

ASSIGNMENT-2

Assignment Date	28 October 2022
Student Name	Ms.Abinaya k
Student Roll Number	192IT107
Maximum Marks	2 Marks
Team ID	PNT2022TMID01939

1810 lines (1810 sloc) | 286 KB

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```
In [4]: import pandas as pd
import os
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
df = pd.read_csv("/content/drive/MyDrive/Churn_Modelling.csv")
df.head()
```

Out[4]:

	RowNumber	CustomerId	Surname	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	1	15634602	Hargrave	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	2	15647311	Hill	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	3	15619304	Onio	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	4	15701354	Boni	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	5	15737888	Mitchell	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

```
In [6]: df['HasCrCard'] = df['HasCrCard'].astype('category')
```

```
In [7]: df['IsActiveMember'] = df['IsActiveMember'].astype('category')
```

```
In [7]: df['IsActiveMember'] = df['IsActiveMember'].astype('category')
df['Exited'] = df['Exited'].astype('category')
```

```
In [8]: df = df.drop(columns=['RowNumber', 'CustomerId', 'Surname'])
```

```
In [9]: df.head()
```

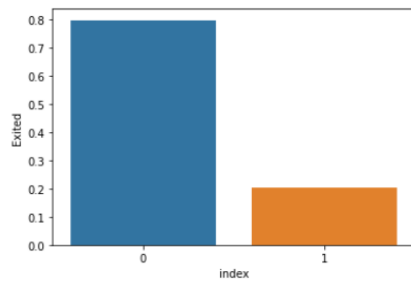
Out[9]:

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary	Exited
0	619	France	Female	42	2	0.00	1	1	1	101348.88	1
1	608	Spain	Female	41	1	83807.86	1	0	1	112542.58	0
2	502	France	Female	42	8	159660.80	3	1	0	113931.57	1
3	699	France	Female	39	1	0.00	2	0	0	93826.63	0
4	850	Spain	Female	43	2	125510.82	1	1	1	79084.10	0

```
In [10]: import seaborn as sns
density = df['Exited'].value_counts(normalize=True).reset_index()
sns.barplot(data=density, x='index', y='Exited', );
density
```

```
Out[10]:
```

	index	Exited
0	0	0.7963
1	1	0.2037



```
In [ ]: import matplotlib.pyplot as plt
```

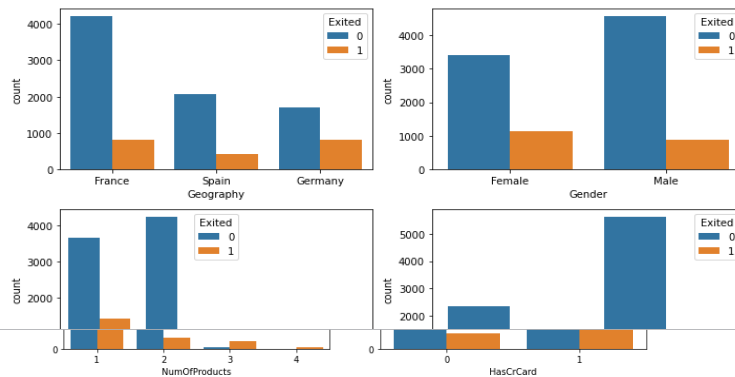
```
In [11]: categorical = df.drop(columns=['CreditScore', 'Age', 'Tenure', 'Balance', 'EstimatedSalary'])
rows = int(np.ceil(categorical.shape[1] / 2)) - 1

# create sub-plots and title them
fig, axes = plt.subplots(nrows=rows, ncols=2, figsize=(10,6))
axes = axes.flatten()

for row in range(rows):
    cols = min(2, categorical.shape[1] - row*2)
    for col in range(cols):
        col_name = categorical.columns[2 * row + col]
        ax = axes[row*2 + col]

        sns.countplot(data=categorical, x=col_name, hue="Exited", ax=ax);

plt.tight_layout()
```



```
In [12]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 11 columns):
#   Column             Non-Null Count  Dtype
---  -
0   CreditScore         10000 non-null  int64
1   Geography           10000 non-null  object
2   Gender              10000 non-null  object
3   Age                 10000 non-null  int64
4   Tenure              10000 non-null  int64
5   Balance              10000 non-null  float64
6   NumOfProducts       10000 non-null  int64
7   HasCrCard           10000 non-null  category
8   IsActiveMember      10000 non-null  category
9   EstimatedSalary      10000 non-null  float64
10  Exited              10000 non-null  category
dtypes: category(3), float64(2), int64(4), object(2)
memory usage: 654.8+ KB
```

```
In [13]: df.describe()
```

```
Out[13]:
```

	CreditScore	Age	Tenure	Balance	NumOfProducts	EstimatedSalary
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean	650.528800	38.921800	5.012800	76485.889288	1.530200	100090.239881
std	96.653299	10.487806	2.892174	62397.405202	0.581654	57510.492818
min	350.000000	18.000000	0.000000	0.000000	1.000000	11.580000
25%	584.000000	32.000000	3.000000	0.000000	1.000000	51002.110000

```
In [13]: df.describe()
```

```
Out[13]:
```

	CreditScore	Age	Tenure	Balance	NumOfProducts	EstimatedSalary
count	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000	10000.000000
mean	650.528800	38.921800	5.012800	76485.889288	1.530200	100090.239881
std	96.653299	10.487806	2.892174	62397.405202	0.581654	57510.492818
min	350.000000	18.000000	0.000000	0.000000	1.000000	11.580000
25%	584.000000	32.000000	3.000000	0.000000	1.000000	51002.110000
50%	652.000000	37.000000	5.000000	97198.540000	1.000000	100193.915000
75%	718.000000	44.000000	7.000000	127644.240000	2.000000	149388.247500
max	850.000000	92.000000	10.000000	250898.090000	4.000000	199992.480000

```
In [14]: df.isna().sum()
```

```
Out[14]:
```

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

```
In [15]: for i in df:  
         if df[i].dtype=='object' or df[i].dtype=='category':  
             print("unique of "+i+" is "+str(len(set(df[i])))+" they are "+str(set(df[i])))
```

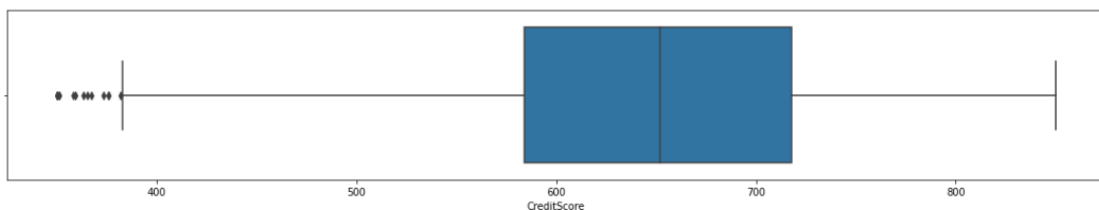
```
In [15]: for i in df:  
         if df[i].dtype=='object' or df[i].dtype=='category':  
             print("unique of "+i+" is "+str(len(set(df[i])))+" they are "+str(set(df[i])))
```

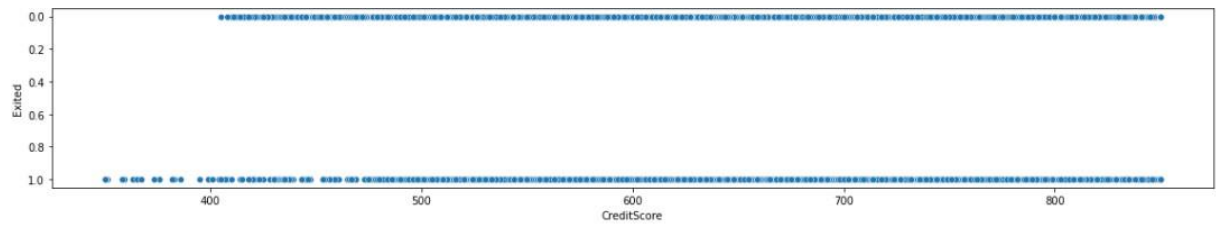
```
unique of Geography is 3 they are {'Spain', 'France', 'Germany'}  
unique of Gender is 2 they are {'Female', 'Male'}  
unique of HasCrCard is 2 they are {0, 1}  
unique of IsActiveMember is 2 they are {0, 1}  
unique of Exited is 2 they are {0, 1}
```

```
In [16]: def box_scatter(data, x, y):  
         fig, (ax1, ax2) = plt.subplots(nrows=2, ncols=1, figsize=(16,6))  
         sns.boxplot(data=data, x=x, ax=ax1)  
         sns.scatterplot(data=data, x=x, y=y, ax=ax2)
```

```
In [17]: box_scatter(df, 'CreditScore', 'Exited');  
         plt.tight_layout()  
         print(f"# of Bivariate Outliers: {len(df.loc[df['CreditScore'] < 400])}")
```

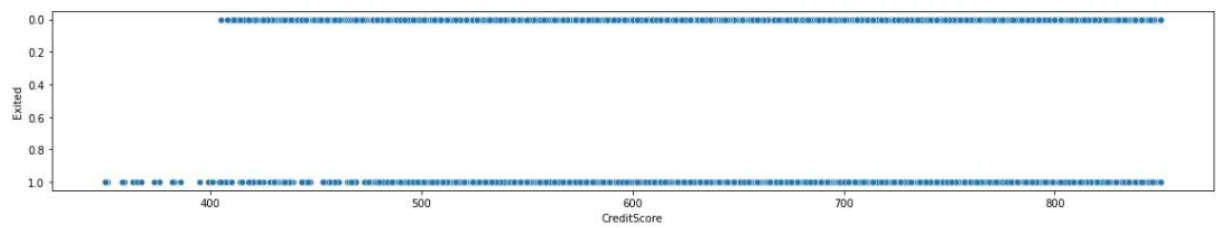
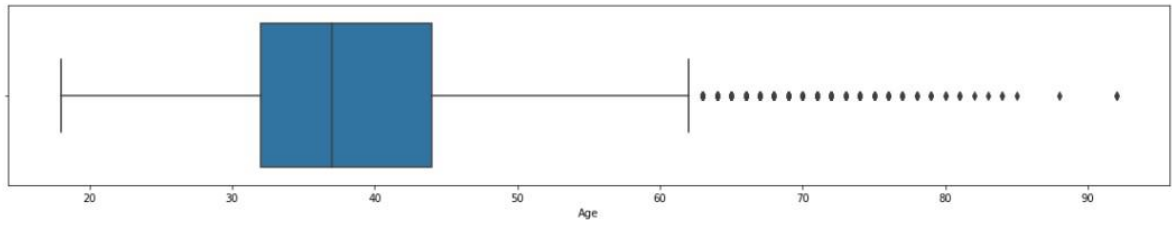
of Bivariate Outliers: 19





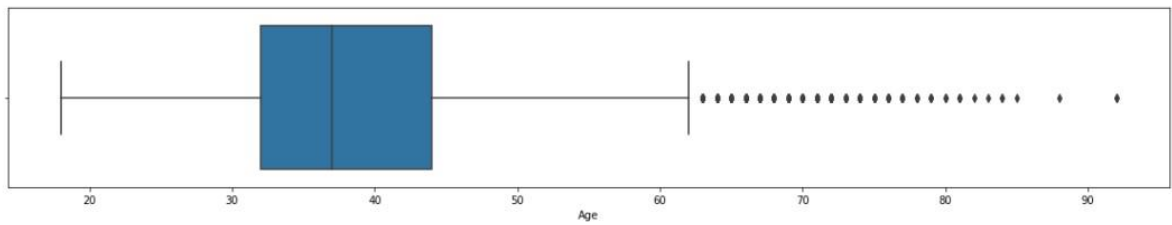
```
] box_scatter(df, 'Age', 'Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Age'] > 87])}")
```

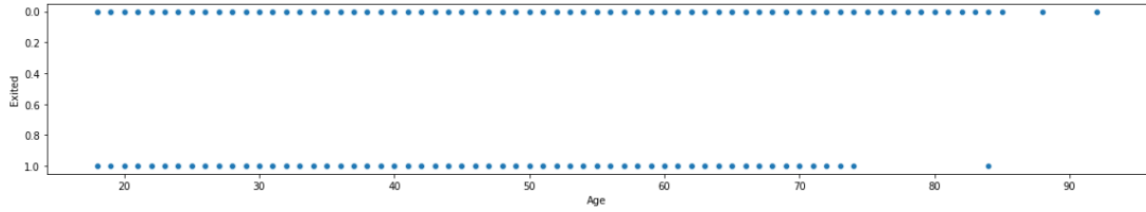
of Bivariate Outliers: 3



```
] box_scatter(df, 'Age', 'Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Age'] > 87])}")
```

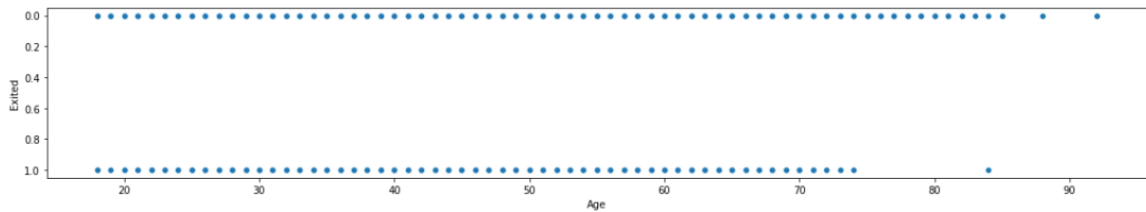
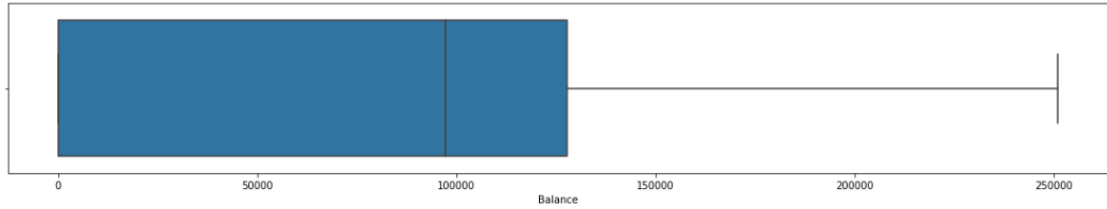
of Bivariate Outliers: 3





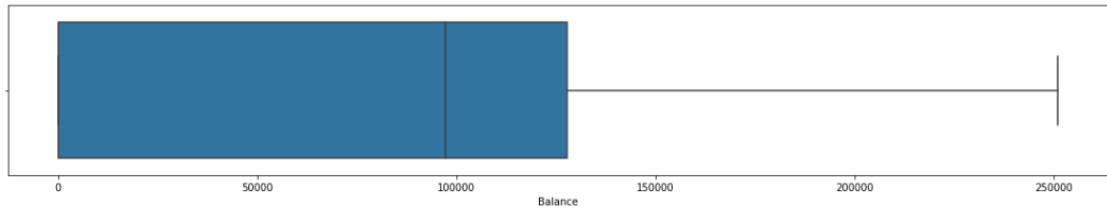
```
In [19]: box_scatter(df,'Balance','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Balance'] > 220000])}")
```

of Bivariate Outliers: 4



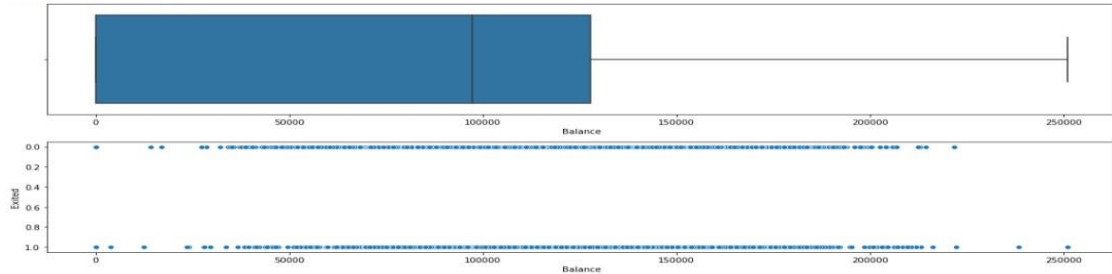
```
In [19]: box_scatter(df,'Balance','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Balance'] > 220000])}")
```

of Bivariate Outliers: 4



```
In [19]: box_scatter(df,'Balance','Exited');
plt.tight_layout()
print(f"# of Bivariate Outliers: {len(df.loc[df['Balance'] > 220000])}")
```

of Bivariate Outliers: 4



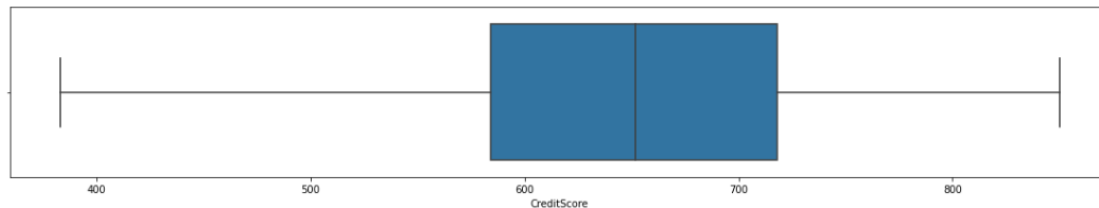
```
In [20]: box_scatter(df,'EstimatedSalary','Exited');
plt.tight_layout()
```

3 new noti

```
In [21]: for i in df:
        if df[i].dtype=='int64' or df[i].dtypes=='float64':
            q1=df[i].quantile(0.25)
            q3=df[i].quantile(0.75)
            iqr=q3-q1
            upper=q3+1.5*iqr
            lower=q1-1.5*iqr
            df[i]=np.where(df[i] >upper, upper, df[i])
            df[i]=np.where(df[i] <lower, lower, df[i])
```

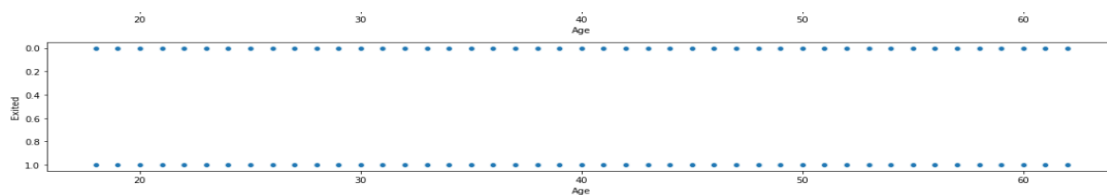
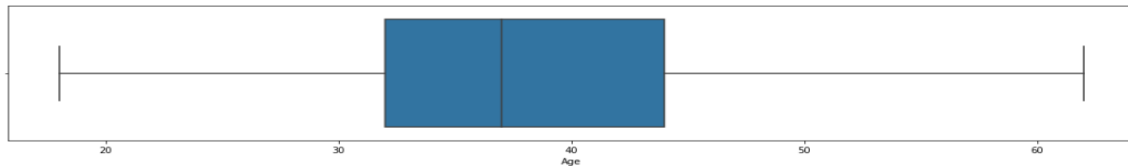
```
In [22]: box_scatter(df,'CreditScore','Exited');
        plt.tight_layout()
        print(f"# of Bivariate Outliers: {len(df.loc[df['CreditScore'] < 400])}")
```

of Bivariate Outliers: 19



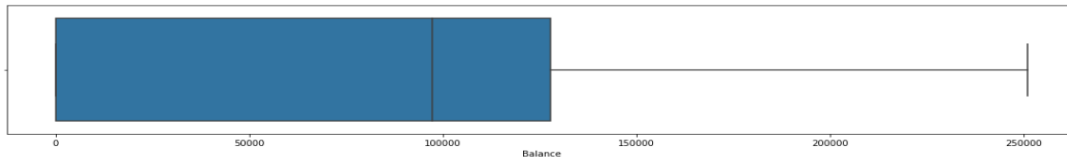
```
[23]: box_scatter(df,'Age','Exited');
        plt.tight_layout()
        print(f"# of Bivariate Outliers: {len(df.loc[df['Age'] > 87])}")
```

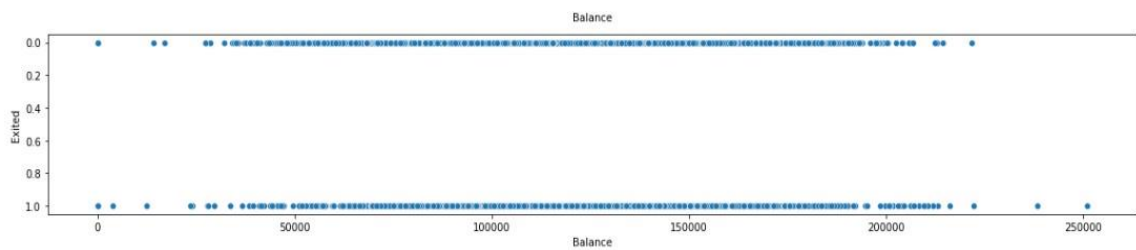
of Bivariate Outliers: 0



