D8

In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

In [2]: df=pd.read_csv(r"C:\Users\user\Downloads\12_mobile_prices_2023.csv")

df

Out[2]:

	Phone Name	Rating ?/5	Number of Ratings	RAM	ROM/Storage	Back/Rare Camera	Front Camera	Battery	Processor	P
0	POCO C50 (Royal Blue, 32 GB)	4.2	33,561	2 GB RAM	32 GB ROM	8MP Dual Camera	5MP Front Camera	5000 mAh	Mediatek Helio A22 Processor, Upto 2.0 GHz Pro	:
1	POCO M4 5G (Cool Blue, 64 GB)	4.2	77,128	4 GB RAM	64 GB ROM	50MP + 2MP	8MP Front Camera	5000 mAh	Mediatek Dimensity 700 Processor	₹
2	POCO C51 (Royal Blue, 64 GB)	4.3	15,175	4 GB RAM	64 GB ROM	8MP Dual Rear Camera	5MP Front Camera	5000 mAh	Helio G36 Processor	ŧ
3	POCO C55 (Cool Blue, 64 GB)	4.2	22,621	4 GB RAM	64 GB ROM	50MP Dual Rear Camera	5MP Front Camera	5000 mAh	Mediatek Helio G85 Processor	ŧ
4	POCO C51 (Power Black, 64 GB)	4.3	15,175	4 GB RAM	64 GB ROM	8MP Dual Rear Camera	5MP Front Camera	5000 mAh	Helio G36 Processor	;
									•••	
1831	Infinix Note 7 (Forest Green, 64 GB)	4.3	25,582	4 GB RAM	64 GB ROM	48MP + 2MP + 2MP + Al Lens Camera	16MP Front Camera	5000 mAh	MediaTek Helio G70 Processor	₹
1832	Infinix Note 7 (Bolivia Blue, 64 GB)	4.3	25,582	4 GB RAM	64 GB ROM	48MP + 2MP + 2MP + Al Lens Camera	16MP Front Camera	5000 mAh	MediaTek Helio G70 Processor	₹
1833	Infinix Note 7 (Aether Black, 64 GB)	4.3	25,582	4 GB RAM	64 GB ROM	48MP + 2MP + 2MP + Al Lens Camera	16MP Front Camera	5000 mAh	MediaTek Helio G70 Processor	₹
1834	Infinix Zero 8i (Silver Diamond, 128 GB)	4.2	7,117	8 GB RAM	128 GB ROM	48MP + 8MP + 2MP + Al Lens Camera	16MP + 8MP Dual Front Camera	4500 mAh	MediaTek Helio G90T Processor	₹
1835	Infinix S5 (Quetzal Cyan, 64 GB)	4.3	15,701	4 GB RAM	64 GB ROM	16MP + 5MP + 2MP + Low Light Sensor	32MP Front Camera	4000 mAh	Helio P22 (MTK6762) Processor	₹

1836 rows × 11 columns

In [3]: df.head(10)

Out[3]:

