```
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\C5_health care diabetes - C5_health care diabetes.csv
df

Out[2]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	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 [3]: 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 [4]: df['Outcome'].value_counts()
```

Out[4]: 0 500 1 268

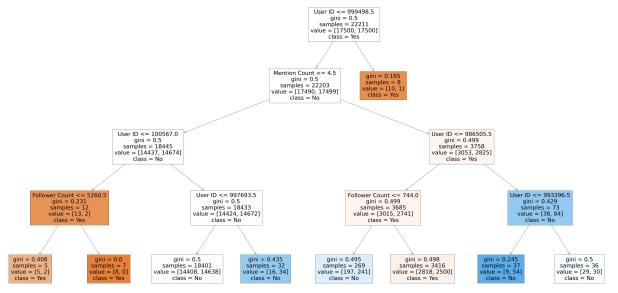
Name: Outcome, dtype: int64

In [5]: df1=df[['Pregnancies','Glucose','BloodPressure','SkinThickness','Insulin','BMI','Age','Outcome

```
In [6]: x=df1.drop('Outcome',axis=1)
          y=df['Outcome']
         g1={"1":{'0':1}}
 In [7]:
          df=df.replace(g1)
          print(df)
                                                                                BMI \
                                      BloodPressure SkinThickness
               Pregnancies
                            Glucose
                                                                    Insulin
                                                                 35
          0
                         6
                                 148
                                                 72
                                                                               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
                                                                 35
                         0
                                 137
                                                 40
                                                                         168
                                                                              43.1
                                                                . . .
          763
                                 101
                                                                         180
                                                                               32.9
                        10
                                                 76
                                                                 48
          764
                                 122
                                                 70
                                                                 27
                                                                              36.8
                         2
                                                                           0
          765
                                                 72
                         5
                                 121
                                                                 23
                                                                         112
                                                                               26.2
          766
                         1
                                 126
                                                 60
                                                                  0
                                                                           0
                                                                               30.1
                                                                 31
                                                                               30.4
          767
                         1
                                 93
                                                 70
               DiabetesPedigreeFunction
                                          Age
                                               Outcome
          0
                                   0.627
          1
                                   0.351
                                           31
                                                     0
                                   0.672
          2
                                           32
                                                      1
          3
                                   0.167
                                           21
                                                     0
          4
                                   2.288
                                           33
                                                      1
                                     . . .
                                          . . .
          763
                                   0.171
                                           63
                                                     0
          764
                                   0.340
                                           27
                                                     0
          765
                                   0.245
                                           30
                                                     0
          766
                                   0.349
                                           47
                                                     1
          767
                                   0.315
                                           23
                                                     0
          [768 rows x 9 columns]
 In [8]: from sklearn.model_selection import train_test_split
          x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.30)
 In [9]: from sklearn.ensemble import RandomForestClassifier
          rfc=RandomForestClassifier()
          rfc.fit(x_train,y_train)
 Out[9]: RandomForestClassifier()
In [10]:
         parameters={'max_depth':[1,2,3,4,5],
                     'min_samples_leaf':[5,10,15,20,25],
                     'n_estimators':[10,20,30,40,50]}
In [11]: 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[11]: 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 [12]: grid_search.best_score_
Out[12]: 0.7747045441935305
In [13]: rfc_best=grid_search.best_estimator_
```

```
In [14]: from sklearn.tree import plot tree
         plt.figure(figsize=(80,40))
         plot_tree(rfc_best.estimators_[5],feature_names=x.columns,class_names=['Yes','No'],filled=True
Out[14]: [Text(2511.0, 1956.96, 'User ID <= 999498.5\ngini = 0.5\nsamples = 22211\nvalue = [17500, 17
         500]\nclass = Yes'),
          Text(2232.0, 1522.0800000000002, 'Mention Count <= 4.5\ngini = 0.5\nsamples = 22203\nvalue
         = [17490, 17499]\nclass = No'),
          Text(1116.0, 1087.2, 'User ID <= 100567.0\ngini = 0.5\nsamples = 18445\nvalue = [14437, 146
         741\nclass = No'),
          Text(558.0, 652.3200000000002, 'Follower Count <= 5260.5\ngini = 0.231\nsamples = 12\nvalue
         = [13, 2]\nclass = Yes'),
          Text(279.0, 217.44000000000000, 'gini = 0.408\nsamples = 5\nvalue = [5, 2]\nclass = Yes'),
          Text(837.0, 217.44000000000005, 'gini = 0.0 \\ \noindent = 7 \\ \noindent = [8, 0] \\ \noindent = Yes'),
          Text(1674.0, 652.3200000000002, 'User ID <= 997693.5\ngini = 0.5\nsamples = 18433\nvalue =
          [14424, 14672]\nclass = No'),
          Text(1395.0, 217.44000000000005, 'gini = 0.5\nsamples = 18401\nvalue = [14408, 14638]\nclas
         s = No'),
          Text(1953.0, 217.44000000000005, 'gini = 0.435\nsamples = 32\nvalue = [16, 34]\nclass = N
         o'),
          Text(3348.0, 1087.2, 'User ID <= 986505.5\ngini = 0.499\nsamples = 3758\nvalue = [3053, 282
         5]\nclass = Yes'),
          Text(2790.0, 652.3200000000002, 'Follower Count <= 744.0\ngini = 0.499\nsamples = 3685\nval
         ue = [3015, 2741]\nclass = Yes'),
          Text(2511.0, 217.440000000000005, 'gini = 0.495\nsamples = 269\nvalue = [197, 241]\nclass =
         No'),
          Text(3069.0, 217.4400000000005, 'gini = 0.498\nsamples = 3416\nvalue = [2818, 2500]\nclass
         = Yes'),
          Text(3906.0, 652.3200000000002, 'User ID <= 993396.5\ngini = 0.429\nsamples = 73\nvalue =
          [38, 84]\nclass = No'),
          Text(3627.0, 217.44000000000005, 'gini = 0.245\nsamples = 37\nvalue = [9, 54]\nclass = N
          Text(4185.0, 217.4400000000005, 'gini = 0.5\nsamples = 36\nvalue = [29, 30]\nclass = No'),
          Text(2790.0, 1522.0800000000002, 'gini = 0.165\nsamples = 8\nvalue = [10, 1]\nclass = Ye
         s')]
```



In []: