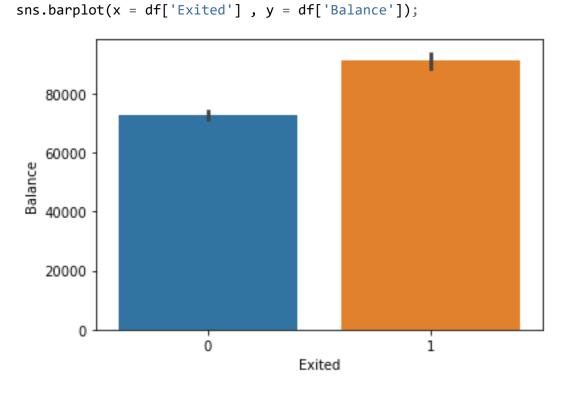
NAME:	Abdul Jabbar		
DEGREE:	Bachelor of Engineering		
DEPARTMENT:	Electronics and communication Engineering		
ASSIGNMENT:	2		
PROJECT NAME:	Emerging Methods for Early Detection of Forest Fires		

```
import panda as pd
import numpy as np
df = pd.read_csv('Churn_Modelling.csv')
df
      RowNumber
                  CustomerId
                                  Surname
                                           CreditScore Geography
                                                                     Gender
                                                                              Age
                                                                                   \
0
                     15634602
                                 Hargrave
                                                    619
                                                            France
                                                                     Female
               1
                                                                               42
1
               2
                    15647311
                                     Hill
                                                     608
                                                             Spain Female
                                                                               41
2
               3
                                                     502
                    15619304
                                     Onio
                                                            France Female
                                                                               42
3
               4
                    15701354
                                     Boni
                                                    699
                                                            France Female
                                                                               39
4
               5
                    15737888
                                 Mitchell
                                                     850
                                                             Spain Female
                                                                               43
             . . .
                                                     . . .
                                                                              . . .
. . .
                          . . .
            9996
                                 Obijiaku
                                                                       Male
                                                                               39
9995
                    15606229
                                                    771
                                                            France
9996
            9997
                    15569892
                               Johnstone
                                                     516
                                                            France
                                                                       Male
                                                                               35
9997
            9998
                    15584532
                                      Liu
                                                    709
                                                            France Female
                                                                               36
9998
            9999
                    15682355
                               Sabbatini
                                                    772
                                                           Germany
                                                                       Male
                                                                               42
9999
           10000
                    15628319
                                   Walker
                                                    792
                                                            France Female
                                                                               28
      Tenure
                 Balance
                           NumOfProducts
                                           HasCrCard IsActiveMember
0
            2
                    0.00
                                        1
                                                    1
                                                                      1
1
            1
                83807.86
                                        1
                                                    0
                                                                      1
2
            8
                                        3
                                                    1
                                                                      0
               159660.80
3
            1
                                        2
                                                    0
                                                                      0
                    0.00
4
            2
                                        1
                                                     1
                                                                      1
               125510.82
. . .
                      . . .
                                       . . .
          . . .
                                                                    . . .
9995
            5
                                        2
                                                    1
                                                                      0
                    0.00
9996
           10
                57369.61
                                        1
                                                    1
                                                                      1
9997
            7
                    0.00
                                        1
                                                    0
                                                                      1
            3
                                        2
                                                    1
9998
                75075.31
                                                                      0
9999
               130142.79
                                        1
                                                     1
      EstimatedSalary
                         Exited
0
             101348.88
1
             112542.58
                              0
2
                              1
             113931.57
3
                              0
              93826.63
4
              79084.10
                              0
```

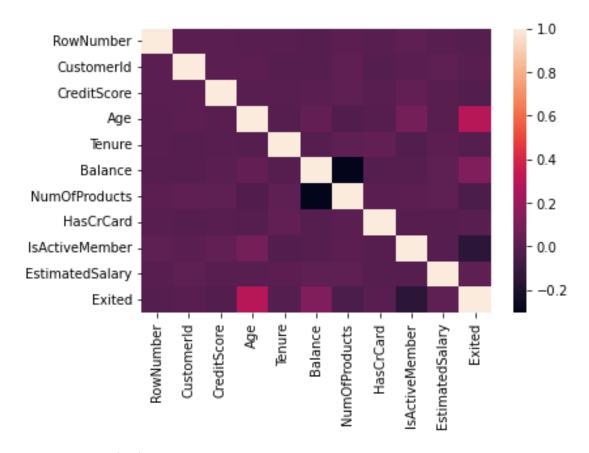
```
9995
             96270.64
                            0
9996
            101699.77
                            0
9997
             42085.58
                            1
9998
             92888.52
                            1
9999
             38190.78
[10000 rows x 14 columns]
##visulaizations
##univariate
df['Age'].mean()
38.9218
df['Balance'].median()
97198.54000000001
##bivariate
import matplotlib.pyplot as plt
```



#multivariate

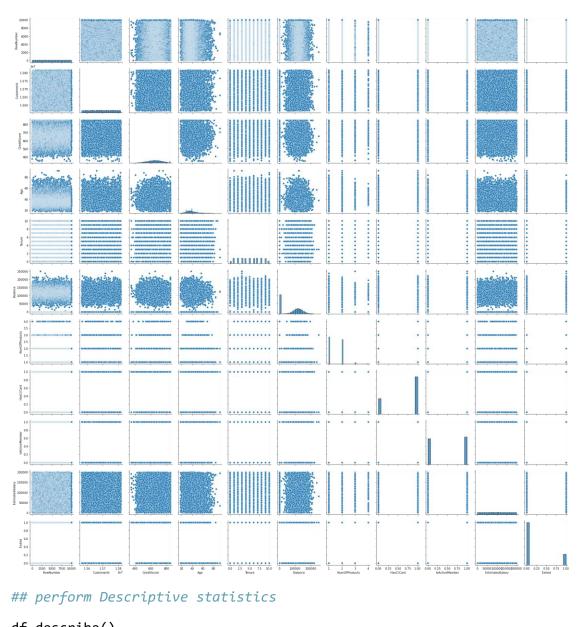
```
sns.heatmap(df.corr());
```

import seaborn as sns



sns.pairplot(df)

<seaborn.axisgrid.PairGrid at 0x29387e55970>



perform Descriptive statistics

df.describe()

count	RowNumber	CustomerId	CreditScore	Age 10000.000000	Tenure 10000.000000	\
mean	5000.50000	1.569094e+07	650.528800	38.921800	5.012800	
std	2886.89568	7.193619e+04	96.653299	10.487806	2.892174	
min	1.00000	1.556570e+07	350.000000	18.000000	0.000000	
25%	2500.75000	1.562853e+07	584.000000	32.000000	3.000000	
50%	5000.50000	1.569074e+07	652.000000	37.000000	5.000000	
75%	7500.25000	1.575323e+07	718.000000	44.000000	7.000000	
max	10000.00000	1.581569e+07	850.000000	92.000000	10.000000	

	Balance	NumOfProducts	HasCrCard	IsActiveMember	\
count	10000.000000	10000.000000	10000.00000	10000.000000	
mean	76485.889288	1.530200	0.70550	0.515100	

std min 25% 50% 75% max	62397.405202 0.000000 0.000000 97198.540000 127644.240000 250898.090000	0.581654 1.000000 1.000000 1.000000 2.000000 4.000000	0.45584 0.00000 0.00000 1.00000 1.00000	0.499797 0.000000 0.000000 1.000000 1.000000
	EstimatedSalary	Exited		
count	10000.000000	10000.000000		
mean	100090.239881	0.203700		
std	57510.492818	0.402769		
min	11.580000	0.000000		
25%	51002.110000	0.000000		
50%	100193.915000	0.000000		
75%	149388.247500	0.000000		
max	199992.480000	1.000000		

#handle missing values

df.isnull().sum()

RowNumber 0 CustomerId 0 Surname CreditScore 0 Geography 0 Gender 0 0 Age Tenure 0 Balance 0 NumOfProducts 0 HasCrCard IsActiveMember EstimatedSalary 0 Exited 0 dtype: int64

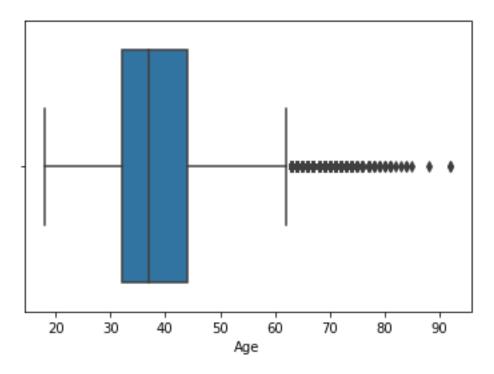
#find the outliers and replace it

```
sns.boxplot(df['Age']);
```

C:\Users\BALAJI POWER MART\anaconda3\lib\site-

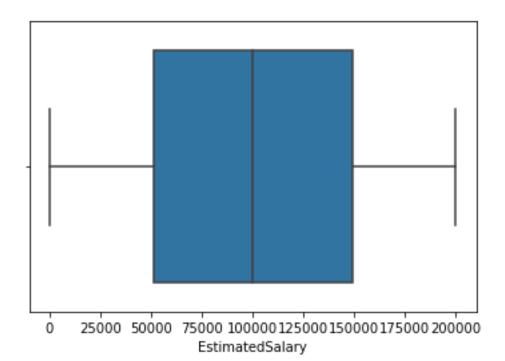
packages\seaborn_decorators.py:36: FutureWarning: Pass the following variable as a keyword arg: x. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

warnings.warn(



sns.boxplot(df['EstimatedSalary']);

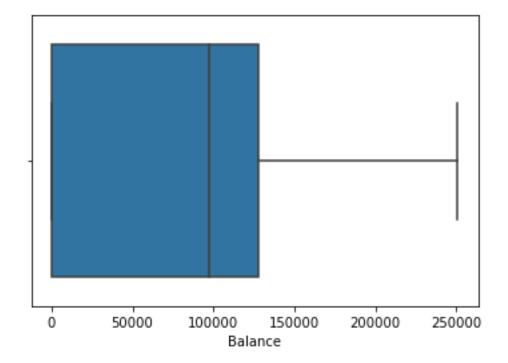
C:\Users\BALAJI POWER MART\anaconda3\lib\sitepackages\seaborn_decorators.py:36: FutureWarning: Pass the following
variable as a keyword arg: x. From version 0.12, the only valid positional
argument will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
 warnings.warn(



sns.boxplot(df['Balance'])

C:\Users\BALAJI POWER MART\anaconda3\lib\sitepackages\seaborn_decorators.py:36: FutureWarning: Pass the following
variable as a keyword arg: x. From version 0.12, the only valid positional
argument will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
warnings.warn(

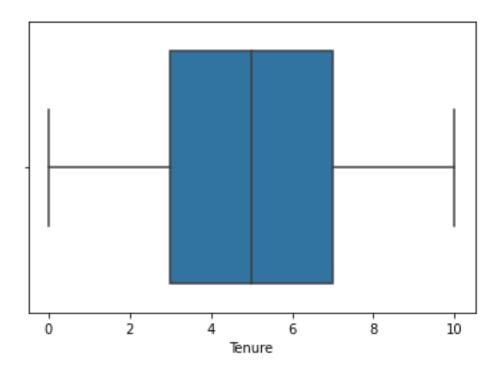
<AxesSubplot:xlabel='Balance'>



sns.boxplot(df['Tenure'])

C:\Users\BALAJI POWER MART\anaconda3\lib\sitepackages\seaborn_decorators.py:36: FutureWarning: Pass the following
variable as a keyword arg: x. From version 0.12, the only valid positional
argument will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
warnings.warn(

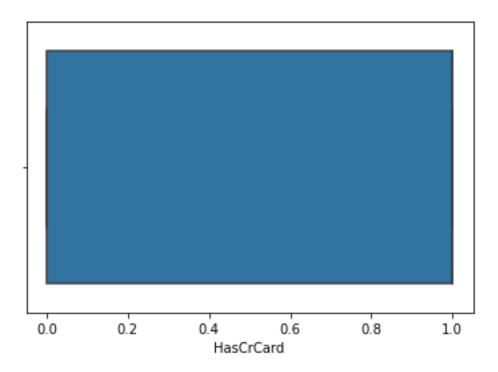
<AxesSubplot:xlabel='Tenure'>



sns.boxplot(df['HasCrCard'])

C:\Users\BALAJI POWER MART\anaconda3\lib\sitepackages\seaborn_decorators.py:36: FutureWarning: Pass the following
variable as a keyword arg: x. From version 0.12, the only valid positional
argument will be `data`, and passing other arguments without an explicit
keyword will result in an error or misinterpretation.
 warnings.warn(

<AxesSubplot:xlabel='HasCrCard'>



#CHECK FOR CATEGORICAL COLUMNS AND PERFORM ENCODING

df.info()

Age

Tenure

7

<class 'pandas.core.frame.DataFrame'> RangeIndex: 10000 entries, 0 to 9999 Data columns (total 14 columns): # Column Non-Null Count Dtype _____ -----0 RowNumber 10000 non-null int64 CustomerId 1 10000 non-null int64 2 Surname 10000 non-null object 3 CreditScore 10000 non-null int64 4 10000 non-null object Geography 5 Gender 10000 non-null object 6 10000 non-null int64

8 Balance 10000 non-null float64 NumOfProducts 10000 non-null int64 HasCrCard 10000 non-null int64 IsActiveMember 10000 non-null int64 11 12 10000 non-null EstimatedSalary float64 13 Exited 10000 non-null int64 dtypes: float64(2), int64(9), object(3)

10000 non-null int64

memory usage: 1.1+ MB

from sklearn.preprocessing import LabelEncoder

```
le = LabelEncoder()
df['Geography'] = le.fit_transform(df['Geography'])
df['CreditScore'] = le.fit transform(df['CreditScore'])
df
      RowNumber
                  CustomerId
                                          CreditScore
                                                        Geography
                                 Surname
                                                                    Gender
                                                                             Age
\
0
               1
                    15634602
                                Hargrave
                                                   228
                                                                    Female
                                                                              42
1
               2
                                                   217
                                                                 2
                                                                    Female
                                                                              41
                    15647311
                                    Hill
2
               3
                                                                    Female
                                                                              42
                    15619304
                                    Onio
                                                   111
3
               4
                    15701354
                                    Boni
                                                   308
                                                                 0
                                                                    Female
                                                                              39
4
               5
                                                   459
                                                                 2
                    15737888
                                Mitchell
                                                                    Female
                                                                              43
                                                   . . .
9995
           9996
                    15606229
                                Obijiaku
                                                   380
                                                                 0
                                                                      Male
                                                                              39
9996
           9997
                               Johnstone
                                                                      Male
                    15569892
                                                   125
                                                                 0
                                                                              35
9997
           9998
                    15584532
                                     Liu
                                                   318
                                                                 0
                                                                    Female
                                                                              36
9998
           9999
                                                                 1
                                                                      Male
                                                                              42
                    15682355
                               Sabbatini
                                                   381
9999
                                                   401
          10000
                    15628319
                                  Walker
                                                                    Female
                                                                              28
      Tenure
                 Balance
                          NumOfProducts
                                          HasCrCard IsActiveMember
0
                    0.00
           2
                                       1
                                                   1
                                                                    1
1
           1
               83807.86
                                       1
                                                   0
                                                                    1
2
           8
              159660.80
                                       3
                                                   1
                                                                    0
3
                                       2
           1
                    0.00
                                                   0
                                                                    0
4
           2
              125510.82
                                       1
                                                   1
                                                                    1
           5
                                       2
9995
                    0.00
                                                   1
                                                                    0
9996
                57369.61
                                       1
                                                   1
                                                                    1
          10
9997
           7
                    0.00
                                       1
                                                   0
                                                                    1
                                       2
                                                   1
9998
           3
                75075.31
                                                                    0
9999
              130142.79
                                       1
                                                   1
      EstimatedSalary Exited
0
            101348.88
                              1
1
            112542.58
                              0
2
            113931.57
                              1
3
             93826.63
                              0
4
              79084.10
                              0
             96270.64
                              0
9995
                              0
9996
            101699.77
9997
              42085.58
                              1
                              1
9998
             92888.52
             38190.78
9999
                              0
```

[10000 rows x 14 columns]

```
#SPLIT THE DATA INTO DEPENDENT AND INDEPENDENT VARIABLES
```

```
X = df.iloc[:, :-1].values
Χ
array([[1, 15634602, 'Hargrave', ..., 1, 1, 101348.88],
       [2, 15647311, 'Hill', ..., 0, 1, 112542.58],
       [3, 15619304, 'Onio', ..., 1, 0, 113931.57],
       [9998, 15584532, 'Liu', ..., 0, 1, 42085.58],
       [9999, 15682355, 'Sabbatini', ..., 1, 0, 92888.52],
       [10000, 15628319, 'Walker', ..., 1, 0, 38190.78]], dtype=object)
#SCALE THE INDEPENDENT VARIABLES
 from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
Y = sc.fit_transform(X)
У
array([2, 2, 0, 2, 2, 0, 1, 0, 1, 1, 2, 1, 1, 2, 1, 1, 2, 0, 2, 0, 2, 1,
       1, 0, 1, 2, 1, 2, 0, 2, 2, 2, 0, 0, 0, 1, 2, 2, 0, 0, 1, 1, 0, 2,
       1, 2, 2, 0, 1, 0, 1, 1, 1, 1, 2, 1, 1, 2, 1, 2, 2, 0, 1, 2, 0, 1,
       2, 0, 2, 1, 0, 0, 2, 0, 0, 0, 2, 0, 0, 2, 2, 2, 1, 0, 2, 0, 2, 1,
       0, 1, 0, 0, 0, 2, 1, 2, 0, 1, 0, 1, 0, 0, 2, 1, 1, 1, 2, 2, 1,
       1, 1, 2, 2, 0, 2, 1, 1, 1, 2, 0, 0, 1, 1, 2, 2, 1, 0, 2, 1, 0, 1,
       1, 2, 0, 0, 0, 2, 1, 0, 1, 1, 0, 2, 0, 0, 2, 0, 1, 0, 1, 2, 2, 1,
       0, 0, 1, 0, 2, 1, 0, 2, 0, 1, 2, 1, 2, 1, 1, 1, 0, 0, 2, 1, 2, 2,
       2, 0, 1, 2, 2, 1, 1, 2, 2, 0, 0, 2, 1, 1, 2, 2, 1, 1, 1, 0, 2, 0,
       0, 0, 0, 0, 1, 1, 2, 2, 2, 1, 0, 1, 1, 2, 0, 1, 0, 0, 0, 2, 1, 2,
       1, 1, 2, 1, 0, 0, 0, 0, 0, 2, 0, 2, 0, 2, 0, 1, 2, 1, 1, 1, 0, 0,
       1, 1, 1, 1, 1, 2, 1, 1, 2, 0, 1, 1, 0, 1, 2, 0, 1, 2, 2, 2, 0, 0,
       1, 1, 0, 1, 2, 1, 1, 2, 2, 2, 2, 0, 0, 1, 2, 1, 1, 1, 2, 1, 2, 1,
       2, 1, 1, 0, 0, 1, 2, 2, 2, 1, 2, 0, 0, 0, 1, 2, 2, 2, 0, 0, 2, 0,
       1, 2, 2, 2, 2, 2, 0, 1, 1, 2, 2, 0, 0, 1, 1, 1, 2, 1, 1, 0, 2, 0,
       0, 2, 1, 2, 1, 1, 1, 0, 1, 2, 2, 1, 2, 0, 2, 1, 0, 0, 1, 0, 2, 1,
       2, 2, 2, 2, 2, 1, 1, 0, 1, 2, 0, 2, 1, 2, 2, 2, 0, 0, 1, 2, 2, 0,
       0, 2, 0, 0, 1, 2, 2, 0, 0, 1, 1, 2, 2, 2, 0, 0, 2, 1, 1, 2, 0, 1,
       2, 0, 2, 0, 0, 1, 2, 2, 2, 0, 2, 2, 2, 0, 2, 1, 0, 0, 1, 1, 1,
       0, 2, 0, 0, 1, 2, 2, 1, 2, 2, 1, 2, 2, 2, 1, 0, 1, 0, 1, 0, 0, 1,
       2, 2, 0, 2, 2, 0, 0, 2, 1, 1, 2, 2, 1, 1, 0, 2, 2, 0, 0, 0, 0, 1,
       0, 1, 2, 1, 2, 0, 0, 0, 1, 2, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 2, 1, 1,
       0, 1, 1, 1, 0, 0, 2, 2, 0, 1, 2, 1, 2, 2, 1, 0, 0, 2, 2, 1, 1, 2,
       0, 2, 1, 0, 1, 2, 1, 2, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 2, 2,
       1, 0, 2, 0, 0, 1, 0, 2, 2, 1, 0, 0, 0, 1, 0, 2, 1, 1, 1, 1, 2, 0,
       0, 0, 0, 2, 0, 2, 1, 1, 2, 1, 0, 0, 0, 0, 2, 1, 1, 1, 2, 0, 0, 2,
       1, 0, 2, 0, 0, 0, 2, 2, 0, 1, 0, 1, 1, 0, 0, 1, 2, 0, 1, 0, 0, 0,
       0, 1, 1, 1, 1, 0, 0, 2, 0, 1, 0, 1, 2, 0, 0, 0, 2, 0, 0, 1, 0, 1,
       0, 1, 2, 2, 2, 0, 1, 0, 0, 0, 0, 2, 1, 2, 2, 0, 1, 2, 2, 0, 1, 0,
       1, 2, 1, 2, 2, 1, 2, 0, 1, 2, 1, 1, 0, 0, 2, 1, 0, 1, 2, 1, 0, 1,
       2, 2, 0, 1, 1, 0, 0, 1, 1, 2, 1, 0, 0, 0, 1, 2, 2, 1, 0, 0, 2, 0,
```

```
2, 0, 1, 2, 2, 0, 1, 1, 1, 1, 0, 0, 1, 2, 0, 0, 1, 2, 0, 0, 2, 2, 1, 1, 2, 0, 0, 0, 1, 0, 1, 1, 1, 0, 1, 2, 1, 2, 1, 2, 2, 0, 2, 0, 2, 0, 1, 2, 1, 2, 1, 2, 2, 2, 2, 1, 1, 0, 1, 0, 0, 2, 2, 2, 2, 1, 1, 0, 1, 0, 0, 2, 2, 2, 2, 1, 1, 0, 1, 2, 1, 2, 2, 2, 2, 1, 1, 0, 1, 2, 1, 2, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 2, 2, 2, 2, 1, 1, 1, 2, 1, 2, 0, 2, 1, 1, 0, 1, 2, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 2, 2, 2, 1, 1, 1, 1, 1, 1, 2, 2, 0, 2, 1, 1, 0, 2, 0, 1, 2, 1, 1, 1, 1, 1, 1, 2, 2, 0, 2, 2, 1, 1, 2, 2, 0, 2, 2, 1, 1, 1, 1, 1, 1, 2, 2, 0, 1, 2, 1, 1, 2, 2, 0, 2, 2, 1, 1, 1, 1, 1, 2, 2, 1, 2, 0, 1, 0, 0, 2, 1, 0, 1, 1, 2, 2, 1, 2, 0, 1, 0, 0, 2, 1, 0, 1, 1, 2, 2, 1, 1, 1, 1, 1, 1, 1, 0, 1, 2, 0, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 0, 0, 0, 2, 2, 1, 1, 1, 1, 1, 1, 0, 1, 2, 0, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 0, 0, 0, 2, 2, 1, 1, 1, 1, 1, 0, 0, 2, 2, 2, 2, 0, 1, 1, 0, 0, 2, 2, 2, 1, 1, 1, 0, 0, 2, 2, 2, 2, 0, 1, 1, 0, 0, 2, 2, 2, 2, 2, 0, 1, 1, 0, 0, 2, 2, 2, 2, 0, 1, 1, 0, 0, 2, 0, 0, 2, 2, 0, 0, 2, 1, 1, 1, 0, 0, 2, 0, 0, 2, 2, 0, 0, 2, 1, 1, 1, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 1, 1, 1, 0, 0, 2, 0, 1, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 1, 0, 0, 0, 2, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 0, 2, 0, 2, 0, 2, 0, 2,
```

split data into training and testing

```
from sklearn.datasets import make_blobs
from sklearn.model_selection import train_test_split
X, y = make_blobs(n_samples=1000)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33)
print(X_train.shape, X_test.shape, y_train.shape, y_test.shape)

(670, 2) (330, 2) (670,) (330,)
```