## Verzeo Internship - Major Project - ML June Batch

## Problem Statement:

Pima Indian Diabetes dataset is originally from the National Institute of Diabetes and Digestive and Kidney Diseases. The objective of the dataset is to diagnostically predict whether or not a patient has diabetes, based on certain diagnostic measurements included in the dataset. Several constraints were placed on the selection of these instances from a larger database. In particular, all patients here are females at least 21 years old of Pima Indian heritage.

Do classification by at least 2 Algorithms.

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import sklearn.metrics as metrics
from sklearn.metrics import classification_report
from sklearn.metrics import plot_confusion_matrix
from sklearn.metrics import average_precision_score
In [2]: df = pd.read_csv("diabetes.csv")

In [3]: df
```

Out[3]:		Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunc
	0	6	148	72	35	0	33.6	C
	1	1	85	66	29	0	26.6	(
	2	8	183	64	0	0	23.3	C
	3	1	89	66	23	94	28.1	(
	4	0	137	40	35	168	43.1	2
	•••		•••		•••	•••		
	763	10	101	76	48	180	32.9	(
	764	2	122	70	27	0	36.8	0
	765	5	121	72	23	112	26.2	0
	766	1	126	60	0	0	30.1	0
	767	1	93	70	31	0	30.4	(

768 rows × 9 columns

```
In [4]: ## statistic of dataset
    diabetes = df[df['Outcome']==1].shape
```

non\_diabetes = df[df['Outcome']==0].shape
print("Has Diabetes {}, Does not have Diabetes {}".format(diabetes,non\_diabetedf.describe()

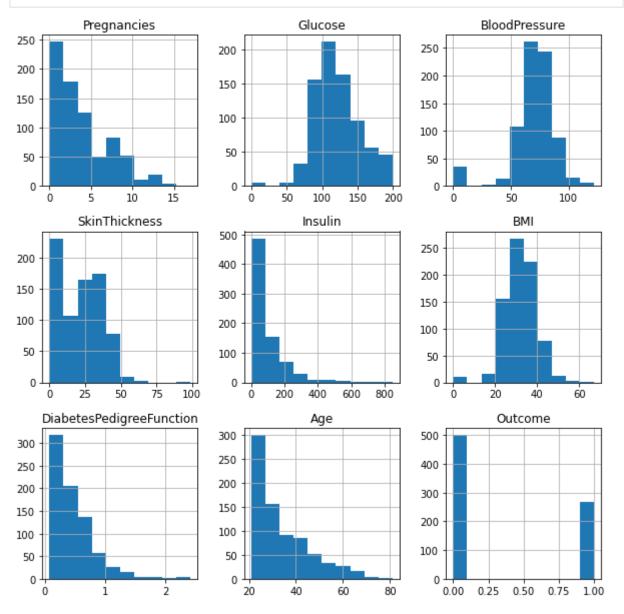
Has Diabetes (268, 9), Does not have Diabetes (500, 9)

Out[4]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	Diab
count	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	
std	3.369578	31.972618	19.355807	15.952218	115.244002	7.884160	
min	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	1.000000	99.000000	62.000000	0.000000	0.000000	27.300000	
50%	3.000000	117.000000	72.000000	23.000000	30.500000	32.000000	
75%	6.000000	140.250000	80.000000	32.000000	127.250000	36.600000	
max	17.000000	199.000000	122.000000	99.000000	846.000000	67.100000	

In [5]:

# x-axis range , y-axis abundance
histo = df.hist(figsize = (10,10))



```
In [6]:
         ## count of zero entries
         bp zeros = df[df['BloodPressure']==0].shape
         bmi_zeros = df[df['BMI']==0].shape
         insulin_zeros = df[df['Insulin']==0].shape
         glucose_zeros = df[df['Glucose']==0].shape
         skin zeros = df[df['SkinThickness']==0].shape
         print('Zero Counts of BP :{},BMI : {} , Insulin : {} , Glucose : {} , skin :
               .format(bp_zeros,bmi_zeros,insulin_zeros,glucose_zeros,skin_zeros))
        Zero Counts of BP: (35, 9), BMI: (11, 9), Insulin: (374, 9), Glucose: (5,
        9) , skin : (227, 9)
In [7]:
         #### pre-process Insulin & SkinThickness
         df_copy = df.copy(deep=True)
         df_copy[['Insulin','SkinThickness']] = df_copy[['Insulin','SkinThickness']].re
         ## The median is the middle point of a number set, in which half the numbers
         ## replace null value with median
         df_copy['Insulin'].fillna(df_copy['Insulin'].median(),inplace=True)
         df_copy['SkinThickness'].fillna(df_copy['SkinThickness'].median(),inplace=True
         ## test result
         # print(df_copy[df_copy['Insulin']==0].shape,df_copy[df_copy['SkinThickness']
         \# pre-process Blood Pressure , BMI & Glucose invalid values
         # as blood pressure & glucose are critical for determing diabeties it should
         # their null fraction is very small so it should better to remove invalid ent
         df_copy = df_copy[(df_copy['BloodPressure']!=0) & (df_copy['BMI']!=0) & (df_copy['BMI']!=0)
         # test result
         # print(df_copy[df_copy['BloodPressure']==0].shape,df_copy[df_copy['BMI']==0]
In [8]:
         from sklearn.model selection import train test split
In [9]:
         ## split dataset on outcome from 70% to 30%
         X = df copy.drop('Outcome',axis=1)
         y = df copy['Outcome']
         X train, X test, y train, y test = train test split(X, y, test size=0.30, random st
         print(X_train.shape)
         print(X test.shape)
         print(y_train.shape)
         print(y test.shape)
        (506, 8)
        (218, 8)
        (506,)
        (218,)
```

## Classification 1: Naive Bayes Classification

```
In [10]: from sklearn.naive_bayes import GaussianNB
In [11]: nb_model = GaussianNB()
In [12]: nb_model.fit(X_train,y_train)
```

```
Out[12]: GaussianNB()
In [13]:
          predicted_nb_model = nb_model.predict(X_test)
In [14]:
          cm nb model = metrics.confusion_matrix(y_test,predicted_nb_model)
In [15]:
          accuracy_nb_model = metrics.accuracy_score(y_test,predicted_nb_model)
In [16]:
          plot_confusion_matrix(nb_model, X_test, y_test)
          plt.show()
                                                120
                                                100
                    125
                                   15
            0
          Frue label
                                                60
            1
                    0
                                   1
                       Predicted label
In [17]:
          print("Confusion Matrix: \n",cm_nb_model)
          print("Accuracy: \n",accuracy_nb_model)
          average precision = average precision score(y test, predicted nb model)
          print("Average Precision = ",average_precision)
          Confusion Matrix:
           [[125 15]
           [ 34 44]]
          Accuracy:
           0.7752293577981652
          Average Precision = 0.5766499607270872
In [18]:
          print(classification report(y test,nb model.predict(X test)))
                        precision
                                      recall f1-score
                                                          support
                              0.79
                                        0.89
                                                   0.84
                                                              140
                              0.75
                                        0.56
                                                   0.64
                                                               78
                                                   0.78
                                                              218
              accuracy
             macro avg
                              0.77
                                        0.73
                                                   0.74
                                                              218
          weighted avg
                              0.77
                                        0.78
                                                   0.77
                                                              218
```

According to confusion matrix of Naive Bayes classifier, there are 44 who has Diabetes, 125 who does not have Diabetes and 49 which are misclassified.

## Classification 2: Support Vector Machine Classification

```
In [19]:
           from sklearn.svm import SVC
In [20]:
          svc = SVC(random_state=0, kernel='rbf')
In [21]:
          svc.fit(X_train, y_train)
Out[21]:
          SVC(random_state=0)
In [22]:
          svm_y_pred = svc.predict(X_test)
In [23]:
           cm svc = metrics.confusion_matrix(y_test,svm_y_pred)
In [24]:
           accuracy_svc = metrics.accuracy_score(y_test,svm_y_pred)
In [25]:
           plot_confusion_matrix(svc, X_test, y_test)
          plt.show()
                                                120
                    132
            0
                                                100
                                                - 80
          Frue label
                                                60
            1
                                                40
                     0
                                   1
                       Predicted label
In [26]:
          print("Confusion Matrix: \n",cm_svc)
          print("Accuracy: \n",accuracy_svc)
          average_precision = average_precision_score(y_test, svm_y_pred)
          print("Average Precision = ",average_precision)
          Confusion Matrix:
           [[132
                   8 ]
           [ 45 33]]
          Accuracy:
           0.7568807339449541
          Average Precision = 0.5469473466788302
In [27]:
          print(classification report(y test,svc.predict(X test)))
```

	precision	recall	f1-score	support
0	0.75	0.94	0.83	140
1	0.80	0.42	0.55	78
accuracy			0.76	218
macro avg	0.78	0.68	0.69	218
weighted avg	0.77	0.76	0.73	218

According to confusion matrix of Support Vector Machine classifier, there are 33 who has Diabetes, 132 who does not have Diabetes and 53 which are misclassified.