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

Out[2]:		id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	 рс	px_height	px_width	ram	sc_h	sc _.
-	0	1	1043	1	1.8	1	14	0	5	0.1	193	 16	226	1412	3476	12	
	1	2	841	1	0.5	1	4	1	61	0.8	191	 12	746	857	3895	6	
	2	3	1807	1	2.8	0	1	0	27	0.9	186	 4	1270	1366	2396	17	
	3	4	1546	0	0.5	1	18	1	25	0.5	96	 20	295	1752	3893	10	
	4	5	1434	0	1.4	0	11	1	49	0.5	108	 18	749	810	1773	15	
	995	996	1700	1	1.9	0	0	1	54	0.5	170	 17	644	913	2121	14	
	996	997	609	0	1.8	1	0	0	13	0.9	186	 2	1152	1632	1933	8	
	997	998	1185	0	1.4	0	1	1	8	0.5	80	 12	477	825	1223	5	
	998	999	1533	1	0.5	1	0	0	50	0.4	171	 12	38	832	2509	15	
	999	1000	1270	1	0.5	0	4	1	35	0.1	140	 19	457	608	2828	9	

1000 rows × 21 columns

In [3]: test_df=pd.read_csv(r"C:\Users\yasoda\Documents\202U1A05C1\Mobile_Price_Classification_train.csv")
 test_df

Out[3]:		battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	 px_height	px_width	ram	sc_h	sc _.
_	0	842	0	2.2	0	1	0	7	0.6	188	2	 20	756	2549	9	
	1	1021	1	0.5	1	0	1	53	0.7	136	3	 905	1988	2631	17	
	2	563	1	0.5	1	2	1	41	0.9	145	5	 1263	1716	2603	11	
	3	615	1	2.5	0	0	0	10	8.0	131	6	 1216	1786	2769	16	
	4	1821	1	1.2	0	13	1	44	0.6	141	2	 1208	1212	1411	8	
	1995	794	1	0.5	1	0	1	2	8.0	106	6	 1222	1890	668	13	
	1996	1965	1	2.6	1	0	0	39	0.2	187	4	 915	1965	2032	11	
	1997	1911	0	0.9	1	1	1	36	0.7	108	8	 868	1632	3057	9	
	1998	1512	0	0.9	0	4	1	46	0.1	145	5	 336	670	869	18	
	1999	510	1	2.0	1	5	1	45	0.9	168	6	 483	754	3919	19	

2000 rows × 21 columns

•

In [4]: train_df.info()

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

Data	COTUMNIS (COCAT	ZI COIUMIIS).	
#	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
dtvne	es: float64(2).	int64(19)	

dtypes: float64(2), int64(19)

memory usage: 164.2 KB

```
In [5]: test df.info()
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 2000 entries, 0 to 1999
        Data columns (total 21 columns):
             Column
                            Non-Null Count Dtype
             battery power 2000 non-null
                                            int64
                            2000 non-null
             blue
                                            int64
             clock_speed
                            2000 non-null
                                            float64
             dual sim
                            2000 non-null
                                            int64
                            2000 non-null
         4
             fc
                                            int64
             four g
                            2000 non-null
                                            int64
                            2000 non-null
             int memory
                                            int64
                            2000 non-null
                                            float64
             m dep
             mobile wt
                            2000 non-null
                                            int64
             n cores
                            2000 non-null
                                            int64
                            2000 non-null
         10
             рс
                                            int64
         11 px height
                            2000 non-null
                                            int64
         12 px width
                            2000 non-null
                                            int64
         13 ram
                            2000 non-null
                                            int64
                            2000 non-null
             sc h
         14
                                            int64
         15 sc w
                            2000 non-null
                                            int64
                            2000 non-null
         16 talk time
                                            int64
                            2000 non-null
         17 three g
                                            int64
         18 touch screen
                            2000 non-null
                                            int64
         19 wifi
                            2000 non-null
                                            int64
         20 price_range
                            2000 non-null
                                            int64
        dtypes: float64(2), int64(19)
        memory usage: 328.2 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]: dual_sim
    1    517
    0    483
    Name: count, dtype: int64

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

Out[9]: blue
    0    1010
    1    990
    Name: count, dtype: int64
```

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

0 1 1043 1 1.8 1 14 0 5 1 1 2 841 1 0.5 1 4 1 61 2 3 1807 1 2.8 0 1 0 27 3 4 1546 0 0.5 1 18 1 25 4 5 1434 0 1.4 0 11 1 49		id	battery_po	wer	blue	clock_spe	ed dual_s:	im fc	four_	g int	_memory	,
2 3 1807 1 2.8 0 1 0 27 3 4 1546 0 0.5 1 18 1 25 4 5 1434 0 1.4 0 11 1 49 995 996 1700 1 1.9 0 0 1 54 996 997 609 0 1.8 1 0 0 13 997 998 1185 0 1.4 0 1 1 8 998 999 1533 1 0.5 1 0 0 50 999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w	0	1	1	.043	1	1	.8	1 14			5	\
3 4 1546 0 0.5 1 18 1 25 4 5 1434 0 1.4 0 11 1 49 995 996 1700 1 1.9 0 0 1 54 996 997 609 0 1.8 1 0 0 13 997 998 1185 0 1.4 0 1 1 8 998 999 1533 1 0.5 1 0 0 50 999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w	1	2		841	1	0	.5	1 4		1	61	
4 5 1434 0 1.4 0 11 1 49	2	3	1	807	1	2	.8	0 1		0	27	ļ
	3	4	1	546	0	0	.5	1 18		1	25)
995 996 1700 1 1.9 0 0 1 54 996 997 609 0 1.8 1 0 0 13 997 998 1185 0 1.4 0 1 1 8 998 999 1533 1 0.5 1 0 0 50 999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w	4	5	1	434	0	1	.4	0 11		1	49	1
996 997 609 0 1.8 1 0 0 13 997 998 1185 0 1.4 0 1 1 8 998 999 1533 1 0.5 1 0 0 50 999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w					• • •							
997 998 1185 0 1.4 0 1 1 8 998 999 1533 1 0.5 1 0 0 50 999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w						1	.9			1	54	•
998 999 1533 1 0.5 1 0 0 50 999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w		997		609	0	1	.8			0		
999 1000 1270 1 0.5 0 4 1 35 m_dep mobile_wt pc px_height px_width ram sc_h sc_w	997	998	1	185	0	1	.4			1	8	i
<pre>m_dep mobile_wt pc px_height px_width ram sc_h sc_w</pre>	998	999	1	.533	1	0	.5	1 0		0	50	į.
	999	1000	1	270	1	0	.5	0 4		1	35	,
0 0.1 193 16 226 1412 34/6 12 / \	•				-					_	,	
4 00 101 10 746 057 2005 6 0											\	
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 10 0												
4 0.5 108 18 749 810 1773 15 8	4									8		
										• • •		
995 0.5 170 17 644 913 2121 14 8												
996 0.9 186 2 1152 1632 1933 8 1												
997 0.5 80 12 477 825 1223 5 0												
998 0.4 171 12 38 832 2509 15 11												
999 0.1 140 19 457 608 2828 9 2	999	0.1	140	• • •	19	457	608	2828	9	2		
talk_time three_g touch_screen wifi		talk ·	time three	σt	ouch	screen wi	fi					
0 2 0 1 0	9	- Cu_										
1 7 1 0 0												
2 10 0 1 1												
3 7 1 1 0												
4 7 1 0 1												
• • • • • • • • • • • • • • • • • • • •							••					
995 15 1 1 0				1		1	0					
996 19 0 1 1												
997 14 1 0 0												
998 6 0 1 0												
999 3 1 0 1												

[1000 rows x 21 columns]

```
In [12]: T={"Home Owner":{"Yes":1,"No":0}}
test_df=test_df.replace(T)
print(test_df)
```

0 1 2 3 4 	16 5 6 18	wer blue 342 0 321 1 563 1 515 1 321 1 	clock_	_speed 2.2 0.5 0.5 2.5 1.2 	dual_sim 0 1 1 0 	fc four 1 0 2 0 13	_g in		ry 7 \ 53 41 10 44 	
1996		965 1		2.6	1	0	0		39	
1997		911 0		0.9	1	1	1		36	
1998	15	512 0		0.9	0	4	1		46	
1999		510 1		2.0	1	5	1		45	
			_cores		x_height	px_width	ram	sc_h	SC_W	
0	0.6	 188	2		20	· _ 756	2549	_ 9	_ 7	\
1	0.7	136	3		905	1988	2631	17	3	
2	0.9	145	5		1263	1716	2603	11	2	
3	0.8	131	6		1216	1786	2769	16	8	
4	0.6	141	2		1208	1212	1411	8	2	
	• • •	• • •	• • •			• • •				
1995	0.8	106	6	• • •	1222	1890	668	13	4	
1996	0.2	187	4	• • •	915	1965	2032	11	10	
1997	0.7	108	8	• • •	868	1632	3057	9	1	
1998	0.1	145	5	• • •	336	670	869	18	10	
1999	0.9	168	6	• • •	483	754	3919	19	4	
	talk_time	three_g	touch_s			ice_range				
0	19	0		0	1	1				
1	7	1		1	0	2				
2	9	1		1	0	2				
3	11	1		0	0	2				
4	15	1		1	0	1				
1005		• • •			•••	• • • •				
1995	19 16	1		1 1	0	0				
1996 1997	16 5	1 1		1	1 0	2				
1997	5 19	1		1	1					
1998	2	1		1	1	9 3				
エフフフ	2	1		т.	7	3				

[2000 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]: ((1400, 20), (600, 20))

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

Out[16]: RandomForestClassifier()
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
```

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

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [17]: rf = RandomForestClassifier()

In [18]: 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]}
```

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

```
In [20]: grid_search.best_score_
Out[20]: 0.5228571428571429
In [21]: rf_best = grid_search.best_estimator_
    print(rf_best)

RandomForestClassifier(max depth=5, min samples leaf=100, n estimators=50)
```

```
In [22]: 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=True)

Out[22]: [Text(0.5909090909090909, 0.875, 'ram <= 2731.5\ngini = 0.499\nsamples = 864\nvalue = [723, 677]\nclass = Yes'),</pre>
```

ram <= 2731.5 gini = 0.499 samples = 864 value = [723, 677] class = Yes

talk_time <= 10.5 gini = 0.495 samples = 571 value = [501, 408] class = Yes

sc_w <= 5.5 gini = 0.495 samples = 293 value = [222, 269] class = No

ram <= 1848.0 gini = 0.5 samples = 269 value = [212, 209] class = Yes talk_time <= 14.5 gini = 0.483 samples = 302 value = [289, 199] class = Yes

gini = 0.5 samples = 166 value = [140, 144] class = No

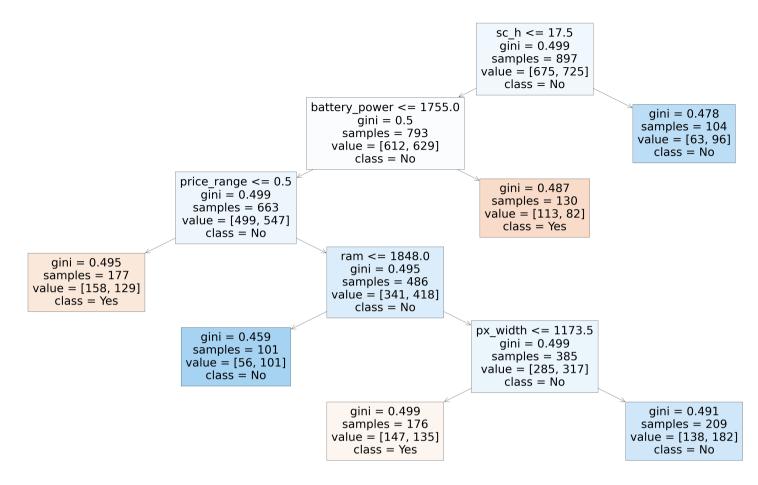
gini = 0.478 samples = 127 value = [82, 125] class = No

gini = 0.499 samples = 169 value = [124, 138] class = No gini = 0.494 samples = 100 value = [88, 71] class = Yes

gini = 0.444 samples = 108 value = [116, 58] class = Yes gini = 0.495 samples = 194 value = [173, 141] class = Yes

Text(0.666666666666666, 0.58333333333333333334, 'gini = 0.487\nsamples = 130\nvalue = [113, 82]\nclass = Yes'),

Text(0.83333333333334, 0.75, 'gini = 0.478\nsamples = 104\nvalue = [63, 96]\nclass = No')]



```
In [24]: rf_best.feature_importances_
Out[24]: array([0.07717307, 0.00879 , 0.04434522, 0.01953574, 0.03032118,
```

```
0.01928442, 0.08756478, 0.03099338, 0.05000981, 0.02300338, 0.06283458, 0.16018601, 0.07803434, 0.06519901, 0.05008368, 0.04774743, 0.10287966, 0.00509412, 0.00930888, 0.02761129])
```

```
In [25]: imp_df = pd.DataFrame({"Vername": x_train.columns,"Imp": rf_best.feature_importances_})
imp_df.sort_values(by="Imp", ascending=False)
```

Out[25]:

	Vername	lmp
11	px_height	0.160186
16	talk_time	0.102880
6	int_memory	0.087565
12	px_width	0.078034
0	battery_power	0.077173
13	ram	0.065199
10	рс	0.062835
14	sc_h	0.050084
8	mobile_wt	0.050010
15	sc_w	0.047747
2	clock_speed	0.044345
7	m_dep	0.030993
4	fc	0.030321
19	price_range	0.027611
9	n_cores	0.023003
3	dual_sim	0.019536
5	four_g	0.019284
18	touch_screen	0.009309
1	blue	0.008790
17	three_g	0.005094