.90
                                            2
         8
                            1
                                                       104.80
                                                                     3046.05
         9
                            0
                                            0
                                                        56.15
                                                                     3487.95
                                            3
                                                        49.95
                                                                      587.45
         10
                            1
                                            1
         11
                            0
                                                        18.95
                                                                      326.80
         12
                                            1
                                                       100.35
                                                                     5681.10
         13
                                            0
                                                       103.70
                                                                     5036.30
                            1
         14
                            1
                                            2
                                                       105.50
                                                                     2686.05
In [72]: from sklearn.metrics import classification_report
         def reports(truelab, predlabels):
             print(confusion_matrix(truelab, predlabels))
             print("Accuracy ",round(accuracy_score(truelab, predlabels),2)*100)
             print (classification_report(y_pred=predlabels,y_true=truelab))
             return ( round(accuracy_score(truelab, predlabels),2)*100)
In [102]: from sklearn.model_selection import train_test_split
          X_train, X_test, y_train, y_test = train_test_split(data, label, test_size = 0.3, rain
In [103]: from sklearn.preprocessing import StandardScaler
          sc_X = StandardScaler()
          X_train = sc_X.fit_transform(X_train)
          X_test = sc_X.transform(X_test)
   GNB
1
In [104]: from sklearn.naive_bayes import GaussianNB
          gaussiannb= GaussianNB()
          gaussiannb.fit(X_train, y_train)
          gaussiannbpred = gaussiannb.predict(X_test)
          probs = gaussiannb.predict(X_test)
          GAUSIAN = reports(y_test,gaussiannbpred)
[[1190 356]
[ 160 407]]
Accuracy 76.0
              precision recall f1-score
                                               support
```

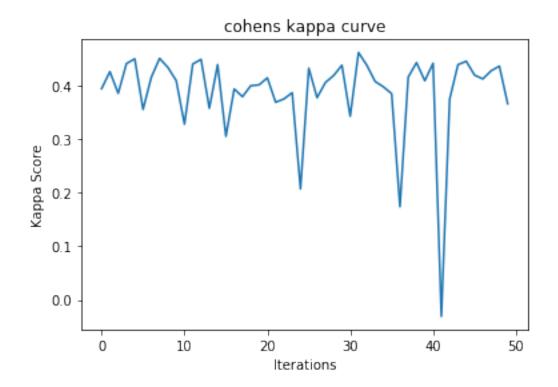
```
No
                    0.88
                              0.77
                                         0.82
                                                    1546
                               0.72
         Yes
                    0.53
                                         0.61
                                                     567
                                         0.76
                                                    2113
    accuracy
   macro avg
                    0.71
                              0.74
                                         0.72
                                                    2113
weighted avg
                    0.79
                               0.76
                                         0.77
                                                    2113
```

In []:

2 Boot Strap

```
In [179]: from imblearn.over_sampling import SMOTE
          sm = SMOTE(random_state=42)
          X_train,y_train = sm.fit_sample(X_train,y_train)
In [180]: X = copy.deepcopy(X_train)
          Y = copy.deepcopy(y_train)
In [181]: X_sparse = coo_matrix(X_train)
In [182]: epoch = 50
In [183]: # X_train
In [184]: kappascores = []
          index = np.arange(len(X_train))
          for i in range(epoch):
              temp = np.random.choice(index,size=100,replace=True)
              gaussiannb= GaussianNB()
                print (temp)
              train_2 = []
              y_train_2 = y_train[temp]
              for col in range(len(X_train.T)):
                  train_2.append(X_train.T[col][temp])
              train_2 = np.array(train_2).T
              gaussiannb.fit(train_2, y_train_2)
              gaussiannbpred = gaussiannb.predict(X_test)
```

Out[187]: Text(0.5, 1.0, 'cohens kappa curve')



In []: