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

In [2]: train_df=pd.read_csv(r"C:\Users\Mastan Reddy\Downloads\random forest1.csv")
 train_df

Out[2]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	 px_hei
0	842	0	2.2	0	1	0	7	0.6	188	2	
1	1021	1	0.5	1	0	1	53	0.7	136	3	 •
2	563	1	0.5	1	2	1	41	0.9	145	5	 12
3	615	1	2.5	0	0	0	10	0.8	131	6	 12
4	1821	1	1.2	0	13	1	44	0.6	141	2	 12
								•••			
1995	794	1	0.5	1	0	1	2	0.8	106	6	 12
1996	1965	1	2.6	1	0	0	39	0.2	187	4	 {
1997	1911	0	0.9	1	1	1	36	0.7	108	8	 }
1998	1512	0	0.9	0	4	1	46	0.1	145	5	 ;
1999	510	1	2.0	1	5	1	45	0.9	168	6	 2

2000 rows × 21 columns

In [3]: test_df=pd.read_csv(r"C:\Users\Mastan Reddy\Downloads\random forest2.csv")
test_df

Out[3]:

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	 рс	px_hei
0	1	1043	1	1.8	1	14	0	5	0.1	193	 16	
1	2	841	1	0.5	1	4	1	61	8.0	191	 12	7
2	3	1807	1	2.8	0	1	0	27	0.9	186	 4	12
3	4	1546	0	0.5	1	18	1	25	0.5	96	 20	1
4	5	1434	0	1.4	0	11	1	49	0.5	108	 18	7
995	996	1700	1	1.9	0	0	1	54	0.5	170	 17	(
996	997	609	0	1.8	1	0	0	13	0.9	186	 2	1'
997	998	1185	0	1.4	0	1	1	8	0.5	80	 12	4
998	999	1533	1	0.5	1	0	0	50	0.4	171	 12	
999	1000	1270	1	0.5	0	4	1	35	0.1	140	 19	2

1000 rows × 21 columns

```
In [4]: train_df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):
```

#	Column	Non-Null Count	Dtype
0	battery_power	2000 non-null	int64
1	blue	2000 non-null	int64
2	clock_speed	2000 non-null	float64
3	dual_sim	2000 non-null	int64
4	fc	2000 non-null	int64
5	four_g	2000 non-null	int64
6	int_memory	2000 non-null	int64
7	m_dep	2000 non-null	float64
8	mobile_wt	2000 non-null	int64
9	n_cores	2000 non-null	int64
10	рс	2000 non-null	int64
11	px_height	2000 non-null	int64
12	px_width	2000 non-null	int64
13	ram	2000 non-null	int64
14	sc_h	2000 non-null	int64
15	SC_W	2000 non-null	int64
16	talk_time	2000 non-null	int64
17	three_g	2000 non-null	int64
18	touch_screen	2000 non-null	int64
19	wifi	2000 non-null	int64
20	price_range	2000 non-null	int64
dtyp	es: float64(2),	int64(19)	
memo	ry usage: 328.2	KB	

In [5]: test_df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 21 columns):

Data	COTUMNIS (COCAT	. 21 CO1411113).							
#	Column	Non-Null Count	Dtype						
0	id	1000 non-null	int64						
1	battery_power	1000 non-null	int64						
2	blue	1000 non-null	int64						
3	clock_speed	1000 non-null	float64						
4	dual_sim	1000 non-null	int64						
5	fc	1000 non-null	int64						
6	four_g	1000 non-null	int64						
7	int_memory	1000 non-null	int64						
8	m_dep	1000 non-null	float64						
9	mobile_wt	1000 non-null	int64						
10	n_cores	1000 non-null	int64						
11	рс	1000 non-null	int64						
12	px_height	1000 non-null	int64						
13	px_width	1000 non-null	int64						
14	ram	1000 non-null	int64						
15	sc_h	1000 non-null	int64						
16	SC_W	1000 non-null	int64						
17	talk_time	1000 non-null	int64						
18	three_g	1000 non-null	int64						
19	touch_screen	1000 non-null	int64						
20	wifi	1000 non-null	int64						
<pre>dtypes: float64(2),</pre>		int64(19)							

memory usage: 164.2 KB

```
In [6]: x=train_df.drop('wifi',axis=1)
    y=train_df['wifi']

In [7]: x=test_df.drop('wifi',axis=1)
    y=test_df['wifi']

In [8]: train_df['dual_sim'].value_counts()

Out[8]: 1    1019
    0    981
    Name: dual_sim, dtype: int64

In [9]: test_df['blue'].value_counts()

Out[9]: 1    516
    0    484
    Name: blue, dtype: int64
```

```
In [10]: T={"Home Owner":{"Yes":1,"No":0}}
train_df=train_df.replace(T)
print(train_df)
```

<pre>print(train_df)</pre>													
	battery	_power	blue	clock	_speed	dual_	sim	fc	four	_g ir	nt_memo	ry \	
0		842	0		2.2		0	1		0		7	
1		1021	1		0.5		1	0		1		53	
2		563	1		0.5		1	2		1		41	
3		615	1		2.5		0	0		0		10	
4		1821	1		1.2		0	13		1		44	
 1995		 794	1		0.5			0	•	1	•	2	
1996		1965	1		2.6		1	0		0		39	
1997		1903	0		0.9		1	1		1		36	
1998		1512	0		0.9		0	4		1		46	
1999		510	1		2.6		1	5		1		45	
1999		310	_		2.6		_	,		_		40	
	m_dep	mobile_w	t n_	cores		px_heig	ht	px_wi	idth	ram	sc_h	SC_W	\
0	0.6	18	8	2			20		756	2549	9	7	
1	0.7	13	6	3		9	05	1	L988	2631	17	3	
2	0.9	14	5	5			63	1	L716	2603	11	2	
3	0.8	13		6		12	16		L786	2769	16	8	
4	0.6	14	1	2	• • •	12	.08	1	L212	1411	8	2	
• • •	• • •	• •		• • •	• • •		• •		• • •	• • •	• • •	• • •	
1995	0.8	10		6	• • •		22		L890	668	13	4	
1996	0.2	18		4	• • •		15		L965	2032	11	10	
1997	0.7	10		8	• • •		68	1	L632	3057	9	1	
1998	0.1	14		5	• • •		36		670	869	18	10	
1999	0.9	16	8	6	• • •	4	83		754	3919	19	4	
	talk ti	me thre	e g	touch_	screer	wifi	pri	ice ra	ange				
0	_	19	_0	_	. 6		•	_	1				
1		7	1		1	. 0			2				
2		9	1		1				2				
3		11	1		e	0			2				
4		15	1		1	. 0			1				
			• • •										
1995		19	1		1				0				
1996		16	1		1				2				
1997		5	1		1				3				
1998		19	1		1				0				
1999		2	1		1	. 1			3				

[2000 rows x 21 columns]

```
In [11]: T={"Home Owner":{"Yes":1,"No":0}}
train_df=train_df.replace(T)
print(train_df)
```

