



```
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
```

Data Analysis :

- 1. Pregnancies : PG - No. of Pregnencies
- 2. Glucose : GL - Glucose Level (mg/dL)
- 3. BloodPressure : BP - Blood Pressure Value
- 4. SkinThickness : ST - in micrometers(um)
- 5. Insulin : INS - Insulin Level in blood (µIU/mL)
- 6. Body Mass Index : BMI
- 7. DiabetesPedigreeFunction : DPF - % of diabetes
- 8. Age : AGE
- 9. Outcome : Outcome - Possiblity of diabetes (Yes - 1, No - 0)

```
#loading the data
dataset = pd.read_csv('/content/diabetes.csv')
```


```
dataset.head()
```

	PG	GL	BP	ST	INS	BMI	DPF	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	

```
# no. of rows and column in dataset
dataset.shape
```

(768, 9)

```
# stats. data
dataset.describe()
```

	PG	GL	BP	ST	INS	BMI	DPF	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	0.348958	
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	0.331329	11.760232	0.476951	
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	

```
dataset['Outcome'].value_counts()
# 0 --> Non diabetic
# 1 --> Diabetic

0    500
1    268
Name: Outcome, dtype: int64
```

```
dataset.groupby('Outcome').mean()
# mean value of the inputs for a particular outcomes
```

	PG	GL	BP	ST	INS	BMI	DPF	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	

```
# Checking Co-relation
correlation_matrix = dataset.corr()
label_correlations = correlation_matrix['Outcome']
print(label_correlations)
```

```
PG      0.221898
GL      0.466581
BP      0.065068
ST      0.074752
INS     0.130548
BMI     0.292695
DPF     0.173844
AGE     0.238356
Outcome 1.000000
Name: Outcome, dtype: float64
```

Note : Since cor. for DPF, BP, ST, INS is significantly lower, we will be dropping these features.

```
# separating data and labels
X = dataset.drop(columns = 'Outcome', axis = 1)
Y = dataset['Outcome']

# columns with low cor.
columns_to_drop = ['BP', 'ST', 'INS', 'DPF']
X = X.drop(columns=columns_to_drop)
```

```
print(X)
print(Y)
```

```
      PG  GL  BMI  AGE
0      6  148  33.6  50
1      1   85  26.6  31
2      8  183  23.3  32
3      1   89  28.1  21
4      0  137  43.1  33
..    ..  ...  ...  ...
763  10  101  32.9  63
764   2  122  36.8  27
765   5  121  26.2  30
766   1  126  30.1  47
767   1   93  30.4  23
```

```
[768 rows x 4 columns]
```

```
0      1
1      0
2      1
3      0
4      1
..
763    0
764    0
765    0
766    1
767    0
```

```
Name: Outcome, Length: 768, dtype: int64
```

Data Standardization :

```
scaler = StandardScaler()
```

```
scaler.fit(X)
X=scaler.transform(X)
```

```
print(X)
```

```
[[ 0.63994726  0.84832379  0.20401277  1.4259954 ]
 [-0.84488505 -1.12339636 -0.68442195 -0.19067191]
 [ 1.23388019  1.94372388 -1.10325546 -0.10558415]
 ...
 [ 0.3429808  0.00330087 -0.73518964 -0.27575966]
 [-0.84488505  0.1597866  -0.24020459  1.17073215]
 [-0.84488505 -0.8730192  -0.20212881 -0.87137393]]
```

Splitting Data :

```
x_train, x_test, y_train, y_test = train_test_split(X, Y, test_size = 0.2, stratify = Y, random_state = 42)
```

```
print(X.shape, x_test.shape, x_train.shape)
```

```
(768, 4) (154, 4) (614, 4)
```

Training Model :

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

```
classifier.fit(x_train, y_train)
```

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

Model Evaluation : Accuracy Score for training data and test data

```
x_train_prediction = classifier.predict(x_train)
train_accuracy = accuracy_score(x_train_prediction, y_train)
print('Accuracy Score for SVC on training data is : ', train_accuracy)
```

```
Accuracy Score for SVC on training data is :  0.7866449511400652
```

```
x_test_prediction = classifier.predict(x_test)
test_accuracy = accuracy_score(x_test_prediction, y_test)
print('Accuracy Score for SVC on testing data is : ', test_accuracy)
```

```
Accuracy Score for SVC on testing data is :  0.7012987012987013
```

Creating .joblib Model file

```
from joblib import dump
```

```
filename = 'diabetes_model.joblib'
dump(classifier, filename)
```

```
['diabetes_model.joblib']
```

Checking the model :

```
from joblib import load
model = load('/content/diabetes_model.joblib')
```

```
input_data = (2, 103, 23.3, 22)
```

```
# Changing to a numpy array
input_numpyarray = np.asarray(input_data)
```

```
# Reshaping
input_resaped = input_numpyarray.reshape(1, -1)
```

```
# Fit the scaler on your training data (if available)
# You should replace X_train with your actual training data
# scaler.fit(X_train)
```

```
# Standardize the input data
inp = scaler.transform(input_resaped)
```

```
inp = scaler.transform(input_resnapeu)
y_pred = model.predict(inp)
```

```
if int(y_pred[0]) == 1:
    print('Diabetic')
else:
    print('Not Diabetic')
```

```
Not Diabetic
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but StandardScaler was fi
warnings.warn(
```

```
input_data2 = (2, 175, 23.3, 55)
```

```
# Changing to a numpy array
input_numpyarray2 = np.asarray(input_data2)
```

```
# Reshaping
input_resnaped2 = input_numpyarray2.reshape(1, -1)
```

```
# Fit the scaler on your training data (if available)
# You should replace X_train with your actual training data
# scaler.fit(X_train)
```

```
# Standardize the input data
inp2 = scaler.transform(input_resnaped2)
y_pred = model.predict(inp2)
```

```
if int(y_pred[0]) == 1:
    print('Diabetic')
else:
    print('Not Diabetic')
```



```
Diabetic
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but StandardScaler was fi
warnings.warn(
```