```
In [82]: import pandas as pd
        import numpy as np
        from sklearn.model_selection import train_test_split
        from sklearn.linear_model import LogisticRegression
        from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score
In [83]: df = pd.read_csv("Social_Network_Ads.csv")
        df
              User ID Gender Age EstimatedSalary Purchased
                                                  0
          0 15624510 Male 19
                                      19000
          1 15810944 Male 35
                                      20000
         2 15668575 Female 26
                                      43000
                                                  0
         3 15603246 Female 27
                                      57000
          4 15804002 Male 19
                                      76000
                       ... ...
        395 15691863 Female 46
                                      41000
        396 15706071 Male 51
                                      23000
        397 15654296 Female 50
                                      20000
        398 15755018 Male 36
                                      33000
        399 15594041 Female 49
                                      36000
        400 rows × 5 columns
In [84]: df["Gender"] = df["Gender"].replace({"Male": 0, "Female": 1})
        df
       C:\Users\Ritesh Kolte\AppData\Local\Temp\ipykernel_19336\3117893451.py:1: FutureWarning: Downcasting behavior, explicitly call `result.infer_objects(copy=False)`. To opt
       -in to the future behavior, set `pd.set_option('future.no_silent_downcasting', True)`
        df["Gender"] = df["Gender"].replace({"Male": 0, "Female": 1})
Out[84]:
              User ID Gender Age EstimatedSalary Purchased
          0 15624510
                        0 19
                                      19000
          1 15810944
                        0 35
                                      20000
          2 15668575
                        1 26
                                      43000
         3 15603246
                        1 27
                                      57000
          4 15804002
                        0 19
                                      76000
        395 15691863
                        1 46
                                      41000
        396 15706071
                        0 51
                                      23000
        397 15654296
                        1 50
                                      20000
        398 15755018
                        0 36
                                      33000
        399 15594041
        400 rows × 5 columns
In [85]: df.columns
Out[85]: Index(['User ID', 'Gender', 'Age', 'EstimatedSalary', 'Purchased'], dtype='object')
In [86]: x = df[['User ID', 'Gender', 'Age', 'EstimatedSalary']]
        y = df['Purchased']
In [87]: x
Out[87]:
              User ID Gender Age EstimatedSalary
          0 15624510
                        0 19
                                      19000
          1 15810944
                        0 35
                                      20000
          2 15668575
                        1 26
                                      43000
         3 15603246
                                      57000
                        1 27
          4 15804002
                        0 19
                                      76000
        395 15691863
                        1 46
                                      41000
        396 15706071
                        0 51
                                      23000
        397 15654296
                        1 50
                                      20000
        398 15755018
                        0 36
                                      33000
        399 15594041
                        1 49
                                      36000
        400 rows × 4 columns
In [88]: y
Out[88]: 0
        395
        396
        397
        398
        399 1
        Name: Purchased, Length: 400, dtype: int64
In [89]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.25,random_state=29)
In [90]: model = LogisticRegression()
        model.fit(x_train,y_train)
Out[90]
         ▼ LogisticRegression 🤍
        LogisticRegression()
In [91]: y_pred = model.predict(x_test)
In [92]: y_pred
Out[92]: array([0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
               0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0,
               0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
               1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1])
In [93]: model.score(x_train,y_train)
Out[93]: 0.86333333333333333
In [94]: model.score(x,y)
Out[94]: 0.85
In [96]: cm = confusion_matrix(y_test, y_pred)
Out[96]: array([[63, 6],
               [13, 18]])
In [99]: tn, fp, fn, tp = confusion_matrix(y_test, y_pred).ravel()
In [100... print(tn,fp,fn,tp)
       63 6 13 18
In [101... a = accuracy_score(y_test,y_pred)
Out[101... 0.81
In [103... e = 1 - a #error value
Out[103... 0.1899999999999999
In [104... precision_score(y_test,y_pred)
Out[104... 0.75
```

In [105... recall\_score(y\_test,y\_pred)

Out[105... 0.5806451612903226

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