

```
In [6]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

```
In [7]: df = pd.read_csv('laptop_data.csv')
```

```
In [8]: df.head()
```

Out[8]:

	Unnamed: 0	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price
0	0	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 2.3GHz	8GB	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37kg	71378.6832
1	1	Apple	Ultrabook	13.3	1440x900	Intel Core i5 1.8GHz	8GB	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34kg	47895.5232
2	2	HP	Notebook	15.6	Full HD 1920x1080	Intel Core i5 7200U 2.5GHz	8GB	256GB SSD	Intel HD Graphics 620	No OS	1.86kg	30636.0000
3	3	Apple	Ultrabook	15.4	IPS Panel Retina Display 2880x1800	Intel Core i7 2.7GHz	16GB	512GB SSD	AMD Radeon Pro 455	macOS	1.83kg	135195.3360
4	4	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 3.1GHz	8GB	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37kg	96095.8080

```
In [9]: df.shape
```

Out[9]: (1303, 12)

```
In [10]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1303 entries, 0 to 1302
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  ---                ---
0   Unnamed: 0            1303 non-null  int64
1   Company              1303 non-null  object
2   TypeName             1303 non-null  object
3   Inches              1303 non-null  float64
4   ScreenResolution     1303 non-null  object
5   Cpu                 1303 non-null  object
6   Ram                 1303 non-null  object
7   Memory             1303 non-null  object
8   Gpu                 1303 non-null  object
9   OpSys              1303 non-null  object
10  Weight             1303 non-null  object
11  Price              1303 non-null  float64
dtypes: float64(2), int64(1), object(9)
memory usage: 122.3+ KB
```

```
In [11]: df.duplicated().sum()
```

Out[11]: 0

```
In [12]: df.isnull().sum()
```

Out[12]:

Unnamed: 0	0
Company	0
TypeName	0
Inches	0
ScreenResolution	0
Cpu	0
Ram	0
Memory	0
Gpu	0
OpSys	0
Weight	0
Price	0

dtype: int64

```
In [13]: df.drop(columns=['Unnamed: 0'],inplace=True)
```

```
In [14]: df.head()
```

Out[14]:

	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price
0	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 2.3GHz	8GB	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37kg	71378.6832
1	Apple	Ultrabook	13.3	1440x900	Intel Core i5 1.8GHz	8GB	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34kg	47895.5232
2	HP	Notebook	15.6	Full HD 1920x1080	Intel Core i5 7200U 2.5GHz	8GB	256GB SSD	Intel HD Graphics 620	No OS	1.86kg	30636.0000
3	Apple	Ultrabook	15.4	IPS Panel Retina Display 2880x1800	Intel Core i7 2.7GHz	16GB	512GB SSD	AMD Radeon Pro 455	macOS	1.83kg	135195.3360
4	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 3.1GHz	8GB	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37kg	96095.8080

In [15]:

```
df['Ram'] = df['Ram'].str.replace('GB','')
df['Weight'] = df['Weight'].str.replace('kg','')
```

In [16]:

```
df.head()
```

Out[16]:

	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price
0	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832
1	Apple	Ultrabook	13.3	1440x900	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232
2	HP	Notebook	15.6	Full HD 1920x1080	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000
3	Apple	Ultrabook	15.4	IPS Panel Retina Display 2880x1800	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360
4	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080

In [17]:

```
df['Ram'] = df['Ram'].astype('int32')
df['Weight'] = df['Weight'].astype('float32')
```

In [18]:

```
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1303 entries, 0 to 1302
Data columns (total 11 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Company                1303 non-null  object
1   TypeName               1303 non-null  object
2   Inches                 1303 non-null  float64
3   ScreenResolution       1303 non-null  object
4   Cpu                    1303 non-null  object
5   Ram                    1303 non-null  int32
6   Memory                 1303 non-null  object
7   Gpu                    1303 non-null  object
8   OpSys                  1303 non-null  object
9   Weight                 1303 non-null  float32
10  Price                  1303 non-null  float64
dtypes: float32(1), float64(2), int32(1), object(7)
memory usage: 101.9+ KB
```

In [19]:

```
import seaborn as sns
```

In [20]:

```
sns.distplot(df['Price'])
```

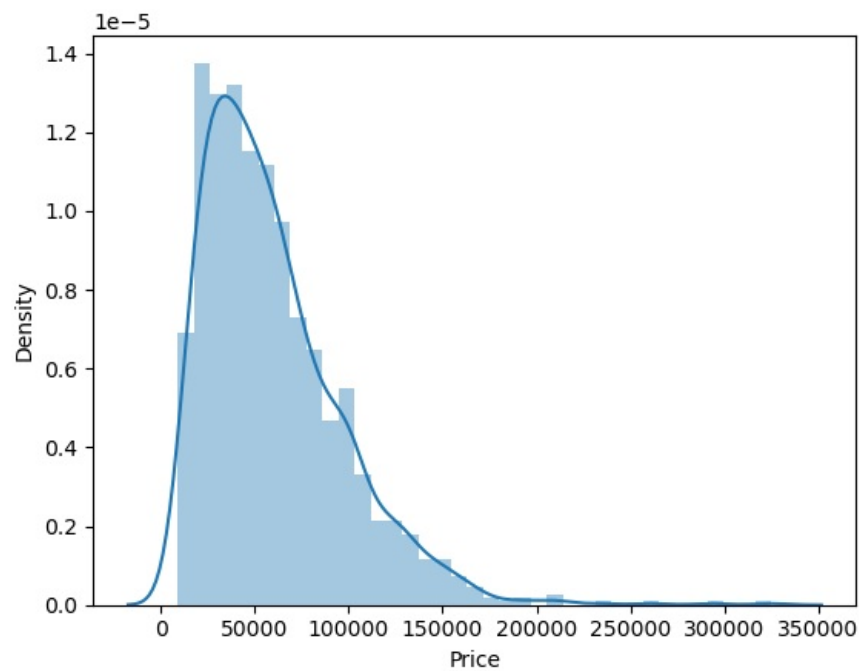
C:\Users\Admin\AppData\Local\Temp\ipykernel\_3208\834922981.py:1: UserWarning:

`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).

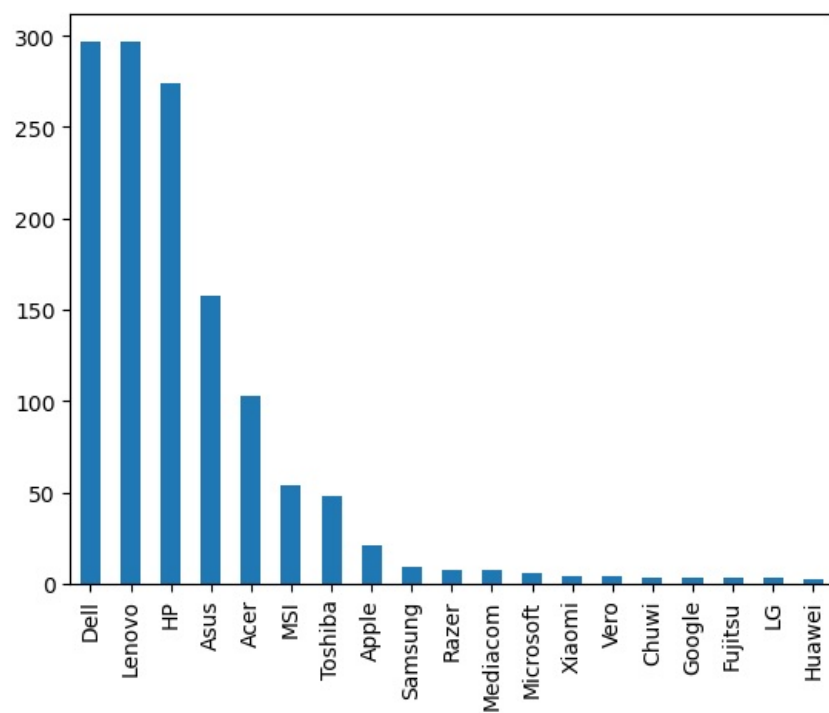
For a guide to updating your code to use the new functions, please see
https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(df['Price'])
<Axes: xlabel='Price', ylabel='Density'>

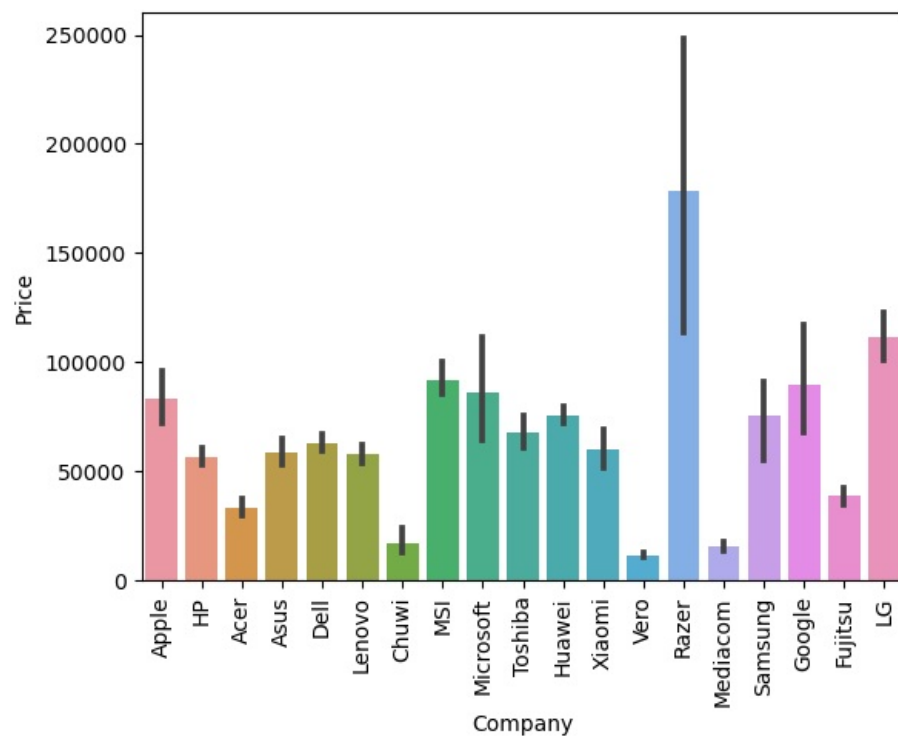


```
In [21]: df['Company'].value_counts().plot(kind='bar')
```

```
Out[21]: <Axes: >
```

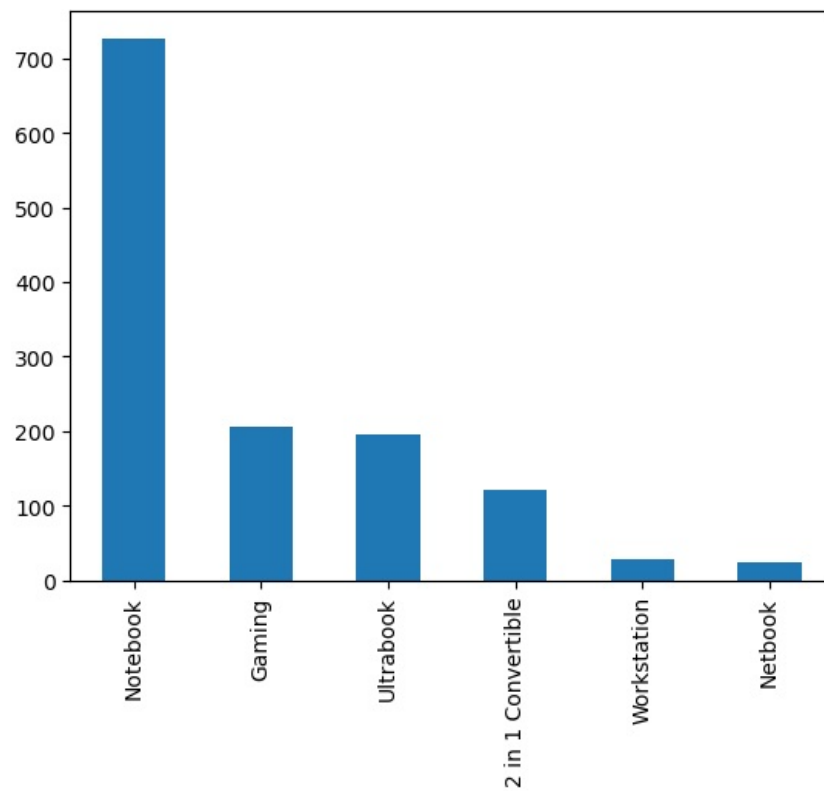


```
In [22]: sns.barplot(x=df['Company'],y=df['Price'])
plt.xticks(rotation='vertical')
plt.show()
```

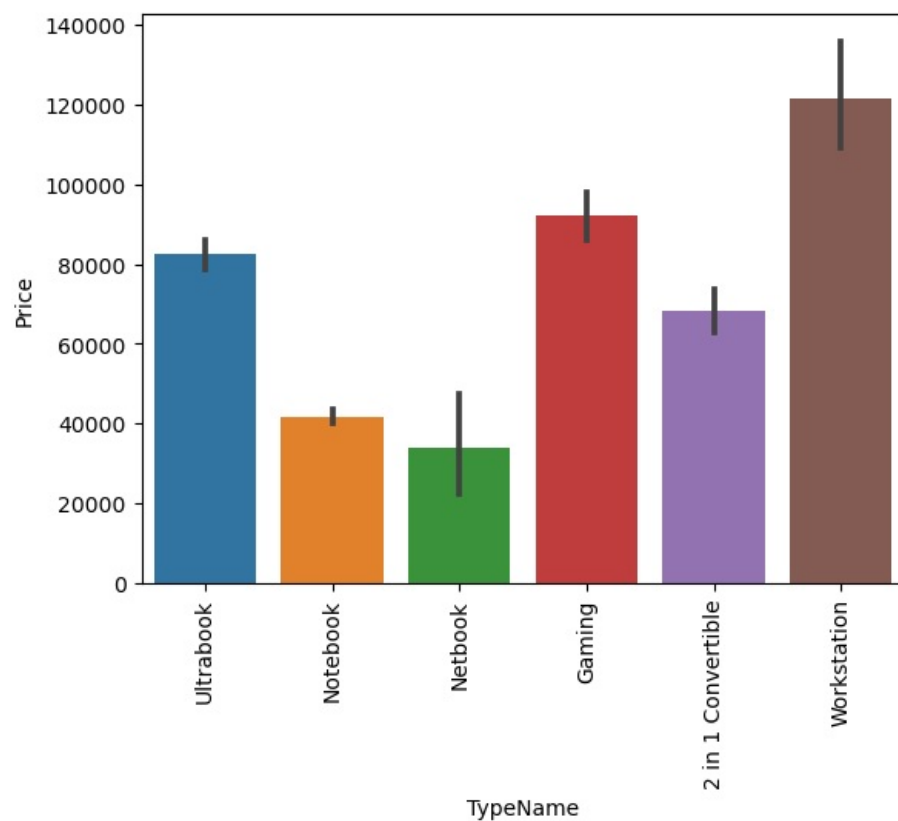


```
In [23]: df['TypeName'].value_counts().plot(kind='bar')
```

Out[23]: <Axes: >



```
In [24]: sns.barplot(x=df['TypeName'],y=df['Price'])
plt.xticks(rotation='vertical')
plt.show()
```



In [25]: `sns.distplot(df['Inches'])`

C:\Users\Admin\AppData\Local\Temp\ipykernel\_3208\1439577752.py:1: UserWarning:

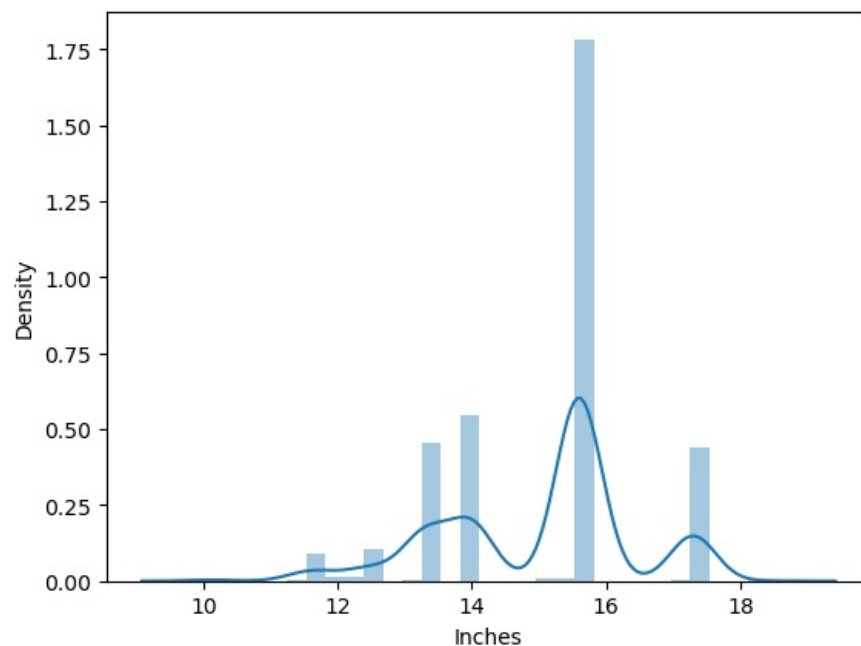
``distplot` is a deprecated function and will be removed in seaborn v0.14.0.`

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>

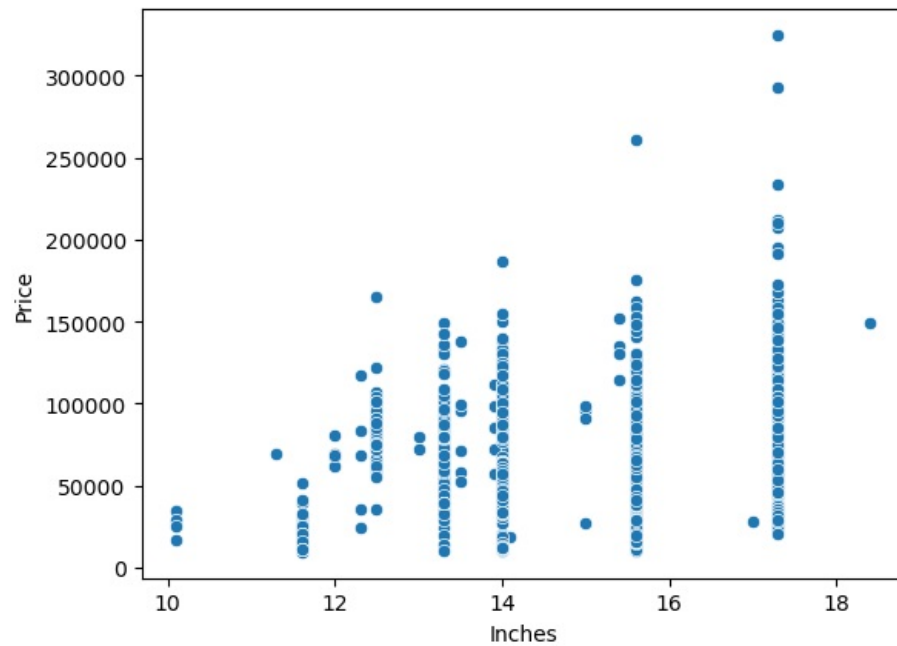
`sns.distplot(df['Inches'])`

Out[25]: `<Axes: xlabel='Inches', ylabel='Density'>`



In [26]: `sns.scatterplot(x=df['Inches'],y=df['Price'])`

Out[26]: `<Axes: xlabel='Inches', ylabel='Price'>`



```
In [27]: df['ScreenResolution'].value_counts()
```

```
Out[27]: Full HD 1920x1080          507
1366x768          281
IPS Panel Full HD 1920x1080      230
IPS Panel Full HD / Touchscreen 1920x1080  53
Full HD / Touchscreen 1920x1080    47
1600x900           23
Touchscreen 1366x768            16
Quad HD+ / Touchscreen 3200x1800    15
IPS Panel 4K Ultra HD 3840x2160     12
IPS Panel 4K Ultra HD / Touchscreen 3840x2160  11
4K Ultra HD / Touchscreen 3840x2160    10
4K Ultra HD 3840x2160              7
Touchscreen 2560x1440              7
IPS Panel 1366x768                 7
IPS Panel Quad HD+ / Touchscreen 3200x1800    6
IPS Panel Retina Display 2560x1600          6
IPS Panel Retina Display 2304x1440          6
Touchscreen 2256x1504                 6
IPS Panel Touchscreen 2560x1440            5
IPS Panel Retina Display 2880x1800          4
IPS Panel Touchscreen 1920x1200            4
1440x900                                 4
IPS Panel 2560x1440                     4
IPS Panel Quad HD+ 2560x1440              3
Quad HD+ 3200x1800                       3
1920x1080                                3
Touchscreen 2400x1600                    3
2560x1440                                3
IPS Panel Touchscreen 1366x768            3
IPS Panel Touchscreen / 4K Ultra HD 3840x2160  2
IPS Panel Full HD 2160x1440               2
IPS Panel Quad HD+ 3200x1800              2
IPS Panel Retina Display 2736x1824        1
IPS Panel Full HD 1920x1200               1
IPS Panel Full HD 2560x1440              1
IPS Panel Full HD 1366x768               1
Touchscreen / Full HD 1920x1080           1
Touchscreen / Quad HD+ 3200x1800          1
Touchscreen / 4K Ultra HD 3840x2160       1
IPS Panel Touchscreen 2400x1600           1
Name: ScreenResolution, dtype: int64
```

```
In [28]: df['Touchscreen'] = df['ScreenResolution'].apply(lambda x:1 if 'Touchscreen' in x else 0)
```

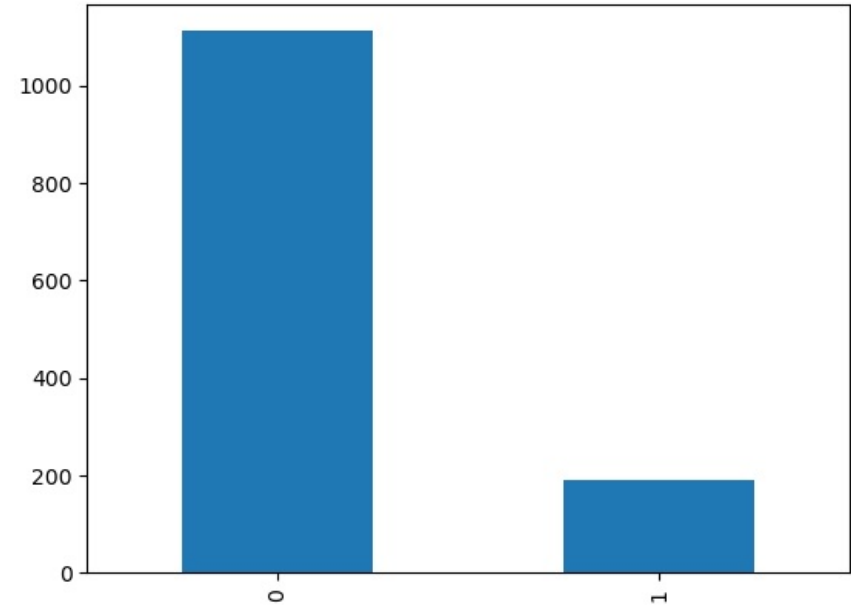
```
In [29]: df.sample(5)
```

Out[29]:

	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen
282	Lenovo	Gaming	15.6	IPS Panel Full HD 1920x1080	Intel Core i5 7300HQ 2.5GHz	8	256GB SSD	Nvidia GeForce GTX 1050	Windows 10	2.5	44169.120	0
540	Dell	Ultrabook	13.3	Quad HD+ / Touchscreen 3200x1800	Intel Core i7 8550U 1.8GHz	8	256GB SSD	Intel UHD Graphics 620	Windows 10	1.2	74538.720	1
251	Asus	Gaming	17.3	Full HD 1920x1080	Intel Core i7 7700HQ 2.8GHz	16	256GB SSD + 1TB HDD	Nvidia GeForce GTX 980M	Windows 10	4.3	95850.720	0
1021	Toshiba	Ultrabook	13.3	Full HD 1920x1080	Intel Core i5 6200U 2.3GHz	8	256GB SSD	Intel HD Graphics 520	Windows 10	1.2	84715.200	0
340	Dell	Notebook	15.6	1366x768	Intel Celeron Dual Core N3060 1.6GHz	4	500GB HDD	Intel HD Graphics	Linux	2.2	14646.672	0

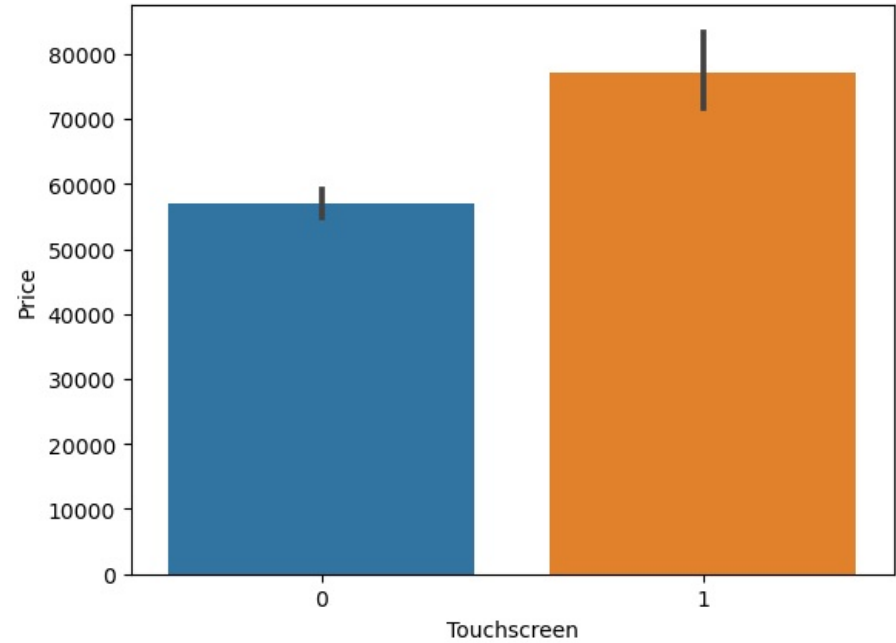
In [30]: df['Touchscreen'].value\_counts().plot(kind='bar')

Out[30]: <Axes: >



In [31]: sns.barplot(x=df['Touchscreen'], y=df['Price'])

Out[31]: <Axes: xlabel='Touchscreen', ylabel='Price'>



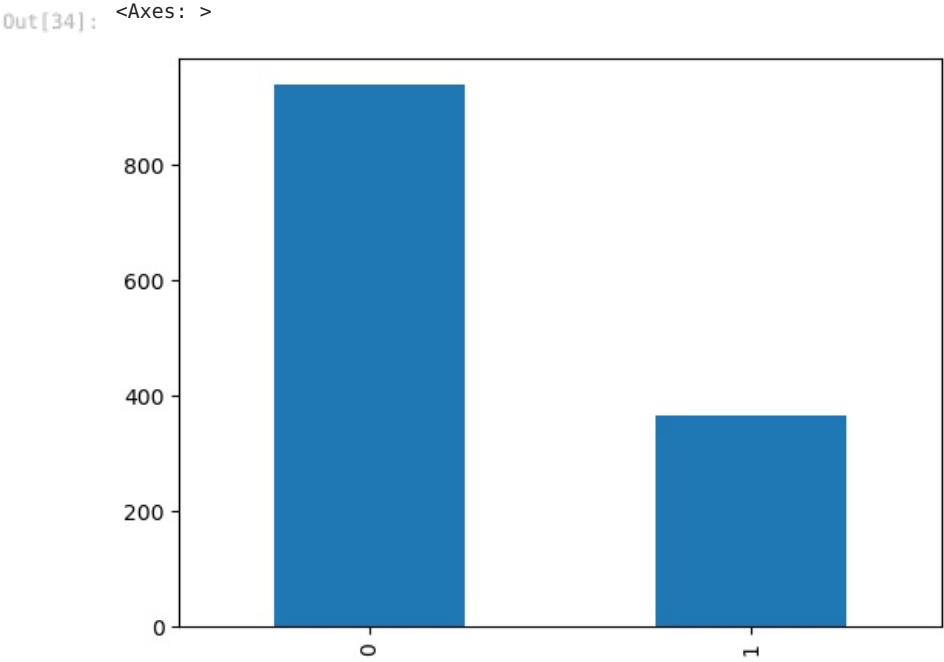
In [32]: df['Ips'] = df['ScreenResolution'].apply(lambda x:1 if 'IPS' in x else 0)

In [33]: df.head()

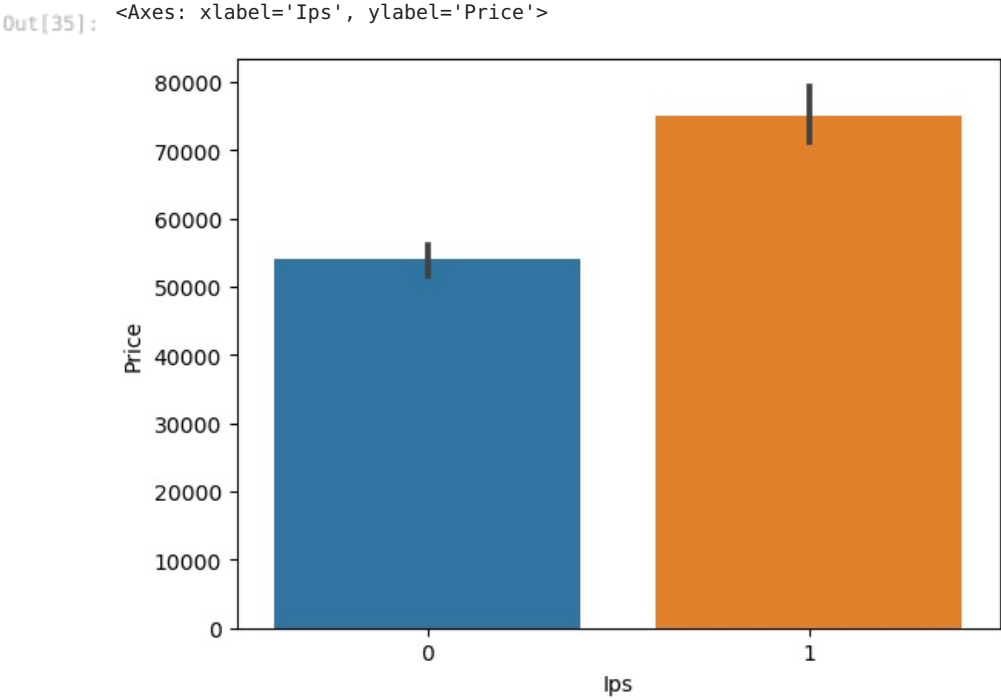
Out[33]:

	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips
0	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1
1	Apple	Ultrabook	13.3	1440x900	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0
2	HP	Notebook	15.6	Full HD 1920x1080	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0
3	Apple	Ultrabook	15.4	IPS Panel Retina Display 2880x1800	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1
4	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1

In [34]: df['Ips'].value\_counts().plot(kind='bar')



In [35]: sns.barplot(x=df['Ips'],y=df['Price'])



In [36]: new = df['ScreenResolution'].str.split('x',n=1,expand=True)

In [37]: df['X\_res'] = new[0]



```
df['Y_res'] = new[1]
```

```
In [38]: df.sample(5)
```

Out[38]:	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips
197	HP	Notebook	13.3	Full HD 1920x1080	Intel Core i5 8250U 1.6GHz	8	512GB SSD	Intel UHD Graphics 620	Windows 10	1.490	51095.5200	0	0
670	Microsoft	Ultrabook	13.5	Touchscreen 2256x1504	Intel Core i7 7600U 2.8GHz	8	256GB SSD	Intel Iris Plus Graphics 640	Windows 10 S	1.252	99519.0480	1	0
843	Dell	Notebook	15.6	1366x768	Intel Core i5 7200U 2.5GHz	8	128GB SSD	Intel HD Graphics 620	Windows 10	2.180	38041.3872	0	0
944	Lenovo	2 in 1 Convertible	11.3	IPS Panel Full HD / Touchscreen 1920x1080	Intel Core M m7- 6Y75 1.2GHz	8	256GB SSD	Intel HD Graphics 515	Windows 10	1.100	69210.7200	1	1
557	Lenovo	Notebook	17.3	1600x900	Intel Core i7 7500U 2.7GHz	6	128GB SSD + 1TB HDD	Nvidia GeForce 940MX	Windows 10	2.800	50562.7200	0	0

```
In [39]: df['X_res'] = df['X_res'].str.replace(',','').str.findall(r'(\d+\.\d+)').apply(lambda x:x[0])
```

```
In [40]: df.head()
```

Out[40]:	Company	TypeName	Inches	ScreenResolution	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	X_res
0	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	2560
1	Apple	Ultrabook	13.3	1440x900	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	1440
2	HP	Notebook	15.6	Full HD 1920x1080	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	1920
3	Apple	Ultrabook	15.4	IPS Panel Retina Display 2880x1800	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	2880
4	Apple	Ultrabook	13.3	IPS Panel Retina Display 2560x1600	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	2560

```
In [41]: df['X_res'] = df['X_res'].astype('int')
df['Y_res'] = df['Y_res'].astype('int')
```

```
In [42]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1303 entries, 0 to 1302
Data columns (total 15 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Company               1303 non-null   object
1   TypeName              1303 non-null   object
2   Inches                1303 non-null   float64
3   ScreenResolution      1303 non-null   object
4   Cpu                   1303 non-null   object
5   Ram                   1303 non-null   int32
6   Memory                1303 non-null   object
7   Gpu                   1303 non-null   object
8   OpSys                 1303 non-null   object
9   Weight                1303 non-null   float32
10  Price                 1303 non-null   float64
11  Touchscreen           1303 non-null   int64
12  Ips                   1303 non-null   int64
13  X_res                 1303 non-null   int32
14  Y_res                 1303 non-null   int32
dtypes: float32(1), float64(2), int32(3), int64(2), object(7)
memory usage: 132.5+ KB
```

```
In [43]: df.corr()['Price']
```

```
C:\Users\Admin\AppData\Local\Temp\ipykernel_3208\815546952.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
df.corr()['Price']
```

```
Out[43]: Inches      0.068197
Ram          0.743007
Weight       0.210370
Price        1.000000
Touchscreen  0.191226
Ips          0.252208
X_res        0.556529
Y_res        0.552809
Name: Price, dtype: float64
```

```
In [44]: df['ppi'] = (((df['X_res']**2) + (df['Y_res']**2))**0.5/df['Inches']).astype('float')
```

```
In [45]: df.corr()['Price']
```

```
C:\Users\Admin\AppData\Local\Temp\ipykernel_3208\815546952.py:1: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.
df.corr()['Price']
```

```
Out[45]: Inches      0.068197
Ram          0.743007
Weight       0.210370
Price        1.000000
Touchscreen  0.191226
Ips          0.252208
X_res        0.556529
Y_res        0.552809
ppi          0.473487
Name: Price, dtype: float64
```

```
In [46]: df.drop(columns=['ScreenResolution'],inplace=True)
```

```
In [47]: df.head()
```

```
Out[47]:
```

	Company	TypeName	Inches	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	X_res	Y_res	ppi
0	Apple	Ultrabook	13.3	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	2560	1600	226.983005
1	Apple	Ultrabook	13.3	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	1440	900	127.677940
2	HP	Notebook	15.6	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	1920	1080	141.211998
3	Apple	Ultrabook	15.4	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	2880	1800	220.534624
4	Apple	Ultrabook	13.3	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	2560	1600	226.983005

```
In [48]: df.drop(columns=['Inches','X_res','Y_res'],inplace=True)
```

```
In [49]: df.head()
```

```
Out[49]:
```

	Company	TypeName	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi
0	Apple	Ultrabook	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	226.983005
1	Apple	Ultrabook	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	127.677940
2	HP	Notebook	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	141.211998
3	Apple	Ultrabook	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	220.534624
4	Apple	Ultrabook	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	226.983005

```
In [50]: df['Cpu'].value_counts()
```

```
Out[50]: Intel Core i5 7200U 2.5GHz      190
Intel Core i7 7700HQ 2.8GHz      146
Intel Core i7 7500U 2.7GHz      134
Intel Core i7 8550U 1.8GHz       73
Intel Core i5 8250U 1.6GHz       72

...
Intel Core M M3-6Y30 0.9GHz      1
AMD A9-Series 9420 2.9GHz        1
Intel Core i3 6006U 2.2GHz        1
AMD A6-Series 7310 2GHz           1
Intel Xeon E3-1535M v6 3.1GHz     1
Name: Cpu, Length: 118, dtype: int64
```

```
In [51]: df['Cpu Name'] = df['Cpu'].apply(lambda x: " ".join(x.split()[0:3]))
```

```
In [52]: df.head()
```

Out[52]:

	Company	TypeName	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu Name
0	Apple	Ultrabook	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	226.983005	Intel Core i5
1	Apple	Ultrabook	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	127.677940	Intel Core i5
2	HP	Notebook	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	141.211998	Intel Core i5
3	Apple	Ultrabook	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	220.534624	Intel Core i7
4	Apple	Ultrabook	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	226.983005	Intel Core i5

```
In [53]: def fetch_processor(text):
if text == 'Intel Core i7' or text == 'Intel Core i5' or text == 'Intel Core i3':
    return text
else:
    if text.split()[0] == 'Intel':
        return 'Other Intel Processor'
    else:
        return 'AMD Processor'
```

```
In [54]: df['Cpu brand'] = df['Cpu Name'].apply(fetch_processor)
```

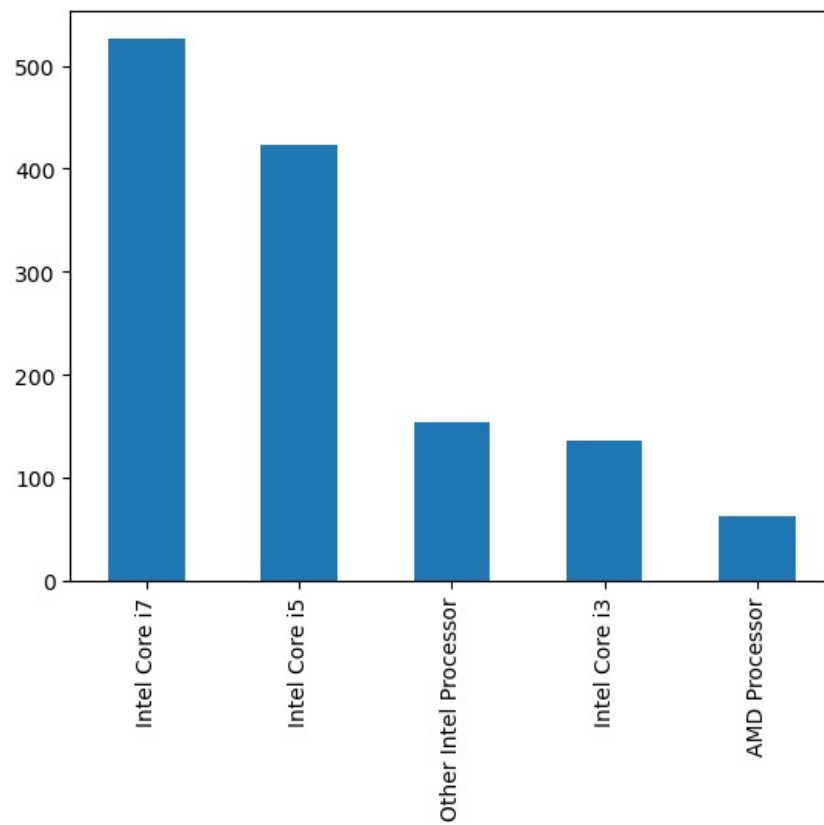
```
In [55]: df.head()
```

Out[55]:

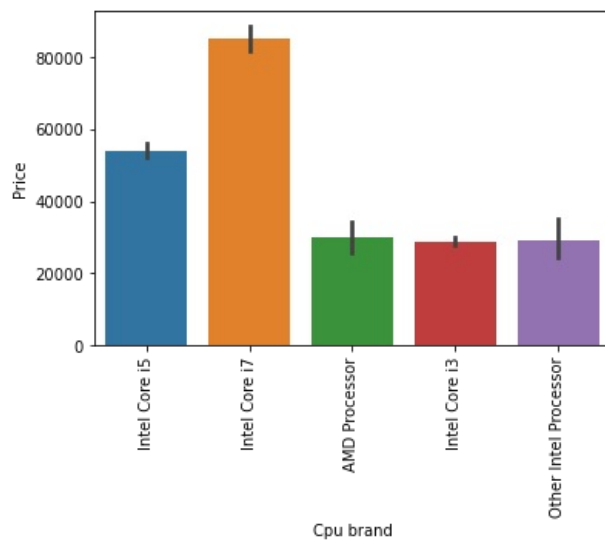
	Company	TypeName	Cpu	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu Name	Cpu brand
0	Apple	Ultrabook	Intel Core i5 2.3GHz	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	226.983005	Intel Core i5	Intel Core i5
1	Apple	Ultrabook	Intel Core i5 1.8GHz	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	127.677940	Intel Core i5	Intel Core i5
2	HP	Notebook	Intel Core i5 7200U 2.5GHz	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	141.211998	Intel Core i5	Intel Core i5
3	Apple	Ultrabook	Intel Core i7 2.7GHz	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	220.534624	Intel Core i7	Intel Core i7
4	Apple	Ultrabook	Intel Core i5 3.1GHz	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	226.983005	Intel Core i5	Intel Core i5

```
In [56]: df['Cpu brand'].value_counts().plot(kind='bar')
```

```
Out[56]: <Axes: >
```



```
In [87]: sns.barplot(x=df['Cpu brand'],y=df['Price'])
plt.xticks(rotation='vertical')
plt.show()
```



```
In [88]: df.drop(columns=['Cpu','Cpu Name'],inplace=True)
```

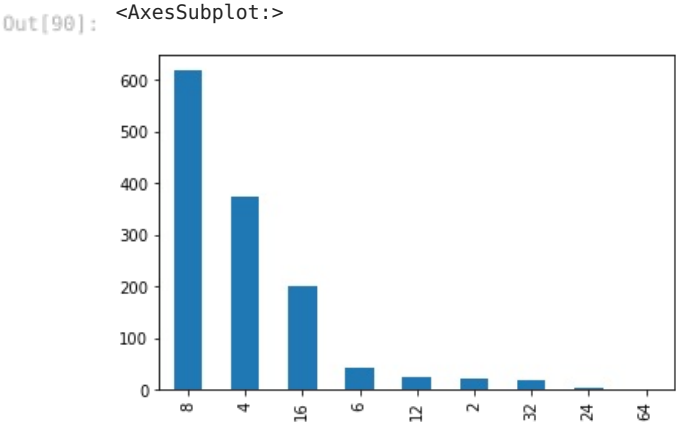
```
In [89]: df.head()
```

Out[89]:

	Company	TypeName	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	
0	Apple	Ultrabook	8	128GB SSD	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832		0	1	226.983005	Intel Core i5
1	Apple	Ultrabook	8	128GB Flash Storage	Intel HD Graphics 6000	macOS	1.34	47895.5232		0	0	127.677940	Intel Core i5
2	HP	Notebook	8	256GB SSD	Intel HD Graphics 620	No OS	1.86	30636.0000		0	0	141.211998	Intel Core i5
3	Apple	Ultrabook	16	512GB SSD	AMD Radeon Pro 455	macOS	1.83	135195.3360		0	1	220.534624	Intel Core i7
4	Apple	Ultrabook	8	256GB SSD	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080		0	1	226.983005	Intel Core i5

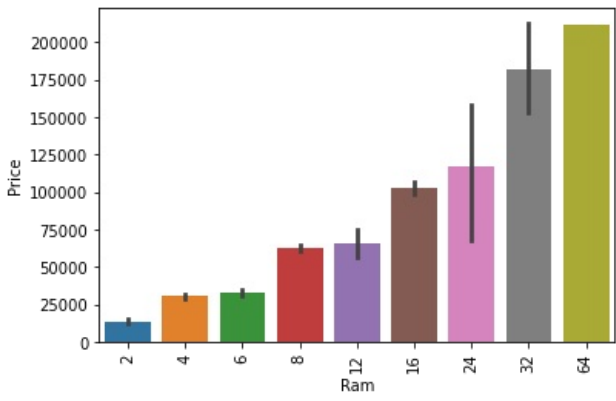
In [90]:

```
df['Ram'].value_counts().plot(kind='bar')
```



In [91]:

```
sns.barplot(x=df['Ram'],y=df['Price'])
plt.xticks(rotation='vertical')
plt.show()
```



In [92]:

```
df['Memory'].value_counts()
```

```
Out[92]:
```

256GB SSD	412
1TB HDD	223
500GB HDD	132
512GB SSD	118
128GB SSD + 1TB HDD	94
128GB SSD	76
256GB SSD + 1TB HDD	73
32GB Flash Storage	38
2TB HDD	16
64GB Flash Storage	15
512GB SSD + 1TB HDD	14
1TB SSD	14
256GB SSD + 2TB HDD	10
1.0TB Hybrid	9
256GB Flash Storage	8
16GB Flash Storage	7
32GB SSD	6
180GB SSD	5
128GB Flash Storage	4
16GB SSD	3
512GB SSD + 2TB HDD	3
256GB SSD + 256GB SSD	2
128GB SSD + 2TB HDD	2
256GB SSD + 500GB HDD	2
512GB Flash Storage	2
1TB SSD + 1TB HDD	2
32GB HDD	1
64GB SSD	1
1.0TB HDD	1
512GB SSD + 256GB SSD	1
512GB SSD + 1.0TB Hybrid	1
8GB SSD	1
240GB SSD	1
128GB HDD	1
1TB HDD + 1TB HDD	1
512GB SSD + 512GB SSD	1
256GB SSD + 1.0TB Hybrid	1
508GB Hybrid	1
64GB Flash Storage + 1TB HDD	1

Name: Memory, dtype: int64

```
In [93]: df['Memory'] = df['Memory'].astype(str).replace('\.0', '', regex=True)
df["Memory"] = df["Memory"].str.replace('GB', '')
df["Memory"] = df["Memory"].str.replace('TB', '000')
new = df["Memory"].str.split("+", n = 1, expand = True)

df["first"] = new[0]
df["first"] = df["first"].str.strip()

df["second"] = new[1]

df["Layer1HDD"] = df["first"].apply(lambda x: 1 if "HDD" in x else 0)
df["Layer1SSD"] = df["first"].apply(lambda x: 1 if "SSD" in x else 0)
df["Layer1Hybrid"] = df["first"].apply(lambda x: 1 if "Hybrid" in x else 0)
df["Layer1Flash_Storage"] = df["first"].apply(lambda x: 1 if "Flash Storage" in x else 0)

df['first'] = df['first'].str.replace(r'\D', '')

df["second"].fillna("0", inplace = True)

df["Layer2HDD"] = df["second"].apply(lambda x: 1 if "HDD" in x else 0)
df["Layer2SSD"] = df["second"].apply(lambda x: 1 if "SSD" in x else 0)
df["Layer2Hybrid"] = df["second"].apply(lambda x: 1 if "Hybrid" in x else 0)
df["Layer2Flash_Storage"] = df["second"].apply(lambda x: 1 if "Flash Storage" in x else 0)

df['second'] = df['second'].str.replace(r'\D', '')

df["first"] = df["first"].astype(int)
df["second"] = df["second"].astype(int)

df["HDD"] = (df["first"] * df["Layer1HDD"] + df["second"] * df["Layer2HDD"])
df["SSD"] = (df["first"] * df["Layer1SSD"] + df["second"] * df["Layer2SSD"])
df["Hybrid"] = (df["first"] * df["Layer1Hybrid"] + df["second"] * df["Layer2Hybrid"])
df["Flash_Storage"] = (df["first"] * df["Layer1Flash_Storage"] + df["second"] * df["Layer2Flash_Storage"])

df.drop(columns=['first', 'second', 'Layer1HDD', 'Layer1SSD', 'Layer1Hybrid',
                 'Layer1Flash_Storage', 'Layer2HDD', 'Layer2SSD', 'Layer2Hybrid',
                 'Layer2Flash_Storage'], inplace=True)
```

```
<ipython-input-93-10829db803de>:16: FutureWarning: The default value of regex will change from True to False in a future version.
df['first'] = df['first'].str.replace(r'\D', '')
<ipython-input-93-10829db803de>:25: FutureWarning: The default value of regex will change from True to False in a future version.
df['second'] = df['second'].str.replace(r'\D', '')
```

```
In [98]: df.sample(5)
```

Out[98]:

	Company	TypeName	Ram	Memory	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Hybrid
1247	Asus	Gaming	16	256 SSD + 1000 HDD	Nvidia GeForce GTX 1070	Windows 10	2.34	123876.000	0	1	141.211998	Intel Core i7	1000	256	0
505	Lenovo	Notebook	8	256 SSD	Intel HD Graphics 620	Windows 10	1.44	50562.720	0	0	165.632118	Intel Core i5	0	256	0
820	Lenovo	Notebook	4	500 HDD	Intel HD Graphics 520	Windows 10	2.10	26101.872	0	0	100.454670	Intel Core i3	500	0	0
21	Lenovo	Gaming	8	128 SSD + 1000 HDD	Nvidia GeForce GTX 1050	Windows 10	2.50	53226.720	0	1	141.211998	Intel Core i5	1000	128	0
301	Asus	Gaming	16	256 SSD + 1000 HDD	Nvidia GeForce GTX 1070	Windows 10	2.90	113060.160	0	0	127.335675	Intel Core i7	1000	256	0

In [99]: df.drop(columns=['Memory'],inplace=True)

In [100]: df.head()

Out[100]:

	Company	TypeName	Ram	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Hybrid	Flash_Storage
0	Apple	Ultrabook	8	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	226.983005	Intel Core i5	0	128	0	0
1	Apple	Ultrabook	8	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	127.677940	Intel Core i5	0	0	0	0
2	HP	Notebook	8	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	141.211998	Intel Core i5	0	256	0	0
3	Apple	Ultrabook	16	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	220.534624	Intel Core i7	0	512	0	0
4	Apple	Ultrabook	8	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	226.983005	Intel Core i5	0	256	0	0

In [101]: df.corr()['Price']

Out[101]:

```
Ram          0.743007
Weight       0.210370
Price        1.000000
Touchscreen  0.191226
Ips          0.252208
ppi          0.473487
HDD         -0.096441
SSD          0.670799
Hybrid       0.007989
Flash_Storage -0.040511
Name: Price, dtype: float64
```

In [102]: df.drop(columns=['Hybrid','Flash\_Storage'],inplace=True)

In [103]: df.head()

Out[103]:

	Company	TypeName	Ram	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD
0	Apple	Ultrabook	8	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	226.983005	Intel Core i5	0	128
1	Apple	Ultrabook	8	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	127.677940	Intel Core i5	0	0
2	HP	Notebook	8	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	141.211998	Intel Core i5	0	256
3	Apple	Ultrabook	16	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	220.534624	Intel Core i7	0	512
4	Apple	Ultrabook	8	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	226.983005	Intel Core i5	0	256

In [104]: df['Gpu'].value\_counts()

```
Out[104]: Intel HD Graphics 620      281
Intel HD Graphics 520      185
Intel UHD Graphics 620      68
Nvidia GeForce GTX 1050     66
Nvidia GeForce GTX 1060     48
...
Intel HD Graphics 540        1
AMD FirePro W6150M          1
AMD Radeon R5 M315          1
AMD Radeon R7 M360          1
AMD FirePro W5130M          1
Name: Gpu, Length: 110, dtype: int64
```

```
In [106]: df['Gpu brand'] = df['Gpu'].apply(lambda x:x.split()[0])
```

```
In [107]: df.head()
```

```
Out[107]:
```

	Company	TypeName	Ram	Gpu	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Gpu brand
0	Apple	Ultrabook	8	Intel Iris Plus Graphics 640	macOS	1.37	71378.6832	0	1	226.983005	Intel Core i5	0	128	Intel
1	Apple	Ultrabook	8	Intel HD Graphics 6000	macOS	1.34	47895.5232	0	0	127.677940	Intel Core i5	0	0	Intel
2	HP	Notebook	8	Intel HD Graphics 620	No OS	1.86	30636.0000	0	0	141.211998	Intel Core i5	0	256	Intel
3	Apple	Ultrabook	16	AMD Radeon Pro 455	macOS	1.83	135195.3360	0	1	220.534624	Intel Core i7	0	512	AMD
4	Apple	Ultrabook	8	Intel Iris Plus Graphics 650	macOS	1.37	96095.8080	0	1	226.983005	Intel Core i5	0	256	Intel

```
In [108]: df['Gpu brand'].value_counts()
```

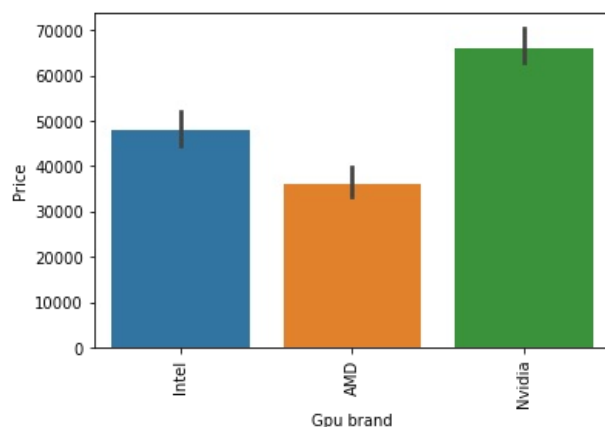
```
Out[108]: Intel      722
Nvidia    400
AMD       180
ARM        1
Name: Gpu brand, dtype: int64
```

```
In [111]: df = df[df['Gpu brand'] != 'ARM']
```

```
In [112]: df['Gpu brand'].value_counts()
```

```
Out[112]: Intel      722
Nvidia    400
AMD       180
Name: Gpu brand, dtype: int64
```

```
In [115]: sns.barplot(x=df['Gpu brand'],y=df['Price'],estimator=np.median)
plt.xticks(rotation='vertical')
plt.show()
```



```
In [116]: df.drop(columns=['Gpu'],inplace=True)
```

C:\Users\91842\anaconda3\lib\site-packages\pandas\core\frame.py:4308: SettingWithCopyWarning:  
A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
return super().drop()
```

```
In [117]: df.head()
```



Out[117]:

	Company	TypeName	Ram	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Gpu brand
0	Apple	Ultrabook	8	macOS	1.37	71378.6832	0	1	226.983005	Intel Core i5	0	128	Intel
1	Apple	Ultrabook	8	macOS	1.34	47895.5232	0	0	127.677940	Intel Core i5	0	0	Intel
2	HP	Notebook	8	No OS	1.86	30636.0000	0	0	141.211998	Intel Core i5	0	256	Intel
3	Apple	Ultrabook	16	macOS	1.83	135195.3360	0	1	220.534624	Intel Core i7	0	512	AMD
4	Apple	Ultrabook	8	macOS	1.37	96095.8080	0	1	226.983005	Intel Core i5	0	256	Intel

In [118..

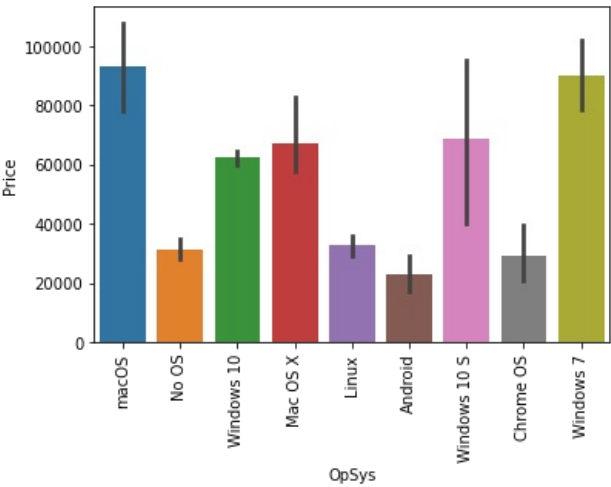
```
df['OpSys'].value_counts()
```

Out[118]:

```
Windows 10      1072
No OS           66
Linux           62
Windows 7       45
Chrome OS       26
macOS           13
Windows 10 S     8
Mac OS X        8
Android         2
Name: OpSys, dtype: int64
```

In [120..

```
sns.barplot(x=df['OpSys'],y=df['Price'])
plt.xticks(rotation='vertical')
plt.show()
```



In [121..

```
def cat_os(inp):
    if inp == 'Windows 10' or inp == 'Windows 7' or inp == 'Windows 10 S':
        return 'Windows'
    elif inp == 'macOS' or inp == 'Mac OS X':
        return 'Mac'
    else:
        return 'Others/No OS/Linux'
```

In [122..

```
df['os'] = df['OpSys'].apply(cat_os)
```

<ipython-input-122-38671a3c07bd>:1: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.  
Try using `.loc[row_indexer,col_indexer] = value` instead

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
df['os'] = df['OpSys'].apply(cat_os)
```

In [123..

```
df.head()
```

Out[123]:

	Company	TypeName	Ram	OpSys	Weight	Price	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Gpu brand	os	
0	Apple	Ultrabook	8	macOS	1.37	71378.6832		0	1	226.983005	Intel Core i5	0	128	Intel	Mac
1	Apple	Ultrabook	8	macOS	1.34	47895.5232		0	0	127.677940	Intel Core i5	0	0	Intel	Mac
2	HP	Notebook	8	No OS	1.86	30636.0000		0	0	141.211998	Intel Core i5	0	256	Intel	Others/No OS/Linux
3	Apple	Ultrabook	16	macOS	1.83	135195.3360		0	1	220.534624	Intel Core i7	0	512	AMD	Mac
4	Apple	Ultrabook	8	macOS	1.37	96095.8080		0	1	226.983005	Intel Core i5	0	256	Intel	Mac

In [124..

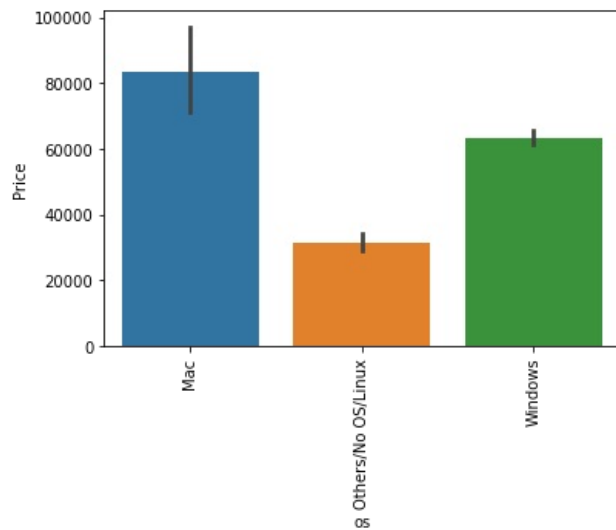
```
df.drop(columns=['OpSys'],inplace=True)
```

```
C:\Users\91842\anaconda3\lib\site-packages\pandas\core\frame.py:4308: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
```

See the caveats in the documentation: [https://pandas.pydata.org/pandas-docs/stable/user\\_guide/indexing.html#returning-a-view-versus-a-copy](https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy)

```
return super().drop()
```

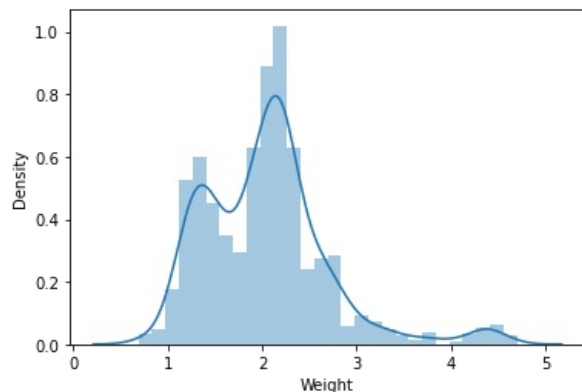
```
In [125]: sns.barplot(x=df['os'],y=df['Price'])
plt.xticks(rotation='vertical')
plt.show()
```



