



```
#load the file
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
df=pd.read_csv('/content/train.csv')
df
```




| | battery_power | blue | clock_speed | dual_sim | fc | four_g | int_memory | m_dep | mobile_wt | n_cores | ... |
|------|---------------|------|-------------|----------|-----|--------|------------|-------|-----------|---------|-----|
| 0 | 842 | 0 | 2.2 | 0 | 1 | 0 | 7 | 0.6 | 188 | 2 | .. |
| 1 | 1021 | 1 | 0.5 | 1 | 0 | 1 | 53 | 0.7 | 136 | 3 | .. |
| 2 | 563 | 1 | 0.5 | 1 | 2 | 1 | 41 | 0.9 | 145 | 5 | .. |
| 3 | 615 | 1 | 2.5 | 0 | 0 | 0 | 10 | 0.8 | 131 | 6 | .. |
| 4 | 1821 | 1 | 1.2 | 0 | 13 | 1 | 44 | 0.6 | 141 | 2 | .. |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | .. |
| 1995 | 794 | 1 | 0.5 | 1 | 0 | 1 | 2 | 0.8 | 106 | 6 | .. |
| 1996 | 1965 | 1 | 2.6 | 1 | 0 | 0 | 39 | 0.2 | 187 | 4 | .. |
| 1997 | 1911 | 0 | 0.9 | 1 | 1 | 1 | 36 | 0.7 | 108 | 8 | .. |
| 1998 | 1512 | 0 | 0.9 | 0 | 4 | 1 | 46 | 0.1 | 145 | 5 | .. |
| 1999 | 510 | 1 | 2.0 | 1 | 5 | 1 | 45 | 0.9 | 168 | 6 | .. |

2000 rows × 21 columns




```
#First 5 rows of data
df.head()
```




| | battery_power | blue | clock_speed | dual_sim | fc | four_g | int_memory | m_dep | mobile_wt | n_cores | ... | p |
|---|---------------|------|-------------|----------|----|--------|------------|-------|-----------|---------|-----|---|
| 0 | 842 | 0 | 2.2 | 0 | 1 | 0 | 7 | 0.6 | 188 | 2 | ... | |
| 1 | 1021 | 1 | 0.5 | 1 | 0 | 1 | 53 | 0.7 | 136 | 3 | ... | |
| 2 | 563 | 1 | 0.5 | 1 | 2 | 1 | 41 | 0.9 | 145 | 5 | ... | |
| 3 | 615 | 1 | 2.5 | 0 | 0 | 0 | 10 | 0.8 | 131 | 6 | ... | |
| 4 | 1821 | 1 | 1.2 | 0 | 13 | 1 | 44 | 0.6 | 141 | 2 | ... | |

5 rows × 21 columns




```
#Last 5 rows of data
df.tail()
```




| | battery_power | blue | clock_speed | dual_sim | fc | four_g | int_memory | m_dep | mobile_wt | n_cores | ... |
|------|---------------|------|-------------|----------|----|--------|------------|-------|-----------|---------|-----|
| 1995 | 794 | 1 | 0.5 | 1 | 0 | 1 | 2 | 0.8 | 106 | 6 | .. |
| 1996 | 1965 | 1 | 2.6 | 1 | 0 | 0 | 39 | 0.2 | 187 | 4 | .. |
| 1997 | 1911 | 0 | 0.9 | 1 | 1 | 1 | 36 | 0.7 | 108 | 8 | .. |
| 1998 | 1512 | 0 | 0.9 | 0 | 4 | 1 | 46 | 0.1 | 145 | 5 | .. |
| 1999 | 510 | 1 | 2.0 | 1 | 5 | 1 | 45 | 0.9 | 168 | 6 | .. |

5 rows × 21 columns




```
#Shape of the dataset
df.shape
```



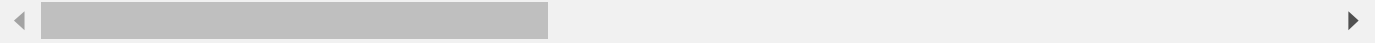
(2000, 21)

```
#Description of the dataset
df.describe()
```



| | battery_power | blue | clock_speed | dual_sim | fc | four_g | int_memory | m_ |
|-------|---------------|-----------|-------------|-------------|-------------|-------------|-------------|-------------|
| count | 2000.000000 | 2000.0000 | 2000.000000 | 2000.000000 | 2000.000000 | 2000.000000 | 2000.000000 | 2000.000000 |
| mean | 1238.518500 | 0.4950 | 1.522250 | 0.509500 | 4.309500 | 0.521500 | 32.046500 | 0.501 |
| std | 439.418206 | 0.5001 | 0.816004 | 0.500035 | 4.341444 | 0.499662 | 18.145715 | 0.288 |
| min | 501.000000 | 0.0000 | 0.500000 | 0.000000 | 0.000000 | 0.000000 | 2.000000 | 0.100 |
| 25% | 851.750000 | 0.0000 | 0.700000 | 0.000000 | 1.000000 | 0.000000 | 16.000000 | 0.200 |
| 50% | 1226.000000 | 0.0000 | 1.500000 | 1.000000 | 3.000000 | 1.000000 | 32.000000 | 0.500 |
| 75% | 1615.250000 | 1.0000 | 2.200000 | 1.000000 | 7.000000 | 1.000000 | 48.000000 | 0.800 |
| max | 1998.000000 | 1.0000 | 3.000000 | 1.000000 | 19.000000 | 1.000000 | 64.000000 | 1.000 |

8 rows × 21 columns



```
#checking null values of the data
df.isnull().sum()
```



0

| | |
|---------------|---|
| battery_power | 0 |
| blue | 0 |
| clock_speed | 0 |
| dual_sim | 0 |
| fc | 0 |
| four_g | 0 |
| int_memory | 0 |
| m_dep | 0 |
| mobile_wt | 0 |
| n_cores | 0 |
| pc | 0 |
| px_height | 0 |
| px_width | 0 |
| ram | 0 |
| sc_h | 0 |
| sc_w | 0 |
| talk_time | 0 |
| three_g | 0 |
| touch_screen | 0 |
| wifi | 0 |
| price_range | 0 |

dtype: int64

```
#Info of the dataset
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

```

```
#change column Names
```

```
df=df.rename(columns={"blue":"bluetooth","clock_speed":"time_speed","touch_screen":"smart_screen"})
df.head()
```



| | battery_power | bluetooth | time_speed | dual_sim | fc | four_g | int_memory | m_dep | mobile_wt | n_cores | .. |
|---|---------------|-----------|------------|----------|----|--------|------------|-------|-----------|---------|----|
| 0 | 842 | 0 | 2.2 | 0 | 1 | 0 | 7 | 0.6 | 188 | 2 | . |
| 1 | 1021 | 1 | 0.5 | 1 | 0 | 1 | 53 | 0.7 | 136 | 3 | . |
| 2 | 563 | 1 | 0.5 | 1 | 2 | 1 | 41 | 0.9 | 145 | 5 | . |
| 3 | 615 | 1 | 2.5 | 0 | 0 | 0 | 10 | 0.8 | 131 | 6 | . |
| 4 | 1821 | 1 | 1.2 | 0 | 13 | 1 | 44 | 0.6 | 141 | 2 | . |


5 rows × 21 columns

```
#change the data
```

```

mobile_df=pd.DataFrame()
mobile_df["bluetooth"]=np.where(df["bluetooth"]<1,"No","yes")
mobile_df["dual_sim"]=np.where(df["dual_sim"]<1,"No","yes")
mobile_df["four_g"]=np.where(df["four_g"]<1,"No","yes")
mobile_df["three_g"]=np.where(df["three_g"]<1,"No","yes")
mobile_df["smart_screen"]=np.where(df["smart_screen"]<1,"No","yes")
mobile_df["wifi"]=np.where(df["wifi"]<1,"No","yes")
print("Column Data has updated Successfully")

```


 Column Data has updated Successfully

```
mobile_df.head()
```

```

df["price_range"].replace({"Low Cost":1, "Median Cost":2, "High Cost":3})
print("Column Data has updated Successfully")

```

 Column Data has updated Successfully

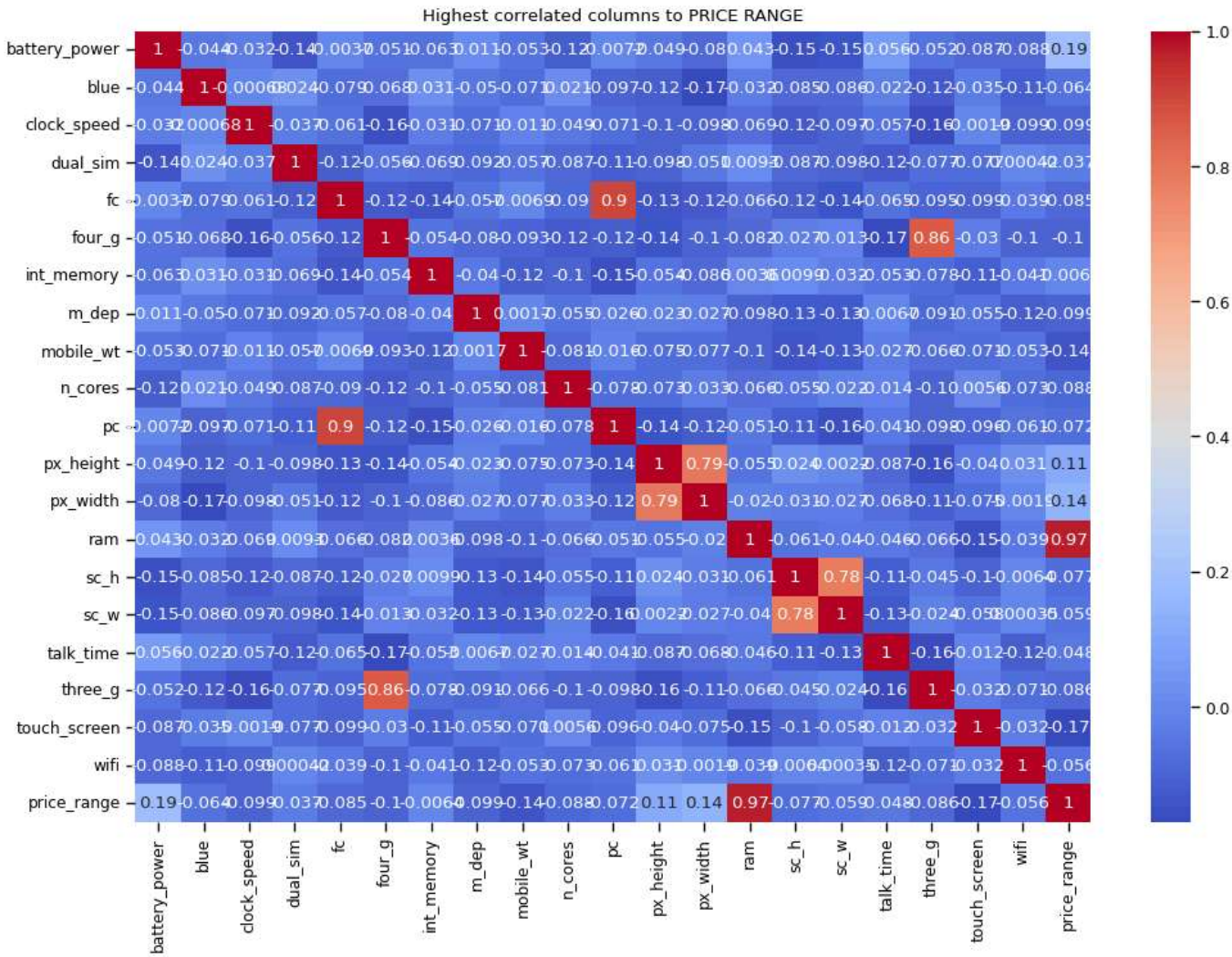
```
df.head()
```



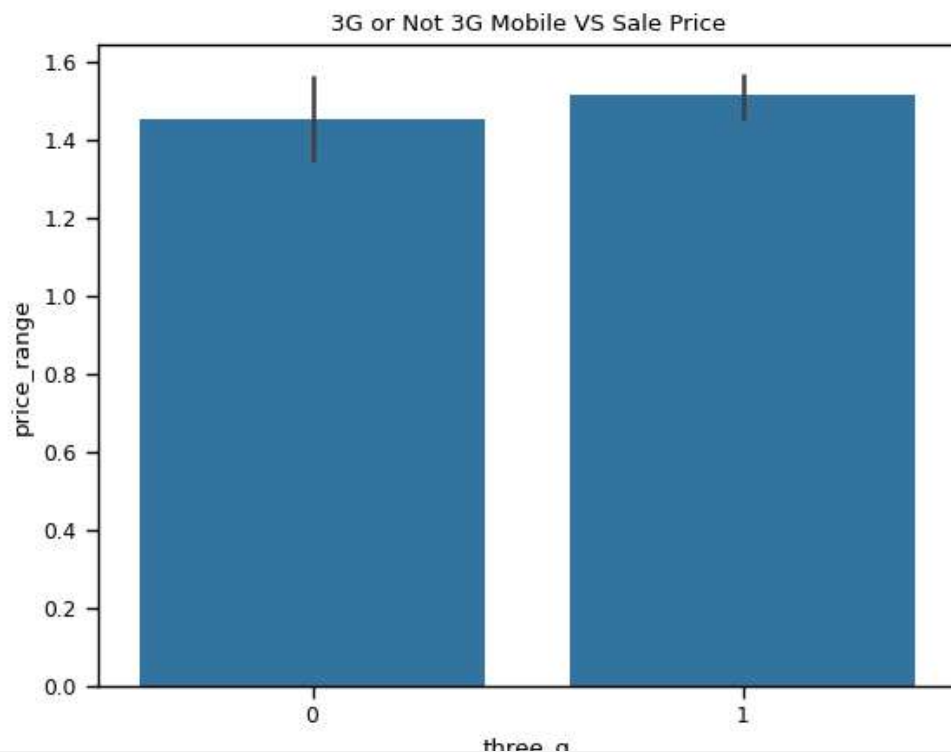
| a_speed | dual_sim | fc | four_g | int_memory | m_dep | mobile_wt | n_cores | .. |
|---------|----------|----|--------|------------|-------|-----------|---------|----|
| 2.2 | 0 | 1 | 0 | 7 | 0.6 | 188 | 2 | .. |
| 0.5 | 1 | 0 | 1 | 53 | 0.7 | 136 | 3 | .. |
| 0.5 | 1 | 2 | 1 | 41 | 0.9 | 145 | 5 | .. |
| 2.5 | 0 | 0 | 0 | 10 | 0.8 | 131 | 6 | .. |
| 1.2 | 0 | 13 | 1 | 44 | 0.6 | 141 | 2 | .. |



```
#VISUALIZATION
#Get the highest correlated columns to price range
import seaborn as sns
import matplotlib.pyplot as plt
plt.figure(figsize=(12,8))
sns.heatmap(mobile_corr.corr(),cmap="coolwarm",annot=True)
plt.title("Highest correlated columns to PRICE RANGE")
plt.show()
```



```
#show the 3G or not 3G mobile vs sale price using bar plot  
sns.barplot(x='three_g',y='price_range',data=df)  
plt.title("3G or Not 3G Mobile VS Sale Price")  
plt.show()
```

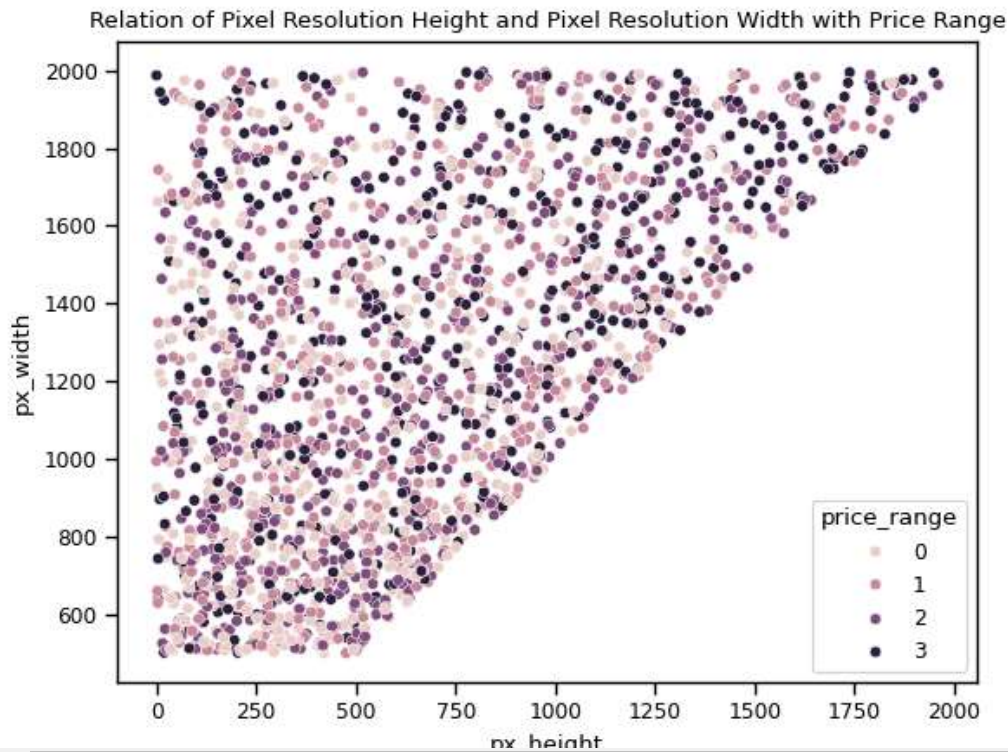


```
#Show the count plot for supporting bluetooth or not vs price  
sns.countplot(x='bluetooth',hue='price_range',data=df)  
plt.title("Supporting Bluetooth or Not VS Price")  
plt.show()
```



Supporting Bluetooth or Not VS Price

```
#Use scatterplot to show the relation of pixel resolution height and pixel resolution width
#with price range
sns.scatterplot(x="px_height",y="px_width",hue="price_range",data=df)
plt.title("Relation of Pixel Resolution Height and Pixel Resolution Width with Price Range")
plt.show()
```



```
#Use scatterplot to show the relation of screen height and screen width
#with price ranges
sns.scatterplot(x="sc_h",y="sc_w",hue='price_range',data=df)
plt.title("Relation of Screen Height and Screen Width with Price Range")
plt.show()
```

