Mobile price classification using machine learning

In [2]:

import numpy as np
import pandas as pd

In [3]:

import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline

In [4]:

df=pd.read_csv("C:\\Users\\astha\\Downloads\\train.csv")
##df
df.head()

Out[4]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt
0	842	0	2.2	0	1	0	7	0.6	188
1	1021	1	0.5	1	0	1	53	0.7	136
2	563	1	0.5	1	2	1	41	0.9	145
3	615	1	2.5	0	0	0	10	8.0	131
4	1821	1	1.2	0	13	1	44	0.6	141

5 rows × 21 columns

In [5]:

df.shape

Out[5]:

(2000, 21)

In [6]:

df.describe()

Out[6]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_m
count	2000.000000	2000.0000	2000.000000	2000.000000	2000.000000	2000.000000	2000.0
mean	1238.518500	0.4950	1.522250	0.509500	4.309500	0.521500	32.0
std	439.418206	0.5001	0.816004	0.500035	4.341444	0.499662	18.1
min	501.000000	0.0000	0.500000	0.000000	0.000000	0.000000	2.0
25%	851.750000	0.0000	0.700000	0.000000	1.000000	0.000000	16.0
50%	1226.000000	0.0000	1.500000	1.000000	3.000000	1.000000	32.0
75%	1615.250000	1.0000	2.200000	1.000000	7.000000	1.000000	48.0
max	1998.000000	1.0000	3.000000	1.000000	19.000000	1.000000	64.(

8 rows × 21 columns

In [7]:

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2000 entries, 0 to 1999
Data columns (total 21 columns):

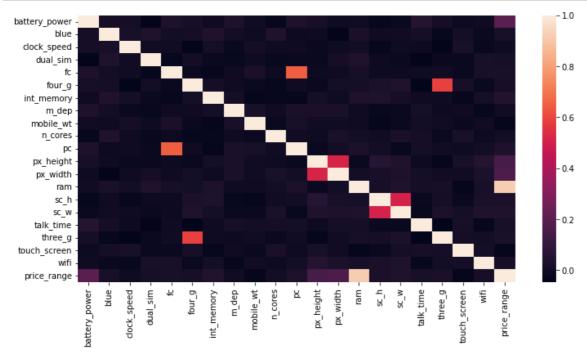
# Column Non-Null Count	Dtype
0 battery_power 2000 non-null	int64
1 blue 2000 non-null	int64
2 clock_speed 2000 non-null	float64
3 dual_sim 2000 non-null	int64
4 fc 2000 non-null	int64
5 four_g 2000 non-null	int64
6 int_memory 2000 non-null	int64
7 m_dep 2000 non-null	float64
8 mobile_wt 2000 non-null	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 [8]:

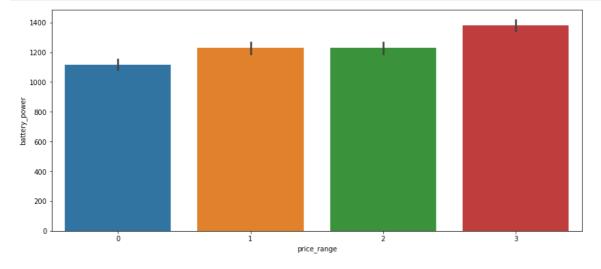
```
## HEAT MAP
plt.figure(figsize=(12,6))
sns.heatmap(df.corr())
plt.show()
```



plotting Relation between price_range and Battery power

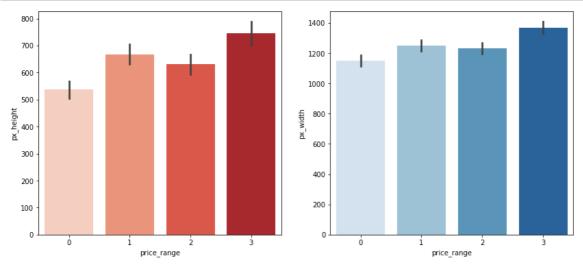
In [9]:

```
plt.figure(figsize=(14,6))
sns.barplot(x ='price_range',y ='battery_power',data=df)
plt.show()
```



In [10]:

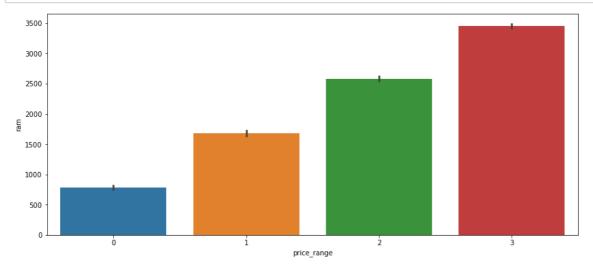
```
plt.figure(figsize=(14,6))
plt.subplot(1,2,1)
sns.barplot(x ='price_range',y ='px_height',data=df,palette='Reds')
plt.subplot(1,2,2)
sns.barplot(x ='price_range',y ='px_width',data=df,palette='Blues')
plt.show()
```



relation between price_range and ram

In [11]:

```
plt.figure(figsize=(14,6))
sns.barplot(x ='price_range',y ='ram',data=df)
plt.show()
```



Data preprocessing

```
In [20]:
x=df.drop(['price_range'],axis=1)
y=df['price_range']
In [21]:
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=0)
knn
In [22]:
from sklearn.neighbors import KNeighborsClassifier
knn=KNeighborsClassifier(n_neighbors=10)
knn.fit(x_train,y_train)
Out[22]:
KNeighborsClassifier(n_neighbors=10)
In [23]:
knn.score(x_train,y_train)
Out[23]:
0.9457142857142857
In [24]:
predictions=knn.predict(x_test)
In [25]:
from sklearn.metrics import accuracy_score
accuracy_score(y_test,predictions)
```

predcting values for test csv

Out[25]:

0.935

```
In [31]:
```

```
test_df=pd.read_csv("C:\\Users\\astha\\Downloads\\test (2).csv")
test_df.head()
```

Out[31]:

	id	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_
0	1	1043	1	1.8	1	14	0	5	0.1	1
1	2	841	1	0.5	1	4	1	61	0.8	1
2	3	1807	1	2.8	0	1	0	27	0.9	1
3	4	1546	0	0.5	1	18	1	25	0.5	
4	5	1434	0	1.4	0	11	1	49	0.5	1

5 rows × 21 columns

←

In [32]:

```
test_df.shape
```

Out[32]:

(1000, 21)

In [33]:

```
test_df=test_df.drop(['id'],axis=1)
test_df.shape
```

Out[33]:

(1000, 20)

In [34]:

```
test_pred=knn.predict(test_df)
```

In [35]:

```
test_df['predicted_price']=test_pred
```

In [36]:

test_df.head()

Out[36]:

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt
0	1043	1	1.8	1	14	0	5	0.1	193
1	841	1	0.5	1	4	1	61	0.8	191
2	1807	1	2.8	0	1	0	27	0.9	186
3	1546	0	0.5	1	18	1	25	0.5	96
4	1434	0	1.4	0	11	1	49	0.5	108

5 rows × 21 columns

1

In []: