

```
In [1]: import numpy as np
import pandas as pd
df=pd.read_csv('Social_Network_Ads.csv')
df
```

```
Out[1]:   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
...
395   15691863  Female    46        41000          1
396   15706071    Male    51        23000          1
397   15654296  Female    50        20000          1
398   15755018    Male    36        33000          0
399   15594041  Female    49        36000          1
```

400 rows × 5 columns

```
In [2]: df.head()
```

```
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
```

```
In [3]: features=df.iloc[:,[2,3]].values
label=df.iloc[:,4].values
features
```

```
Out[3]: array([[ 19, 19000],  
 [ 35, 20000],  
 [ 26, 43000],  
 [ 27, 57000],  
 [ 19, 76000],  
 [ 27, 58000],  
 [ 27, 84000],  
 [ 32, 150000],  
 [ 25, 33000],  
 [ 35, 65000],  
 [ 26, 80000],  
 [ 26, 52000],  
 [ 20, 86000],  
 [ 32, 18000],  
 [ 18, 82000],  
 [ 29, 80000],  
 [ 47, 25000],  
 [ 45, 26000],  
 [ 46, 28000],  
 [ 48, 29000],  
 [ 45, 22000],  
 [ 47, 49000],  
 [ 48, 41000],  
 [ 45, 22000],  
 [ 46, 23000],  
 [ 47, 20000],  
 [ 49, 28000],  
 [ 47, 30000],  
 [ 29, 43000],  
 [ 31, 18000],  
 [ 31, 74000],  
 [ 27, 137000],  
 [ 21, 16000],  
 [ 28, 44000],  
 [ 27, 90000],  
 [ 35, 27000],  
 [ 33, 28000],  
 [ 30, 49000],  
 [ 26, 72000],  
 [ 27, 31000],  
 [ 27, 17000],  
 [ 33, 51000],  
 [ 35, 108000],  
 [ 30, 15000],  
 [ 28, 84000],  
 [ 23, 20000],  
 [ 25, 79000],  
 [ 27, 54000],  
 [ 30, 135000],  
 [ 31, 89000],  
 [ 24, 32000],  
 [ 18, 44000],  
 [ 29, 83000],  
 [ 35, 23000],  
 [ 27, 58000],  
 [ 24, 55000],
```

```
[ 23, 48000],  
[ 28, 79000],  
[ 22, 18000],  
[ 32, 117000],  
[ 27, 20000],  
[ 25, 87000],  
[ 23, 66000],  
[ 32, 120000],  
[ 59, 83000],  
[ 24, 58000],  
[ 24, 19000],  
[ 23, 82000],  
[ 22, 63000],  
[ 31, 68000],  
[ 25, 80000],  
[ 24, 27000],  
[ 20, 23000],  
[ 33, 113000],  
[ 32, 18000],  
[ 34, 112000],  
[ 18, 52000],  
[ 22, 27000],  
[ 28, 87000],  
[ 26, 17000],  
[ 30, 80000],  
[ 39, 42000],  
[ 20, 49000],  
[ 35, 88000],  
[ 30, 62000],  
[ 31, 118000],  
[ 24, 55000],  
[ 28, 85000],  
[ 26, 81000],  
[ 35, 50000],  
[ 22, 81000],  
[ 30, 116000],  
[ 26, 15000],  
[ 29, 28000],  
[ 29, 83000],  
[ 35, 44000],  
[ 35, 25000],  
[ 28, 123000],  
[ 35, 73000],  
[ 28, 37000],  
[ 27, 88000],  
[ 28, 59000],  
[ 32, 86000],  
[ 33, 149000],  
[ 19, 21000],  
[ 21, 72000],  
[ 26, 35000],  
[ 27, 89000],  
[ 26, 86000],  
[ 38, 80000],  
[ 39, 71000],  
[ 37, 71000],
```

```
[ 38, 61000],  
[ 37, 55000],  
[ 42, 80000],  
[ 40, 57000],  
[ 35, 75000],  
[ 36, 52000],  
[ 40, 59000],  
[ 41, 59000],  
[ 36, 75000],  
[ 37, 72000],  
[ 40, 75000],  
[ 35, 53000],  
[ 41, 51000],  
[ 39, 61000],  
[ 42, 65000],  
[ 26, 32000],  
[ 30, 17000],  
[ 26, 84000],  
[ 31, 58000],  
[ 33, 31000],  
[ 30, 87000],  
[ 21, 68000],  
[ 28, 55000],  
[ 23, 63000],  
[ 20, 82000],  
[ 30, 107000],  
[ 28, 59000],  
[ 19, 25000],  
[ 19, 85000],  
[ 18, 68000],  
[ 35, 59000],  
[ 30, 89000],  
[ 34, 25000],  
[ 24, 89000],  
[ 27, 96000],  
[ 41, 30000],  
[ 29, 61000],  
[ 20, 74000],  
[ 26, 15000],  
[ 41, 45000],  
[ 31, 76000],  
[ 36, 50000],  
[ 40, 47000],  
[ 31, 15000],  
[ 46, 59000],  
[ 29, 75000],  
[ 26, 30000],  
[ 32, 135000],  
[ 32, 100000],  
[ 25, 90000],  
[ 37, 33000],  
[ 35, 38000],  
[ 33, 69000],  
[ 18, 86000],  
[ 22, 55000],  
[ 35, 71000],
```

```
[ 29, 148000],  
[ 29, 47000],  
[ 21, 88000],  
[ 34, 115000],  
[ 26, 118000],  
[ 34, 43000],  
[ 34, 72000],  
[ 23, 28000],  
[ 35, 47000],  
[ 25, 22000],  
[ 24, 23000],  
[ 31, 34000],  
[ 26, 16000],  
[ 31, 71000],  
[ 32, 117000],  
[ 33, 43000],  
[ 33, 60000],  
[ 31, 66000],  
[ 20, 82000],  
[ 33, 41000],  
[ 35, 72000],  
[ 28, 32000],  
[ 24, 84000],  
[ 19, 26000],  
[ 29, 43000],  
[ 19, 70000],  
[ 28, 89000],  
[ 34, 43000],  
[ 30, 79000],  
[ 20, 36000],  
[ 26, 80000],  
[ 35, 22000],  
[ 35, 39000],  
[ 49, 74000],  
[ 39, 134000],  
[ 41, 71000],  
[ 58, 101000],  
[ 47, 47000],  
[ 55, 130000],  
[ 52, 114000],  
[ 40, 142000],  
[ 46, 22000],  
[ 48, 96000],  
[ 52, 150000],  
[ 59, 42000],  
[ 35, 58000],  
[ 47, 43000],  
[ 60, 108000],  
[ 49, 65000],  
[ 40, 78000],  
[ 46, 96000],  
[ 59, 143000],  
[ 41, 80000],  
[ 35, 91000],  
[ 37, 144000],  
[ 60, 102000],
```

```
[ 35, 60000],  
[ 37, 53000],  
[ 36, 126000],  
[ 56, 133000],  
[ 40, 72000],  
[ 42, 80000],  
[ 35, 147000],  
[ 39, 42000],  
[ 40, 107000],  
[ 49, 86000],  
[ 38, 112000],  
[ 46, 79000],  
[ 40, 57000],  
[ 37, 80000],  
[ 46, 82000],  
[ 53, 143000],  
[ 42, 149000],  
[ 38, 59000],  
[ 50, 88000],  
[ 56, 104000],  
[ 41, 72000],  
[ 51, 146000],  
[ 35, 50000],  
[ 57, 122000],  
[ 41, 52000],  
[ 35, 97000],  
[ 44, 39000],  
[ 37, 52000],  
[ 48, 134000],  
[ 37, 146000],  
[ 50, 44000],  
[ 52, 90000],  
[ 41, 72000],  
[ 40, 57000],  
[ 58, 95000],  
[ 45, 131000],  
[ 35, 77000],  
[ 36, 144000],  
[ 55, 125000],  
[ 35, 72000],  
[ 48, 90000],  
[ 42, 108000],  
[ 40, 75000],  
[ 37, 74000],  
[ 47, 144000],  
[ 40, 61000],  
[ 43, 133000],  
[ 59, 76000],  
[ 60, 42000],  
[ 39, 106000],  
[ 57, 26000],  
[ 57, 74000],  
[ 38, 71000],  
[ 49, 88000],  
[ 52, 38000],  
[ 50, 36000],
```

```
[ 59, 88000],  
[ 35, 61000],  
[ 37, 70000],  
[ 52, 21000],  
[ 48, 141000],  
[ 37, 93000],  
[ 37, 62000],  
[ 48, 138000],  
[ 41, 79000],  
[ 37, 78000],  
[ 39, 134000],  
[ 49, 89000],  
[ 55, 39000],  
[ 37, 77000],  
[ 35, 57000],  
[ 36, 63000],  
[ 42, 73000],  
[ 43, 112000],  
[ 45, 79000],  
[ 46, 117000],  
[ 58, 38000],  
[ 48, 74000],  
[ 37, 137000],  
[ 37, 79000],  
[ 40, 60000],  
[ 42, 54000],  
[ 51, 134000],  
[ 47, 113000],  
[ 36, 125000],  
[ 38, 50000],  
[ 42, 70000],  
[ 39, 96000],  
[ 38, 50000],  
[ 49, 141000],  
[ 39, 79000],  
[ 39, 75000],  
[ 54, 104000],  
[ 35, 55000],  
[ 45, 32000],  
[ 36, 60000],  
[ 52, 138000],  
[ 53, 82000],  
[ 41, 52000],  
[ 48, 30000],  
[ 48, 131000],  
[ 41, 60000],  
[ 41, 72000],  
[ 42, 75000],  
[ 36, 118000],  
[ 47, 107000],  
[ 38, 51000],  
[ 48, 119000],  
[ 42, 65000],  
[ 40, 65000],  
[ 57, 60000],  
[ 36, 54000],
```

```
[ 58, 144000],  
[ 35, 79000],  
[ 38, 55000],  
[ 39, 122000],  
[ 53, 104000],  
[ 35, 75000],  
[ 38, 65000],  
[ 47, 51000],  
[ 47, 105000],  
[ 41, 63000],  
[ 53, 72000],  
[ 54, 108000],  
[ 39, 77000],  
[ 38, 61000],  
[ 38, 113000],  
[ 37, 75000],  
[ 42, 90000],  
[ 37, 57000],  
[ 36, 99000],  
[ 60, 34000],  
[ 54, 70000],  
[ 41, 72000],  
[ 40, 71000],  
[ 42, 54000],  
[ 43, 129000],  
[ 53, 34000],  
[ 47, 50000],  
[ 42, 79000],  
[ 42, 104000],  
[ 59, 29000],  
[ 58, 47000],  
[ 46, 88000],  
[ 38, 71000],  
[ 54, 26000],  
[ 60, 46000],  
[ 60, 83000],  
[ 39, 73000],  
[ 59, 130000],  
[ 37, 80000],  
[ 46, 32000],  
[ 46, 74000],  
[ 42, 53000],  
[ 41, 87000],  
[ 58, 23000],  
[ 42, 64000],  
[ 48, 33000],  
[ 44, 139000],  
[ 49, 28000],  
[ 57, 33000],  
[ 56, 60000],  
[ 49, 39000],  
[ 39, 71000],  
[ 47, 34000],  
[ 48, 35000],  
[ 48, 33000],  
[ 47, 23000],
```

```
[ 45, 45000],  
[ 60, 42000],  
[ 39, 59000],  
[ 46, 41000],  
[ 51, 23000],  
[ 50, 20000],  
[ 36, 33000],  
[ 49, 36000]])
```

In [4]: label

```
In [5]: from sklearn.model_selection import train_test_split  
       from sklearn.linear_model import LogisticRegression
```

```
In [ ]: for i in range(1,401):
    x_train,x_test,y_train,y_test=train_test_split(features,label,test_size=0.2, random_state=i)
    model=LogisticRegression()
    model.fit(x_train,y_train)
    train_score=model.score(x_train,y_train)
    test_score=model.score(x_test,y_test)
    if test_score>train_score:
        print("Test {} Train{} Random State {}".format(test_score,train_score,i))
```

```
In [11]: x_train,x_test,y_train,y_test=train_test_split(features,label,test_size=0.2, random_state=42)
finalModel=LogisticRegression()
finalModel.fit(x_train,y_train)
```

Out[11]: LogisticRegression

## ► Parameters

```
In [12]: print(finalModel.score(x_train,y_train))
          print(finalModel.score(x_test,y_test))
```

0.8375  
0.8875

In [13]:

```
from sklearn.metrics import classification_report
print(classification_report(label,finalModel.predict(features)))
```

	precision	recall	f1-score	support
0	0.85	0.93	0.89	257
1	0.85	0.70	0.77	143
accuracy			0.85	400
macro avg	0.85	0.81	0.83	400
weighted avg	0.85	0.85	0.84	400

In [ ]: