

Random Forest

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

```
In [2]: df=pd.read_csv(r"C:\Users\user\Downloads\health.csv")[0:1000]
df
```

```
Out[2]:
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI	DiabetesPedigreeFunction	Age	Outcome
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
...
763	10	101	76	48	180	32.9	0.171	63	0
764	2	122	70	27	0	36.8	0.340	27	0
765	5	121	72	23	112	26.2	0.245	30	0
766	1	126	60	0	0	30.1	0.349	47	1
767	1	93	70	31	0	30.4	0.315	23	0

768 rows × 9 columns

```
In [9]: 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   BloodPressure         768 non-null   int64  
3   SkinThickness         768 non-null   int64  
4   Insulin               768 non-null   int64  
5   BMI                  768 non-null   float64 
6   DiabetesPedigreeFunction 768 non-null   float64 
7   Age                  768 non-null   int64  
8   Outcome              768 non-null   int64  
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
```

```
In [10]: df.columns
```

```
Out[10]: Index(['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin',
               'BMI', 'DiabetesPedigreeFunction', 'Age', 'Outcome'],
              dtype='object')
```

```
In [11]: d['Outcome'].value_counts()
```

```
Out[11]: 0    500
         1    268
         Name: Outcome, dtype: int64
```

```
In [12]: x=d.drop('Outcome',axis=1)
         y=d['Outcome']
```

```
In [13]: g1={"Outcome":{"":0,0:1}}
         d=d.replace(g1)
         print(d)
```

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	BMI \
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
..
763	10	101	76	48	180	32.9
764	2	122	70	27	0	36.8
765	5	121	72	23	112	26.2
766	1	126	60	0	0	30.1
767	1	93	70	31	0	30.4

	DiabetesPedigreeFunction	Age	Outcome
0	0.627	50	1
1	0.351	31	1
2	0.672	32	1
3	0.167	21	1
4	2.288	33	1
..
763	0.171	63	1
764	0.340	27	1
765	0.245	30	1
766	0.349	47	1
767	0.315	23	1

[768 rows x 9 columns]

```
In [14]: from sklearn.model_selection import train_test_split
```

```
In [15]: x_train,x_test,y_train,y_test=train_test_split(x,y,train_size=0.70)
```

```
In [16]: from sklearn.ensemble import RandomForestClassifier
```

```
In [17]: rfc=RandomForestClassifier()
         rfc.fit(x_train,y_train)
```

```
Out[17]: RandomForestClassifier()
```

```
In [18]: parameters={'max_depth':[1,2,3,4,5],
                    'min_samples_leaf':[5,10,15,20,25],
                    'n_estimators':[10,20,30,40,50]}
         }
```

```
In [19]: 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[19]: 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 [20]: grid_search.best_score_
```

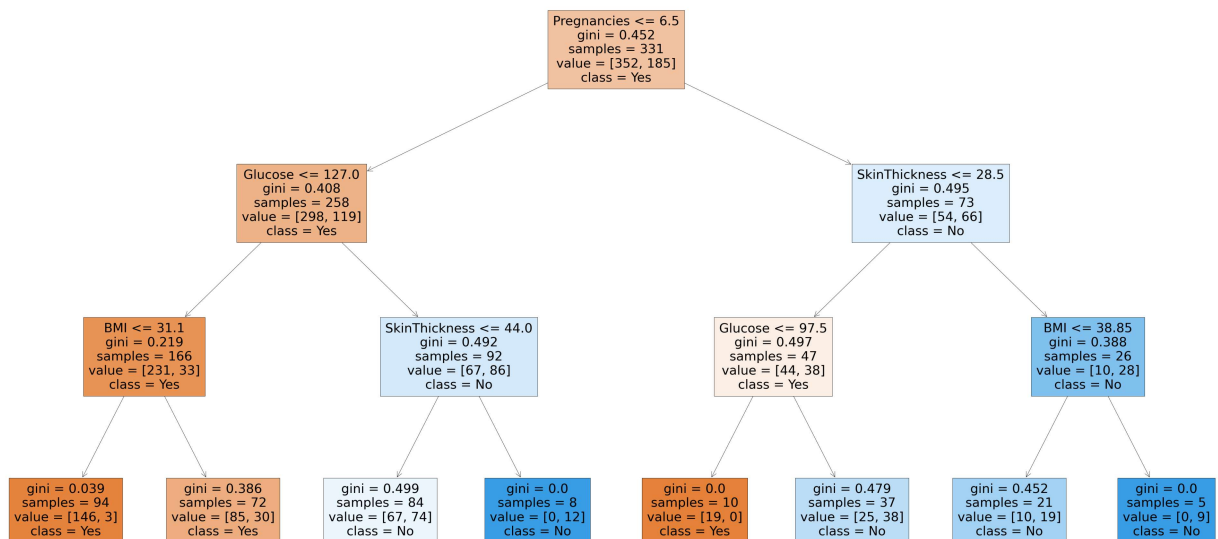
```
Out[20]: 0.7802599456250346
```

```
In [21]: from sklearn.tree import plot_tree
```

```
In [22]: rfc_best=grid_search.best_estimator_
```

```
In [23]: plt.figure(figsize=(80,40))
plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['Yes','No'],filled=True)
```

```
Out[23]: [Text(2232.0, 1902.6000000000001, 'Pregnancies <= 6.5\ngini = 0.452\nsamples = 331\nvalue = [352, 185]\nnclass = Yes'),
Text(1116.0, 1359.0, 'Glucose <= 127.0\ngini = 0.408\nsamples = 258\nvalue = [298, 119]\nnclass = Yes'),
Text(558.0, 815.4000000000001, 'BMI <= 31.1\ngini = 0.219\nsamples = 166\nvalue = [231, 33]\nnclass = Yes'),
Text(279.0, 271.79999999999995, 'gini = 0.039\nsamples = 94\nvalue = [146, 3]\nnclass = Yes'),
Text(837.0, 271.79999999999995, 'gini = 0.386\nsamples = 72\nvalue = [85, 30]\nnclass = Yes'),
Text(1674.0, 815.4000000000001, 'SkinThickness <= 44.0\ngini = 0.492\nsamples = 92\nvalue = [67, 86]\nnclass = No'),
Text(1395.0, 271.79999999999995, 'gini = 0.499\nsamples = 84\nvalue = [67, 74]\nnclass = No'),
Text(1953.0, 271.79999999999995, 'gini = 0.0\nsamples = 8\nvalue = [0, 12]\nnclass = No'),
Text(3348.0, 1359.0, 'SkinThickness <= 28.5\ngini = 0.495\nsamples = 73\nvalue = [54, 66]\nnclass = No'),
Text(2790.0, 815.4000000000001, 'Glucose <= 97.5\ngini = 0.497\nsamples = 47\nvalue = [44, 38]\nnclass = Yes'),
Text(2511.0, 271.79999999999995, 'gini = 0.0\nsamples = 10\nvalue = [19, 0]\nnclass = Yes'),
Text(3069.0, 271.79999999999995, 'gini = 0.479\nsamples = 37\nvalue = [25, 38]\nnclass = No'),
Text(3906.0, 815.4000000000001, 'BMI <= 38.85\ngini = 0.388\nsamples = 26\nvalue = [10, 28]\nnclass = No'),
Text(3627.0, 271.79999999999995, 'gini = 0.452\nsamples = 21\nvalue = [10, 19]\nnclass = No'),
Text(4185.0, 271.79999999999995, 'gini = 0.0\nsamples = 5\nvalue = [0, 9]\nnclass = No')]
```



```
In [ ]:
```