

In [1]:

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

In [2]:

```
df=pd.read_csv(r"C:\Users\DELL E5490\Downloads\Mobile_Price_Classification_test.csv")
df
```

Out[2]:

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	m
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 [3]:

```
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   pc                    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 [4]:

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

In [5]:

```
df['dual_sim'].value_counts()
```

Out[5]:

```
dual_sim
1      517
0      483
Name: count, dtype: int64
```

In [6]:

```
m={"three_g":{"yes":1,"No":0}}
df=df.replace(m)
print(df)
```

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_me
mory								
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								

	m_dep	mobile_wt	...	pc	px_height	px_width	ram	sc_h	sc_w
0	0.1	193	...	16	226	1412	3476	12	7 \
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
..
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

	talk_time	three_g	touch_screen	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	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 [7]:

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

In [8]:

```
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[8]:

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

In [9]:

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

Out[9]:

```
▼ RandomForestClassifier
RandomForestClassifier()
```

In [10]:

```
rf=RandomForestClassifier()
```

In [11]:

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

In [12]:

```
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[12]:

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

In [13]:

```
grid_search.best_score_
```

Out[13]:

0.5557142857142857

In [14]:

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

RandomForestClassifier(max_depth=3, min_samples_leaf=10, n_estimators=10)

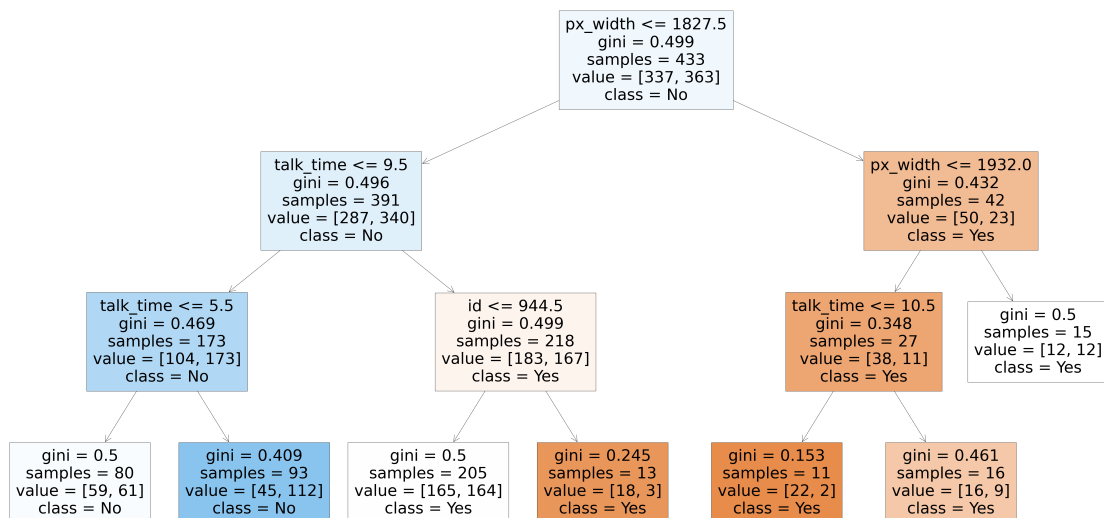
In [15]:

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

RandomForestClassifier(max_depth=3, min_samples_leaf=10, n_estimators=10)

In [16]:

```
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
```



In [17]:

```
rf_best.feature_importances_
```

Out[17]:

```
array([0.0175179 , 0.04594338, 0.          , 0.1026275 , 0.00491417,  
       0.01966802, 0.          , 0.06177251, 0.04614801, 0.19817225,  
       0.          , 0.02153329, 0.03253249, 0.07874952, 0.0943978 ,  
       0.05990252, 0.09720844, 0.11891219, 0.          , 0.          ])
```

In [18]:

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

Out[18]:

	Variance	Imp
9	mobile_wt	0.198172
17	talk_time	0.118912
3	clock_speed	0.102628
16	sc_w	0.097208
14	ram	0.094398
13	px_width	0.078750
7	int_memory	0.061773
15	sc_h	0.059903
8	m_dep	0.046148
1	battery_power	0.045943
12	px_height	0.032532
11	pc	0.021533
5	fc	0.019668
0	id	0.017518
4	dual_sim	0.004914
18	three_g	0.000000
10	n_cores	0.000000
6	four_g	0.000000
2	blue	0.000000
19	touch_screen	0.000000

Train Data

In [19]:

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

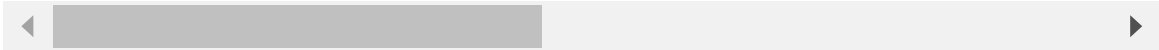
In [20]:

```
df=pd.read_csv(r"C:\Users\DELL E5490\Downloads\Mobile_Price_Classification_train (1).csv")
df
```

Out[20]:

	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	1
1	1021	1	0.5	1	0	1	53	0.7	1
2	563	1	0.5	1	2	1	41	0.9	1
3	615	1	2.5	0	0	0	10	0.8	1
4	1821	1	1.2	0	13	1	44	0.6	1
...	
1995	794	1	0.5	1	0	1	2	0.8	1
1996	1965	1	2.6	1	0	0	39	0.2	1
1997	1911	0	0.9	1	1	1	36	0.7	1
1998	1512	0	0.9	0	4	1	46	0.1	1
1999	510	1	2.0	1	5	1	45	0.9	1

2000 rows × 21 columns



In [21]:

```
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   pc              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 [22]:

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

In [23]:

```
df['dual_sim'].value_counts()
```

Out[23]:

```
dual_sim
1    1019
0     981
Name: count, dtype: int64
```


In [24]:

```
m={"three_g":{"yes":1,"No":0}}
df=df.replace(m)
print(df)
```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory
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	1	0	1	2
1996	1965	1	2.6	1	0	0	39
1997	1911	0	0.9	1	1	1	36
1998	1512	0	0.9	0	4	1	46
1999	510	1	2.0	1	5	1	45

	m_dep	mobile_wt	n_cores	...	px_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_screen	wifi	price_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
...
1995	19	1	1	0	0
1996	16	1	1	1	2
1997	5	1	1	0	3
1998	19	1	1	1	0
1999	2	1	1	1	3

[2000 rows x 21 columns]

In [25]:

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

In [26]:

```
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[26]:

```
((1400, 20), (600, 20))
```

In [27]:

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

Out[27]:

```
▼ RandomForestClassifier
RandomForestClassifier()
```

In [28]:

```
rf=RandomForestClassifier()
```

In [29]:

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

In [30]:

```
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[30]:

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

In [31]:

```
grid_search.best_score_
```

Out[31]:

0.5257142857142857

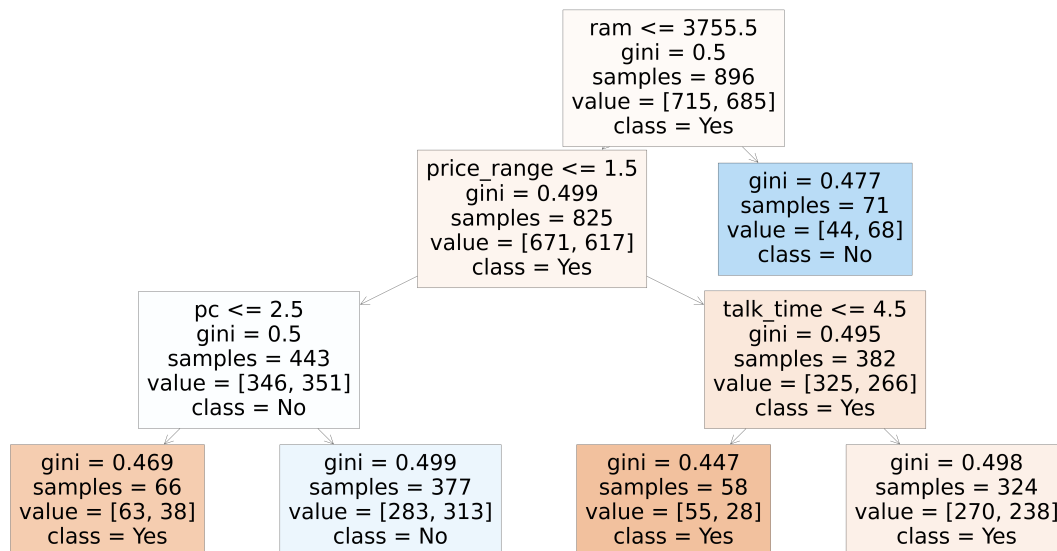
In [32]:

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

RandomForestClassifier(max_depth=3, min_samples_leaf=50, n_estimators=10)

In [33]:

```
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
```



In [34]:

```
rf_best.feature_importances_
```

Out[34]:

```
array([0.05325087, 0.01412726, 0.03137883, 0.          , 0.1464907 ,  
       0.          , 0.08165167, 0.0332542 , 0.10146693, 0.00419767,  
       0.09932768, 0.09396769, 0.04695801, 0.09348731, 0.00499327,  
       0.02108103, 0.08269192, 0.01371565, 0.03443456, 0.04352475])
```

In [35]:

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

Out[35]:

	Varname	IMP
4	fc	0.146491
8	mobile_wt	0.101467
10	pc	0.099328
11	px_height	0.093968
13	ram	0.093487
16	talk_time	0.082692
6	int_memory	0.081652
0	battery_power	0.053251
12	px_width	0.046958
19	price_range	0.043525
18	touch_screen	0.034435
7	m_dep	0.033254
2	clock_speed	0.031379
15	sc_w	0.021081
1	blue	0.014127
17	three_g	0.013716
14	sc_h	0.004993
9	n_cores	0.004198
5	four_g	0.000000
3	dual_sim	0.000000