

In []: Name: Akash Varade
Roll No: A-04

In [1]: `import pandas as pd`
`import numpy as np`
`from matplotlib import pyplot as plt`
`%matplotlib inline`

In [2]: `df = pd.read_csv("/home/kj-comp/Akash Varade/GCR/DB/Social_Network_Ads(1).csv")`
`df.head(10)`

Out[2]:

	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0
5	15728773	Male	27	58000	0
6	15598044	Female	27	84000	0
7	15694829	Female	32	150000	1
8	15600575	Male	25	33000	0
9	15727311	Female	35	65000	0

In [3]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   User ID               400 non-null   int64
1   Gender                400 non-null   object
2   Age                   400 non-null   int64
3   EstimatedSalary       400 non-null   int64
4   Purchased             400 non-null   int64
dtypes: int64(4), object(1)
memory usage: 15.8+ KB
```

In [4]: `df.describe()`

Out[4]:

	User ID	Age	EstimatedSalary	Purchased
count	4.000000e+02	400.000000	400.000000	400.000000
mean	1.569154e+07	37.655000	69742.500000	0.357500
std	7.165832e+04	10.482877	34096.960282	0.479864
min	1.556669e+07	18.000000	15000.000000	0.000000
25%	1.562676e+07	29.750000	43000.000000	0.000000
50%	1.569434e+07	37.000000	70000.000000	0.000000
75%	1.575036e+07	46.000000	88000.000000	1.000000
max	1.581524e+07	60.000000	150000.000000	1.000000

```
In [5]: X = df.iloc[:,[2,3]].values
        y = df.iloc[:,4].values
```

```
In [6]: X
```

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```

In [7]: y

```
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```

```
In [10]: 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,rand
```

```
In [11]: from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X_train = sc.fit_transform(X_train)
X_test = sc.transform(X_test)
```

```
In [12]: X_train
```

```
Out[12]: array([[ 0.58164944, -0.88670699],
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```

```
In [13]: from sklearn.linear_model import LogisticRegression
classifier = LogisticRegression(random_state=0)
classifier.fit(X_train,y_train)
```

```
Out[13]: LogisticRegression(random_state=0)
```

```
In [14]: y_pred = classifier.predict(X_test)
```

```
In [15]: y_pred
```

```
Out[15]: array([0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1,
                0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0,
                1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1,
                0, 0, 0, 0, 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])
```

```
In [16]: from sklearn.metrics import confusion_matrix,classification_report
cm = confusion_matrix(y_test , y_pred)
```

```
In [17]: cm
```

```
Out[17]: array([[65,  3],
                [ 8, 24]])
```

```
In [18]: c1_report = classification_report(y_test,y_pred)
```

```
In [19]: c1_report
```

```
Out[19]: '          precision    recall  f1-score   support\n\n      0          0.96      0.92      0.94         68\n      1          0.89      0.89      0.89         10\n\n   accuracy: 0.89      100\n   macro avg: 0.89      0.89      0.89      100\n   weighted avg: 0.89      0.89      0.89      100'
```

```
In [20]: tp , fn ,fp , tn = confusion_matrix(y_test,y_pred,labels=[0,1]).reshape(-1)
print('Outcome values : \n' , tp , fn , fp ,tn)
```

```
Outcome values :
65 3 8 24
```

```
In [21]: accuracy_cm = (tp+tn)/(tp+fp+tn+fn)
precision_cm = tp/(tp+fp)
recall_cm = tp/(tp+fn)
f1_score = 2/((1/recall_cm)+(1/precision_cm))
```

```
In [22]: print("Accuracy :",accuracy_cm)
print("Precision :",precision_cm)
print("Recall :",recall_cm)
print("F1-Score :",f1_score)
```

```
Accuracy : 0.89
Precision : 0.8904109589041096
Recall : 0.9558823529411765
F1-Score : 0.9219858156028368
```

```
In [ ]:
```