RANDOM FOREST USING MOBILE DATASET

In [14]:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.tree import DecisionTreeClassifier
```

In [15]:

 $traindf=pd.read_csv(r"C:\Users\prajapath\ Arjun\Downloads\Mobile_Price_Classification_train.csv") traindf$

Out[15]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	 px_height
0	842	0	2.2	0	1	0	7	0.6	188	2	 20
1	1021	1	0.5	1	0	1	53	0.7	136	3	 905
2	563	1	0.5	1	2	1	41	0.9	145	5	 1263
3	615	1	2.5	0	0	0	10	8.0	131	6	 1216
4	1821	1	1.2	0	13	1	44	0.6	141	2	 1208
1995	794	1	0.5	1	0	1	2	8.0	106	6	 1222
1996	1965	1	2.6	1	0	0	39	0.2	187	4	 915
1997	1911	0	0.9	1	1	1	36	0.7	108	8	 868
1998	1512	0	0.9	0	4	1	46	0.1	145	5	 336
1999	510	1	2.0	1	5	1	45	0.9	168	6	 483

2000 rows × 21 columns

localhost:8888/notebooks/RANDOM FOREST.ipynb#RANDOM-FOREST-USING-MOBILE-DATASET

In [16]:

testdf=pd.read_csv(r"C:\Users\prajapath Arjun\Downloads\Mobile_Price_Classification_test.csv")
testdf

Out[16]:

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	 рс	px_height
0	1	1043	1	1.8	1	14	0	5	0.1	193	 16	226
1	2	841	1	0.5	1	4	1	61	8.0	191	 12	746
2	3	1807	1	2.8	0	1	0	27	0.9	186	 4	1270
3	4	1546	0	0.5	1	18	1	25	0.5	96	 20	295
4	5	1434	0	1.4	0	11	1	49	0.5	108	 18	749
		•••									 	
995	996	1700	1	1.9	0	0	1	54	0.5	170	 17	644
996	997	609	0	1.8	1	0	0	13	0.9	186	 2	1152
997	998	1185	0	1.4	0	1	1	8	0.5	80	 12	477
998	999	1533	1	0.5	1	0	0	50	0.4	171	 12	38
999	1000	1270	1	0.5	0	4	1	35	0.1	140	 19	457

1000 rows × 21 columns

Data columns (total 21 columns):

In [17]:

traindf.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999

#	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	<pre>mobile_wt</pre>	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 <u>g</u>	2000 non-null	int64
18	touch_screen	2000 non-null	int64
19	wifi	2000 non-null	int64
20	price_range	2000 non-null	int64
d+vn	oc. float64(2)	in+61(10)	

dtypes: float64(2), int64(19)
memory usage: 328.3 KB

In [18]:

```
testdf.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 1000 entries, 0 to 999 Data columns (total 21 columns): Non-Null Count Dtype # Column --------0 1000 non-null int64 id battery_power 1000 non-null int64 1 2 blue 1000 non-null int64 3 clock_speed 1000 non-null float64 4 dual_sim 1000 non-null int64 5 1000 non-null int64 fc four_g 1000 non-null int64 6 int_memory 7 1000 non-null int64 8 m_dep 1000 non-null float64 9 mobile_wt 1000 non-null int64 10 1000 non-null int64 n_cores 1000 non-null int64 11 рс 12 px_height 1000 non-null int64 1000 non-null 13 px_width int64 14 ram 1000 non-null int64 1000 non-null 15 sc_h 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

1000 non-null

int64

dtypes: float64(2), int64(19)

memory usage: 164.2 KB

In [19]:

20 wifi

traindf.shape,testdf.shape

Out[19]:

((2000, 21), (1000, 21))

In [20]:

traindf=traindf.head(1000)
traindf

Out[20]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	 px_height	p
0	842	0	2.2	0	1	0	7	0.6	188	2	 20	
1	1021	1	0.5	1	0	1	53	0.7	136	3	 905	
2	563	1	0.5	1	2	1	41	0.9	145	5	 1263	
3	615	1	2.5	0	0	0	10	0.8	131	6	 1216	
4	1821	1	1.2	0	13	1	44	0.6	141	2	 1208	
995	1456	0	1.6	1	5	0	49	0.2	193	3	 1285	
996	774	0	0.5	1	2	1	10	0.5	188	2	 1480	
997	1068	0	0.5	1	0	1	19	0.9	197	8	 322	
998	1373	1	1.9	1	1	1	29	0.9	141	6	 1220	
999	1777	1	3.0	0	3	0	20	0.6	188	6	 511	