<pre>print(train_df)</pre>													
	battery	_power	blue	clock	_speed	dual_	sim	fc	four	_g ir	nt_memo	ry \	
0		842	0		2.2		0	1		0		7	
1		1021	1		0.5		1	0		1		53	
2		563	1		0.5		1	2		1		41	
3		615	1		2.5		0	0		0		10	
4		1821	1		1.2		0	13		1		44	
 1995		 794	1		0.5			0	•	1	•	2	
1996		1965	1		2.6		1	0		0		39	
1997		1903	0		0.9		1	1		1		36	
1998		1512	0		0.9		0	4		1		46	
1999		510	1		2.6		1	5		1		45	
1999		310	_		2.6		_	,		_		40	
	m_dep	mobile_w	t n_	cores		px_heig	ht	px_wi	idth	ram	sc_h	SC_W	\
0	0.6	18	8	2			20		756	2549	9	7	
1	0.7	13	6	3		9	05	1	L988	2631	17	3	
2	0.9	14	5	5			63	1	L716	2603	11	2	
3	0.8	13		6		12	16		L786	2769	16	8	
4	0.6	14	1	2	• • •	12	.08	1	L212	1411	8	2	
• • •	• • •	• •		• • •	• • •		• •		• • •	• • •	• • •	• • •	
1995	0.8	10		6	• • •		22		L890	668	13	4	
1996	0.2	18		4	• • •		15		L965	2032	11	10	
1997	0.7	10		8	• • •		68	1	L632	3057	9	1	
1998	0.1	14		5	• • •		36		670	869	18	10	
1999	0.9	16	8	6	• • •	4	83		754	3919	19	4	
	talk ti	me thre	e g	touch_	screer	wifi	pri	ice ra	ange				
0	_	19	_0	_	. 6		•	_	1				
1		7	1		1	. 0			2				
2		9	1		1				2				
3		11	1		e	0			2				
4		15	1		1	. 0			1				
			• • •										
1995		19	1		1				0				
1996		16	1		1				2				
1997		5	1		1				3				
1998		19	1		1				0				
1999		2	1		1	. 1			3				

[2000 rows x 21 columns]

```
In [12]: T={"Home Owner":{"Yes":1,"No":0}}
         test_df=test_df.replace(T)
         print(test_df)
                id battery_power blue clock_speed dual_sim fc four_g int_memory
         0
                             1043
                                                  1.8
                                                                                      5
         1
                              841
                                                  0.5
                                                                                     61
         2
                 3
                             1807
                                      1
                                                  2.8
                                                                1
                                                                                     27
         3
                 4
                             1546
                                      0
                                                  0.5
                                                              1 18
                                                                                     25
         4
                 5
                             1434
                                      0
                                                  1.4
                                                              0
                                                                 11
                                                                          1
                                                                                     49
                              . . .
                                                  . . .
                                                                                     . . .
          . .
         995
               996
                             1700
                                      1
                                                  1.9
                                                              0
                                                                  0
                                                                          1
                                                                                     54
         996
               997
                             609
                                      а
                                                  1.8
                                                              1
                                                                0
                                                                          а
                                                                                     13
         997
               998
                             1185
                                      0
                                                  1.4
                                                                1
                                                                                      8
         998
               999
                             1533
                                                  0.5
                                                                                     50
         999
              1000
                             1270
                                      1
                                                  0.5
                                                                                     35
              m_dep mobile_wt ...
                                      pc px_height px_width
                                                                ram
                                                                     sc_h
                                                                           SC_W
         0
                0.1
                           193
                                     16
                                                226
                                                         1412
                                                               3476
         1
                0.8
                           191 ... 12
                                               746
                                                          857
                                                               3895
                                                                        6
                                                                              0
         2
                0.9
                           186 ...
                                     4
                                               1270
                                                         1366 2396
                                                                       17
                                                                             10
         3
                0.5
                            96 ... 20
                                               295
                                                         1752
                                                               3893
                0.5
                           108 ...
                                     18
                                               749
                                                          810 1773
                                                                       15
                                                . . .
                . . .
                           . . .
                                      . .
                                                          . . .
                                                                . . .
         995
                0.5
                           170
                                      17
                                                               2121
                                               644
                                                         913
                                                                       14
                                                                              8
         996
                0.9
                           186 ...
                                      2
                                               1152
                                                         1632
                                                               1933
                                                                        8
                                                                              1
         997
                0.5
                            80 ...
                                     12
                                               477
                                                          825
                                                               1223
                                                                        5
                                                                              0
                                                          832 2509
         998
                0.4
                           171 ...
                                    12
                                                38
                                                                       15
                                                                             11
         999
                0.1
                           140 ...
                                    19
                                                457
                                                          608 2828
                                                                              2
              talk_time three_g touch_screen wifi
         0
                      2
                               0
         1
                      7
                               1
                                                    0
         2
                     10
                                             1
                                                    1
         3
                      7
                               1
                                             1
                                                    0
                      7
         4
                               1
                                              0
                                                    1
                     15
                                             1
         996
                     19
                               0
                                             1
                                                   1
         997
                     14
                               1
                                              0
                                                    0
         998
                      6
                                              1
                                                    0
         999
         [1000 rows x 21 columns]
In [13]: x=train df.drop('wifi',axis=1)
         y=train_df['wifi']
In [14]:
         x=test_df.drop('wifi',axis=1)
         y=test_df['wifi']
In [15]: | from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test = train_test_split(x,y,train_size=0.7,random_state=42)
         x_train.shape,x_test.shape
Out[15]: ((700, 20), (300, 20))
```

In a Jupyter environment, please rerun this cell to show the HTML representation or

```
In [16]: from sklearn.ensemble import RandomForestClassifier
    rfc = RandomForestClassifier()
    rfc.fit(x_train,y_train)

Out[16]: RandomForestClassifier()

In [17]: params={"max_depth":[2,3,5,10,20],'min_samples_leaf':[5,10,20,50,100,200],'n_estimators':[10,

In [18]: from sklearn.ensemble import RandomForestClassifier
    rfc = RandomForestClassifier()
    rfc.fit(x_train,y_train)

Out[18]: RandomForestClassifier()
```

In a Jupyter environment, please rerun this cell to show the HTML representation or

```
In [19]: rf = RandomForestClassifier()
 In [ ]: params = {'max_depth': [2,3,5,10,20],
          'min_samples_leaf': [5,10,20,50,100,200],
          'n_estimators': [10,25,30,50,100,200]}
In [22]: from sklearn.model selection import GridSearchCV
         grid search = GridSearchCV(estimator=rf,param grid=params,cv = 2, scoring='accuracy')
         grid_search.fit(x_train,y_train)
Out[22]: GridSearchCV(cv=2, estimator=RandomForestClassifier(),
                      param_grid={'max_depth': [2, 3, 5, 10, 20],
                                   'min_samples_leaf': [5, 10, 20, 50, 100, 200],
                                   'n_estimators': [10, 25, 30, 50, 100, 200]},
                      scoring='accuracy')
In [23]: grid_search.best_score_
Out[23]: 0.5585714285714286
In [24]: rf best = grid search.best estimator
         print(rf best)
```

RandomForestClassifier(max_depth=5, min_samples_leaf=100, n_estimators=200)

```
In [25]: from sklearn.tree import plot tree
          plt.figure(figsize=(80,40))
          plot tree(rf best.estimators [5], feature names = x.columns,class names=['Yes',"No"],filled=T
Out[25]: [Text(0.5, 0.83333333333333334, 'n_cores <= 4.5\ngini = 0.5\nsamples = 453\nvalue = [361, 33
          9]\nclass = Yes'),
           Text(0.25, 0.5, 'ram <= 1788.5 / ngini = 0.494 / nsamples = 240 / nvalue = [207, 167] / nclass = Ye
           Text(0.125, 0.1666666666666666, 'gini = 0.5\nsamples = 102\nvalue = [79, 77]\nclass = Ye
          s'),
           Text(0.375, 0.1666666666666666, 'gini = 0.485\nsamples = 138\nvalue = [128, 90]\nclass = Y
          es'),
           Text(0.75, 0.5, 'battery_power <= 1188.0\ngini = 0.498\nsamples = 213\nvalue = [154, 172]\n
          class = No'),
           Text(0.625, 0.1666666666666666, 'gini = 0.47\nsamples = 100\nvalue = [55, 91]\nclass = N
          o'),
           Text(0.875, 0.1666666666666666, 'gini = 0.495\nsamples = 113\nvalue = [99, 81]\nclass = Ye
          s')]
                                                    n_cores <= 4.5
                                                      gini = 0.5
                                                    samples = 453
                                                  value = [361, 339]
                                                      class = Yes
                           ram <= 1788.5
                                                                       battery power <= 1188.0
                             gini = 0.494
                                                                             gini = 0.498
                            samples = 240
                                                                            samples = 213
                          value = [207, 167]
                                                                           value = [154, 172]
                             class = Yes
                                                                              class = No
                                                                                          qini = 0.495
                  qini = 0.5
                                         qini = 0.485
                                                                  qini = 0.47
               samples = 102
                                        samples = 138
                                                                samples = 100
                                                                                        samples = 113
               value = [79, 77]
                                       value = [128, 90]
                                                                value = [55, 91]
                                                                                        value = [99, 81]
```

class = Yes

class = No

class = Yes

class = Yes

```
In [26]: from sklearn.tree import plot tree
         plt.figure(figsize=(80,40))
         plot tree(rf best.estimators [7],feature names=x.columns,class names=["Yes","No"],filled=True
Out[26]: [Text(0.5, 0.83333333333333334, 'four_g <= 0.5\ngini = 0.5\nsamples = 443\nvalue = [358, 342]
         \nclass = Yes'),
         Text(0.25, 0.5, 'blue <= 0.5\ngini = 0.499\nsamples = 227\nvalue = [171, 186]\nclass = N
         Text(0.125, 0.1666666666666666, 'gini = 0.488\nsamples = 111\nvalue = [73, 100]\nclass = N
         Text(0.375, 0.1666666666666666, 'gini = 0.498\nsamples = 116\nvalue = [98, 86]\nclass = Ye
         Text(0.75, 0.5, 'id <= 540.5\ngini = 0.496\nsamples = 216\nvalue = [187, 156]\nclass = Ye
         Text(0.625, 0.166666666666666, 'gini = 0.482\nsamples = 113\nvalue = [107, 73]\nclass = Y
         Text(0.875, 0.16666666666666666, 'gini = 0.5\nsamples = 103\nvalue = [80, 83]\nclass = N
         o')]
                                             four g <= 0.5
                                                aini = 0.5
                                             samples = 443
                                           value = [358, 342]
                                               class = Yes
                        blue \leq = 0.5
                                                                    id <= 540.5
                        gini = 0.499
                                                                    gini = 0.496
                       samples = 227
                                                                   samples = 216
                     value = [171, 186]
                                                                 value = [187, 156]
                         class = No
                                                                     class = Yes
             gini = 0.488
                                   gini = 0.498
                                                         gini = 0.482
                                                                                 gini = 0.5
            samples = 111
                                                                              samples = 103
                                  samples = 116
                                                        samples = 113
                                 value = [98, 86]
           value = [73, 100]
                                                      value = [107, 73]
                                                                             value = [80, 83]
              class = No
                                    class = Yes
                                                          class = Yes
                                                                                class = No
```

```
In [28]: imp_df = pd.DataFrame({"Vername": x_train.columns,"Imp": rf_best.feature_importances_})
imp_df.sort_values(by="Imp", ascending=False)
```

Out[28]:

	Vername	Imp
13	px_width	0.177158
9	mobile_wt	0.095870
5	fc	0.088410
8	m_dep	0.072922
3	clock_speed	0.070815
1	battery_power	0.067856
7	int_memory	0.067008
14	ram	0.057543
12	px_height	0.056196
0	id	0.052923
17	talk_time	0.045026
16	sc_w	0.034381
6	four_g	0.028103
11	рс	0.026042
15	sc_h	0.020477
4	dual_sim	0.015370
2	blue	0.011788
10	n_cores	0.010056
19	touch_screen	0.002056
18	three_g	0.000000

In []: