## **Project Name: Al Based Diabetes Prediction System Project Code:203476**

## **Coding:**

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
```

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import pandas as pd
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```

```
df=pd.read_csv('/content/diabetes (1).csv')
df.head()
```

Pregnancies         Glucose         BloodPressure         SkinThickness         Insulin         BMI         DiabetesPedigreeFunction         Age         Output           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		df=pd.read_csv('/content/diabetes (1).csv') df.head()								
1 1 85 66 29 0 26.6 0.351 31		Pregnancies	Glucose B	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
	C	0 6	148	72	35	0	33.6	0.627	50	1
<b>2</b> 8 183 64 0 0 23.3 0.672 32	1	1 1	85	66	29	0	26.6	0.351	31	0
	2	2 8	183	64	0	0	23.3	0.672	32	1
<b>3</b> 1 89 66 23 94 28.1 0.167 21	3	3 1	89	66	23	94	28.1	0.167	21	0
<b>4</b> 0 137 40 35 168 43.1 2.288 33	4	4 0	137	40	35	168	43.1	2.288	33	1

#outlier remove

```
Q1=df.quantile(0.25)
Q3=df.quantile(0.75)
IQR=Q3-Q1

print("---Q1--- \n",Q1)
print("\n---Q3--- \n",Q3)
print("\n---IQR---\n",IQR)

df_out = df[~((df < (Q1 - 1.5 * IQR)) | (df > (Q3 + 1.5 * IQR))).any(axis=1)]
df.shape,df_out.shape
X=df_out.drop(columns=['Outcome'])
y=df_out['Outcome']
#Splitting train test data 80 20 ratio
```

```
#outlier remove
    Q1=df.quantile(0.25)
    Q3=df.quantile(0.75)
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    df.shape,df_out.shape
    X=df_out.drop(columns=['Outcome'])
    y=df_out['Outcome']
    #Splitting train test data 80 20 ratio
→ ---Q1---
     Pregnancies
    Glucose
    BloodPressure
                                62.00000
    SkinThickness
                                0.00000
    Insulin
                                0.00000
    BMI
                                27,30000
    DiabetesPedigreeFunction
                                0.24375
    Outcome
                                0.00000
    Name: 0 25 dtyne: float64
```

```
Outcome
                           0.00000
 Name: 0.25, dtype: float64
      ---03---
       Pregnancies
                            6.00000
      Glucose
                          140.25000
      BloodPressure
                           80.00000
      SkinThickness
                           32,00000
      Insulin
                          127.25000
      BMI
                           36,60000
      DiabetesPedigreeFunction
                           0.62625
                          41.00000
      Outcome
                           1.00000
      Name: 0.75, dtype: float64
      ---IQR---
       Pregnancies
                            5.0000
      Glucose
                          41.2500
      BloodPressure
                           18,0000
      SkinThickness
                           32.0000
      Insulin
                          127,2500
      BMI
                           9.3000
      DiabetesPedigreeFunction
                           0.3825
                          17.0000
      Outcome
                           1.0000
      dtype: float64
from sklearn.model_selection import train_test_split
train_X,test_X,train_y,test_y=train_test_split(X,y,test_size=0.2)
train_X.shape,test_X.shape,train_y.shape,test_y.shape
    [5] from sklearn.model selection import train test split
          train_X,test_X,train_y,test_y=train_test_split(X,y,test_size=0.2)
          train_X.shape,test_X.shape,train_y.shape,test_y.shape
          ((511, 8), (128, 8), (511,), (128,))
from sklearn.metrics import confusion_matrix,accuracy_score,make_scorer
from sklearn.model selection import cross validate
def tn(y_true, y_pred): return confusion_matrix(y_true, y_pred)[0, 0]
def fp(y_true, y_pred): return confusion_matrix(y_true, y_pred)[0, 1]
def fn(y_true, y_pred): return confusion_matrix(y_true, y_pred)[1, 0]
def tp(y_true, y_pred): return confusion_matrix(y_true, y_pred)[1, 1]
```

```
#cross validation purpose
scoring = {'accuracy': make_scorer(accuracy_score), 'prec': 'precision'}
scoring = {'tp': make_scorer(tp), 'tn': make_scorer(tn),
           'fp': make_scorer(fp), 'fn': make_scorer(fn)}
def display_result(result):
    print("TP: ",result['test_tp'])
    print("TN: ",result['test_tn'])
    print("FN: ",result['test_fn'])
    print("FP: ",result['test_fp'])
#Logistic Regression
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score
acc=[]
roc=[]
clf=LogisticRegression()
clf.fit(train_X,train_y)
y_pred=clf.predict(test_X)
#find accuracy
```

```
ac=accuracy_score(test_y,y_pred)
acc.append(ac)

#find the ROC_AOC curve
rc=roc_auc_score(test_y,y_pred)
roc.append(rc)
print("\nAccuracy {0} ROC {1}".format(ac,rc))

#cross val score
result=cross_validate(clf,train_X,train_y,scoring=scoring,cv=10)
display_result(result)
```

```
Accuracy 0.8515625 ROC 0.7956821026282853
                                                                                                                                                                                                    1
       TP: [ 3 8 10 12 7 11 8 9 9 9]
TN: [30 30 28 30 25 33 31 31 34 34]
FN: [14 9 7 5 10 6 8 7 7 7]
FP: [ 5 4 6 4 9 1 4 4 1 1]
       //usr/local/lib/python3.1p/d/st-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: 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(
/usr/local/lib/pythons.le/dist-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed to converge (status=1):
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https://scikit-leam.org/stable/modules/linear_model.html@logistic-regression
n.iter; = _check_optimize_result(
//usr/local/lib/python3.18/dist-packages/sklearn/linear_model/_logistic.py:458: ConvergenceWarning: lbfgs failed to converge (status=1):
       STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
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    https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
    n_iter_i = _check_optimize_result(
/usr/local/lib/ovthon3.10/dist-packages/sklearn/linear_model/ logistic.pv:458: ConvergenceWarning: lbfgs failed to converge (status=1):
#Naive Bayes Theorem
#import library
from sklearn.naive_bayes import GaussianNB
clf=GaussianNB()
clf.fit(train_X,train_y)
y_pred=clf.predict(test_X)
#find accuracy
ac=accuracy_score(test_y,y_pred)
acc.append(ac)
#find the ROC AOC curve
rc=roc_auc_score(test_y,y_pred)
roc.append(rc)
print("\nAccuracy {0} ROC {1}".format(ac,rc))
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Accuracy 0.796875 ROC 0.7819072313454336
TP: [10 11 8 10 7 8 10 11 7 11]
TN: [32 26 26 32 28 28 31 31 31 27]
FN: [7 5 8 6 9 8 6 5 9 5]
FP: [3 9 9 3 7 7 4 4 4 8]
```