In [1]:

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

In [2]:

 $\label{train_df_pd_read_csv} train_df=pd.read_csv(r"C:\Users\DELL\Downloads\Mobile_Price_Classification_train.csv") train_df$

Out[2]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_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	0.8	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	0.8	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

In [3]:

test_df=pd.read_csv(r"C:\Users\DELL\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	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	0.8	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										

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```
In [4]:
                                                                                                                H
x=train_df.drop('price_range',axis=1)
y=train_df['price_range']
In [5]:
                                                                                                                H
train_df['four_g'].value_counts()
Out[5]:
four_g
     1043
1
      957
Name: count, dtype: int64
In [6]:
                                                                                                                M
x=train_df.drop('price_range',axis=1)
y=train_df['price_range']
In [7]:
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[7]:
((1400, 20), (600, 20))
In [8]:
                                                                                                                H
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(x_train,y_train)
Out[8]:
 ▼ RandomForestClassifier
RandomForestClassifier()
                                                                                                                M
In [9]:
rf=RandomForestClassifier()
In [10]:
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 [13]:
                                                                                                                H
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[13]:
             GridSearchCV
 ▶ estimator: RandomForestClassifier
       ▶ RandomForestClassifier
```

In [14]:

grid_search.best_score_

Out[14]:

0.8335714285714286

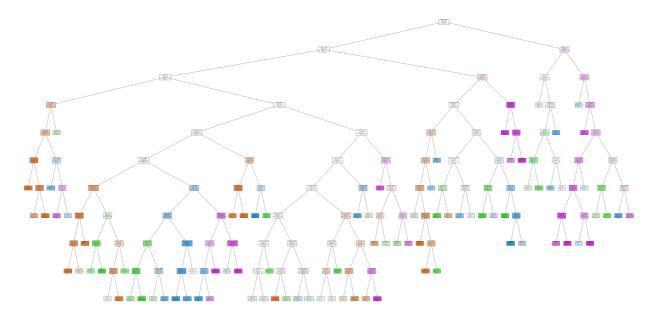
In [15]:

rf_best=grid_search.best_estimator_
print(rf_best)

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

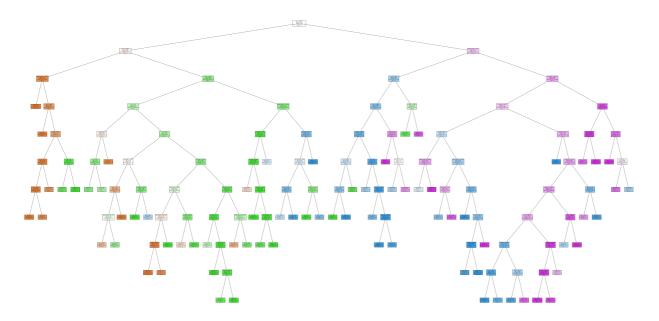
In [16]:

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



In [17]:

from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[7],feature_names=x.columns,class_names=['0','1','2','3'],filled=True);



Out[19]:

	Varname	lmp
13	ram	0.604540
0	battery_power	0.073189
11	px_height	0.048439
12	px_width	0.047258
8	mobile_wt	0.028895
6	int_memory	0.028666
16	talk_time	0.021548
10	рс	0.019843
2	clock_speed	0.019489
15	sc_w	0.018559
14	sc_h	0.017002
7	m_dep	0.016904
4	fc	0.016049
9	n_cores	0.014945
18	touch_screen	0.005634
5	four_g	0.004465
3	dual_sim	0.004094
1	blue	0.003913
19	wifi	0.003775
17	three_g	0.002793

```
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                                                        Mobile price prediction - Jupyter Notebook
  In [20]:
                                                                                                                            H
  imp_df=pd.DataFrame({"Varname":x_train.columns,"Imp":rf_best.feature_importances_})
  imp_df.sort_values(by="Imp",ascending=False)
  Out[20]:
          Varname
                       Imp
   13
                   0.604540
    0
      battery_power 0.073189
   11
          px_height 0.048439
   12
          px_width 0.047258
    8
          mobile_wt 0.028895
    6
        int_memory 0.028666
          talk_time 0.021548
   16
   10
               pc 0.019843
    2
        clock_speed 0.019489
   15
             sc_w 0.018559
              sc_h 0.017002
   14
    7
            m dep 0.016904
    4
                fc 0.016049
           n_cores 0.014945
    9
       touch_screen 0.005634
   18
             four_g 0.004465
    5
    3
           dual_sim 0.004094
    1
              blue 0.003913
   19
               wifi 0.003775
   17
            three_g 0.002793
  In [21]:
                                                                                                                            H
  test_df['four_g'].value_counts()
  Out[21]:
  four_g
  0
       513
  1
       487
  Name: count, dtype: int64
                                                                                                                            M
  In [22]:
  x=test_df.drop('wifi',axis=1)
  y=test_df['wifi']
  In [23]:
                                                                                                                            H
  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
```

```
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```

Out[23]:

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

```
In [25]:
from sklearn.ensemble import RandomForestClassifier
rfc=RandomForestClassifier()
rfc.fit(x_test,y_test)
Out[25]:
▼ RandomForestClassifier
RandomForestClassifier()
In [26]:
                                                                                                               M
rf=RandomForestClassifier()
In [27]:
                                                                                                               H
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 [35]:
                                                                                                               H
from sklearn.model_selection import GridSearchCV
grid_search=GridSearchCV(estimator=rf,param_grid=params,cv=2,scoring="accuracy")
grid_search.fit(x_test,y_test)
Out[35]:
             GridSearchCV
 ▶ estimator: RandomForestClassifier
       ▶ RandomForestClassifier
In [29]:
                                                                                                               H
grid_search.best_score_
Out[29]:
0.55
                                                                                                               H
In [30]:
rf best=grid search.best estimator
print(rf_best)
```

RandomForestClassifier(max_depth=10, min_samples_leaf=50, n_estimators=25)

```
In [31]:
from sklearn.tree import plot_tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[5],feature_names=x.columns,class_names=['0','1','2','3'],filled=True);
                                        px width <= 1570.5
                                            gini = 0.499
                                          samples = 199
                                         value = [145, 155]
                                              class = 1
                        talk time \leq 10.5
                                                              gini = 0.45
                           gini = 0.497
                                                            samples = 59
                          samples = 140
                                                           value = [28, 54]
                       value = [117, 101]
                                                               class = 1
                             class = 0
          gini = 0.462
                                            gini = 0.493
         samples = 67
                                           samples = 73
        value = [67, 38]
                                          value = [50, 63]
            class = 0
                                              class = 1
In [32]:
from sklearn.tree import plot tree
plt.figure(figsize=(80,40))
plot_tree(rf_best.estimators_[7],feature_names=x.columns,class_names=['0','1','2','3'],filled=True);
                                             pc <= 12.5
                                            gini = 0.499
                                          samples = 190
                                         value = [156, 144]
                                              class = 0
                        talk time \leq 10.5
                                                             gini = 0.455
                           gini = 0.489
                                                            samples = 81
                          samples = 109
                                                           value = [82, 44]
                        value = [74, 100]
                                                               class = 0
                             class = 1
          gini = 0.498
                                            gini = 0.449
                                           samples = 59
         samples = 50
        value = [41, 36]
                                          value = [33, 64]
            class = 0
                                              class = 1
In [33]:
                                                                                               M
rf_best.feature_importances_
Out[33]:
```

```
In [34]:
imp_df=pd.DataFrame({"Varname":x_train.columns,"Imp":rf_best.feature_importances_})
imp_df.sort_values(by="Imp",ascending=False)
```

Out[34]:

In []:

	Varname	Imp
0	id	0.158910
7	int_memory	0.139026
13	px_width	0.120815
15	sc_h	0.080734
2	blue	0.077962
14	ram	0.067617
16	sc_w	0.044349
11	рс	0.043817
8	m_dep	0.040414
5	fc	0.040000
12	px_height	0.039605
19	touch_screen	0.036016
3	clock_speed	0.031668
17	talk_time	0.031102
10	n_cores	0.027207
1	battery_power	0.020759
9	mobile_wt	0.000000
6	four_g	0.000000
4	dual_sim	0.000000
18	three_g	0.000000
In []:	