In [1]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt,seaborn as sns
```

In [4]:

train_df=pd.read_csv(r"C:\Users\poorn\Downloads\test.csv")
train_df

Out[4]:

	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	8.0	
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 [5]:

test_df=pd.read_csv(r"C:\Users\poorn\Downloads\train.csv")
test_df

Out[5]:

	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	1:
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	0.8	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

In [6]:

```
train_df.info()
```

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

#	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

dtypes: float64(2), int64(19)

memory usage: 164.2 KB

In [7]:

```
test 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
                     2000 non-null
                                     int64
 2
     clock_speed
                     2000 non-null
                                     float64
 3
                                     int64
     dual sim
                     2000 non-null
 4
     fc
                     2000 non-null
                                     int64
 5
     four_g
                     2000 non-null
                                     int64
                     2000 non-null
                                     int64
 6
     int_memory
 7
     m dep
                     2000 non-null
                                     float64
                                     int64
 8
     mobile_wt
                     2000 non-null
     n_cores
 9
                     2000 non-null
                                     int64
 10
                     2000 non-null
                                     int64
     рс
                     2000 non-null
                                     int64
 11
     px_height
 12
                     2000 non-null
                                     int64
     px_width
 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
                     2000 non-null
                                     int64
 17
     three_g
     touch_screen
                     2000 non-null
                                     int64
 19
     wifi
                     2000 non-null
                                     int64
     price_range
                     2000 non-null
                                     int64
dtypes: float64(2), int64(19)
memory usage: 328.3 KB
In [8]:
x=train_df.drop('wifi',axis=1)
y=train_df['wifi']
In [9]:
x=test_df.drop('wifi',axis=1)
y=test_df['wifi']
In [10]:
train_df['dual_sim'].value_counts()
Out[10]:
dual sim
1
     517
     483
Name: count, dtype: int64
```

```
In [11]:
```

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

Out[11]:

blue

0 10101 990

Name: count, dtype: int64

In [12]:

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

	id	batt	ery_pow	ver	blue	clock_s	peed	dual_si	m fc	four_	g i	nt_mem
ory 0	1		16	943	1		1.8		1 14		0	
5 \ 1	2		8	841	1		0.5		1 4		1	
61 2	3			807	1		2.8		0 1		0	
27												
3 25	4			46	0		0.5		1 18		1	
4 49	5		14	134	0		1.4		0 11		1	
••	• • •		•	• •	•••		•••	• •		• •	•	
995 54	996		17	700	1		1.9		0 0		1	
996 13	997		6	609	0		1.8		1 0		0	
997	998		11	.85	0		1.4		0 1		1	
8 998	999		15	33	1		0.5		1 0		0	
50 999	1000		12	270	1		0.5		0 4		1	
35												
	m_dep		ile_wt					x_width		_	_	•
0	0.1		193				26	1412		12		7 \
1	0.8		191	• • •			46	857	3895	6		0
2	0.9		186	• • •		12		1366	2396	17		0
3	0.5		96	• • •			95	1752	3893	10		0
4	0.5		108	• • •	18	74	49	810	1773	15		8
• •	• • •		• • •	• • •			• •	• • •	• • •	• • •	• •	
995	0.5		170	• • •			44	913	2121	14		8
996	0.9		186	• • •		11:		1632	1933	8		1
997	0.5		80	• • •	12		77	825		5		0
998	0.4		171	• • •			38	832	2509	15	1	
999	0.1		140	• • •	19	4:	57	608	2828	9		2
	talk_	time	three_	_g t	ouch_	screen i	wifi					
0		2		0		1	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 [13]:

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

	id	battery_po	ower	blue	clock_spe	ed o	dual_sim	fc	four_g	int	_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	<u>:</u>	1546	0	0.	. 5	1	18	1		
25 4	5	:	1434	0	1.	.4	0	11	1		
49 ••					• •	•					
 995	996	<u>-</u>	1700	1	1.	.9	0	0	1		
54 996	997		609	0	1.		1		0		
13 997	998		1185	0	1.		0		1		
8 998	999		1533	1	0.		1		0		
50 999	1000		1270	1	0.		0		1		
35	1000	-	1270	_	0.		O	4	1		
	m_dep	-		•	px_height	px_	_width	ram	_	sc_w	
0	0.1				226			3476	12	7	\
1	0.8				746			3895	6	0	
2	0.9				1270			2396	17	10	
3	0.5				295			3893	10	0	
4	0.5				749			1773	15	8	
005		17/					012		11	•••	
995 996	0.5 0.9			_	644 1152			2121 1933	14 8	8 1	
997	0.5			12	477			1223	5	0	
998	0.4				38			2509	15	11	
999	0.1			. 19	457			2828	9	2	
	talk_	time three	e g	touch	screen wid	fi					
0	_	2	0	_	1	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	0					
996		15 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 [14]:

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

0	battery_p	ower blu 842	ue clock 0	_speed 2.2	dual_sim 0		_g ir 0	nt_memo	ory 7
\ 1		1021	1	0.5	1	0	1		53
2	•	563	1	0.5	1		1		41
3		615	1	2.5	0		0		10
4		1821	1	1.2	0		1		44
• • •				• • •	• • •	••			
1995		794	1	0.5	1		1		2
1996		1965	1	2.6	1		0		39
1997 1998		1911 1512	0 0	0.9 0.9	1 0		1 1		36 46
1999	•	510	1	2.0	1		1		45
	m_dep mol		n_cores			px_width		sc_h	SC_
W		_	_	••• }				_	36_
0 7 \	0.6	188	2	• • •	20	756	2549	9	
1 3	0.7	136	3	• • •	905	1988	2631	17	
2	0.9	145	5	• • •	1263	1716	2603	11	
2	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	0.2	187	4	• • •	915	1965	2032	11	1
1997 1	0.7	108	8	• • •	868	1632	3057	9	
1998 0	0.1	145	5	• • •	336	670	869	18	1
1999 4	0.9	168	6	•••	483	754	3919	19	
	talk time	three s	g touch_	screen	wifi pr	ice_range			
0	19)	0	1	1			
1	7	1	L	1	0	2			
2	9	1	l	1	0	2			
3	11	2		0	0	2			
4	15	-	L	1	0	1			
 1995	 19		•	1	 Ø				
1995	19	-		1	1	2			
1997	5	-		1	0	3			
1998	19	-		1	1	0			
1999	2	-		1	1	3			
			_						

[2000 rows x 21 columns]

```
In [15]:
x=train df.drop('wifi',axis=1)
y=train_df['wifi']
In [16]:
x=test_df.drop('wifi',axis=1)
y=test_df['wifi']
In [17]:
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[17]:
((1400, 20), (600, 20))
In [18]:
from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier()
rfc.fit(x_train,y_train)
Out[18]:
 ▼ RandomForestClassifier
RandomForestClassifier()
```

In a jupyter environment, please return this cell to show the HTML representation or

```
In [19]:

rf = RandomForestClassifier()

In [20]:

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 [21]:

```
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[21]:

```
► GridSearchCV
► estimator: RandomForestClassifier
► RandomForestClassifier
```

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

```
In [23]:
grid_search.best_score_
Out[23]:
0.52499999999999
In [24]:
rf_best = grid_search.best_estimator_
print(rf_best)
```

RandomForestClassifier(max_depth=10, min_samples_leaf=200, n_estimators=2
5)

In [27]:

```
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"],fil
```

Out[27]:

px_width <= 1358.0 gini = 0.499 samples = 907 value = [724, 676] class = Yes

sc_w <= 5.5 gini = 0.494 samples = 523 value = [453, 365] class = Yes

gini = 0.498 samples = 384 value = [271, 311] class = No

gini = 0.499 samples = 299 value = [234, 213] class = Yes gini = 0.484 samples = 224 value = [219, 152] class = Yes

In [28]:

```
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"],fil
```

Out[28]:

px_width <= 1358.0 gini = 0.499 samples = 907 value = [724, 676] class = Yes

sc_w <= 5.5 gini = 0.494 samples = 523 value = [453, 365] class = Yes

gini = 0.498 samples = 384 value = [271, 311] class = No

gini = 0.499 samples = 299 value = [234, 213] class = Yes gini = 0.484 samples = 224 value = [219, 152] class = Yes

In [29]:

```
rf_best.feature_importances_
```

Out[29]:

```
array([0.02475867, 0. , 0.03013278, 0.01426141, 0.01822027, 0. , 0.01232766, 0.06114952, 0.03368426, 0.02091575, 0.08832281, 0.17423659, 0.24921247, 0.06521139, 0. , 0.0437792 , 0.13098155, 0. , 0. , 0.03280566])
```

In [31]:

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

Out[31]:

	Vername	Imp
12	px_width	0.249212
11	px_height	0.174237
16	talk_time	0.130982
10	рс	0.088323
13	ram	0.065211
7	m_dep	0.061150
15	sc_w	0.043779
8	mobile_wt	0.033684
19	price_range	0.032806
2	clock_speed	0.030133
0	battery_power	0.024759
9	n_cores	0.020916
4	fc	0.018220
3	dual_sim	0.014261
6	int_memory	0.012328
1	blue	0.000000
5	four_g	0.000000
14	sc_h	0.000000
17	three_g	0.000000
18	touch_screen	0.000000

In []: