------- بسم الله الرحمن الرحيم-------

```
import pandas as pd
import numpy as np
import seaborn as sns
from sklearn import svm
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score as ac
import pickle
from matplotlib import pyplot as plt
```

Understanding the data

```
In [2]: df = pd.read_csv("diabetes.csv")
    df.head()
    datafr = df
```

In [3]: df.tail()

Out[3]:		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Age	Outco
	763	10	101	76	48	180	32.9	0.171	63	
	764	2	122	70	27	0	36.8	0.340	27	
	765	5	121	72	23	112	26.2	0.245	30	
	766	1	126	60	0	0	30.1	0.349	47	
	767	1	93	70	31	0	30.4	0.315	23	

```
In [4]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):

#	Column	Non-Null Count	Dtype
0	Pregnancies	768 non-null	int64
1	Glucose	768 non-null	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	Age	768 non-null	int64
8	Outcome	768 non-null	int64
ж.	C1		

dtypes: float64(2), int64(7)
memory usage: 54.1 KB

```
In [5]: df.shape
```

Out[5]: (768, 9)

In [6]: df.describe()

```
768.000000
                             768.000000
                                             768.000000
                                                           768.000000
                                                                       768.000000
                                                                                   768.000000
                                                                                                            768.00000
          count
          mean
                    3.845052 120.894531
                                              69.105469
                                                            20.536458
                                                                        79.799479
                                                                                    31.992578
                                                                                                              0.4718
                    3.369578
            std
                              31.972618
                                              19.355807
                                                            15.952218
                                                                      115.244002
                                                                                     7.884160
                                                                                                              0.33132
                    0.000000
           min
                               0.000000
                                               0.000000
                                                             0.000000
                                                                         0.000000
                                                                                     0.000000
                                                                                                              0.07800
           25%
                    1.000000
                               99.000000
                                              62.000000
                                                             0.000000
                                                                         0.000000
                                                                                    27.300000
                                                                                                              0.2437!
           50%
                    3.000000 117.000000
                                              72.000000
                                                            23.000000
                                                                        30.500000
                                                                                    32.000000
                                                                                                              0.37250
           75%
                    6.000000
                            140.250000
                                              80.000000
                                                            32.000000
                                                                       127.250000
                                                                                    36.600000
                                                                                                              0.6262!
                   17.000000 199.000000
                                             122.000000
                                                            99.000000
                                                                       846.000000
                                                                                                              2.42000
           max
                                                                                    67.100000
In [7]:
          df.isnull().sum()
                                           0
          Pregnancies
Out[7]:
          Glucose
                                           0
          BloodPressure
                                           0
          SkinThickness
                                           0
          Insulin
                                           0
          BMI
                                           0
          DiabetesPedigreeFunction
                                           0
                                           0
                                           0
          Outcome
          dtype: int64
          1 -----> diabtic 0 -----> undiabtic
          df.Outcome.value_counts()
In [8]:
                500
          0
Out[8]:
                268
          Name: Outcome, dtype: int64
          df.groupby("Outcome").mean()
In [9]:
Out[9]:
                    Pregnancies
                                   Glucose BloodPressure SkinThickness
                                                                              Insulin
                                                                                           BMI DiabetesPedigreeFunc
          Outcome
                       3.298000 109.980000
                 0
                                                 68.184000
                                                               19.664000
                                                                           68.792000 30.304200
                                                                                                                0.429
                       4.865672 141.257463
                                                 70.824627
                                                               22.164179 100.335821 35.142537
                                                                                                                0.550
         perprocessing
```

BloodPressure SkinThickness

Insulin

DiabetesPedigreeFunction

```
In [10]: df.columns =df.columns.str.lower()

In [11]: str_col = df.columns
    str_target = ['outcome']
    X = df[str_col[:-1]]
    Y = df[str_target]
In [12]: X.head()
```

Out[6]:

Pregnancies

Glucose

Out[12]:		pregnancies	glucose	bloodpressure	skinthickness	insulin	bmi	diabetespedigreefunction	age
	0	6	148	72	35	0	33.6	0.627	50
	1	1	85	66	29	0	26.6	0.351	31
	2	8	183	64	0	0	23.3	0.672	32
	3	1	89	66	23	94	28.1	0.167	21
	4	0	137	40	35	168	43.1	2.288	33

```
In [13]: Y.tail()
```

Out[13]:		outcome
	763	0
	764	0
	765	0
	766	1
	767	0

Data Standardization

```
In [14]:
         scaler = StandardScaler()
In [15]:
         scaler.fit(X)
Out[15]: ▼ StandardScaler
         StandardScaler()
         standardized_data = scaler.transform(X)
In [16]:
In [17]:
         standardized_data
         array([[ 0.63994726, 0.84832379,
                                            0.14964075, ..., 0.20401277,
Out[17]:
                  0.46849198, 1.4259954],
                [-0.84488505, -1.12339636, -0.16054575, \ldots, -0.68442195,
                 -0.36506078, -0.19067191],
                [ 1.23388019, 1.94372388, -0.26394125, ..., -1.10325546,
                  0.60439732, -0.10558415],
                [ 0.3429808 , 0.00330087, 0.14964075, ..., -0.73518964,
                 -0.68519336, -0.27575966],
                [-0.84488505, 0.1597866, -0.47073225, ..., -0.24020459,
                 -0.37110101, 1.17073215],
                [-0.84488505, -0.8730192 , 0.04624525, ..., -0.20212881,
                 -0.47378505, -0.87137393]])
In [18]: X = standardized_data
In [19]:
         X.shape
         (768, 8)
Out[19]:
```

Training the model

Model evaluation

```
In [24]: prediction = classifier.predict(X_train )
prediction[2]
Out[24]: 1
```

accuracy score

DONE

Building a predictive system

all our data is numerical so our input will be numerical as well it mighit be an array, a list or even a dictionary

```
In [27]: # as atuple
x11 = (1,126,60,0,0,30.1,0.349,47)
```

Save the function of the model

```
In [30]: pickle.dump(is_diabtic,open("diabetes_disease.pkl","wb"))
In [31]: fun = pickle.load(open("diabetes_disease.pkl",'rb'))
In [32]: fun(x11)
```

This person has diabetes

DONE ALHAMDULLAH

```
In [33]: df

Out[33]: pregnancies glucose bloodpressure skinthickness insulin bmi diabetespedigreefunction age outcome
```

		pregnancies	glucose	bloodpressure	skinthickness	insulin	bmi	diabetespedigreefunction	age	outcome
	0	6	148	72	35	0	33.6	0.627	50	1
	1	1	85	66	29	0	26.6	0.351	31	0
	2	8	183	64	0	0	23.3	0.672	32	1
	3	1	89	66	23	94	28.1	0.167	21	0
	4	0	137	40	35	168	43.1	2.288	33	1
	763	10	101	76	48	180	32.9	0.171	63	0
	764	2	122	70	27	0	36.8	0.340	27	0
	765	5	121	72	23	112	26.2	0.245	30	0
	766	1	126	60	0	0	30.1	0.349	47	1
	767	1	93	70	31	0	30.4	0.315	23	0

768 rows × 9 columns

```
In [34]: datafr.head()
```

```
Out[34]:
            pregnancies glucose bloodpressure skinthickness insulin bmi diabetespedigreefunction age outcome
          0
                           148
                                         72
                                                      35
                                                              0 33.6
                                                                                     0.627
                                                                                            50
                                                                                                      1
                     1
                                                      29
                                                              0 26.6
          1
                            85
                                         66
                                                                                     0.351
                                                                                            31
                                                                                                      0
          2
                     8
                           183
                                         64
                                                       0
                                                              0 23.3
                                                                                     0.672
                                                                                            32
                                                                                                      1
          3
                     1
                            89
                                         66
                                                      23
                                                             94
                                                                28.1
                                                                                            21
                                                                                                      0
                                                                                     0.167
          4
                     0
                                         40
                                                      35
                                                                                                      1
                           137
                                                            168 43.1
                                                                                     2.288
                                                                                            33
          def Is_Diabtic() :
In [35]:
             print("this folloeing data must be filledout by the doctor
             Pregnancies = float(input("pregnanancies : "))
             Glucose = float(input("glucose : "))
             BloodPressure = float(input("glucose : "))
             SkinThickness= float(input("SkinThickness: "))
             Insulin = float(input("Insulin : "))
             BMI= float(input("BMI : "))
             DiabetesPedigreeFunction = float(input("glucose : "))
             Age = float(input("Age : "))
             listt = [Pregnancies, Glucose, BloodPressure, SkinThickness, Insulin, BMI, DiabetesPedigree
             fun(listt)
In [ ]: Is_Diabtic()
          this folloeing data must be filledout by the doctor
 In [ ]:
```