

```

•[1]: #M.vishal
      #240701598
      #7/10/2025
      import numpy as np
      import pandas as pd
      df=pd.read_csv('Social_Network_Ads.csv')
      df

```

```

[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

```

•[2]: #M.vishal
      #240701598
      #7/10/2025
      df.head()

```

```

[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

```

```

•[3]: #M.vishal
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      features=df.iloc[:,[2,3]].values
      label=df.iloc[:,4].values
      features

```

```

[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],

```

```
•[4]: #M.vishal
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      label
```

```
[4]: array([[0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1,
          1, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
          0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
          0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
          0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1,
          0, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0,
          1, 1, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0,
          1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1,
          0, 1, 0, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1,
          1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1,
          0, 1, 0, 0, 1, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 0,
          1, 0, 1, 1, 1, 0, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 1, 0, 1,
          0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1,
          1, 1, 0, 1]])
```

```
•[5]: #M.vishal
      #240701598
      #7/10/2025
      from sklearn.model_selection import train_test_split
      from sklearn.linear_model import LogisticRegression
```

```
•[7]: #M.vishal
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      #7/10/2025
      for i in range(1, 401):
          x_train, x_test, y_train, y_test = train_test_split(features, label, test_size=0.3, 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))
```

```
Test 0.8833333333333333 Train0.8285714285714286 Random State 4
Test 0.8666666666666667 Train0.8464285714285714 Random State 5
Test 0.875 Train0.8464285714285714 Random State 6
Test 0.875 Train0.8392857142857143 Random State 7
Test 0.875 Train0.8464285714285714 Random State 10
Test 0.8583333333333333 Train0.8285714285714286 Random State 13
Test 0.8666666666666667 Train0.8535714285714285 Random State 14
Test 0.8833333333333333 Train0.8464285714285714 Random State 15
Test 0.8833333333333333 Train0.8464285714285714 Random State 16
Test 0.8583333333333333 Train0.8535714285714285 Random State 17
Test 0.8833333333333333 Train0.8285714285714286 Random State 18
```

```
•[8]: #M.vishal
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      #7/10/2025
      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)
```

```
[8]: ▾ LogisticRegression 3 3
      ► Parameters
```

```
•[9]: #M.vishal
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      #7/10/2025
      print(finalModel.score(x_train,y_train))
      print(finalModel.score(x_test,y_test))

      0.8375
      0.8875
```

•[10]:

```
#M.vishal  
#240701598  
#7/10/2025  
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

[]: