

## Importing the libraries

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

## Importing the dataset

```
df =  
pd.read_csv('https://github.com/YBI-Foundation/Dataset/raw/main/Custom  
er%20Purchase.csv')
```

## Get Information of Dataframe

```
df.head()
```

|   | Customer ID | Age | Gender | Education | Review  | Purchased |
|---|-------------|-----|--------|-----------|---------|-----------|
| 0 | 1021        | 30  | Female | School    | Average | No        |
| 1 | 1022        | 68  | Female | UG        | Poor    | No        |
| 2 | 1023        | 70  | Female | PG        | Good    | No        |
| 3 | 1024        | 72  | Female | PG        | Good    | No        |
| 4 | 1025        | 16  | Female | UG        | Average | No        |

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 50 entries, 0 to 49  
Data columns (total 6 columns):  
 #   Column      Non-Null Count  Dtype     
---  --    
 0   Customer ID  50 non-null    int64    
 1   Age          50 non-null    int64    
 2   Gender        50 non-null    object    
 3   Education     50 non-null    object    
 4   Review        50 non-null    object    
 5   Purchased     50 non-null    object    
 dtypes: int64(2), object(4)  
memory usage: 2.5+ KB
```

```
df.isnull().sum()
```

|              | 0 |
|--------------|---|
| Customer ID  | 0 |
| Age          | 0 |
| Gender       | 0 |
| Education    | 0 |
| Review       | 0 |
| Purchased    | 0 |
| dtype: int64 |   |

```
df.columns
```

```
Index(['Customer ID', 'Age', 'Gender', 'Education', 'Review',  
'Purchased'], dtype='object')
```

## Setting X and y

```
X= df[['Age', 'Gender', 'Education', 'Review']]
```

```
y=df['Purchased']
```

```
X[:]
```

|    | Age | Gender | Education | Review  |
|----|-----|--------|-----------|---------|
| 0  | 30  | Female | School    | Average |
| 1  | 68  | Female | UG        | Poor    |
| 2  | 70  | Female | PG        | Good    |
| 3  | 72  | Female | PG        | Good    |
| 4  | 16  | Female | UG        | Average |
| 5  | 31  | Female | School    | Average |
| 6  | 18  | Male   | School    | Good    |
| 7  | 60  | Female | School    | Poor    |
| 8  | 65  | Female | UG        | Average |
| 9  | 74  | Male   | UG        | Good    |
| 10 | 98  | Female | UG        | Good    |
| 11 | 74  | Male   | UG        | Good    |
| 12 | 51  | Male   | School    | Poor    |
| 13 | 57  | Female | School    | Average |
| 14 | 15  | Male   | PG        | Poor    |
| 15 | 75  | Male   | UG        | Poor    |
| 16 | 59  | Male   | UG        | Poor    |
| 17 | 22  | Female | UG        | Poor    |
| 18 | 19  | Male   | School    | Good    |
| 19 | 97  | Male   | PG        | Poor    |
| 20 | 57  | Female | School    | Average |
| 21 | 32  | Male   | PG        | Average |
| 22 | 18  | Female | PG        | Poor    |
| 23 | 96  | Female | School    | Good    |
| 24 | 16  | Female | PG        | Average |
| 25 | 57  | Female | School    | Good    |
| 26 | 53  | Female | PG        | Poor    |
| 27 | 69  | Female | PG        | Poor    |
| 28 | 48  | Male   | School    | Poor    |
| 29 | 83  | Female | UG        | Average |
| 30 | 73  | Male   | UG        | Average |
| 31 | 22  | Female | School    | Poor    |
| 32 | 92  | Male   | UG        | Average |
| 33 | 89  | Female | PG        | Good    |
| 34 | 86  | Male   | School    | Average |
| 35 | 74  | Male   | School    | Poor    |
| 36 | 34  | Female | UG        | Good    |

|    |    |        |        |         |
|----|----|--------|--------|---------|
| 37 | 94 | Male   | PG     | Average |
| 38 | 45 | Female | School | Good    |
| 39 | 76 | Male   | PG     | Poor    |
| 40 | 39 | Male   | School | Good    |
| 41 | 23 | Male   | PG     | Good    |
| 42 | 30 | Female | PG     | Good    |
| 43 | 27 | Male   | PG     | Poor    |
| 44 | 77 | Female | UG     | Average |
| 45 | 61 | Male   | PG     | Poor    |
| 46 | 64 | Female | PG     | Poor    |
| 47 | 38 | Female | PG     | Good    |
| 48 | 39 | Female | UG     | Good    |
| 49 | 25 | Female | UG     | Good    |

## Ordinal Encoder for rank categorical data

```
from sklearn.preprocessing import OrdinalEncoder
oe=OrdinalEncoder()
X[['Gender', 'Education', 'Review']] = oe.fit_transform(X[['Gender',
'Education', 'Review']])
#X[:,1:] = oe.fit_transform(X[:,1:])
X
```

/usr/local/lib/python3.7/dist-packages/pandas/core/frame.py:3678:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row\_indexer,col\_indexer] = value instead

See the caveats in the documentation:

[https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
    self[col] = igetitem(value, i)
```

|    | Age | Gender | Education | Review |
|----|-----|--------|-----------|--------|
| 0  | 30  | 0.0    | 1.0       | 0.0    |
| 1  | 68  | 0.0    | 2.0       | 2.0    |
| 2  | 70  | 0.0    | 0.0       | 1.0    |
| 3  | 72  | 0.0    | 0.0       | 1.0    |
| 4  | 16  | 0.0    | 2.0       | 0.0    |
| 5  | 31  | 0.0    | 1.0       | 0.0    |
| 6  | 18  | 1.0    | 1.0       | 1.0    |
| 7  | 60  | 0.0    | 1.0       | 2.0    |
| 8  | 65  | 0.0    | 2.0       | 0.0    |
| 9  | 74  | 1.0    | 2.0       | 1.0    |
| 10 | 98  | 0.0    | 2.0       | 1.0    |
| 11 | 74  | 1.0    | 2.0       | 1.0    |
| 12 | 51  | 1.0    | 1.0       | 2.0    |
| 13 | 57  | 0.0    | 1.0       | 0.0    |
| 14 | 15  | 1.0    | 0.0       | 2.0    |
| 15 | 75  | 1.0    | 2.0       | 2.0    |

```

16   59    1.0    2.0    2.0
17   22    0.0    2.0    2.0
18   19    1.0    1.0    1.0
19   97    1.0    0.0    2.0
20   57    0.0    1.0    0.0
21   32    1.0    0.0    0.0
22   18    0.0    0.0    2.0
23   96    0.0    1.0    1.0
24   16    0.0    0.0    0.0
25   57    0.0    1.0    1.0
26   53    0.0    0.0    2.0
27   69    0.0    0.0    2.0
28   48    1.0    1.0    2.0
29   83    0.0    2.0    0.0
30   73    1.0    2.0    0.0
31   22    0.0    1.0    2.0
32   92    1.0    2.0    0.0
33   89    0.0    0.0    1.0
34   86    1.0    1.0    0.0
35   74    1.0    1.0    2.0
36   34    0.0    2.0    1.0
37   94    1.0    0.0    0.0
38   45    0.0    1.0    1.0
39   76    1.0    0.0    2.0
40   39    1.0    1.0    1.0
41   23    1.0    0.0    1.0
42   30    0.0    0.0    1.0
43   27    1.0    0.0    2.0
44   77    0.0    2.0    0.0
45   61    1.0    0.0    2.0
46   64    0.0    0.0    2.0
47   38    0.0    0.0    1.0
48   39    0.0    2.0    1.0
49   25    0.0    2.0    1.0

```

## Display Encoded Categories

```
oe.categories_
```

```
[array(['Female', 'Male'], dtype=object),
 array(['PG', 'School', 'UG'], dtype=object),
 array(['Average', 'Good', 'Poor'], dtype=object)]
```

## Reverse Encode Integer array to Categorical Features

```
oe.inverse_transform([[0,0,0]])
```

```
array([[15, 'Female', 'PG', 'Average']], dtype=object)
```

```
oe.inverse_transform([[1,1,1]])
```

```
array([['Male', 'School', 'Good']], dtype=object)
oe.inverse_transform([[1,2,2]])
array([['Male', 'UG', 'Poor']], dtype=object)
```

## Encoding Categorical features as an integer array of your choice

```
X=df[['Gender', 'Education', 'Review']]
oe=OrdinalEncoder(categories=[['Male','Female'],['School','UG','PG'],
['Poor','Average','Good']])
X=oe.fit_transform(X)
oe.categories_
[array(['Male', 'Female'], dtype=object),
 array(['School', 'UG', 'PG'], dtype=object),
 array(['Poor', 'Average', 'Good'], dtype=object)]
```