1000 rows × 21 columns

```
In [21]:
traindf.shape,testdf.shape
Out[21]:
((1000, 21), (1000, 21))
In [22]:
X=testdf
y=traindf['price_range']
X_{\text{train}}, X_{\text{test}}, y_{\text{train}}, y_{\text{test}} = \text{train\_test\_split}(X, y, \text{train\_size} = 0.7, \text{random\_state} = 42)
In [23]:
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(X_train,y_train)
Out[23]:
 ▼ RandomForestClassifier
 RandomForestClassifier()
In [24]:
rf=RandomForestClassifier()
In [25]:
rams={'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 [26]:
from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rf,param_grid=params,cv=2,scoring="accuracy")
In [27]:
grid_search.fit(X_train,y_train)
grid_search.best_score_
Out[27]:
0.2914285714285714
In [28]:
rf_best=grid_search.best_estimator_
rf_best
```

Out[28]:

```
RandomForestClassifier
RandomForestClassifier(max_depth=2, min_samples_leaf=100, n_estimators=10)
```

```
In [29]:
```

```
traindf['price_range'].value_counts()
```

Out[29]:

In [30]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[4],feature_names=X.columns,class_names=['3','2','1','0'],filled=True);
```

```
\begin{array}{c} \text{px\_height} <= 852.0 \\ \text{gini} = 0.749 \\ \text{samples} = 451 \\ \text{value} = [175, 160, 192, 173] \\ \text{class} = 1 \end{array}
```

```
sc_h \le 10.5

gini = 0.748

samples = 317

value = [135, 115, 114, 135]

class = 3
```

gini = 0.724 samples = 134 value = [40, 45, 78, 38] class = 1

```
gini = 0.741
samples = 119
value = [46, 43, 43, 65]
class = 0
```

gini = 0.747 samples = 198 value = [89, 72, 71, 70] class = 3

In [32]:

```
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[5],feature_names=X.columns,class_names=['3','2','1','0'],filled=True);
```

```
battery_power <= 1610.0

gini = 0.748

samples = 452

value = [159, 160, 184, 197]

class = 0
```

```
talk_time <= 14.5

gini = 0.749

samples = 343

value = [139, 119, 144, 127]

class = 1
```

gini = 0.707 samples = 109 value = [20, 41, 40, 70] class = 0

```
gini = 0.743
samples = 234
value = [111, 80, 90, 73]
class = 3
```

```
gini = 0.734
samples = 109
value = [28, 39, 54, 54]
class = 1
```

```
In [33]:
```

```
rf_best.feature_importances_
```

Out[33]:

```
array([0. , 0.12924496, 0.0332677 , 0.0242826 , 0.07767695, 0.13821046, 0. , 0. , 0.12434641, 0. , 0.0351815 , 0.05937534, 0.26316113, 0. , 0.02745117, 0.02390157, 0. , 0.0393994 , 0. , 0. , 0.0245008 ])
```

In [34]:

```
imp_df=pd.DataFrame({"Varname":X_train.columns,"Imp":rf_best.feature_importances_})
```

In [35]:

```
imp_df.sort_values(by="Imp",ascending=False)
```

Out[35]:

	Varname	lmp
12	px_height	0.263161
5	fc	0.138210
1	battery_power	0.129245
8	m_dep	0.124346
4	dual_sim	0.077677
11	рс	0.059375
17	talk_time	0.039399
10	n_cores	0.035182
2	blue	0.033268
14	ram	0.027451
20	wifi	0.024501
3	clock_speed	0.024283
15	sc_h	0.023902
9	mobile_wt	0.000000
7	int_memory	0.000000
13	px_width	0.000000
6	four_g	0.000000
16	sc_w	0.000000
18	three_g	0.000000
19	touch_screen	0.000000
0	id	0.000000