


```
"""
Developer : Saran B ( 2205102 )
Date : 14/12/2024
Topic : Diabetes Prediction Model
Data Source : https://www.kaggle.com/datasets/uciml/pima-indians-diabetes-database
Credits : Lex Fridman, Jeremy Howards, Siddharthan
"""
```

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


Providing Data (diabetes.csv)

```
# Importing the data
DiabetesData_1 = pd.read_csv('/content/diabetes.csv')
# Check the Number of rows and columns in the data set
DiabetesData_1.shape
```

 (768, 9)

We have 768 rows and 9 columns. to check the column name..


```
#displaying the 5 rows of the data
DiabetesData_1.head()
```



	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

Checking the basic Statistics of the Data using description()

```
DiabetesData_1.describe()
```



	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
count	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000	768.000000
mean	3.845052	120.894531	69.105469	20.536458	79.799479	31.992578	0.471876	33.240885	0.348611
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	11.760232	0.486590
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.078000	21.000000	0.000000
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	0.243750	24.000000	0.000000
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	0.372500	29.000000	0.000000
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	0.626250	41.000000	1.000000
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	2.420000	81.000000	1.000000

```
DiabetesData_1.mean()
```




	0
Pregnancies	3.845052
Glucose	120.894531
BloodPressure	69.105469
SkinThickness	20.536458
Insulin	79.799479
BMI	31.992578
DiabetesPedigreeFunction	0.471876
Age	33.240885
Outcome	0.348958

dtype: float64

Grouping the Outcome and getting mean value of those will give us a Idea that how the Prediction is going to be.

Here the 0 is the " Non Diabetic people " and the 1 is the " Diabetic people ".

```
DiabetesData_1.groupby("Outcome").mean()
```




	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age
Outcome								
0	3.298000	109.980000	68.184000	19.664000	68.792000	30.304200	0.429734	31.190000
1	4.865672	141.257463	70.824627	22.164179	100.335821	35.142537	0.550500	37.067164

dtype: float64

Dividing the data set into features and labels and store them into the variables "Features" and "Label" respectively

```
Features = DiabetesData_1.drop(columns='Outcome', axis=1)
Label = DiabetesData_1['Outcome']
```

```
Features.head()
```



	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

dtype: float64

```
Label.head()
```



	Outcome
0	1
1	0
2	1
3	0
4	1

dtype: int64

Now let's do standardize the data. In the next few steps, we are going to 1). Fit the data 2). Standardize (transform the data into a small range) these steps will make our model to predict high

```
Stool = StandardScaler()
```

```
Stool.fit_transform(Features)
```

```
array([[ 0.63994726,  0.84832379,  0.14964075, ...,  0.20401277,
         0.46849198,  1.4259954 ],
       [-0.84488505, -1.12339636, -0.16054575, ..., -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]])
```

```
Standardized_Features = Stool.fit_transform(Features)
```

```
Features = Standardized_Features
Label = Label
```

Now all the basic data standardization is completed, Let's move to the Training section.

before that, we need to spllit the data into training portion and testing portion. Here is the steps..

```
Features_train, Features_test, Label_train, Label_test = train_test_split(Features, Label, test_size=0.2, stratify=Label, ra
print(Features.shape, Features_train.shape, Features_test.shape)
```

```
(768, 8) (614, 8) (154, 8)
```

```
print(Label.shape, Label_train.shape, Label_test.shape)
```

```
(768,) (614,) (154,)
```

The data splitted successfully, now get into the training session

```
Classifier = svm.SVC(kernel='linear')
```

```
Classifier.fit(Features_train, Label_train)
```

```
SVC
SVC(kernel='linear')
```

Hello shiva sir, since we have a small data set. the training wont take much time to complete

```
#Testing and evaluvating
```

```
Features_train_prediction = Classifier.predict(Features_train)
Training_data_accuracy = accuracy_score(Features_train_prediction, Label_train)
print("Traning data accuracy is", Training_data_accuracy)
```

```
Traning data accuracy is 0.7866449511400652
```

```
Features_test_prediction = Classifier.predict(Features_test)
Testing_data_accuracy = accuracy_score(Features_test_prediction, Label_test)
print("Testing data accuracy is", Testing_data_accuracy)
```

```
Testing data accuracy is 0.7727272727272727
```

We evaluated our modal and that is performing good after some oprimization in data.

Now lets build a predictive system by user's input

```
input_data = (2,112,66,22,0,25,0.307,24)
input_data_as_numpy_array = np.asarray(input_data)
input_data_resaped = input_data_as_numpy_array.reshape(1,-1)
```

```
Stooled_data = Stool.transform(input_data_resaped)
print(Stooled_data)
```

```
prediction = Classifier.predict(Stooled_data)
print(prediction)
```

```
if (prediction[0] == 0):
```

```

    print('The person is not diabetic')
else:
    print('The person is diabetic')

[-0.54791859 -0.27837344 -0.16054575  0.09180513 -0.69289057 -0.88749274
 -0.497946   -0.78628618]]
[0]
The person is not diabetic
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:493: UserWarning: X does not have valid feature names, but StandardScaler.fit_transform will assume an ordering of the features:
warnings.warn(

```

Saving the Model

```

file_name = "trained_model.sav"
pickle.dump(Classifier, open(file_name,"wb"))

loaded_model = pickle.load(open("trained_model.sav","rb"))

input_data = (2,112,66,22,0,25,0.307,24)
input_data_as_numpy_array = np.asarray(input_data)
input_data_reshaped = input_data_as_numpy_array.reshape(1,-1)

Stooled_data = Stool.transform(input_data_reshaped)
print(Stooled_data)

prediction = loaded_model.predict(Stooled_data)
print(prediction)

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

[-0.54791859 -0.27837344 -0.16054575  0.09180513 -0.69289057 -0.88749274
 -0.497946   -0.78628618]]
[0]
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```