	Phone Name	Rating ?/5	Number of Ratings	RAM	ROM/Storage	Back/Rare Camera	Front Camera	Battery	Processor	Price ir INF
0	POCO C50 (Royal Blue, 32 GB)	4.2	33,561	2 GB RAM	32 GB ROM	8MP Dual Camera	5MP Front Camera	5000 mAh	Mediatek Helio A22 Processor, Upto 2.0 GHz Pro	₹5,64\$
1	POCO M4 5G (Cool Blue, 64 GB)	4.2	77,128	4 GB RAM	64 GB ROM	50MP + 2MP	8MP Front Camera	5000 mAh	Mediatek Dimensity 700 Processor	₹11,999
2	POCO C51 (Royal Blue, 64 GB)	4.3	15,175	4 GB RAM	64 GB ROM	8MP Dual Rear Camera	5MP Front Camera	5000 mAh	Helio G36 Processor	₹6,999
3	POCO C55 (Cool Blue, 64 GB)	4.2	22,621	4 GB RAM	64 GB ROM	50MP Dual Rear Camera	5MP Front Camera	5000 mAh	Mediatek Helio G85 Processor	₹7,749
4	POCO C51 (Power Black, 64 GB)	4.3	15,175	4 GB RAM	64 GB ROM	8MP Dual Rear Camera	5MP Front Camera	5000 mAh	Helio G36 Processor	₹6,99\$
5	POCO M4 5G (Power Black, 64 GB)	4.2	77,128	4 GB RAM	64 GB ROM	50MP + 2MP	8MP Front Camera	5000 mAh	Mediatek Dimensity 700 Processor	₹11,99§
6	POCO C55 (Power Black, 64 GB)	4.2	22,621	4 GB RAM	64 GB ROM	50MP Dual Rear Camera	5MP Front Camera	5000 mAh	Mediatek Helio G85 Processor	₹7,74\$
7	POCO C55 (Forest Green, 64 GB)	4.2	22,621	4 GB RAM	64 GB ROM	50MP Dual Rear Camera	5MP Front Camera	5000 mAh	Mediatek Helio G85 Processor	₹7,74\$
8	POCO C55 (Cool Blue, 128 GB)	4.1	13,647	6 GB RAM	128 GB ROM	50MP Dual Rear Camera	5MP Front Camera	5000 mAh	Mediatek Helio G85 Processor	₹9,24€
9	POCO M4 5G (Yellow, 128 GB)	4.2	40,525	6 GB RAM	128 GB ROM	50MP + 2MP	8MP Front Camera	5000 mAh	Mediatek Dimensity 700 Processor	₹13,99€
4										>

```
In [4]: df.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 1836 entries, 0 to 1835 Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype		
0	Phone Name	1836 non-null	object		
1	Rating ?/5	1836 non-null	float64		
2	Number of Ratings	1836 non-null	object		
3	RAM	1836 non-null	object		
4	ROM/Storage	1662 non-null	object		
5	Back/Rare Camera	1827 non-null	object		
6	Front Camera	1435 non-null	object		
7	Battery	1826 non-null	object		
8	Processor	1781 non-null	object		
9	Price in INR	1836 non-null	object		
10	Date of Scraping	1836 non-null	object		
dtypes: float64(1), object(10)					

memory usage: 157.9+ KB

In [5]: dff=df.dropna()

In [6]: dff.info()

<class 'pandas.core.frame.DataFrame'> Int64Index: 1291 entries, 0 to 1835 Data columns (total 11 columns):

#	Column	Non-Null Count	Dtype
0	Phone Name	1291 non-null	object
1	Rating ?/5	1291 non-null	float64
2	Number of Ratings	1291 non-null	object
3	RAM	1291 non-null	object
4	ROM/Storage	1291 non-null	object
5	Back/Rare Camera	1291 non-null	object
6	Front Camera	1291 non-null	object
7	Battery	1291 non-null	object
8	Processor	1291 non-null	object
9	Price in INR	1291 non-null	object
10	Date of Scraping	1291 non-null	object

dtypes: float64(1), object(10)

memory usage: 121.0+ KB

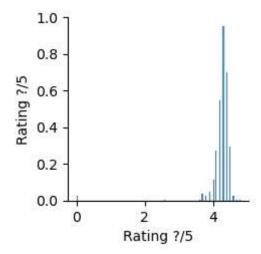
```
In [7]: dff.describe()
```

Out[7]:		Rating ?/5
	count	1291.000000
	mean	4.241208
	std	0.427166
	min	0.000000
	25%	4.200000
	50%	4.300000
	75%	4.400000
	max	4.800000

```
In [6]: dff.columns
```

In [7]: sns.pairplot(dff)

Out[7]: <seaborn.axisgrid.PairGrid at 0x29d331be6d0>



