```
In [ ]: #Data.csv
```

Step 1: Importing the libraries

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

Step 2: Importing dataset

```
In [2]: data = pd.read_csv('Data.csv')
   data
```

```
Out[2]:
             Country Age
                           Salary Purchased
         0
                      44.0 72000.0
              France
                                           No
         1
               Spain
                      27.0 48000.0
                                           Yes
         2 Germany
                      30.0 54000.0
                                           No
         3
               Spain
                      38.0 61000.0
                                           No
         4 Germany
                      40.0
                              NaN
                                           Yes
                      35.0 58000.0
              France
                                           Yes
                           52000.0
               Spain
                      NaN
                                           No
               France
                      48.0
                           79000.0
                                           Yes
         8 Germany
                      50.0 83000.0
                                           No
               France
                     37.0 67000.0
                                           Yes
```

Step 3: Handling the missing data

print(X)

```
In [3]: X = data.iloc[:, :-1].values
        y = data.iloc[:, 3].values
        print("X = ", X, '\n')
        print("y = ", y, '\n')
        X = [['France' 44.0 72000.0]]
         ['Spain' 27.0 48000.0]
         ['Germany' 30.0 54000.0]
         ['Spain' 38.0 61000.0]
         ['Germany' 40.0 nan]
         ['France' 35.0 58000.0]
         ['Spain' nan 52000.0]
         ['France' 48.0 79000.0]
         ['Germany' 50.0 83000.0]
         ['France' 37.0 67000.0]]
        y = ['No' 'Yes' 'No' 'No' 'Yes' 'Yes' 'No' 'Yes' 'No' 'Yes']
In [5]: from sklearn.impute import SimpleImputer
        imputer = SimpleImputer(missing_values = np.nan, strategy = 'mean')
        imputer = imputer.fit(X[:, 1:3])
        X[:, 1:3] = imputer.transform(X[:, 1:3])
```

```
[['France' 44.0 72000.0]
['Spain' 27.0 48000.0]
['Germany' 30.0 54000.0]
['Spain' 38.0 61000.0]
['Germany' 40.0 63777.77777777778]
['France' 35.0 58000.0]
['Spain' 38.77777777777778 52000.0]
['France' 48.0 79000.0]
['Germany' 50.0 83000.0]
['France' 37.0 67000.0]]
```

Step 4: Encoding categorical data

```
In [9]: from sklearn.preprocessing import LabelEncoder, OneHotEncoder
        from sklearn.compose import ColumnTransformer
        labelencoder_X = LabelEncoder()
        X[:, 0] = labelencoder_X.fit_transform(X[:, 0])
        #onehotencoder = OneHotEncoder(categorical_features = [0])
        ct = ColumnTransformer([("Country", OneHotEncoder(), [1])], remainder = 'passthrough')
        X = ct.fit_transform(X).toarray()
        print('X = ', X)
        labelencoder_y = LabelEncoder()
        y = labelencoder_y.fit_transform(y)
        print('y = ', y)
        X = \begin{bmatrix} 0.000000000e+00 & 0.00000000e+00 & 0.00000000e+00 & 0.00000000e+00 \end{bmatrix}
          0.0000000e+00 0.0000000e+00 0.0000000e+00 1.0000000e+00
          0.00000000e+00 0.00000000e+00 0.00000000e+00 7.20000000e+04]
         [1.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.0000000e+00 0.0000000e+00 0.0000000e+00 0.0000000e+00
          0.00000000e+00 0.00000000e+00 2.00000000e+00 4.80000000e+04]
         [0.00000000e+00 1.00000000e+00 0.00000000e+00 0.00000000e+00
          0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.00000000e+00 0.00000000e+00 1.00000000e+00 5.40000000e+04]
         [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          1.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.00000000e+00 0.00000000e+00 2.00000000e+00 6.10000000e+04]
         [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.0000000e+00 0.00000000e+00 1.00000000e+00 0.00000000e+00
          0.0000000e+00 0.0000000e+00 1.0000000e+00 6.3777778e+04]
         [0.00000000e+00 0.00000000e+00 1.00000000e+00 0.00000000e+00
          0.0000000e+00 0.0000000e+00 0.0000000e+00 0.0000000e+00
          0.0000000e+00 0.0000000e+00 0.0000000e+00 5.80000000e+04]
         [0.00000000e+00 0.00000000e+00 0.00000000e+00 0.00000000e+00
          0.00000000e+00 1.00000000e+00 0.0000000e+00 0.00000000e+00
          0.0000000e+00 0.0000000e+00 2.0000000e+00 5.20000000e+04]
         [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          1.00000000e+00 0.00000000e+00 0.00000000e+00 7.90000000e+04]
         [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
          0.00000000e+00 1.00000000e+00 1.00000000e+00 8.30000000e+04]
         [0.00000000e+00 0.00000000e+00 0.0000000e+00 1.00000000e+00
          0.0000000e+00 0.0000000e+00 0.0000000e+00 0.0000000e+00
          0.00000000e+00 0.0000000e+00 0.0000000e+00 6.70000000e+04]]
        y = [0100110101]
```

Step 5: Creating a dummy variable

Step 6: Splitting the datasets into training sets and Test sets

```
In [11]: from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)
         print("X train = ", X_train, "\n")
         print("X test = ", X_test, "\n")
         X train = [[0.00000000e+00 0.00000000e+00 0.00000000e+00 0.00000000e+00
           0.0000000e+00 0.0000000e+00 1.0000000e+00 0.0000000e+00
           0.00000000e+00 0.00000000e+00 1.00000000e+00 6.37777778e+04]
          [0.00000000e+00 0.00000000e+00 0.0000000e+00 1.00000000e+00
           0.00000000e+00 0.00000000e+00 0.00000000e+00 0.00000000e+00
           0.00000000e+00 0.00000000e+00 0.00000000e+00 6.70000000e+04]
          [1.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.00000000e+00 0.00000000e+00 2.00000000e+00 4.80000000e+04]
          [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.00000000e+00 1.00000000e+00 0.0000000e+00 0.00000000e+00
           0.0000000e+00 0.0000000e+00 2.0000000e+00 5.20000000e+04]
          [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           1.00000000e+00 0.00000000e+00 0.00000000e+00 7.90000000e+04]
          [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           1.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.0000000e+00 0.0000000e+00 2.0000000e+00 6.10000000e+04]
          [0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.00000000e+00 0.00000000e+00 0.0000000e+00 1.00000000e+00
           0.00000000e+00 0.00000000e+00 0.00000000e+00 7.20000000e+04]
          [0.00000000e+00 0.00000000e+00 1.00000000e+00 0.00000000e+00
           0.00000000e+00 0.00000000e+00 0.0000000e+00 0.00000000e+00
           0.0000000e+00 0.0000000e+00 0.0000000e+00 5.80000000e+04]]
         X test = [[0.0e+00 1.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00
           0.0e+00 1.0e+00 5.4e+04]
          [0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00 0.0e+00
           1.0e+00 1.0e+00 8.3e+04]]
```

Step 7: Feature Scaling

```
In [12]: from sklearn.preprocessing import StandardScaler
    sc_X = StandardScaler()
    X_train = sc_X.fit_transform(X_train)
    X_test = sc_X.transform(X_test)
    print("X train = ", X_train)
```

```
X train = [[-0.37796447 0. -0.37796447 -0.37796447 -0.37796447 -0.37796447
  2.64575131 -0.37796447 -0.37796447 0. 0.13483997 0.12381479]
 [-0.37796447 0. -0.37796447 2.64575131 -0.37796447 -0.37796447
 -0.37796447 -0.37796447 -0.37796447 0. -0.94387981 0.46175632]
                -0.37796447 -0.37796447 -0.37796447 -0.37796447
 [ 2.64575131 0.
 -0.37796447 -0.37796447 -0.37796447 0.
                                           1.21355975 -1.53093341]
 [-0.37796447 0. -0.37796447 -0.37796447 2.64575131
 -0.37796447 -0.37796447 -0.37796447 0.
                                          1.21355975 -1.11141978]
 [-0.37796447 0. -0.37796447 -0.37796447 -0.37796447 -0.37796447
 -0.37796447 -0.37796447 2.64575131 0. -0.94387981 1.7202972 ]
 [-0.37796447 0.
                 -0.37796447 -0.37796447 2.64575131 -0.37796447
 -0.37796447 -0.37796447 -0.37796447 0. 1.21355975 -0.16751412]
                 -0.37796447 -0.37796447 -0.37796447 -0.37796447
 [-0.37796447 0.
 -0.37796447 2.64575131 -0.37796447 0. -0.94387981 0.98614835]
 [-0.37796447 0. 2.64575131 -0.37796447 -0.37796447 -0.37796447
  -0.37796447 -0.37796447 -0.37796447 0. -0.94387981 -0.48214934]]
```

In []: