Importing the necessary libraries.

```
import numpy as np
import pandas as pd
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.svm import SVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, classification_report
import warnings
warnings.filterwarnings("ignore")
```

Data Collection and Analysis

```
diabetes_data = pd.read_csv('diabetes.csv')
```

Displaying the first 5 rows of dataframe

diabetes	_data.hea	ad()				
Pregn BMI \	ancies (Glucose Bl	oodPressure	SkinThickness	Insulin	
0	6	148	72	35	0	33.6
1	1	85	66	29	0	26.6
2	8	183	64	0	0	23.3
3	1	89	66	23	94	28.1
4	Θ	137	40	35	168	43.1
Diaha	+ D - d		A O +			
0 1 2 3	respealgi	reeFunctior 0.627 0.351 0.672 0.167 2.288	50 31 32 21	ome 1 0 1 0		

Displaying the last 5 rows of dataframe

```
diabetes_data.tail()
    Pregnancies Glucose BloodPressure SkinThickness Insulin BMI
763     10     101     76     48     180     32.9
```

764		2	122		70	27	0	36
765		5	121		72	23	112	26
766		1	126		60	0	0	30
767		1	93		70	31	0	30.
760	DiabetesPe	digree		Age	Outcome			
763			0.171	63	0			
764			0.340	27	0			
765			0.245	30	Θ			
766			0.349	47	1			
767			0.315	23	0			

No. of rows and columns

```
shape = diabetes_data.shape
print("Rows", shape[0])
print("Columns", shape[1])
Rows 768
Columns 9
diabetes data.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
 #
     Column
                                Non-Null Count
                                                 Dtype
- - -
                                768 non-null
 0
     Pregnancies
                                                 int64
     Glucose
                                768 non-null
 1
                                                 int64
 2
     BloodPressure
                                768 non-null
                                                 int64
 3
     SkinThickness
                                 768 non-null
                                                 int64
 4
     Insulin
                                768 non-null
                                                 int64
 5
     BMI
                                768 non-null
                                                 float64
 6
     DiabetesPedigreeFunction
                                768 non-null
                                                 float64
 7
                                 768 non-null
                                                 int64
     Age
     Outcome
                                 768 non-null
                                                 int64
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
```

Check for missing values

```
diabetes_data.isnull().sum()
```

Pregnancies	0
Glucose	0
BloodPressure	0
SkinThickness	0
Insulin	0
BMI	0
DiabetesPedigreeFunction	0
Age	0
Outcome	0
dtype: int64	

Check for duplicate values

```
diabetes_data.duplicated()
0
       False
1
       False
2
       False
3
       False
4
       False
763
       False
764
       False
765
       False
766
       False
767
       False
Length: 768, dtype: bool
diabetes data.duplicated().sum()
0
```

Statistical measures of data

diabetes_data.describe() Glucose BloodPressure SkinThickness Pregnancies Insulin \ 768.000000 768.000000 768.000000 768.000000 count 768.000000 3.845052 120.894531 69.105469 20.536458 mean 79.799479 std 3.369578 31.972618 19.355807 15.952218 115.244002 0.000000 0.000000 0.000000 0.000000 min 0.000000 25% 1.000000 99.000000 62.000000 0.000000 0.000000 50% 3.000000 117.000000 72.000000 23.000000 30.500000

```
75%
          6.000000 140.250000
                                      80.000000
                                                      32.000000
127.250000
         17.000000
                    199.000000
                                     122.000000
                                                      99.000000
max
846.000000
                    DiabetesPedigreeFunction
                                                               Outcome
              BMI
                                                       Age
       768.000000
                                   768.000000
                                               768.000000
                                                            768.000000
count
mean
        31.992578
                                     0.471876
                                                33.240885
                                                              0.348958
                                     0.331329
                                                11.760232
         7.884160
                                                              0.476951
std
                                     0.078000
                                                21.000000
                                                              0.000000
min
         0.000000
25%
        27.300000
                                     0.243750
                                                24.000000
                                                              0.000000
50%
        32,000000
                                     0.372500
                                                29.000000
                                                              0.000000
75%
        36.600000
                                     0.626250
                                                41.000000
                                                              1.000000
        67.100000
                                     2,420000
                                                81.000000
                                                              1.000000
max
diabetes data['Outcome'].value counts()
Outcome
0
     500
1
     268
Name: count, dtype: int64
```

0 -> Non-Diabetic 1 -> Diabetic

```
diabetes data.groupby('Outcome').mean()
                         Glucose BloodPressure SkinThickness
         Pregnancies
Insulin
Outcome
            3.298000 109.980000
                                       68.184000
                                                      19.664000
68.792000
            4.865672 141.257463
                                       70.824627
                                                      22.164179
100.335821
               BMI
                    DiabetesPedigreeFunction
                                                     Age
Outcome
         30.304200
                                     0.429734
                                               31.190000
0
1
         35.142537
                                     0.550500
                                               37.067164
```

Separate the dataset into independent and dependent variable

```
X = diabetes_data.drop(labels='Outcome',axis=1)
y = diabetes_data['Outcome']

print(X)

    Pregnancies Glucose BloodPressure SkinThickness Insulin BMI
0 6 148 72 35 0 33.6
```

1	1	85	66	29	0	26.6
2	8	183	64	0	0	23.3
3	1	89	66	23	94	28.1
4	0	137	40	35	168	43.1
763	10	101	76	48	180	32.9
764	2	122	70	27	0	36.8
765	5	121	72	23	112	26.2
766	1	126	60	0	0	30.1
767	1	93	70	31	0	30.4
0 1 2 3 4 763 764 765 766 767 [768 rows print(y) 0 1 1 0 2 1 3 0 4 1 763 0 764 0 765 0 766 1 767 0	x 8 column	0.627 0.351 0.672 0.167 2.288 0.171 0.340 0.245 0.349 0.315	Age 50 31 32 21 33 63 27 30 47 23			

```
scaler = StandardScaler()
standardized data = scaler.fit transform(X)
print(standardized data)
[[ 0.63994726  0.84832379  0.14964075  ...  0.20401277  0.46849198
   1.4259954 ]
 [-0.84488505 \ -1.12339636 \ -0.16054575 \ \dots \ -0.68442195 \ -0.36506078
  -0.190671911
 -0.105584151
               0.00330087 \quad 0.14964075 \quad \dots \quad -0.73518964 \quad -0.68519336
 [ 0.3429808
 -0.27575966]
 [-0.84488505 \quad 0.1597866 \quad -0.47073225 \quad \dots \quad -0.24020459 \quad -0.37110101
   1.17073215]
 [-0.84488505 - 0.8730192 \quad 0.04624525 \dots -0.20212881 -0.47378505
 -0.8713739311
X = standardized data
```

Train Test Split

```
X_train, X_test, y_train, y_test =
train_test_split(X,y,test_size=0.2,stratify=y,random_state=2)
print(X.shape, X_train.shape, X_test.shape)
(768, 8) (614, 8) (154, 8)
```

Training the model

1. Logistic Regression

```
reg = LogisticRegression()
reg.fit(X_train,y_train)
LogisticRegression()
X_test_pred = reg.predict(X_test)
accuracy = accuracy_score(y_test, X_test_pred)
print("Accuracy Score is:",accuracy)

print()
print("Report")
report = classification_report(y_test,X_test_pred)
print(report)
```

Accuracy Score is: 0.7597402597402597						
Report						
		precision	recall	f1-score	support	
	0	0.77	0.89	0.83	100	
	1	0.72	0.52	0.60	54	
				0.76	154	
	uracy			0.76	154	
	o avg	0.75	0.70	0.72	154	
weighte	d avg	0.75	0.76	0.75	154	

1. Decision Tree Classifier

```
DTC = DecisionTreeClassifier()
DTC.fit(X_train, y_train)
DecisionTreeClassifier()
X test pred = DTC.predict(X test)
accuracy = accuracy_score(y_test, X_test_pred)
print("Accuracy Score is:",accuracy)
print()
print("Report")
report = classification_report(y_test,X_test_pred)
print(report)
Accuracy Score is: 0.7077922077922078
Report
              precision
                           recall f1-score
                                               support
           0
                   0.75
                             0.83
                                        0.79
                                                   100
           1
                   0.60
                             0.48
                                        0.54
                                                    54
                                        0.71
                                                   154
    accuracy
                   0.68
                             0.66
                                        0.66
                                                   154
   macro avg
weighted avg
                   0.70
                             0.71
                                        0.70
                                                   154
```

1. Support Vector Classifier

```
classifier = SVC()
classifier.fit(X_train, y_train)
SVC()
```

```
X test pred = classifier.predict(X test)
accuracy = accuracy score(y test, X test pred)
print("Accuracy Score is:",accuracy)
print()
print("Report")
report = classification report(y test,X test pred)
print(report)
Accuracy Score is: 0.72727272727273
Report
                            recall f1-score
              precision
                                               support
           0
                   0.75
                              0.88
                                        0.81
                                                    100
           1
                   0.67
                                        0.53
                                                     54
                              0.44
                                        0.73
                                                    154
    accuracy
                                        0.67
                                                    154
                   0.71
                              0.66
   macro avg
weighted avg
                   0.72
                              0.73
                                        0.71
                                                    154
```

1. Random Forest Classifier

```
RFC = RandomForestClassifier(n estimators=100)
RFC.fit(X_train,y_train)
RandomForestClassifier()
X test pred = RFC.predict(X test)
accuracy = accuracy score(y test, X test pred)
print("Accuracy Score is:",accuracy)
print()
print("Report")
report = classification_report(y_test,X_test_pred)
print(report)
Accuracy Score is: 0.7467532467532467
Report
                            recall f1-score
              precision
                                               support
           0
                              0.87
                                                    100
                   0.77
                                        0.82
                                                     54
           1
                   0.68
                              0.52
                                        0.59
                                                    154
    accuracy
                                        0.75
                   0.73
                              0.69
                                        0.70
                                                    154
   macro avg
```

As the Logistic Regression is predicting more accurately so we use it for creating the predictive system

Creating a predictive system

```
input_data = [4,110,92,0,0,37.6,0.191,30] # y = 0

# Convert the list to a numpy array for easy manipulation
input_data = np.array(input_data)

# reshape array as we are predicting for one instance
input_data_reshaped = input_data.reshape(1,-1)

# standardize the input data
std_data = scaler.transform(input_data_reshaped)

prediction = reg.predict(std_data)

if(prediction[0] == 0):
    print("The person is not diabetic")

else:
    print("The person is diabetic")
```