# AI DIABETES PREDICTION SYSTEM

NAME:SIVANESAN.G

REG NO:720421104045

### **INTRODUCTION:**

K-Nearest Neighbours (KNN) is a popular machine learning algorithm used for classification and regression tasks. It is a lazy learning, non-parametric algorithm that uses data with several classes to predict the classification of the new sample point. KNN is non-parametric since it doesn't make any assumptions on the data being studied.

During the training phase, the KNN algorithm stores the entire training dataset as a reference. When implementing an algorithm, you will always need a data set. So, you start by loading the training and the test data. Then, you choose the nearest data points (the value of K). K can be any integer.

The working of KNN Algorithm in Machine Learning can be summarized in three steps:

- 1. Load the data
- 2. Choose the nearest data points (the value of K)
- 3. Do the following, for each test data
  - Calculate the distance between test data and each row of training data
  - Sort the calculated distances in ascending order based on distance values
  - o Get top K rows from sorted array
  - o Get the most frequent class of these rows
  - Return this class as output.

## **PROCESS:**

import pandas as pd

import numpy as np

 $from \ sklearn.model\_selection \ import \ train\_test\_split$ 

from sklearn.preprocessing import StandardScaler

from sklearn.neighbors import KNeighborsClassifier

from sklearn.metrics import confusion\_matrix

from sklearn.metrics import fl\_score

from sklearn.metrics import accuracy\_score

$$dataset = pd.read\_csv\Big( "/kaggle/input/diabetes-data-set/diabetes.csv" \Big)$$
 
$$print\Big(len\Big(dataset\Big)\Big)$$

#### **OUTPUT:**

768

Pregnancies Glucose Blood Pressure Skin<br/>Thickness Insulin  $\,$  BMI  $\,\setminus\,$ 

0	6	148	72	35	0 33.6	
1	1	85	66	29	0 26.6	
2	8	183	64	0	0 23.3	
3	1	89	66	23	94 28.1	
4	0	137	40	35	168 43.1	

DiabetesPedigreeFunction Age Outcome

```
2 \hspace{1.5cm} 0.672 \hspace{.15cm} 32 \hspace{.15cm} 1
```

$$nonzero = \Big\lceil 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI' \Big\rceil$$

for col in nonzero:

$$\begin{split} & dataset \Big[ col \Big] \text{-}dataset \Big[ col \Big] \text{.} replace \Big( 0, np. NaN \Big) \\ & mean = int \Big( dataset \Big[ col \Big] \text{.} mean \Big( skipna = True \Big) \Big) \\ & dataset \Big[ col \Big] \text{-}dataset \Big[ col \Big] \text{.} replace \Big( np. NaN, mean \Big) \\ & print \Big( dataset \Big[ 'Glucose' \Big] \Big) \end{split}$$

#### **OUTPUT:**

•••

```
765 121.0
```

766 126.0

767 93.0

Name: Glucose, Length: 768, dtype: float64

```
x=dataset.iloc[:,0:8]
y=dataset.iloc[:,8]
 x\_train,x\_test,y\_train,y\_test=train\_test\_split \Big(x,y,random\_state=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_size=1,test\_siz
 0.3
 sc=StandardScaler
x_{train} = sc.fit_{transform}(x_{train})
x_{test} = sc.transform(x_{test})
 classifier = KNeighbors Classifier \Big( n\_neighbors = 15, p = 2, metric = 'euclidean' \Big)
 model=classifier.fit(x_train,y_train)
```

```
yp=classifier.predict(x_test)
```

yp

#### **OUTPUT:**

 $CM = confusion\_matrix \Big( y\_test, yp \Big)$ 

print(CM)

#### **OUTPUT:**

$$\begin{bmatrix} 133 & 13 \\ 34 & 51 \end{bmatrix}$$

$$print \Big( "F-Score: ", \Big( fl\_score \Big( y\_test, yp \Big) \Big) \Big)$$

#### **OUTPUT:**

F-Score: 0.6845637583892616

## **CONCLUSION:**

The AI prediction system using the KNN algorithm has shown promise in making

accurate predictions. While it has its strengths, we acknowledge its limitations and recommend further research and improvements to maximize its potential. The system has the potential to contribute to data-driven decision-making and add value in real-world applications.