```
In [24]: box_scatter(df,'Balance','Exited');
        plt.tight_layout()
        print(f"# of Bivariate Outliers: {len(df.loc[df['Balance'] > 220000])}")
```

of Bivariate Outliers: 4





```
In [25]: from sklearn.preprocessing import LabelEncoder
encoder=LabelEncoder()
for i in df:
    if df[i].dtype=='object' or df[i].dtype=='category':
        df[i]=encoder.fit_transform(df[i])
```

```
In [26]: x=df.iloc[:, :-1]
x.head()
```

```
Out[26]:
```

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	619.0	0	0	42.0	2.0	0.00	1.0	1	1	101348.88
1	608.0	2	0	41.0	1.0	83807.86	1.0	0	1	112542.58
2	502.0	0	0	42.0	8.0	159660.80	3.0	1	0	113931.57
3	699.0	0	0	39.0	1.0	0.00	2.0	0	0	93826.63
4	850.0	2	0	43.0	2.0	125510.82	1.0	1	1	79084.10

```
Out[26]:
```

	CreditScore	Geography	Gender	Age	Tenure	Balance	NumOfProducts	HasCrCard	IsActiveMember	EstimatedSalary
0	619.0	0	0	42.0	2.0	0.00	1.0	1	1	101348.88
1	608.0	2	0	41.0	1.0	83807.86	1.0	0	1	112542.58
2	502.0	0	0	42.0	8.0	159660.80	3.0	1	0	113931.57
3	699.0	0	0	39.0	1.0	0.00	2.0	0	0	93826.63
4	850.0	2	0	43.0	2.0	125510.82	1.0	1	1	79084.10

```
In [30]: y=df.iloc[:, -1]
y.head()
```

```
Out[30]:
```

0	1
1	0
2	1
3	0
4	0

Name: Exited, dtype: int64

```
In [27]: from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
x=scaler.fit_transform(x)
```

```
In [27]: from sklearn.preprocessing import StandardScaler
        scaler=StandardScaler()
        x=scaler.fit_transform(x)
```

```
In [28]: x
```

```
Out[28]: array([[ -0.32687761, -0.90188624, -1.09598752, ...,  0.64609167,
         0.97024255,  0.02188649],
        [ -0.44080365,  1.51506738, -1.09598752, ..., -1.54776799,
         0.97024255,  0.21653375],
        [ -1.53863634, -0.90188624, -1.09598752, ...,  0.64609167,
        -1.03067011,  0.2406869 ],
        ...,
        [  0.60524449, -0.90188624, -1.09598752, ..., -1.54776799,
         0.97024255, -1.00864308],
        [  1.25772996,  0.30659057,  0.91241915, ...,  0.64609167,
        -1.03067011, -0.12523071],
        [  1.46486682, -0.90188624, -1.09598752, ...,  0.64609167,
        -1.03067011, -1.07636976]])
```

```
In [31]: from sklearn.model_selection import train_test_split
        x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33)
```

```
In [32]: x_train.shape
```

```
Out[32]: (6700, 10)
```

```
In [31]: from sklearn.model_selection import train_test_split
        x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.33)
```

```
In [32]: x_train.shape
```

```
Out[32]: (6700, 10)
```

```
In [33]: x_test.shape
```

```
Out[33]: (3300, 10)
```

```
In [34]: y_train.shape
```

```
Out[34]: (6700,)
```

```
In [35]: y_test.shape
```

```
Out[35]: (3300,)
```