Aim:

- 1. Logistic Regression
- 2. Differentiate between Linear and Logistic Regression
- 3. Sigmoid Function
- 4. Types of LogisticRegression
- 5. Confusion Matrix Evaluation Metrics

Code:

```
In [43]: import numpy as np
import pandas as pd
import matplotlib as plt
```

```
In [44]: data = pd.read_csv("diabetes.csv")
data
```

Out[44]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction
0	6	148	72	35	0	33.6	0.627
1	1	85	66	29	0	26.6	0.351
2	8	183	64	0	0	23.3	0.672
3	1	89	66	23	94	28.1	0.167
4	0	137	40	35	168	43.1	2.288
					•••		
763	10	101	76	48	180	32.9	0.171
764	2	122	70	27	0	36.8	0.340
765	5	121	72	23	112	26.2	0.245
766	1	126	60	0	0	30.1	0.349
767	1	93	70	31	0	30.4	0.315

768 rows × 9 columns

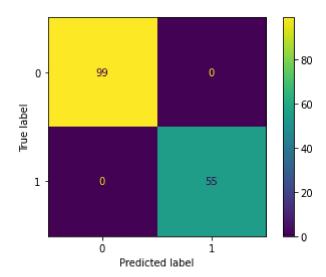
```
In [45]: data.head()
Out[45]:
              Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction A
           0
                        6
                                              72
                                                            35
                                                                     0 33.6
                               148
                                                                                               0.627
           1
                               85
                                              66
                                                            29
                                                                     0 26.6
                                                                                               0.351
                        1
           2
                        8
                              183
                                              64
                                                             0
                                                                     0 23.3
                                                                                               0.672
           3
                        1
                               89
                                              66
                                                            23
                                                                       28.1
                                                                                               0.167
                                                                    94
                                                                                               2.288
                        0
                              137
                                              40
                                                            35
                                                                   168 43.1
In [46]: data.tail()
Out[46]:
                Pregnancies
                            Glucose BloodPressure
                                                    SkinThickness Insulin BMI
                                                                              DiabetesPedigreeFunction
           763
                         10
                                 101
                                                76
                                                              48
                                                                     180
                                                                         32.9
                                                                                                 0.171
           764
                         2
                                122
                                                70
                                                              27
                                                                       0
                                                                         36.8
                                                                                                 0.340
                          5
           765
                                 121
                                                72
                                                              23
                                                                     112 26.2
                                                                                                 0.245
           766
                          1
                                 126
                                                60
                                                               0
                                                                       0
                                                                         30.1
                                                                                                 0.349
           767
                          1
                                 93
                                                70
                                                              31
                                                                       0 30.4
                                                                                                 0.315
          print("The shape of the data is: ")
In [47]:
          data.shape
          The shape of the data is:
Out[47]: (768, 9)
In [48]: print(data.isnull().sum())
                                          0
          Pregnancies
          Glucose
                                          0
          BloodPressure
                                          0
          SkinThickness
                                          0
          Insulin
                                          0
          BMI
                                          0
          DiabetesPedigreeFunction
                                          0
                                          0
          Age
          Outcome
                                          0
```

dtype: int64

```
In [49]: X = data.iloc[:,0:13]
         y = data.iloc[:,-1]
         Χ
         У
Out[49]: 0
                1
         1
                0
         2
                1
         3
                0
         4
                1
         763
                0
         764
         765
                0
         766
                1
         767
         Name: Outcome, Length: 768, dtype: int64
In [50]: | from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20,random_s
In [51]: |print(X_train.shape)
         print(X_test.shape)
         print(y_train.shape)
         print(y_test.shape)
         (614, 9)
         (154, 9)
         (614,)
         (154,)
In [68]: from sklearn.datasets import make classification
         from sklearn.linear_model import LogisticRegression
         from sklearn.model_selection import train_test_split
         from sklearn.pipeline import make_pipeline
         from sklearn.preprocessing import StandardScaler
         X, y = make classification(random state=42)
         X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=42)
         pipe = make_pipeline(StandardScaler(), LogisticRegression())
         pipe.fit(X_train, y_train)
Out[68]: Pipeline(steps=[('standardscaler', StandardScaler()),
                          ('logisticregression', LogisticRegression())])
In [53]: from sklearn.linear_model import LinearRegression
In [54]: from sklearn.preprocessing import StandardScaler
         from sklearn.pipeline import make pipeline
         model = make_pipeline(StandardScaler(with_mean=False), LinearRegression())
         model.fit(X_train, y_train)
Out[54]: Pipeline(steps=[('standardscaler', StandardScaler(with_mean=False)),
                         ('linearregression', LinearRegression())])
```

```
In [55]: model.score(X_test,y_test)
Out[55]: 1.0
In [56]: X_test
Out[56]:
                Pregnancies
                           Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction
           668
                          6
                                 98
                                                58
                                                              33
                                                                     190
                                                                         34.0
                                                                                                0.430
           324
                          2
                                 112
                                                75
                                                              32
                                                                         35.7
                                                                                                0.148
                                                                      0
                          2
           624
                                108
                                                               0
                                                                         30.8
                                                                                                0.158
                                                64
                                                                      0
           690
                          8
                                107
                                                80
                                                               0
                                                                         24.6
                                                                                                0.856
                         7
           473
                                136
                                                90
                                                               0
                                                                         29.9
                                                                      0
                                                                                                0.210
                         •••
           355
                         9
                                165
                                                88
                                                               0
                                                                         30.4
                                                                                                0.302
                                                                      0
           534
                          1
                                 77
                                                56
                                                              30
                                                                      56
                                                                        33.3
                                                                                                1.251
           344
                          8
                                 95
                                                72
                                                                                                0.485
                                                               0
                                                                      0
                                                                         36.8
           296
                          2
                                                70
                                                                         28.0
                                                                                                0.337
                                146
                                                              38
                                                                     360
           462
                          8
                                 74
                                                70
                                                              40
                                                                      49 35.3
                                                                                                0.705
          154 rows × 9 columns
In [57]: y_test
Out[57]: 668
                  0
          324
                  0
          624
                  0
          690
                  0
          473
                  0
                  . .
          355
                  1
          534
                  0
          344
                  0
          296
                  1
          462
          Name: Outcome, Length: 154, dtype: int64
In [63]: | from sklearn.metrics import precision_score,ConfusionMatrixDisplayconfusion_matri
          cm= confusion_matrix(y_test, y_pred)
          disp = ConfusionMatrixDisplay(confusion matrix = cm)
          print("Confusion matrix :")
          print(cm)
          Confusion matrix :
          [[99 0]
           [ 0 55]]
```

In [65]: disp.plot()



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