

## Importing the required libraries

In [1]:

```
import numpy as np
import pandas as pd

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
```

## Reading the dataset

In [3]:

```
data=pd.read_csv('diabetes_data_upload.csv')
data.head()
```

Out[3]:

	Age	Gender	Polyuria	Polydipsia	sudden weight loss	weakness	Polyphagia	Genital thrush	visual blurring	Itching
0	40	Male	No	Yes	No	Yes	No	No	No	Yes
1	58	Male	No	No	No	Yes	No	No	Yes	No
2	41	Male	Yes	No	No	Yes	Yes	No	No	Yes
3	45	Male	No	No	Yes	Yes	Yes	Yes	No	Yes
4	60	Male	Yes	Yes	Yes	Yes	Yes	No	Yes	Yes

## Defining rows and columns

In [4]:

```
data
```

Out[4]:

	Age	Gender	Polyuria	Polydipsia	sudden weight loss	weakness	Polyphagia	Genital thrush	visual blurring	Itchi
0	40	Male	No	Yes	No	Yes	No	No	No	✓
1	58	Male	No	No	No	Yes	No	No	Yes	
2	41	Male	Yes	No	No	Yes	Yes	No	No	✓
3	45	Male	No	No	Yes	Yes	Yes	Yes	No	✓
4	60	Male	Yes	Yes	Yes	Yes	Yes	No	Yes	✓
...	...	...	...	...	...	...	...	...	...	
515	39	Female	Yes	Yes	Yes	No	Yes	No	No	✓
516	48	Female	Yes	Yes	Yes	Yes	Yes	No	No	✓
517	58	Female	Yes	Yes	Yes	Yes	Yes	No	Yes	
518	32	Female	No	No	No	Yes	No	No	Yes	✓
519	42	Male	No	No	No	No	No	No	No	

520 rows × 17 columns



In [5]:

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 520 entries, 0 to 519
Data columns (total 17 columns):
#   Column                                Non-Null Count  Dtype
---  ---                                -
0   Age                                   520 non-null    int64
1   Gender                               520 non-null    object
2   Polyuria                             520 non-null    object
3   Polydipsia                           520 non-null    object
4   sudden weight loss                   520 non-null    object
5   weakness                             520 non-null    object
6   Polyphagia                           520 non-null    object
7   Genital thrush                       520 non-null    object
8   visual blurring                      520 non-null    object
9   Itching                              520 non-null    object
10  Irritability                         520 non-null    object
11  delayed healing                      520 non-null    object
12  partial paresis                      520 non-null    object
13  muscle stiffness                     520 non-null    object
14  Alopecia                             520 non-null    object
15  Obesity                              520 non-null    object
16  class                                520 non-null    object
dtypes: int64(1), object(16)
memory usage: 69.2+ KB
```

## Defining null values

In [6]:

```
print("Number of null values :", data.isnull().sum().sum())
data.describe(include='all')
```

Number of null values : 0

Out[6]:

	Age	Gender	Polyuria	Polydipsia	sudden weight loss	weakness	Polyphagia	Genital thrush	vi blur
count	520.000000	520	520	520	520	520	520	520	
unique	NaN	2	2	2	2	2	2	2	
top	NaN	Male	No	No	No	Yes	No	No	
freq	NaN	328	262	287	303	305	283	404	
mean	48.028846	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
std	12.151466	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
min	16.000000	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
25%	39.000000	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
50%	47.500000	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
75%	57.000000	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
max	90.000000	NaN	NaN	NaN	NaN	NaN	NaN	NaN	



# Defining columns

In [7]:

```
{ column: len(data[column].unique()) for column in data.columns}
```

Out[7]:

```
{'Age': 51,
'Gender': 2,
'Polyuria': 2,
'Polydipsia': 2,
'sudden weight loss': 2,
'weakness': 2,
'Polyphagia': 2,
'Genital thrush': 2,
'visual blurring': 2,
'Itching': 2,
'Irritability': 2,
'delayed healing': 2,
'partial paresis': 2,
'muscle stiffness': 2,
'Alopecia': 2,
'Obesity': 2,
'class': 2}
```

## Data preprocessing

In [29]:

```
def preprocessing(df):
    df= df.copy()

    # Gender column Binary Encoding
    df['Gender'] = df ['Gender'].replace({'Female':0, 'Male':1 })

    #Symptom Column Binary Encoding
    for column in df.columns.drop(['Age', 'Gender', 'class']):
        df[column]= df[column].replace({'No':0 , 'Yes': 1})

    #train
    y=df["class"]
    X=df.drop("class", axis=1)
    X_train, X_test,y_train,y_test = train_test_split(X,y
                                                    ,train_size=0.7,shuffle=True,random_state=1)
    scaler=StandardScaler()
    scaler.fit(X_train)
    X_train=pd.DataFrame(scaler.transform(X_train)
                        ,index=X_train.index , columns=X_train.columns)
    X_test=pd.DataFrame(scaler.transform(X_test)
                      ,index=X_test.index, columns=X_test.columns)

    return X_train,X_test,y_train,y_test
```

In [17]:

```
X_train,X_test,y_train,y_test= preprocessing(data)
```

In [18]:

```
X_train
```

Out[18]:

	Age	Gender	Polyuria	Polydipsia	sudden weight loss	weakness	Polyphagia	Genital thrush	
122	-0.658902	0.740902	-0.994521	1.129159	-0.846747	0.841974	1.104315	-0.560428	-
168	-0.913060	0.740902	-0.994521	-0.885615	-0.846747	0.841974	-0.905539	-0.560428	-
23	0.018852	0.740902	-0.994521	1.129159	1.180990	0.841974	-0.905539	-0.560428	
13	1.120204	0.740902	1.005510	1.129159	1.180990	0.841974	1.104315	1.784351	
61	-1.082499	-1.349706	1.005510	1.129159	1.180990	0.841974	1.104315	-0.560428	
...	...	...	...	...	...	...	...	...	
129	0.018852	0.740902	1.005510	1.129159	1.180990	0.841974	-0.905539	-0.560428	-
144	1.713239	0.740902	1.005510	1.129159	-0.846747	-1.187685	1.104315	-0.560428	
72	1.459081	-1.349706	-0.994521	-0.885615	-0.846747	-1.187685	-0.905539	1.784351	-
235	-1.844973	0.740902	-0.994521	-0.885615	-0.846747	-1.187685	-0.905539	-0.560428	-
37	1.289643	0.740902	1.005510	1.129159	1.180990	0.841974	1.104315	-0.560428	

364 rows × 16 columns



In [19]:

```
y_test
```

Out[19]:

```
273    Negative
272    Negative
329    Negative
480    Negative
173    Positive
...
335    Negative
407    Negative
330    Negative
257    Positive
95     Positive
Name: class, Length: 156, dtype: object
```

# Decision tree algorithm

In [21]:

```
models= {  
    '    DecisionTree ': DecisionTreeClassifier()  
}  
for name,model in models.items():  
    model.fit(X_train,y_train)  
    print(name+ ': trained')
```

DecisionTree : trained

In [22]:

```
for name,model in models.items():  
    print(name+ ": {:.2f}%".format(model.score(X_test,y_test) * 100))
```

DecisionTree : 96.15%

## Random forest algortihm

In [23]:

```
models= {  
    '    Randomforest ': RandomForestClassifier()  
}  
for name,model in models.items():  
    model.fit(X_train,y_train)  
    print(name+ ': trained')
```

Randomforest : trained

In [24]:

```
for name,model in models.items():  
    print(name+ ": {:.2f}%".format(model.score(X_test,y_test) * 100))
```

Randomforest : 98.08%