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\thara\Downloads\Mobile_Price_Classification_train.csv")
train_df

Out[2]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_\
0	842	0	2.2	0	1	0	7	0.6	18
1	1021	1	0.5	1	0	1	53	0.7	10
2	563	1	0.5	1	2	1	41	0.9	14
3	615	1	2.5	0	0	0	10	8.0	10
4	1821	1	1.2	0	13	1	44	0.6	14
1995	794	1	0.5	1	0	1	2	8.0	1(
1996	1965	1	2.6	1	0	0	39	0.2	18
1997	1911	0	0.9	1	1	1	36	0.7	1(
1998	1512	0	0.9	0	4	1	46	0.1	14
1999	510	1	2.0	1	5	1	45	0.9	16

2000 rows × 21 columns

localhost:8888/notebooks/Random Forest-Mobile Price Classification.ipynb

In [3]:

test_df=pd.read_csv(r"C:\Users\thara\Downloads\Mobile_Price_Classification_test.csv")
test_df

Out[3]:

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mc
0	1	1043	1	1.8	1	14	0	5	0.1	
1	2	841	1	0.5	1	4	1	61	0.8	
2	3	1807	1	2.8	0	1	0	27	0.9	
3	4	1546	0	0.5	1	18	1	25	0.5	
4	5	1434	0	1.4	0	11	1	49	0.5	
995	996	1700	1	1.9	0	0	1	54	0.5	
996	997	609	0	1.8	1	0	0	13	0.9	
997	998	1185	0	1.4	0	1	1	8	0.5	
998	999	1533	1	0.5	1	0	0	50	0.4	
999	1000	1270	1	0.5	0	4	1	35	0.1	

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):

	00-0		
#	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

dtypes: float64(2), int64(19)

memory 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):
     Column
                    Non-Null Count
 #
                                     Dtype
                     -----
 0
                     1000 non-null
                                     int64
     id
 1
                    1000 non-null
                                     int64
     battery_power
 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
                    1000 non-null
                                     float64
     m dep
 9
     mobile wt
                    1000 non-null
                                     int64
 10
                     1000 non-null
     n_cores
                                     int64
                     1000 non-null
 11
                                     int64
     рс
 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
                     1000 non-null
                                     int64
     touch_screen
 20
     wifi
                    1000 non-null
                                     int64
dtypes: float64(2), 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]:
dual sim
     1019
1
      981
Name: count, dtype: int64
```

In [9]:

```
test_df['blue'].value_counts()
```

Out[9]:

blue

516
 484

Name: count, dtype: int64

In [10]:

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

0	battery_p	ower b] 842	Lue 0	clock	_speed_ 2.2		sim 0	fc 1	four	_g ir 0	nt_memo	ory 7
\ 1		1021	1		0.	5	1	0		1		53
2		563	1		0.		1	2		1		41
3		615	1		2.		0	0		0		10
4		1821	1		1.	2	0	13		1		44
• • •		• • • •	• • •		• •		• • •	• •	•	• •	•	• •
1995		794	1		0.		1	0		1		2
1996 1997		1965 1911	1 0		2.0 0.9		1 1	0 1		0 1		39 36
1998		1512	0		0.9		0	4		1		46
1999		510	1		2.0		1	5		1		45
W	m_dep mc	bile_wt	n_c	ores	• • •	px_heig	ght	px_wi	dth	ram	sc_h	sc_
0 7 \	0.6	188		2	•••		20	-	756	2549	9	
1 3	0.7	136		3	• • •	9	05	19	988	2631	17	
2	0.9	145		5	• • •	12	263	17	716	2603	11	
3	0.8	131		6	• • •	12	16	17	786	2769	16	
6 4 2	0.6	141		2	• • •	12	208	12	212	1411	8	
•••	• • •	• • •		•••	• • •	•	••		• • •	• • •	• • •	
1995 4	0.8	106		6	•••	12	222	18	390	668	13	
1996 0	0.2	187		4	•••	9	15	19	965	2032	11	1
1997 1	0.7	108		8	• • •	8	868	10	532	3057	9	
1998 0	0.1	145		5	• • •	3	36	(570	869	18	1
1999 4	0.9	168		6	•••	4	183	-	754	3919	19	
	talk_time	three_	_g t	ouch_	scree	n wifi	pri	ice_raı	nge			
0	19		0			9 1			1			
1	7		1			1 0			2			
2	9 11		1 1			1 0 9 0			2 2			
3 4	15		1			0 1 0			1			
•••	• • • •				• •							
1995	19		1		:	1 0			0			
1996	16		1			1 1			2			
1997	5		1			1 0			3			
1998 1999	19 2		1 1			1 1 1 1			0 3			
1777	2	-	1		•	. 1			ر			

[2000 rows x 21 columns]

In [11]:

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

	id	batt	ery_p	ower	blue	clock_sp	eed	dual_si	m fc	four_	g i	nt_mem
ory 0	1			1043	1		1.8		1 14	(0	
5 \ 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 ••											•	
 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	1000			1270	1		0.5		0 4	,	1	
0	m_dep		ile_w		•	px_heigh	-	x_width		_	sc_	
0 1	0.1 0.8		19 19			22 74		1412 857	3476 3895			7 \ 0
2	0.9		18		. 4	127		1366	2396		1	
3	0.5				. 20	29		1752	3893			0
4	0.5		16	8	18	74	19	810	1773	15		8
 995	0.5		 17			64		913	2121	 14	• •	• 8
996	0.9		18		_	115		1632	1933			1
997	0.5			30 .	. 12	47		825				0
998	0.4		17		. 12		88	832			1	1
999	0.1		14	10	19	45	57	608	2828	9		2
0	talk_	time 2	thre	ee_g 0	touch_	screen w	vifi 0					
1		7		1		0	0					
2		10		0		1	1					
3		7		1		1	0					
4		7		1		0	1					
 995		 15		1		1						
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]:
```

