Assignment No-5

Data Analytics II 1. Implement logistic regression using Python/R to perform classification on Social_Network_Ads.csv dataset. 2. Compute Confusion matrix to find TP, FP, TN, FN, Accuracy, Error rate, Precision, Recall on the given dataset

Import libraries and create alias for Pandas, Numpy

```
[1]: import pandas as pd import numpy as np
```

Import the Social Media Adv Dataset

- [2]: from google.colab import files
- []: files.upload()

<IPython.core.display.HTML object>

Saving Social Network Ads.csv to Social Network Ads (1).csv

[]: {'Social Network Ads (1).csv': b'Age,EstimatedSalary,Purchased\r\n19,19000,0\r\n 35,20000,0\r\n26,43000,0\r\n27,57000,0\r\n19,76000,0\r\n27,58000,0\r\n27,84000,0 \r\n32,150000,1\r\n25,33000,0\r\n35,65000,0\r\n26,80000,0\r\n26,52000,0\r\n20,86 6.28000.1\r\n48.29000.1\r\n45.22000.1\r\n47.49000.1\r\n48.41000.1\r\n45.22000.1\ r\n46,23000,1\r\n47,20000,1\r\n49,28000,1\r\n47,30000,1\r\n29,43000,0\r\n31,1800 $0,0\r\n31,74000,0\r\n27,137000,1\r\n21,16000,0\r\n28,44000,0\r\n27,90000,0\r\n35$,27000,0\r\n33,28000,0\r\n30,49000,0\r\n26,72000,0\r\n27,31000,0\r\n27,17000,0\r \n33,51000,0\r\n35,108000,0\r\n30,15000,0\r\n28,84000,0\r\n23,20000,0\r\n25,7900 $0,0\r\n27,54000,0\r\n30,135000,1\r\n31,89000,0\r\n24,32000,0\r\n18,44000,0\r\n29$,83000,0\r\n35,23000,0\r\n27,58000,0\r\n24,55000,0\r\n23,48000,0\r\n28,79000,0\r \n22,18000,0\r\n32,117000,0\r\n27,20000,0\r\n25,87000,0\r\n23,66000,0\r\n32,1200 $00,1\rn59,83000,0\rn24,58000,0\rn24,19000,0\rn23,82000,0\rn22,63000,0\rn31$ $,68000,0\r\n25,80000,0\r\n24,27000,0\r\n20,23000,0\r\n33,113000,0\r\n32,18000,0\$ $r\34,112000,1\r\n18,52000,0\r\n22,27000,0\r\n28,87000,0\r\n26,17000,0\r\n30,800$ $00.0\r\n39.42000.0\r\n20.49000.0\r\n35.88000.0\r\n30.62000.0\r\n31.118000.1\r\n20.49000.0\r\n30.62$ 4,55000,0\r\n28,85000,0\r\n26,81000,0\r\n35,50000,0\r\n22,81000,0\r\n30,116000.0 \r\n26,15000,0\r\n29,28000,0\r\n29,83000,0\r\n35,44000,0\r\n35,25000,0\r\n28,123 000,1\r\n35,73000,0\r\n28,37000,0\r\n27,88000,0\r\n28,59000,0\r\n32,86000,0\r\n3 $3,149000,1\r\n19,21000,0\r\n21,72000,0\r\n26,35000,0\r\n27,89000,0\r\n26,86000,0$ \r\n38.80000.0\r\n39.71000.0\r\n37.71000.0\r\n38.61000.0\r\n37.55000.0\r\n42.800 00,0\r\n40,57000,0\r\n35,75000,0\r\n36,52000,0\r\n40,59000,0\r\n41,59000,0\r\n36 ,75000,0\r\n37,72000,0\r\n40,75000,0\r\n35,53000,0\r\n41,51000,0\r\n39,61000,0\r \n42.65000.0\r\n26.32000.0\r\n30.17000.0\r\n26.84000.0\r\n31.58000.0\r\n33.31000 ,0\r\n30,87000,0\r\n21,68000,0\r\n28,55000,0\r\n23,63000,0\r\n20,82000,0\r\n30,1 07000,1\r\n28,59000,0\r\n19,25000,0\r\n19,85000,0\r\n18,68000,0\r\n35,59000,0\r\ n30,89000,0\r\n34,25000,0\r\n24,89000,0\r\n27,96000,1\r\n41,30000,0\r\n29,61000, 0\r\n20.74000.0\r\n26.15000.0\r\n41.45000.0\r\n31.76000.0\r\n36.50000.0\r\n40.47 $000,0\r\n31,15000,0\r\n46,59000,0\r\n29,75000,0\r\n26,30000,0\r\n32,135000,1\r\n$ $32,100000,1\r\n25,90000,0\r\n37,33000,0\r\n35,38000,0\r\n33,69000,0\r\n18,86000,$ 0\r\n22.55000.0\r\n35.71000.0\r\n29.148000.1\r\n29.47000.0\r\n21.88000.0\r\n34.1 $15000,0\r\n26,118000,0\r\n34,43000,0\r\n34,72000,0\r\n23,28000,0\r\n35,47000,0\r\n36,11800,0\r\n36,118000,0\r$ $\n25,22000,0\r\n24,23000,0\r\n31,34000,0\r\n26,16000,0\r\n31,71000,0\r\n32,11700$ $0.1\r\n33.43000.0\r\n33.60000.0\r\n31.66000.0\r\n20.82000.0\r\n33.41000.0\r\n35.$ $72000,0\r\n24,84000,0\r\n19,26000,0\r\n29,43000,0\r\n19,70000,0\r\n29,43000,0\r\n29,$ n28,89000,0\r\n34,43000,0\r\n30,79000,0\r\n20,36000,0\r\n26,80000,0\r\n35,22000, $0\r\n35,39000,0\r\n49,74000,0\r\n39,134000,1\r\n41,71000,0\r\n58,101000,1\r\n47,$ 47000,0\r\n55,130000,1\r\n52,114000,0\r\n40,142000,1\r\n46,22000,0\r\n48,96000,1 \r\n52,150000,1\r\n59,42000,0\r\n35,58000,0\r\n47,43000,0\r\n60,108000,1\r\n49,6 5000.0\r\n40.78000.0\r\n46.96000.0\r\n59.143000.1\r\n41.80000.0\r\n35.91000.1\r\ n37,144000,1\r\n60,102000,1\r\n35,60000,0\r\n37,53000,0\r\n36,126000,1\r\n56,133 000,1\r\n40,72000,0\r\n42,80000,1\r\n35,147000,1\r\n39,42000,0\r\n40,107000,1\r\ $n49,86000,1\r\n38,112000,0\r\n46,79000,1\r\n40,57000,0\r\n37,80000,0\r\n46,82000$ $0\r\n53,143000,1\r\n42,149000,1\r\n38,59000,0\r\n50,88000,1\r\n56,104000,1\r\n42,149000,1\r\n50,88$ 1.72000.0\r\n51.146000.1\r\n35.50000.0\r\n57.122000.1\r\n41.52000.0\r\n35.97000. 1\r\n44,39000,0\r\n37,52000,0\r\n48,134000,1\r\n37,146000,1\r\n50,44000,0\r\n52, $90000,1\r\n41,72000,0\r\n58,95000,1\r\n45,131000,1\r\n35,77000,0\r\n58,95000,1\r\n45,131000,1\r\n35,77000,0\r\n40,57000,0\r\n4$ \n36,144000,1\r\n55,125000,1\r\n35,72000,0\r\n48,90000,1\r\n42,108000,1\r\n40,75 $000, 0 \\ r \\ n37, 74000, 0 \\ r \\ n47, 144000, 1 \\ r \\ n40, 61000, 0 \\ r \\ n43, 133000, 0 \\ r \\ n59, 76000, 1 \\ r \\ r \\ n40, 1000, 1 \\ r \\ n40, 1400, 1 \\ r \\ n400, 1400, 1400, 1 \\ r \\ n400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400, 1400,$ $n60,42000,1\r\n39,106000,1\r\n57,26000,1\r\n57,74000,1\r\n38,71000,0\r\n49,88000$,1\r\n52,38000,1\r\n50,36000,1\r\n59,88000,1\r\n35,61000,0\r\n37,70000,1\r\n52,2 1000,1\r\n48,141000,0\r\n37,93000,1\r\n37,62000,0\r\n48,138000,1\r\n41,79000,0\r \n37,78000,1\r\n39,134000,1\r\n49,89000,1\r\n55,39000,1\r\n37,77000,0\r\n35,5700 $0.0\r\n36.63000.0\r\n42.73000.1\r\n43.112000.1\r\n45.79000.0\r\n46.117000.1\r\n5$ $8,38000,1\r\n48,74000,1\r\n37,137000,1\r\n37,79000,1\r\n40,60000,0\r\n42,54000,0$ \r\n51,134000,0\r\n47,113000,1\r\n36,125000,1\r\n38,50000,0\r\n42,70000,0\r\n39, 96000.1\r\n38.50000.0\r\n49.141000.1\r\n39.79000.0\r\n39.75000.1\r\n54.104000.1\ $r\35,55000,0\r\n45,32000,1\r\n36,60000,0\r\n52,138000,1\r\n53,82000,1\r\n41,520$ $00,0\r\n48,30000,1\r\n48,131000,1\r\n41,60000,0\r\n41,72000,0\r\n42,75000,0\r\n3$ $6,118000,1\r\n47,107000,1\r\n38,51000,0\r\n48,119000,1\r\n42,65000,0\r\n40,65000$ $0.0\$,0\r\n57,60000,1\r\n36,54000,0\r\n58,144000,1\r\n35,79000,0\r\n38,55000,0\r\n39, 122000,1\r\n53,104000,1\r\n35,75000,0\r\n38,65000,0\r\n47,51000,1\r\n47,105000,1 \r\n41,63000,0\r\n53,72000,1\r\n54,108000,1\r\n39,77000,0\r\n38,61000,0\r\n38,11 $3000,1\r\n37,75000,0\r\n42,90000,1\r\n37,57000,0\r\n36,99000,1\r\n60,34000,1\r\n$

Initialize the data frame

[4]: df=pd.read_csv("/content/drive/MyDrive/Colab Notebooks/Social_Network_Ads.csv")

Perform Data Preprocessing

]: df.head()						
9]:		Age	EstimatedSalary	Purchased		
	0	19	19000	0		
	1	35	20000	0		
	2	26	43000	0		
	3	27	57000	0		
	4	19	76000	0		
1:	df	.tail	()			
Г1:		Δα	e EstimatedSalar	v Purchased		

[]:		Age	EstimatedSalary	Purchased	
	395	46	41000	1	
	396	51	23000	1	
	397	50	20000	1	
	398	36	33000	0	
	399	49	36000	1	

[10]: df

[10]:		Age	EstimatedSalary	Purchased
	0	19	19000	0
	1	35	20000	0
	2	26	43000	0
	3	27	57000	0
	4	19	76000	0
				•••
	395	46	41000	1
	396	51	23000	1
	397	50	20000	1
	398	36	33000	0
	399	49	36000	1

[400 rows x 3 columns]

[11]: df.describe()

[11]:		Age	EstimatedSalary	Purchased	
	count	400.000000	400.000000	400.000000	
	mean	37.655000	69742.500000	0.357500	
	std	10.482877	34096.960282	0.479864	
	min	18.000000	15000.000000	0.000000	
	25%	29.750000	43000.000000	0.000000	
	50%	37.000000	70000.000000	0.000000	
	75%	46.000000	88000.000000	1.000000	
	max	60.000000	150000.000000	1.000000	

Identification and Handling of Null Values

[6]: df.isnull()

[6]:		Age	EstimatedSalary	Purchased
	0	False	False	False
	1	False	False	False
	2	False	False	False
	3	False	False	False
	4	False	False	False
	395	False	False	False
	396	False	False	False
	397	False	False	False
	398	False	False	False
	399	False	False	False

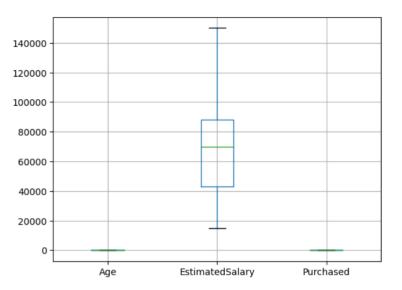
[400 rows x 3 columns]

[7]: df.isna()

[7]:		Age	EstimatedSalary	Purchased
	0	False	False	False
	1	False	False	False
	2	False	False	False
	3	False	False	False
	4	False	False	False
	395	False	False	False
	396	False	False	False
	397	False	False	False
	398	False	False	False
	399	False	False	False

```
[8]: df.isnull().sum()
[8]: Age
                        0
     EstimatedSalary
                        0
     Purchased
     dtype: int64
[12]: df.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 400 entries, 0 to 399
     Data columns (total 3 columns):
                          Non-Null Count Dtype
         Column
                          -----
                          400 non-null
                                          int64
         Age
         EstimatedSalary 400 non-null
                                          int64
      2 Purchased
                          400 non-null
                                          int64
     dtypes: int64(3)
     memory usage: 9.5 KB
[13]: df.dtypes
                        int64
[13]: Age
     EstimatedSalary
                        int64
                        int64
     Purchased
     dtype: object
     Import Seaborn and Matplotlib
[14]: import seaborn as sns
     import matplotlib.pyplot as plt
[17]: df.boxplot()
[17]: <Axes: >
```

[400 rows x 3 columns]



```
[19]: X = df.drop(['Purchased'], axis = 1)
    Y = df['Purchased']
```

Use Logistic regression (Train the Machine) to Create Model:

```
[21]: ytrain
```

```
[21]: 336
           1
     64
           0
     55
           0
     106
           0
     300
           1
     323
           1
     192
     117
           0
     47
           0
     172
           0
```

```
[22]: xtrain
[22]:
         EstimatedSalary
       58
               144000
   336
       59
               83000
   64
               55000
   55
       24
               35000
   106
       26
   300
       58
                38000
   . .
               30000
   323
       48
   192
       29
                43000
               52000
   117
       36
   47
       27
               54000
   172
       26
               118000
   [320 rows x 2 columns]
[23]: import sklearn
   from sklearn.linear_model import LogisticRegression
[24]: logreg = LogisticRegression()
   model=logreg.fit(xtrain, ytrain)
[26]: LogisticRegression()
[26]: LogisticRegression()
   Predict the y pred
[27]: y_pred=logreg.predict(xtest)
[28]: y_pred
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
[29]: y_pred_xtrain=logreg.predict(xtrain)
[30]: y pred xtrain
```

Name: Purchased, Length: 320, dtype: int64

```
[31]: print(xtrain)
    print("-----\n")
    print(ytest)
    print(ytest)
    print("-----\n")
    print(y_pred)
    print("----\n")
    print(y_pred_xtrain)
```

	Age	EstimatedSalary
336	58	144000
64	59	83000
55	24	55000
106	26	35000
300	58	38000
323	48	30000
192	29	43000
117	36	52000
47	27	54000
172	26	118000

[320 rows x 2 columns]

	Age	EstimatedSalar		
132	30	87000		
309	38	50000		
341	35	75000		
196	30	79000		

```
246
  35
       50000
       82000
  18
14
363
  42
       79000
304
  40
       60000
       34000
361
  53
       107000
329
  47
[80 rows x 2 columns]
336
64
  0
55
  0
106
300
323
  1
192
117
  0
47
172
Name: Purchased, Length: 320, dtype: int64
132
  0
309
341
196
  0
246
  0
14
363
  0
304
  0
361
329
Name: Purchased, Length: 80, dtype: int64
0 0 0 0 0 0]
-----
```

2 Evaluation parameters

[39]: 0.275

```
[32]: from sklearn.metrics import
        precision score, confusion matrix, accuracy score, recall score, classification report
     1.Confusion Matrix
[33]: cm= confusion matrix(ytest, y pred)
[34]: cm
[34]: array([[58, 0],
             [22, 0]])
     2. Accuracy score
[35]: print ("Accuracy: ", accuracy score(ytest, y pred))
     Accuracy: 0.725
     3.Precision
[36]: ps = precision score(ytest, y pred)
     /usr/local/lib/python3.10/dist-packages/sklearn/metrics/ classification.py:1344:
     UndefinedMetricWarning: Precision is ill-defined and being set to 0.0 due to no
     predicted samples. Use `zero_division` parameter to control this behavior.
       warn prf(average, modifier, msg start, len(result))
     4.Recall score
[37]: rs = recall_score(ytest, y_pred)
[38]: rs
[38]: 0.0
     5 Error Rate
[39]: error_rate = 1- accuracy_score(ytest, y_pred)
      error_rate
```

6. Classification Report

[40]: print("classification report: ",classification report(ytest, y pred))

classificat	ion rep	ort:		precision	recall	f1-score	support
	0	0.72	1.00	0.84	58		
	1	0.00	0.00	0.00	22		
accurac	у			0.73	80		
macro av	g	0.36	0.50	0.42	80		
weighted av	g	0.53	0.72	0.61	80		

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

_warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are ill-defined and being set to 0.0 in labels with no predicted samples. Use `zero_division` parameter to control this behavior.

warn prf(average, modifier, msg start, len(result))