In [8]: data.info()

```
#
             Column
                                          Non-Null Count Dtype
                                          -----
         0
             Pregnancies
                                          2768 non-null
                                                          int64
          1
                                          2768 non-null
                                                           int64
              Glucose
          2
              BloodPressure
                                          2768 non-null
                                                           int64
          3
              SkinThickness
                                          2768 non-null
                                                           int64
          4
              Insulin
                                          2768 non-null
                                                           int64
                                          2768 non-null
                                                           float64
          6
              DiabetesPedigreeFunction
                                          2768 non-null
                                                           float64
                                          2768 non-null
                                                           int64
              Age
          8
              Outcome
                                          2768 non-null
                                                           int64
         dtypes: float64(2), int64(7)
         memory usage: 194.8 KB
 In [9]: data.describe()
 Out[9]:
                 Pregnancies
                                 Glucose BloodPressure SkinThickness
                                                                           Insulin
                                                                                          BMI DiabetesPedigreeFunction
                                                                                                                              Age
          count 2768.000000
                             2768.000000
                                            2768 000000
                                                          2768 000000
                                                                      2768.000000
                                                                                  2768 000000
                                                                                                           2768.000000 2768.000000
          mean
                    3.742775
                              121.102601
                                              69.134393
                                                            20.824422
                                                                        80.127890
                                                                                     32.137392
                                                                                                              0.471193
                                                                                                                         33.132225
            std
                    3.323801
                               32.036508
                                              19.231438
                                                            16.059596
                                                                        112.301933
                                                                                      8.076127
                                                                                                              0.325669
                                                                                                                         11.777230
                                0.000000
                                               0.000000
                                                                                                              0.078000
            min
                    0.000000
                                                             0.000000
                                                                         0.000000
                                                                                      0.000000
                                                                                                                         21.000000
           25%
                    1.000000
                               99.000000
                                              62.000000
                                                             0.000000
                                                                         0.000000
                                                                                     27.300000
                                                                                                              0.244000
                                                                                                                         24.000000
           50%
                    3.000000
                              117.000000
                                              72.000000
                                                            23.000000
                                                                        37.000000
                                                                                     32.200000
                                                                                                              0.375000
                                                                                                                         29.000000
           75%
                    6.000000
                              141.000000
                                              80.000000
                                                            32.000000
                                                                        130.000000
                                                                                     36.625000
                                                                                                              0.624000
                                                                                                                         40.000000
                   17.000000
                              199.000000
                                             122.000000
                                                                                     80.600000
                                                                                                              2.420000
           max
                                                           110.000000
                                                                        846.000000
                                                                                                                         81.000000
          4
In [10]: data.shape
Out[10]: (2768, 9)
In [11]: data.size
Out[11]: 24912
In [12]: data.isnull().sum()
                                        0
Out[12]: Pregnancies
                                        0
          Glucose
          BloodPressure
                                        0
          SkinThickness
                                        0
          Insulin
                                        0
          BMI
                                        0
          {\tt DiabetesPedigreeFunction}
                                        0
          Age
                                        0
          Outcome
                                        0
          dtype: int64
In [13]: data.Outcome.value counts()
Out[13]: Outcome
          0
                1816
                 952
          Name: count, dtype: int64
          0--> No 1--> Yes
In [14]: No = data[data.Outcome == 0]
          Yes = data[data.Outcome == 1]
In [15]: No.shape
Out[15]: (1816, 9)
In [16]: Yes.shape
Out[16]: (952, 9)
In [17]: No_sample = No.sample(n=952)
          No sample.shape
Out[17]: (952, 9)
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2768 entries, 0 to 2767
Data columns (total 9 columns):

```
In [19]: newdata.sample(15)
                Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age Outcome
Out[19]:
           479
                                 132
                                                               31
                                                                        0 28.0
                                                                                                                     0
                          4
                                                 86
                                                                                                   0.419
                                                                                                          63
           690
                                 107
                                                                0
                                                                        0 24.6
                                                                                                                     0
                          8
                                                 80
                                                                                                   0.856
                                                                                                          34
          2548
                          2
                                 121
                                                 70
                                                               32
                                                                       95 39.1
                                                                                                   0.886
                                                                                                          23
                                                                                                                     0
                          1
          1810
                                  93
                                                 56
                                                               11
                                                                        0 22.5
                                                                                                   0.417
                                                                                                          22
                                                                                                                     0
          1512
                         13
                                 153
                                                 88
                                                               37
                                                                      140 40.6
                                                                                                          39
                                                                                                                     0
                                                                                                   1.174
          2451
                          2
                                  99
                                                  0
                                                                0
                                                                        0 22.2
                                                                                                   0.108
                                                                                                          23
                                                                                                                     0
            33
                          6
                                  92
                                                 92
                                                                0
                                                                        0 19.9
                                                                                                   0.188
                                                                                                           28
                                                                                                                     0
          1074
                         10
                                 161
                                                 68
                                                               23
                                                                      132 25.5
                                                                                                   0.326
                                                                                                           47
                                                                                                                     1
                                                 70
          1753
                          3
                                 163
                                                               18
                                                                      105 31.6
                                                                                                   0.268
                                                                                                          28
                                                                                                                     1
           878
                          3
                                 171
                                                 72
                                                               33
                                                                      135 33.3
                                                                                                   0.199
                                                                                                          24
                                                                                                                     1
          2217
                          0
                                 107
                                                 62
                                                               30
                                                                       74 36.6
                                                                                                   0.757
                                                                                                          25
                                                                                                                     1
           338
                          9
                                 152
                                                 78
                                                               34
                                                                      171 34.2
                                                                                                   0.893
                                                                                                          33
                                                                                                                     1
                                                                                                                     0
          1501
                          2
                                 106
                                                                      165 29.0
                                                                                                          22
                                                 56
                                                               27
                                                                                                   0.426
          1372
                          4
                                 183
                                                  0
                                                                0
                                                                        0 28.4
                                                                                                   0.212
                                                                                                          36
                                                                                                                     1
          2470
                         11
                                 138
                                                 74
                                                               26
                                                                      144 36.1
                                                                                                   0.557
                                                                                                           50
                                                                                                                     1
In [20]: newdata.Outcome.value_counts()
Out[20]: Outcome
          0
                952
                952
          1
          Name: count, dtype: int64
In [21]: X=newdata.drop("Outcome" , axis="columns")
          X.head()
Out[21]:
                Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age
           887
                                                                       51 23.2
                                                                                                   0.223
                                                                                                          21
                          4
          2583
                          0
                                 141
                                                 84
                                                               26
                                                                        0
                                                                           32.4
                                                                                                   0.433
                                                                                                          22
           762
                                                                0
                                                                        0 22.5
                                                                                                          33
                          9
                                  89
                                                 62
                                                                                                   0.142
          1343
                          1
                                 119
                                                 44
                                                               47
                                                                       63 35.5
                                                                                                   0.280
                                                                                                          25
                                                 75
                                                                        0 29.9
                                                                                                   0.434
                                                                                                          28
In [22]:
         Y=newdata.Outcome.values
In [23]:
Out[23]: array([0, 0, 0, ..., 1, 1, 1], dtype=int64)
In [24]: print(Y)
         [0 0 0 ... 1 1 1]
In [25]: from sklearn.model selection import train test split
          X_train , X_test , y_train , y_test = train_test_split(X , Y , test_size= 0.2)
In [26]: len(X_train)
Out[26]: 1523
In [27]: len(X_test)
Out[27]:
          381
In [28]: len(y_train)
Out[28]: 1523
In [29]: len(y_test)
Out[29]: 381
```

In [18]: newdata = pd.concat([No_sample , Yes], axis=0)

```
In [30]: X_train.shape
Out[30]: (1523, 8)
In [31]: y_train.shape
Out[31]: (1523,)
In [32]: X train.head()
               Pregnancies Glucose BloodPressure SkinThickness Insulin BMI DiabetesPedigreeFunction Age
         1928
                        2
                              129
                                              0
                                                           0
                                                                                           0.304
                                                                                                  41
                                                                  0 38.5
                        9
                                                          22
         1698
                              120
                                             72
                                                                  56 20.8
                                                                                           0.733
                                                                                                  48
          896
                        1
                               117
                                             88
                                                          24
                                                                 145 34.5
                                                                                           0.403
                                                                                                  40
           58
                        0
                               146
                                             82
                                                           0
                                                                  0 40.5
                                                                                           1.781
                                                                                                   44
                        0
         1175
                              101
                                             62
                                                           0
                                                                  0 21.9
                                                                                           0.336
                                                                                                  25
In [33]: y train
Out[33]: array([0, 0, 1, ..., 1, 0, 1], dtype=int64)
In [34]: from sklearn.linear model import LogisticRegression
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
In [35]: lrmodel = LogisticRegression()
         lrmodel.fit(X_train , y_train)
        C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\linear model\ logistic.py:460: ConvergenceW
        arning: lbfgs failed to converge (status=1):
        STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
        Increase the number of iterations (max_iter) or scale the data as shown in:
            https://scikit-learn.org/stable/modules/preprocessing.html
        Please also refer to the documentation for alternative solver options:
            https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
          n_iter_i = _check_optimize_result(
Out[35]: ▼ LogisticRegression
         LogisticRegression()
In [36]: lrmodel.predict(X_test)
Out[36]: array([1, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1,
                 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 1,
                 0, 0, 0, 1, 1, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
                 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 1, 1, 0, 0, 0, 0, 1, 1, 1, 0,
                 1, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 1, 1, 1, 1, 0, 0, 0, 1,
                 1, 0, 0, 1, 1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 1,
                 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1,
                                                     0, 0, 1, 1, 1, 0, 1, 1, 0, 1,
                 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 1, 1,
                 1, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 1, 0, 1, 0,
                 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 1, 1, 0, 1, 0, 1, 1, 1,
                 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 1, 1, 0, 1, 0,
                0, 1, 0, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 0,
                 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0,
                 1, 0, 1, 1, 1, 0, 0, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0,
                 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0, 1, 0, 0, 1, 0, 0, 1,
                1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 1,
                0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0,
                1, 1, 0, 1, 0, 0, 1], dtype=int64)
In [37]: from sklearn.metrics import accuracy_score
         X train prediction = lrmodel.predict(X train)
         training data accuracy = accuracy score(X train prediction , y train)
In [38]: training data accuracy
Out[38]: 0.7590282337491793
In [39]: from sklearn.metrics import accuracy score
         X_test_prediction = lrmodel.predict(X_test)
         testing data accuracy = accuracy score(X test prediction , y test)
In [40]: testing data accuracy
```

```
Out[40]: 0.7532808398950132
In [41]: dtmodel = DecisionTreeClassifier(random state=42)
                dtmodel.fit(X train, y train)
Out[41]: v
                                 DecisionTreeClassifier
                DecisionTreeClassifier(random state=42)
In [42]: from sklearn.metrics import accuracy score
                X train prediction = dtmodel.predict(X train)
                training_data_accuracy = accuracy_score(X_train_prediction , y_train)
In [43]: training data accuracy
Out[43]: 1.0
In [44]: from sklearn.metrics import accuracy score
                X_test_prediction = dtmodel.predict(X_test)
                testing data accuracy = accuracy score(X test prediction , y test)
In [45]: testing data accuracy
Out[45]: 0.952755905511811
In [46]: '''
                Pregnancies: To express the Number of pregnancies
                Glucose: To express the Glucose level in blood
                BloodPressure: To express the Blood pressure measurement
                SkinThickness: To express the thickness of the skin
                Insulin: To express the Insulin level in blood
                BMI: To express the Body mass index
                DiabetesPedigreeFunction: To express the Diabetes percentage
                Age: To express the age
                Outcome: To express the final result 1 is Yes and 0 is No
Out[46]: '\nPregnancies: To express the Number of pregnancies\n\nGlucose: To express the Glucose level in blood\n\nBlood
                 Pressure: To express the Blood pressure measurement\n\nSkinThickness: To express the thickness of the skin\n\nI
                 nsulin: To express the Insulin level in blood\n\nBMI: To express the Body mass index\n\nDiabetesPedigreeFunctio
                 n: To express the Diabetes percentage\n\nAge: To express the age\n\nOutcome: To express the final result 1 is Y
                 es and 0 is No\n'
In [47]: dtmodel.predict([[9,122,56,0,0,33.3,1.114,33]])
               \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have various and the packages of the packages 
              alid feature names, but DecisionTreeClassifier was fitted with feature names
               warnings.warn(
Out[47]: array([1], dtype=int64)
In [48]: input data = (9,122,56,0,0,33.3,1.114,33)
                input_data_as_numpy_array = np.asarray(input_data)
                input data reshaped = input data as numpy array.reshape(1 , -1)
                prediction = dtmodel.predict(input_data_reshaped)
                if prediction == [0]:
                      print("The person does not have Diabetes.")
                else:
                       print("The person has Diabetes.")
              The person has Diabetes.
              C:\Users\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py:465: UserWarning: X does not have v
              alid feature names, but DecisionTreeClassifier was fitted with feature names
              warnings.warn(
In [49]: import pickle
                with open("Diabetes_model_pickle" , "wb") as file:
                       pickle.dump(dtmodel , file)
  In [1]: import numpy as np
                import pickle
                with open("Diabetes_model_pickle" , "rb") as file:
                       dtmodel = pickle.load(file)
```

```
In [2]: input data = (0,152,82,39,272,41.5,0.270,27)
                                           input_data_as_numpy_array = np.asarray(input_data)
                                           input data_reshaped = input_data_as_numpy_array.reshape(1 , -1)
                                           prediction = dtmodel.predict(input data reshaped)
                                           if prediction == [0]:
                                                                print("The person does not have Diabetes")
                                                               print("The person has Diabetes")
                                     The person does not have Diabetes
                                      \verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have various and the packages of the packages
```

alid feature names, but DecisionTreeClassifier was fitted with feature names warnings.warn(

```
In [56]: Pregnancies = int(input(""))
         Glucose = int(input(""))
         BloodPressure = int(input(""))
         SkinThickness = int(input(""))
         Insulin = int(input(""))
         BMI = float(input(""))
         DiabetesPedigreeFunction = float(input(""))
         Age = int(input(""))
In [57]: input data = (Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI, DiabetesPedigreeFunction, Age)
```

```
input_data_as_numpy_array = np.asarray(input_data)
input data reshaped = input data as numpy array.reshape(1 , -1)
prediction = model.predict(input_data_reshaped)
if prediction == [0]:
   print("The person does not have Diabetes")
    print("The person has Diabetes")
```

The person has Diabetes

 $\verb|C:\USers\ASUS\AppData\Roaming\Python\Python311\site-packages\sklearn\base.py: 465: UserWarning: X does not have various and the packages of the packages$ alid feature names, but LogisticRegression was fitted with feature names

In []:

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js