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

In [13]:

```
x=test_df.drop('wifi',axis=1)
y=test_df['wifi']
```

In [14]:

```
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[14]:

```
((700, 20), (300, 20))
```

In [15]:

```
from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier()
rfc.fit(x_train,y_train)
```

Out[15]:

```
RandomForestClassifier
RandomForestClassifier()
```

In [16]:

```
rf = RandomForestClassifier()
```

In [17]:

In [18]:

```
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[18]:

```
▶ GridSearchCV▶ estimator: RandomForestClassifier▶ RandomForestClassifier
```

In [19]:

```
grid_search.best_score_
```

Out[19]:

0.5557142857142857

In [20]:

```
rf_best = grid_search.best_estimator_
print(rf_best)
```

RandomForestClassifier(max_depth=3, min_samples_leaf=100, n_estimators=30)

In [21]:

```
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"],fill
```

```
pc <= 14.5
                                  gini = 0.498
                                samples = 436
                              value = [373, 327]
                                  class = Yes
               talk time \leq 13.5
                                                 gini = 0.466
                   gini = 0.5
                                                samples = 132
                 samples = 304
                                              value = [138, 81]
               value = [235, 246]
                                                  class = Yes
                   class = No
   gini = 0.495
                                  gini = 0.477
 samples = 187
                                samples = 117
value = [161, 132]
                               value = [74, 114]
   class = Yes
                                  class = No
```

In [22]:

```
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"],fill
                      px_width <= 935.0
                          gini = 0.5
                        samples = 446
                       value = [350, 350]
                          class = Yes
                                       talk time <= 8.5
          gini = 0.48
                                         gini = 0.496
         samples = 139
                                        samples = 307
        value = [85, 128]
                                      value = [265, 222]
           class = No
                                         class = Yes
                                                     int memory <= 32.5
                         gini = 0.498
                                                        gini = 0.487
                        samples = 105
                                                       samples = 202
                        value = [75, 84]
                                                      value = [190, 138]
                          class = No
                                                         class = Yes
                                         gini = 0.479
                                                                        gini = 0.494
                                        samples = 101
                                                                       samples = 101
                                       value = [97, 64]
                                                                      value = [93, 74]
                                          class = Yes
                                                                        class = Yes
```

In [23]:

```
rf_best.feature_importances_
```

Out[23]:

```
array([0.10650064, 0.0265796 , 0. , 0.01446986, 0. , 0.06388421, 0.00675413, 0.10724736, 0.07044479, 0.10650008, 0. , 0.04469659, 0.06148225, 0.1953958 , 0.03061033, 0.03388301, 0.03575791, 0.06476904, 0.01868639, 0.01233803])
```

In [24]:

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

Out[24]:

	Vername	Imp
13	px_width	0.195396
7	int_memory	0.107247
0	id	0.106501
9	mobile_wt	0.106500
8	m_dep	0.070445
17	talk_time	0.064769
5	fc	0.063884
12	px_height	0.061482
11	рс	0.044697
16	sc_w	0.035758
15	sc_h	0.033883
14	ram	0.030610
1	battery_power	0.026580
18	three_g	0.018686
3	clock_speed	0.014470
19	touch_screen	0.012338
6	four_g	0.006754
4	dual_sim	0.000000
2	blue	0.000000
10	n_cores	0.000000