```
In [8]: sns.distplot(df["Rating ?/5"])
```

C:\Users\user\AppData\Local\Temp\ipykernel_5412\1803240722.py:1: UserWarning:

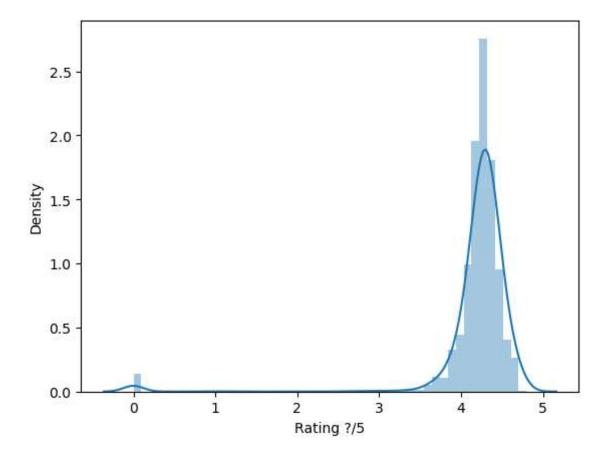
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751 (https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751)

sns.distplot(df["Rating ?/5"])

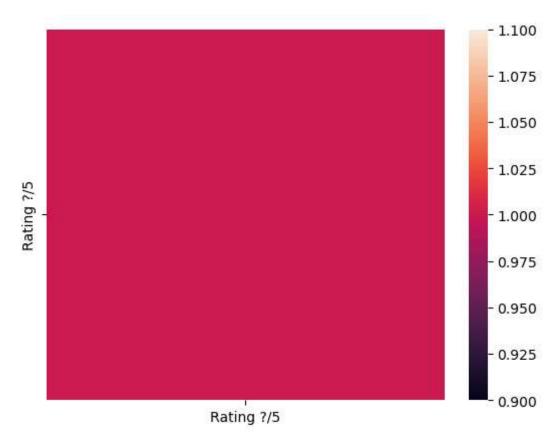
Out[8]: <Axes: xlabel='Rating ?/5', ylabel='Density'>



```
In [10]: sns.heatmap(df1.corr())
```

C:\Users\user\AppData\Local\Temp\ipykernel_5412\781785195.py:1: FutureWarnin
g: The default value of numeric_only in DataFrame.corr is deprecated. In a fu
ture version, it will default to False. Select only valid columns or specify
the value of numeric_only to silence this warning.
 sns.heatmap(df1.corr())

Out[10]: <Axes: >



```
In [15]: x=df1[['Rating ?/5']]
y=df1['Rating ?/5']

In [16]: 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)

In [17]: from sklearn.linear_model import LinearRegression
    lr=LinearRegression()
    lr.fit(x_train,y_train)

Out[17]:    v LinearRegression
    LinearRegression()

In [18]: print(lr.intercept_)
```

-1.7763568394002505e-15

```
coeff = pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
In [19]:
         coeff
```

Out[19]:

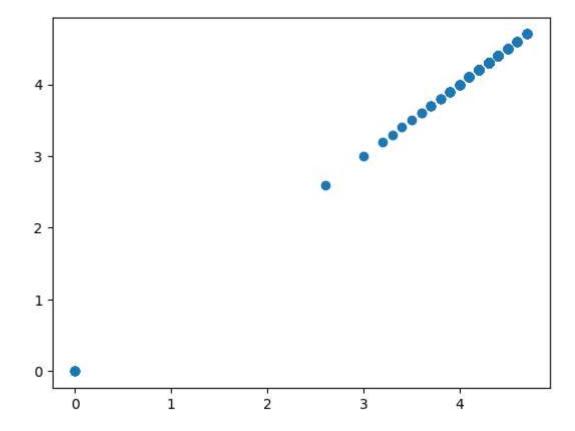
Co-efficient

Rating ?/5

1.0

```
In [20]:
         prediction=lr.predict(x_test)
         plt.scatter(y_test,prediction)
```

Out[20]: <matplotlib.collections.PathCollection at 0x29d3abdcc90>



```
In [21]: print(lr.score(x_test,y_test))
```

1.0

from sklearn.linear_model import Ridge,Lasso In [22]:

rr=Ridge(alpha=10) In [23]: rr.fit(x_train,y_train)

Out[23]: Ridge Ridge(alpha=10)