

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [54]: data=pd.read_csv(r"C:\Users\user\Desktop\vicky\C4_framingham (1).csv")
```

```
In [94]: data.fillna(value=1)
```

Out[94]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	
...
4233	1	50	1.0	1	1.0	0.0	0	1	0	313.0	
4234	1	51	3.0	1	43.0	0.0	0	0	0	207.0	
4235	0	48	2.0	1	20.0	1.0	0	0	0	248.0	
4236	0	44	1.0	1	15.0	0.0	0	0	0	210.0	
4237	0	52	2.0	0	0.0	0.0	0	0	0	269.0	

4238 rows × 16 columns

```
In [95]: data.head()
```

Out[95]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sys
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	10
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	12
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	12
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	15
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	13

In [105]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4238 entries, 0 to 4237
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  -
0   male                   4238 non-null   int64
1   age                   4238 non-null   int64
2   education              4133 non-null   float64
3   currentSmoker          4238 non-null   int64
4   cigsPerDay             4209 non-null   float64
5   BPMeds                 4185 non-null   float64
6   prevalentStroke        4238 non-null   int64
7   prevalentHyp           4238 non-null   int64
8   diabetes               4238 non-null   int64
9   totChol               4188 non-null   float64
10  sysBP                  4238 non-null   float64
11  diaBP                  4238 non-null   float64
12  BMI                    4219 non-null   float64
13  heartRate              4237 non-null   float64
14  glucose                3850 non-null   float64
15  TenYearCHD             4238 non-null   object
dtypes: float64(9), int64(6), object(1)
memory usage: 529.9+ KB
```

In [107]: data1=data[['male','age','diabetes','TenYearCHD']]

In [108]: data1['TenYearCHD'].value_counts()

Out[108]: No 3594
Yes 644
Name: TenYearCHD, dtype: int64

In [109]: x=data1.drop('TenYearCHD',axis=1)
y=data1['TenYearCHD']

In []:

In [110]: g1={"TenYearCHD":{1:"Yes",0:"No",}}
data1=data1.replace(g1)
print(data1)

	male	age	diabetes	TenYearCHD
0	1	39	0	No
1	0	46	0	No
2	1	48	0	No
3	0	61	0	Yes
4	0	46	0	No
...
4233	1	50	0	Yes
4234	1	51	0	No
4235	0	48	0	No
4236	0	44	0	No
4237	0	52	0	No

[4238 rows x 4 columns]

```
In [111]: from sklearn.model_selection import train_test_split
```

```
In [112]: x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [113]: from sklearn.ensemble import RandomForestClassifier
```

```
In [114]: rfc=RandomForestClassifier()  
rfc.fit(x_train,y_train)
```

```
Out[114]: RandomForestClassifier()
```

```
In [115]: parameters = {'max_depth':[1,2,3,4,5],  
                        'min_samples_leaf':[5,10,15,20,25],  
                        'n_estimators':[10,20,30,40,50]  
                        }
```

```
In [116]: from sklearn.model_selection import GridSearchCV  
grid_search=GridSearchCV(estimator=rfc,param_grid=parameters,cv=2,scoring="accuracy")  
grid_search.fit(x_train,y_train)
```

```
Out[116]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),  
                      param_grid={'max_depth': [1, 2, 3, 4, 5],  
                                   'min_samples_leaf': [5, 10, 15, 20, 25],  
                                   'n_estimators': [10, 20, 30, 40, 50]},  
                      scoring='accuracy')
```

```
In [117]: grid_search.best_score_
```

```
Out[117]: 0.844571813890762
```

```
In [118]: from sklearn.tree import plot_tree
```

```
In [119]: rfc_best=grid_search.best_estimator_
```

```
In [120]: plt.figure(figsize=(80,40))
plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['Yes','No'],filled=True)
```

```
Out[120]: [Text(2232.0, 1902.6000000000001, 'age <= 60.5\ngini = 0.267\nsamples = 1879\nvalue = [2495,
471]\nclass = Yes'),
Text(1116.0, 1359.0, 'age <= 48.5\ngini = 0.218\nsamples = 1623\nvalue = [2239, 318]\nclass
= Yes'),
Text(558.0, 815.4000000000001, 'male <= 0.5\ngini = 0.138\nsamples = 901\nvalue = [1320, 10
6]\nclass = Yes'),
Text(279.0, 271.79999999999995, 'gini = 0.12\nsamples = 505\nvalue = [741, 51]\nclass = Ye
s'),
Text(837.0, 271.79999999999995, 'gini = 0.158\nsamples = 396\nvalue = [579, 55]\nclass = Ye
s'),
Text(1674.0, 815.4000000000001, 'age <= 49.5\ngini = 0.305\nsamples = 722\nvalue = [919, 21
2]\nclass = Yes'),
Text(1395.0, 271.79999999999995, 'gini = 0.38\nsamples = 60\nvalue = [70, 24]\nclass = Ye
s'),
Text(1953.0, 271.79999999999995, 'gini = 0.297\nsamples = 662\nvalue = [849, 188]\nclass = Y
es'),
Text(3348.0, 1359.0, 'male <= 0.5\ngini = 0.468\nsamples = 256\nvalue = [256, 153]\nclass =
Yes'),
Text(2790.0, 815.4000000000001, 'diabetes <= 0.5\ngini = 0.468\nsamples = 144\nvalue = [147,
88]\nclass = Yes'),
Text(2511.0, 271.79999999999995, 'gini = 0.46\nsamples = 136\nvalue = [141, 79]\nclass = Ye
s'),
Text(3069.0, 271.79999999999995, 'gini = 0.48\nsamples = 8\nvalue = [6, 9]\nclass = No'),
Text(3906.0, 815.4000000000001, 'diabetes <= 0.5\ngini = 0.468\nsamples = 112\nvalue = [109,
65]\nclass = Yes'),
Text(3627.0, 271.79999999999995, 'gini = 0.442\nsamples = 103\nvalue = [108, 53]\nclass = Ye
s'),
Text(4185.0, 271.79999999999995, 'gini = 0.142\nsamples = 9\nvalue = [1, 12]\nclass = No')]
```

```
In [ ]:
```