### **Mini Project 6: Disease Prediction**

### Step 1: Import library

```
import pandas as pd
import numpy as np
import seaborn as sns
```

# Step 2: Import Data

```
df=pd.read_csv('https://github.com/YBI-Foundation/Dataset/raw/main/Diabetes.csv')
df
```

	pregnancies	glucose	diastolic	triceps	insulin	bmi	dpf	age	diabetes
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

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	diastolic	768 non-null	int64
3	triceps	768 non-null	int64
4	insulin	768 non-null	int64
5	bmi	768 non-null	float64
6	dpf	768 non-null	float64
7	age	768 non-null	int64
8	diabetes	768 non-null	int64

dtypes: float64(2), int64(7)

memory usage: 54.1 KB

df.describe()

	pregnancies	glucose	diastolic	triceps	insulin	bmi	dpf	age	diabetes	
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	
df.shape										
(768, 9	)									
df.columns	df.columns									
<pre>Index(['pregnancies', 'glucose', 'diastolic', 'triceps', 'insulin', 'bmi',</pre>										

# Auxillary Step: Get Unique Values in y

	pregnancies	glucose	diastolic	triceps	insulin	bmi	dpf	age
diabetes								
0	3.298000	109.980000	68.184000	19.664000	68.792000	30.304200	0.429734	31.190000

### Step 3. Define y X

```
y = df['diabetes']
y.shape
     (768,)
У
            1
            0
            1
     763
            0
     764
     765
            0
     766
     767
     Name: diabetes, Length: 768, dtype: int64
X=df.drop(['diabetes'],axis=1)
#or
#X=df[['pregnancies', 'glucose', 'diastolic', 'triceps', 'insulin', 'bmi', 'dpf', 'age']]
```

X.shape

(768, 8)

Χ

	pregnancies	glucose	diastolic	triceps	insulin	bmi	dpf	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
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

768 rows × 8 columns

# Step 5: Standardising X

from sklearn.preprocessing import MinMaxScaler

mm=MinMaxScaler()

#### Step 5: Splitting Data

```
from sklearn.model_selection import train_test_split

X_train,X_test,y_train,y_test=train_test_split(X,y,train_size=0.7,stratify=y,random_state=2529)

X_train.shape,X_test.shape,y_train.shape,y_test.shape

((537, 8), (231, 8), (537,), (231,))
```

### Step 6: Creating Model

```
from sklearn.linear_model import LogisticRegression
model=LogisticRegression()
```

### Step 7: Training Model

```
model.fit(X_train,y_train)
    LogisticRegression()
```

### Step 8: Predicting Model

#### Auxillary Step: Probability for Each Predicted Class

```
model.predict proba(X test)
     array([[0.71101198, 0.28898802],
             [0.80246044, 0.19753956],
            [0.50085081, 0.49914919],
             [0.8745601 , 0.1254399 ],
             [0.84313967, 0.15686033],
             [0.72965238, 0.27034762],
             [0.32611128, 0.67388872],
             [0.82905388, 0.17094612],
             [0.57764733, 0.42235267],
            [0.5794767 , 0.4205233 ],
             [0.90475455, 0.09524545],
             [0.42428281, 0.57571719],
             [0.81659611, 0.18340389],
             [0.86057018, 0.13942982],
             [0.55629153, 0.44370847],
             [0.83208198, 0.16791802],
             [0.40636481, 0.59363519],
             [0.8430081 , 0.1569919 ],
             [0.6035823 , 0.3964177 ],
             [0.51982645, 0.48017355],
             [0.65174255, 0.34825745],
             [0.89662971, 0.10337029],
             [0.88362346, 0.11637654],
             [0.50529753, 0.49470247],
             [0.74048922, 0.25951078],
             [0.38010129, 0.61989871],
             [0.86743064, 0.13256936],
             [0.63051126, 0.36948874],
             [0.54593476, 0.45406524],
             [0.16753486, 0.83246514],
             [0.62023267, 0.37976733],
             [0.45959131, 0.54040869],
```

```
[0.75494925, 0.24505075],
[0.71542708, 0.28457292],
[0.75628148, 0.24371852],
[0.77697399, 0.22302601],
[0.67351753, 0.32648247],
[0.79929534, 0.20070466],
[0.74875692, 0.25124308],
[0.54844936, 0.45155064],
[0.32997346, 0.67002654],
[0.22941801, 0.77058199],
[0.7448794 , 0.2551206 ],
[0.34550793, 0.65449207],
[0.18608056, 0.81391944],
[0.26374953, 0.73625047],
[0.26523806, 0.73476194],
[0.62304765, 0.37695235],
[0.68999299, 0.31000701],
[0.45692191, 0.54307809],
[0.62161158, 0.37838842],
[0.79037029, 0.20962971],
[0.89387086, 0.10612914],
[0.73010557, 0.26989443],
[0.2052682 , 0.7947318 ],
[0.39976501, 0.60023499],
[0.83575072, 0.16424928],
[0.71212552. 0.28787448].
```

#### Step 9: Accuracy

print(classification\_report(y\_test,y\_pred))

	precision	recall	f1-score	support
0	0.79 0.76	0.91 0.54	0.84 0.63	150 81
accuracy macro avg weighted avg	0.77 0.78	0.72 0.78	0.78 0.74 0.77	231 231 231

# Step 10: Sample Prediction

```
df_new=df.sample(1)
```

df\_new

	pregnancies	glucose	diastolic	triceps	insulin	bmi	dpf	age	diabetes
506	0	180	90	26	90	36.5	0.314	35	1

df\_new.shape

(1, 9)

X\_new=df\_new.drop('diabetes',axis=1)

X\_new

```
pregnancies glucose diastolic triceps insulin bmi dpf age
X_new.shape
    (1, 8)

X_new=mm.fit_transform(X_new)

y_pred_new=model.predict(X_new)

y_pred_new
    array([0])

model.predict_proba(X_new)
    array([[0.99508059, 0.00491941]])
```

#### Link of the same:

https://colab.research.google.com/drive/1GP\_\_HNbVCjYpSmbFekx9am001kPTH954?usp=sharing