```
In [126]: sns.distplot(df['Weight'])
```

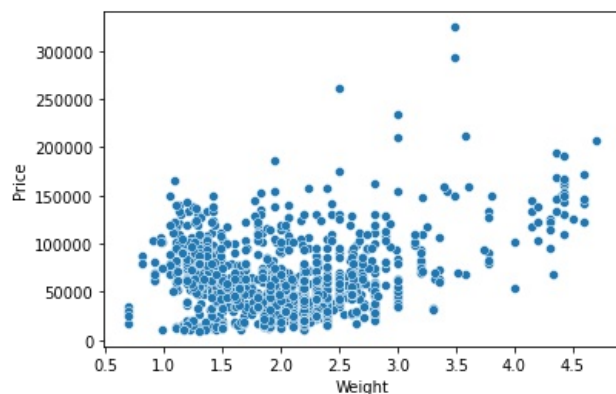
```
C:\Users\91842\anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).
warnings.warn(msg, FutureWarning)
```

```
Out[126]: <AxesSubplot:xlabel='Weight', ylabel='Density'>
```



```
In [127]: sns.scatterplot(x=df['Weight'],y=df['Price'])
```

```
Out[127]: <AxesSubplot:xlabel='Weight', ylabel='Price'>
```

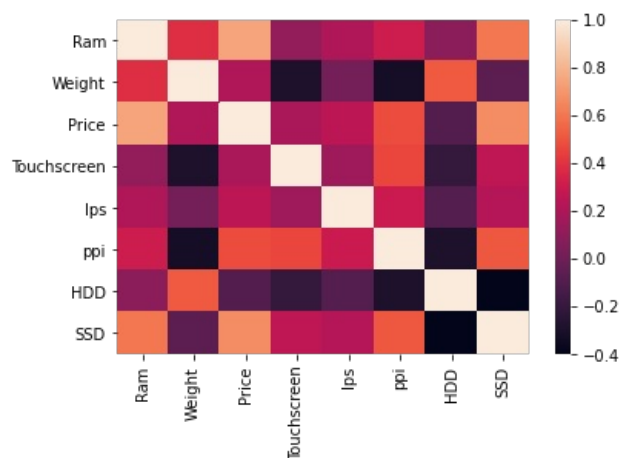


```
In [128]: df.corr()['Price']
```

```
Out[128]: Ram          0.742905
Weight       0.209867
Price        1.000000
Touchscreen  0.192917
Ips          0.253320
ppi          0.475368
HDD          -0.096891
SSD          0.670660
Name: Price, dtype: float64
```

```
In [130]: sns.heatmap(df.corr())
```

```
Out[130]: <AxesSubplot:>
```

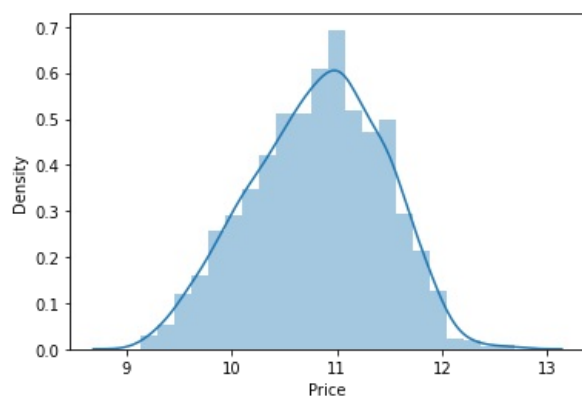


```
In [133]: sns.distplot(np.log(df['Price']))
```

C:\Users\91842\anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

```
Out[133]: <AxesSubplot:xlabel='Price', ylabel='Density'>
```



```
In [134]: X = df.drop(columns=['Price'])
y = np.log(df['Price'])
```

```
In [135]: X
```

Out[135]:

	Company	TypeName	Ram	Weight	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Gpu brand	os
0	Apple	Ultrabook	8	1.37	0	1	226.983005	Intel Core i5	0	128	Intel	Mac
1	Apple	Ultrabook	8	1.34	0	0	127.677940	Intel Core i5	0	0	Intel	Mac
2	HP	Notebook	8	1.86	0	0	141.211998	Intel Core i5	0	256	Intel	Others/No OS/Linux
3	Apple	Ultrabook	16	1.83	0	1	220.534624	Intel Core i7	0	512	AMD	Mac
4	Apple	Ultrabook	8	1.37	0	1	226.983005	Intel Core i5	0	256	Intel	Mac
...	...	...	...	...	...	...	...	...	...	...	...	...
1298	Lenovo	2 in 1 Convertible	4	1.80	1	1	157.350512	Intel Core i7	0	128	Intel	Windows
1299	Lenovo	2 in 1 Convertible	16	1.30	1	1	276.053530	Intel Core i7	0	512	Intel	Windows
1300	Lenovo	Notebook	2	1.50	0	0	111.935204	Other Intel Processor	0	0	Intel	Windows
1301	HP	Notebook	6	2.19	0	0	100.454670	Intel Core i7	1000	0	AMD	Windows
1302	Asus	Notebook	4	2.20	0	0	100.454670	Other Intel Processor	500	0	Intel	Windows

1302 rows × 12 columns

In [136..

y

Out[136]:

0	11.175755
1	10.776777
2	10.329931
3	11.814476
4	11.473101
...	
1298	10.433899
1299	11.288115
1300	9.409283
1301	10.614129
1302	9.886358

Name: Price, Length: 1302, dtype: float64

In [137..

```
from sklearn.model_selection import train_test_split
X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.15,random_state=2)
```

In [138..

X\_train

Out[138]:

	Company	TypeName	Ram	Weight	Touchscreen	Ips	ppi	Cpu brand	HDD	SSD	Gpu brand	os
183	Toshiba	Notebook	8	2.00	0	0	100.454670	Intel Core i5	0	128	Intel	Windows
1141	MSI	Gaming	8	2.40	0	0	141.211998	Intel Core i7	1000	128	Nvidia	Windows
1049	Asus	Netbook	4	1.20	0	0	135.094211	Other Intel Processor	0	0	Intel	Others/No OS/Linux
1020	Dell	2 in 1 Convertible	4	2.08	1	1	141.211998	Intel Core i3	1000	0	Intel	Windows
878	Dell	Notebook	4	2.18	0	0	141.211998	Intel Core i5	1000	128	Nvidia	Windows
...	...	...	...	...	...	...	...	...	...	...	...	...
466	Acer	Notebook	4	2.20	0	0	100.454670	Intel Core i3	500	0	Nvidia	Windows
299	Asus	Ultrabook	16	1.63	0	0	141.211998	Intel Core i7	0	512	Nvidia	Windows
493	Acer	Notebook	8	2.20	0	0	100.454670	AMD Processor	1000	0	AMD	Windows
527	Lenovo	Notebook	8	2.20	0	0	100.454670	Intel Core i3	2000	0	Nvidia	Others/No OS/Linux
1193	Apple	Ultrabook	8	0.92	0	1	226.415547	Other Intel Processor	0	0	Intel	Mac

1106 rows × 12 columns

In [144..

```
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import OneHotEncoder
from sklearn.metrics import r2_score,mean_absolute_error
```

In [57]:

```
from sklearn.linear_model import LinearRegression,Ridge,Lasso
from sklearn.neighbors import KNeighborsRegressor
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor,GradientBoostingRegressor,AdaBoostRegressor,ExtraTreesRegressor
from sklearn.svm import SVR
from xgboost import XGBRegressor
```

```
-----  
ModuleNotFoundError                                Traceback (most recent call last)  
Cell In[57], line 6  
      4 from sklearn.ensemble import  
RandomForestRegressor,GradientBoostingRegressor,AdaBoostRegressor,ExtraTreesRegressor  
      5 from sklearn.svm import SVR  
----> 6 from xgboost import XGBRegressor  
  
ModuleNotFoundError: No module named 'xgboost'
```

## Linear regression

```
In [145.. step1 = ColumnTransformer(transformers=[  
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])  
],remainder='passthrough')  
  
step2 = LinearRegression()  
  
pipe = Pipeline([  
    ('step1',step1),  
    ('step2',step2)  
])  
  
pipe.fit(X_train,y_train)  
  
y_pred = pipe.predict(X_test)  
  
print('R2 score',r2_score(y_test,y_pred))  
print('MAE',mean_absolute_error(y_test,y_pred))  
  
R2 score 0.8073277448418521  
MAE 0.21017827976429174
```

## Ridge Regression

```
In [158.. step1 = ColumnTransformer(transformers=[  
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])  
],remainder='passthrough')  
  
step2 = Ridge(alpha=10)  
  
pipe = Pipeline([  
    ('step1',step1),  
    ('step2',step2)  
])  
  
pipe.fit(X_train,y_train)  
  
y_pred = pipe.predict(X_test)  
  
print('R2 score',r2_score(y_test,y_pred))  
print('MAE',mean_absolute_error(y_test,y_pred))  
  
R2 score 0.8127331031311811  
MAE 0.20926802242582954
```

## Lasso Regression

```
In [171.. step1 = ColumnTransformer(transformers=[  
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])  
],remainder='passthrough')  
  
step2 = Lasso(alpha=0.001)  
  
pipe = Pipeline([  
    ('step1',step1),  
    ('step2',step2)  
])  
  
pipe.fit(X_train,y_train)  
  
y_pred = pipe.predict(X_test)  
  
print('R2 score',r2_score(y_test,y_pred))  
print('MAE',mean_absolute_error(y_test,y_pred))  
  
R2 score 0.8071853945317105  
MAE 0.21114361613472565
```

## KNN

```
In [180.. step1 = ColumnTransformer(transformers=[  
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])  
],remainder='passthrough')
```

```

step2 = KNeighborsRegressor(n_neighbors=3)

pipe = Pipeline([
    ('step1', step1),
    ('step2', step2)
])

pipe.fit(X_train, y_train)

y_pred = pipe.predict(X_test)

print('R2 score', r2_score(y_test, y_pred))
print('MAE', mean_absolute_error(y_test, y_pred))

```

R2 score 0.8021984604448553  
MAE 0.19319716721521116

## Decision Tree

```

In [191.. step1 = ColumnTransformer(transformers=[
    ('col_tnf', OneHotEncoder(sparse=False, drop='first'), [0, 1, 7, 10, 11])
], remainder='passthrough')

step2 = DecisionTreeRegressor(max_depth=8)

pipe = Pipeline([
    ('step1', step1),
    ('step2', step2)
])

pipe.fit(X_train, y_train)

y_pred = pipe.predict(X_test)

print('R2 score', r2_score(y_test, y_pred))
print('MAE', mean_absolute_error(y_test, y_pred))

```

R2 score 0.8466456692979233  
MAE 0.1806340977609143

## SVM

```

In [213.. step1 = ColumnTransformer(transformers=[
    ('col_tnf', OneHotEncoder(sparse=False, drop='first'), [0, 1, 7, 10, 11])
], remainder='passthrough')

step2 = SVR(kernel='rbf', C=10000, epsilon=0.1)

pipe = Pipeline([
    ('step1', step1),
    ('step2', step2)
])

pipe.fit(X_train, y_train)

y_pred = pipe.predict(X_test)

print('R2 score', r2_score(y_test, y_pred))
print('MAE', mean_absolute_error(y_test, y_pred))

```

R2 score 0.8083180902257614  
MAE 0.20239059427481307

## Random Forest

```

In [306.. step1 = ColumnTransformer(transformers=[
    ('col_tnf', OneHotEncoder(sparse=False, drop='first'), [0, 1, 7, 10, 11])
], remainder='passthrough')

step2 = RandomForestRegressor(n_estimators=100,
                             random_state=3,
                             max_samples=0.5,
                             max_features=0.75,
                             max_depth=15)

pipe = Pipeline([
    ('step1', step1),
    ('step2', step2)
])

pipe.fit(X_train, y_train)

y_pred = pipe.predict(X_test)

print('R2 score', r2_score(y_test, y_pred))
print('MAE', mean_absolute_error(y_test, y_pred))

```

R2 score 0.8873402378382488  
MAE 0.15860130110457718

## ExtraTrees

```
In [228.. step1 = ColumnTransformer(transformers=[
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])
],remainder='passthrough')

step2 = ExtraTreesRegressor(n_estimators=100,
                             random_state=3,
                             max_samples=0.5,
                             max_features=0.75,
                             max_depth=15)

pipe = Pipeline([
    ('step1',step1),
    ('step2',step2)
])

pipe.fit(X_train,y_train)

y_pred = pipe.predict(X_test)

print('R2 score',r2_score(y_test,y_pred))
print('MAE',mean_absolute_error(y_test,y_pred))
```

R2 score 0.8753793123440623  
MAE 0.15979519126758127

## AdaBoost

```
In [244.. step1 = ColumnTransformer(transformers=[
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])
],remainder='passthrough')

step2 = AdaBoostRegressor(n_estimators=15,learning_rate=1.0)

pipe = Pipeline([
    ('step1',step1),
    ('step2',step2)
])

pipe.fit(X_train,y_train)

y_pred = pipe.predict(X_test)

print('R2 score',r2_score(y_test,y_pred))
print('MAE',mean_absolute_error(y_test,y_pred))
```

R2 score 0.7929652659237908  
MAE 0.23296532406396742

## Gradient Boost

```
In [260.. step1 = ColumnTransformer(transformers=[
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])
],remainder='passthrough')

step2 = GradientBoostingRegressor(n_estimators=500)

pipe = Pipeline([
    ('step1',step1),
    ('step2',step2)
])

pipe.fit(X_train,y_train)

y_pred = pipe.predict(X_test)

print('R2 score',r2_score(y_test,y_pred))
print('MAE',mean_absolute_error(y_test,y_pred))
```

R2 score 0.8823244736036472  
MAE 0.15929506744611283

## XgBoost

```
In [287.. step1 = ColumnTransformer(transformers=[
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])
],remainder='passthrough')

step2 = XGBRegressor(n_estimators=45,max_depth=5,learning_rate=0.5)

pipe = Pipeline([
```

```

        ('step1',step1),
        ('step2',step2)
    ])

    pipe.fit(X_train,y_train)

    y_pred = pipe.predict(X_test)

    print('R2 score',r2_score(y_test,y_pred))
    print('MAE',mean_absolute_error(y_test,y_pred))

```

R2 score 0.8811773435850243  
MAE 0.16496203512600974

## Voting Regressor

```

In [305.. from sklearn.ensemble import VotingRegressor,StackingRegressor

step1 = ColumnTransformer(transformers=[
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])
],remainder='passthrough')

rf = RandomForestRegressor(n_estimators=350,random_state=3,max_samples=0.5,max_features=0.75,max_depth=15)
gbdt = GradientBoostingRegressor(n_estimators=100,max_features=0.5)
xgb = XGBRegressor(n_estimators=25,learning_rate=0.3,max_depth=5)
et = ExtraTreesRegressor(n_estimators=100,random_state=3,max_samples=0.5,max_features=0.75,max_depth=10)

step2 = VotingRegressor([('rf', rf), ('gbdt', gbdt), ('xgb',xgb), ('et',et)],weights=[5,1,1,1])

pipe = Pipeline([
    ('step1',step1),
    ('step2',step2)
])

pipe.fit(X_train,y_train)

y_pred = pipe.predict(X_test)

print('R2 score',r2_score(y_test,y_pred))
print('MAE',mean_absolute_error(y_test,y_pred))

```

R2 score 0.8901036732986811  
MAE 0.15847265699907628

## Stacking

```

In [296.. from sklearn.ensemble import VotingRegressor,StackingRegressor

step1 = ColumnTransformer(transformers=[
    ('col_tnf',OneHotEncoder(sparse=False,drop='first'),[0,1,7,10,11])
],remainder='passthrough')

estimators = [
    ('rf', RandomForestRegressor(n_estimators=350,random_state=3,max_samples=0.5,max_features=0.75,max_depth=15)
    ('gbdt',GradientBoostingRegressor(n_estimators=100,max_features=0.5)),
    ('xgb', XGBRegressor(n_estimators=25,learning_rate=0.3,max_depth=5))
]

step2 = StackingRegressor(estimators=estimators, final_estimator=Ridge(alpha=100))

pipe = Pipeline([
    ('step1',step1),
    ('step2',step2)
])

pipe.fit(X_train,y_train)

y_pred = pipe.predict(X_test)

print('R2 score',r2_score(y_test,y_pred))
print('MAE',mean_absolute_error(y_test,y_pred))

```

R2 score 0.8816958647512341  
MAE 0.1663048975120589

## Exporting the Model

```

In [308.. import pickle

pickle.dump(df,open('df.pkl','wb'))
pickle.dump(pipe,open('pipe.pkl','wb'))

```

In [307.. df



Out[307]:

	Company	TypeName	Ram	Weight	Price	Touchscreen	lps	ppi	Cpu brand	HDD	SSD	Gpu brand	os
0	Apple	Ultrabook	8	1.37	71378.6832	0	1	226.983005	Intel Core i5	0	128	Intel	Mac
1	Apple	Ultrabook	8	1.34	47895.5232	0	0	127.677940	Intel Core i5	0	0	Intel	Mac
2	HP	Notebook	8	1.86	30636.0000	0	0	141.211998	Intel Core i5	0	256	Intel	Others/No OS/Linux
3	Apple	Ultrabook	16	1.83	135195.3360	0	1	220.534624	Intel Core i7	0	512	AMD	Mac
4	Apple	Ultrabook	8	1.37	96095.8080	0	1	226.983005	Intel Core i5	0	256	Intel	Mac
...	...	...	...	...	...	...	...	...	...	...	...	...	...
1298	Lenovo	2 in 1 Convertible	4	1.80	33992.6400	1	1	157.350512	Intel Core i7	0	128	Intel	Windows
1299	Lenovo	2 in 1 Convertible	16	1.30	79866.7200	1	1	276.053530	Intel Core i7	0	512	Intel	Windows
1300	Lenovo	Notebook	2	1.50	12201.1200	0	0	111.935204	Other Intel Processor	0	0	Intel	Windows
1301	HP	Notebook	6	2.19	40705.9200	0	0	100.454670	Intel Core i7	1000	0	AMD	Windows
1302	Asus	Notebook	4	2.20	19660.3200	0	0	100.454670	Other Intel Processor	500	0	Intel	Windows

1302 rows × 13 columns

In [309]:

X\_train

Out[309]:

	Company	TypeName	Ram	Weight	Touchscreen	lps	ppi	Cpu brand	HDD	SSD	Gpu brand	os
183	Toshiba	Notebook	8	2.00	0	0	100.454670	Intel Core i5	0	128	Intel	Windows
1141	MSI	Gaming	8	2.40	0	0	141.211998	Intel Core i7	1000	128	Nvidia	Windows
1049	Asus	Netbook	4	1.20	0	0	135.094211	Other Intel Processor	0	0	Intel	Others/No OS/Linux
1020	Dell	2 in 1 Convertible	4	2.08	1	1	141.211998	Intel Core i3	1000	0	Intel	Windows
878	Dell	Notebook	4	2.18	0	0	141.211998	Intel Core i5	1000	128	Nvidia	Windows
...	...	...	...	...	...	...	...	...	...	...	...	...
466	Acer	Notebook	4	2.20	0	0	100.454670	Intel Core i3	500	0	Nvidia	Windows
299	Asus	Ultrabook	16	1.63	0	0	141.211998	Intel Core i7	0	512	Nvidia	Windows
493	Acer	Notebook	8	2.20	0	0	100.454670	AMD Processor	1000	0	AMD	Windows
527	Lenovo	Notebook	8	2.20	0	0	100.454670	Intel Core i3	2000	0	Nvidia	Others/No OS/Linux
1193	Apple	Ultrabook	8	0.92	0	1	226.415547	Other Intel Processor	0	0	Intel	Mac

1106 rows × 12 columns

In [ ]:

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