# Data Analytics II - Logistic Regression

```
import pandas as pd
In [1]:
         import numpy as np
         import matplotlib.pyplot as plt
         %matplotlib inline
         df = pd.read_csv('Social_Network_Ads.csv')
In [2]:
                User ID
                        Gender
                                Age
                                     EstimatedSalary Purchased
Out[2]:
                                                             0
             15624510
                          Male
                                 19
                                              19000
           1 15810944
                                                             0
                          Male
                                 35
                                              20000
           2 15668575
                        Female
                                 26
                                              43000
                                                             0
           3 15603246
                        Female
                                 27
                                              57000
                                                             0
           4 15804002
                                 19
                                              76000
                                                             0
                          Male
              15691863
                        Female
                                              41000
         395
                                 46
                                                             1
         396
             15706071
                          Male
                                 51
                                              23000
                                                             1
              15654296
                        Female
                                 50
                                              20000
                                                             1
                                              33000
                                                             0
         398
             15755018
                          Male
                                 36
         399 15594041 Female
                                 49
                                              36000
                                                             1
        400 rows × 5 columns
         df.hist()
In [3]:
         array([[<AxesSubplot:title={'center':'User ID'}>,
Out[3]:
                  <AxesSubplot:title={'center':'Age'}>],
                 [<AxesSubplot:title={'center':'EstimatedSalary'}>,
                  <AxesSubplot:title={'center':'Purchased'}>]], dtype=object)
                    User ID
                                      75
          40
                                      50
          20
                                      25
          0
                                      0
              156
EstimatedSalary 158
                                                            60
                                              Purchased
          75
                                     200
          50
                                    100
          25
                       100000
                50000
                              150000
                                        0.0
                                                  0.5
                                                           1.0
         df.isnull().sum()
In [4]:
```

User ID

Gender

Age

Out[4]:

0

0

0

```
In [5]:
          df.dropna()
                         Gender Age
                                     EstimatedSalary Purchased
 Out[5]:
                 User ID
            0 15624510
                                                             0
                                  19
                                               19000
                           Male
            1 15810944
                           Male
                                  35
                                               20000
                                                             0
                                                             0
            2 15668575
                         Female
                                  26
                                               43000
            3 15603246
                                                             0
                         Female
                                  27
                                               57000
               15804002
                           Male
                                  19
                                               76000
                                                             0
          395
              15691863
                                               41000
                         Female
                                  46
                                                             1
          396 15706071
                           Male
                                  51
                                               23000
                                                             1
          397 15654296
                                                             1
                         Female
                                  50
                                               20000
          398 15755018
                           Male
                                  36
                                               33000
                                                             0
          399 15594041 Female
                                  49
                                               36000
                                                             1
         400 rows × 5 columns
          df['EstimatedSalary'].plot.box() ##whisker boxplot min median max Q1 Q2 Q3
 In [6]:
          <AxesSubplot:>
 Out[6]:
          140000
          120000
          100000
           80000
           60000
           40000
           20000
                                    EstimatedSalary
          df.drop('Gender', axis=1, inplace=True)
 In [7]:
 In [ ]:
          df.drop('User ID',axis=1,inplace=True)
 In [8]:
          x = df.drop('Purchased',axis=1) #drop column
 In [9]:
          y = df['Purchased']
In [10]:
Out[10]:
               Age
                   EstimatedSalary
            0
                19
                             19000
```

EstimatedSalary

Purchased dtype: int64

0

1	35	20000
2	26	43000
3	27	57000
4	19	76000
395	46	41000
396	51	23000
397	50	20000
398	36	33000
399	49	36000

400 rows × 2 columns

```
In [11]:
                 0
Out[11]:
                 0
                 0
                 0
          395
                 1
          396
                  1
          397
                 1
          398
          399
                  1
          Name: Purchased, Length: 400, dtype: int64
In [12]:
          x.shape
          (400, 2)
Out[12]:
In [13]:
          y.shape
          (400,)
Out[13]:
```

## Train Test Split

```
In [14]: from sklearn.model_selection import train_test_split
    x_train, x_test, Y_train, Y_test = train_test_split(x,y,test_size=0.25,random_state=0)

In [15]: x_train.shape

Out[15]: (300, 2)

In [16]: x_test.shape

Out[16]: (100, 2)
```

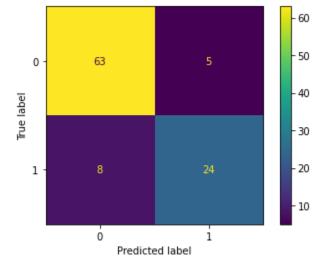
## Feature Scaling

In [17]: **from** sklearn.preprocessing **import** StandardScaler

```
In [19]: x_train = std.fit_transform(x_train)
         x_test = std.fit_transform(x_test)
         Model Training
In [20]:
         from sklearn.linear_model import LogisticRegression
         model = LogisticRegression()
         model.fit(x_train,Y_train)
         LogisticRegression()
Out[20]:
         y_predict = model.predict(x_test)
In [21]:
         y_predict
In [22]:
         array([0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
Out[22]:
                0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                1, 0, 0, 1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
                0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1,
                0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 1])
         Performance Evaluation
In [23]:
         from sklearn.metrics import accuracy_score,precision_score,recall_score
In [24]:
         a_score = accuracy_score(Y_test,y_predict)
         a_score
         0.87
Out[24]:
In [25]:
         pre_score = precision_score(Y_test,y_predict)
         pre_score
         0.8275862068965517
Out[25]:
In [26]:
         rec_score= recall_score(Y_test,y_predict)
         rec_score
         0.75
Out[26]:
         Confusion Matrix
         from sklearn.metrics import confusion_matrix,ConfusionMatrixDisplay
In [27]:
In [28]:
         cm = confusion_matrix(Y_test,y_predict)
In [29]:
         cm
         array([[63, 5],
Out[29]:
                [ 8, 24]])
         cm_d = ConfusionMatrixDisplay(cm).plot()
In [30]:
```

std = StandardScaler()

In [18]:



#### Predict over new value

```
In [31]: p_value = [(46,41000)]
    new_predict = model.predict(p_value)

In [32]: new_predict

Out[32]: array([1])

In [51]: # for in range(len(y_predict)):
    # print(Y_test[i], y_predict[i])

In []:

In [59]: # for i,p in y_predict:
    # print(Y_test[i], y_predict[p])
```

## **Classfication Report**

```
from sklearn.metrics import classification_report
In [42]:
          report = classification_report(Y_test,y_predict)
         print(report)
In [43]:
                        precision
                                      recall f1-score
                                                          support
                     0
                             0.89
                                        0.93
                                                   0.91
                                                               68
                             0.83
                                        0.75
                                                   0.79
                                                               32
                     1
                                                   0.87
                                                              100
             accuracy
                                                   0.85
             macro avg
                             0.86
                                        0.84
                                                              100
         weighted avg
                             0.87
                                        0.87
                                                   0.87
                                                              100
```

```
In [ ]:
```