

logistic-regression

April 4, 2024

0.1 To predict whether a person will purchase from Social Network Ads

```
[59]: ## Importing Necessary Libraries
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report, \
    confusion_matrix
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import LabelEncoder
```

```
[60]: ## Loading dataset
data = pd.read_csv(r'C:\Users\ntpc\Desktop\Social_Network_Ads.csv')
data.head()
```

```
[60]:
```

	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

```
[61]: data.dtypes
```

```
[61]: User ID          int64
Gender          object
Age            int64
EstimatedSalary int64
Purchased       int64
dtype: object
```

```
[62]: #checking each value counts
data['Purchased'].value_counts()
```

```
[62]: Purchased
0      257
1      143
Name: count, dtype: int64
```

```
[63]: #dropping User ID column  
data = data.drop('User ID',axis = 1)  
data.head()
```

```
[63]:   Gender  Age  EstimatedSalary  Purchased  
0    Male   19           19000           0  
1    Male   35           20000           0  
2  Female   26           43000           0  
3  Female   27           57000           0  
4    Male   19           76000           0
```

```
[64]: # Encoding categorical variables  
le = LabelEncoder()  
data['Gender'] = le.fit_transform(data['Gender'])
```

```
[65]: data.head()
```

```
[65]:   Gender  Age  EstimatedSalary  Purchased  
0        1   19           19000           0  
1        1   35           20000           0  
2        0   26           43000           0  
3        0   27           57000           0  
4        1   19           76000           0
```

```
[66]: # Splitting the data into training and testing sets  
X = data.drop('Purchased', axis=1)  
y = data['Purchased']  
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,  
↪random_state=42)
```

```
[67]: ## Doing Feature Scaling for better accuracy  
from sklearn.preprocessing import StandardScaler  
sc = StandardScaler()  
X_train = sc.fit_transform(X_train)  
X_test = sc.transform(X_test)
```

```
[68]: model = LogisticRegression()
```

```
[69]: model.fit(X_train,y_train)
```

```
[69]: LogisticRegression()
```

```
[70]: # Making predictions on the testing data  
y_pred = model.predict(X_test)
```

```
[71]: # Evaluating the model  
accuracy = accuracy_score(y_test, y_pred)
```

```
print(f'Accuracy: {accuracy}')
```

Accuracy: 0.8875

```
[72]: # Printing classification report and confusion matrix
print('Classification Report:')
print(classification_report(y_test, y_pred))
```

Classification Report:

	precision	recall	f1-score	support
0	0.88	0.96	0.92	52
1	0.91	0.75	0.82	28
accuracy			0.89	80
macro avg	0.90	0.86	0.87	80
weighted avg	0.89	0.89	0.88	80

```
[73]: print('Confusion Matrix:')
print(confusion_matrix(y_test, y_pred))
```

Confusion Matrix:

```
[[50  2]
 [ 7 21]]
```