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```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import re as re
from sklearn.preprocessing import LabelEncoder,StandardScaler
from sklearn.linear_model import LinearRegression,Lasso
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error,mean_absolute_error
from sklearn.ensemble import RandomForestRegressor
import warnings
warnings.filterwarnings("ignore")
```

```
#load data
df = pd.read_csv("/content/sample_data/data.csv")
print(df.head())
```



	Unnamed: 0	brand	name	price	\
0	0	HP	Victus 15-fb0157AX Gaming Laptop	49900	
1	1	HP	15s-fq5007TU Laptop	39900	
2	2	Acer	One 14 Z8-415 Laptop	26990	
3	3	Lenovo	Yoga Slim 6 14IAP8 82WU0095IN Laptop	59729	
4	4	Apple	MacBook Air 2020 MGND3HN Laptop	69990	

	spec_rating	processor	CPU	\
0	73.000000	5th Gen AMD Ryzen 5 5600H	Hexa Core, 12 Threads	
1	60.000000	12th Gen Intel Core i3 1215U	Hexa Core (2P + 4E), 8 Threads	
2	69.323529	11th Gen Intel Core i3 1115G4	Dual Core, 4 Threads	
3	66.000000	12th Gen Intel Core i5 1240P	12 Cores (4P + 8E), 16 Threads	
4	69.323529	Apple M1	Octa Core (4P + 4E)	

	Ram	Ram_type	ROM	ROM_type	GPU	display_size	\
0	8GB	DDR4	512GB	SSD	4GB AMD Radeon RX 6500M	15.6	
1	8GB	DDR4	512GB	SSD	Intel UHD Graphics	15.6	
2	8GB	DDR4	512GB	SSD	Intel Iris Xe Graphics	14.0	
3	16GB	LPDDR5	512GB	SSD	Intel Integrated Iris Xe	14.0	
4	8GB	DDR4	256GB	SSD	Apple M1 Integrated Graphics	13.3	

	resolution_width	resolution_height	OS	warranty
0	1920	1080	Windows 11 OS	1
1	1920	1080	Windows 11 OS	1
2	1920	1080	Windows 11 OS	1
3	2240	1400	Windows 11 OS	1
4	2560	1600	Mac OS	1

```
#Using drop to clean data
df = df.drop(['Unnamed: 0'], axis=1)
df.head()
```



	brand	name	price	spec_rating	processor	CPU	Ram	Ram_type	ROM	ROM_1
0	HP	Victus 15-fb0157AX Gaming Laptop	49900	73.000000	5th Gen AMD Ryzen 5 5600H	Hexa Core, 12 Threads	8GB	DDR4	512GB	
1	HP	15s-fq5007TU Laptop	39900	60.000000	12th Gen Intel Core i3 1215U	Hexa Core (2P + 4E), 8 Threads	8GB	DDR4	512GB	
2	Acer	One 14 Z8-415 Laptop	26990	69.323529	11th Gen Intel Core i3 1115G4	Dual Core, 4 Threads	8GB	DDR4	512GB	
3	Lenovo	Yoga Slim 6 14IAP8 82WU0095IN Laptop	59729	66.000000	12th Gen Intel Core i5 1240P	12 Cores (4P + 8E), 16 Threads	16GB	LPDDR5	512GB	
4	Apple	MacBook Air 2020 MGND3HN Laptop	69990	69.323529	Apple M1	Octa Core (4P + 4E)	8GB	DDR4	256GB	

```
#check the duplicated rows if any
print(df.duplicated().sum())
print(df.shape)
```



0
(893, 16)

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



```
brand      0
name       0
price      0
spec_rating 0
processor  0
CPU        0
Ram        0
Ram_type   0
ROM        0
ROM_type   0
GPU        0
display_size 0
resolution_width 0
resolution_height 0
OS         0
warranty   0
dtype: int64
```

```
#descriptions
df.brand.value_counts()
df.shape
df.describe(include='object')
df.describe()
```



	price	spec_rating	display_size	resolution_width	resolution_height	war
count	893.000000	893.000000	893.000000	893.000000	893.000000	893.0
mean	79907.409854	69.379026	15.173751	2035.393057	1218.324748	1.0
std	60880.043823	5.541555	0.939095	426.076009	326.756883	0.3
min	9999.000000	60.000000	11.600000	1080.000000	768.000000	0.0
25%	44500.000000	66.000000	14.000000	1920.000000	1080.000000	1.0
50%	61990.000000	69.323529	15.600000	1920.000000	1080.000000	1.0
75%	90990.000000	71.000000	15.600000	1920.000000	1200.000000	1.0
max	450039.000000	89.000000	18.000000	3840.000000	3456.000000	3.0

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```
# Identify numeric and categorical features
numeric_features = ['price', 'spec_rating', 'Ram', 'ROM', 'display_size', 'resolution_width', 'resolution_height', 'war']
categorical_features = ['brand', 'name', 'processor', 'CPU', 'Ram_type', 'ROM_type', 'GPU', '']

#Extract numbers and units
#Ram
def add_space_between_number_and_unit_Ram(input_string):
    return re.sub(r'(\d+)([a-zA-Z]+)', r'\1 \2', input_string)

df['Ram'] = df['Ram'].apply(add_space_between_number_and_unit_Ram)
print(df)
```



	brand			name	price	\
0	HP		Victus 15-fb0157AX Gaming Laptop	49900		
1	HP		15s-fq5007TU Laptop	39900		
2	Acer		One 14 Z8-415 Laptop	26990		
3	Lenovo		Yoga Slim 6 14IAP8 82WU0095IN Laptop	59729		
4	Apple		MacBook Air 2020 MGN03HN Laptop	69990		
..		
888	Asus		Vivobook 15X 2023 K3504VAB-NJ321WS Laptop	44990		
889	Asus		TUF A15 FA577RM-HQ032WS Laptop	110000		
890	Asus	ROG Zephyrus G14 2023 GA402XV-N2034WS Gaming L...	189990			
891	Asus	TUF Gaming F15 2023 FX507VU-LP083WS Gaming Laptop	129990			
892	Asus	TUF Gaming A15 2023 FA577XU-LP041WS Gaming Laptop	131990			
	spec_rating		processor	\		
0	73.000000		5th Gen AMD Ryzen 5 5600H			
1	60.000000		12th Gen Intel Core i3 1215U			
2	69.323529		11th Gen Intel Core i3 1115G4			
3	66.000000		12th Gen Intel Core i5 1240P			
4	69.323529		Apple M1			
..			
888	69.323529		13th Gen Intel Core i3 1315U			
889	71.000000		6th Gen AMD Ryzen 7 6800H			
890	89.000000		7th Gen AMD Ryzen 9 7940HS			
891	73.000000		13th Gen Intel Core i7 13700H			
892	84.000000		7th Gen AMD Ryzen 9 7940HS			
		CPU	Ram	Ram_type	ROM	ROM_type \
0		Hexa Core, 12 Threads	8 GB	DDR4	512GB	SSD
1		Hexa Core (2P + 4E), 8 Threads	8 GB	DDR4	512GB	SSD
2		Dual Core, 4 Threads	8 GB	DDR4	512GB	SSD
3		12 Cores (4P + 8E), 16 Threads	16 GB	LPDDR5	512GB	SSD

4	Octa Core (4P + 4E)	8 GB	DDR4	256GB	SSD
...
888	Hexa Core (2P + 4E), 8 Threads	8 GB	DDR4	512GB	SSD
889	Octa Core, 16 Threads	16 GB	DDR	1TB	SSD
890	Octa Core, 16 Threads	32 GB	DDR5	1TB	SSD
891	14 Cores (6P + 8E), 20 Threads	16 GB	DDR4	512GB	SSD
892	Octa Core, 16 Threads	16 GB	DDR4	1TB	SSD

	GPU	display_size	resolution_width	\
0	4GB AMD Radeon RX 6500M	15.6	1920	
1	Intel UHD Graphics	15.6	1920	
2	Intel Iris Xe Graphics	14.0	1920	
3	Intel Integrated Iris Xe	14.0	2240	
4	Apple M1 Integrated Graphics	13.3	2560	
...
888	Integrated Intel UHD Graphics	15.6	1920	
889	6GB NVIDIA GeForce RTX 3060	15.6	2560	
890	8GB NVIDIA GeForce RTX 4060	14.0	2560	
891	6GB NVIDIA GeForce RTX 4050	15.6	1920	
892	6GB NVIDIA GeForce RTX 4050	15.6	1920	

	resolution_height	OS	warranty
0	1080	Windows 11 OS	1
1	1080	Windows 11 OS	1
2	1080	Windows 11 OS	1
3	1400	Windows 11 OS	1
4	1600	Mac OS	1

#ROM

```
def add_space_between_number_and_unit_ROM(input_string):
    return re.sub(r'(\d+)([a-zA-Z]+)', r'\1 \2', input_string)
```

```
df['ROM'] = df['ROM'].apply(add_space_between_number_and_unit_ROM)
print(df)
```

	brand	name	price	\
0	HP	Victus 15-fb0157AX Gaming Laptop	49900	
1	HP	15s-fq5007TU Laptop	39900	
2	Acer	One 14 Z8-415 Laptop	26990	
3	Lenovo	Yoga Slim 6 14IAP8 82WU0095IN Laptop	59729	
4	Apple	MacBook Air 2020 MGND3HN Laptop	69990	
...
888	Asus	Vivobook 15X 2023 K3504VAB-NJ321WS Laptop	44990	
889	Asus	TUF A15 FA577RM-HQ032WS Laptop	110000	
890	Asus	ROG Zephyrus G14 2023 GA402XV-N2034WS Gaming L...	189990	
891	Asus	TUF Gaming F15 2023 FX507VU-LP083WS Gaming Laptop	129990	
892	Asus	TUF Gaming A15 2023 FA577XU-LP041WS Gaming Laptop	131990	

	spec_rating	processor	\
0	73.000000	5th Gen AMD Ryzen 5 5600H	
1	60.000000	12th Gen Intel Core i3 1215U	
2	69.323529	11th Gen Intel Core i3 1115G4	
3	66.000000	12th Gen Intel Core i5 1240P	
4	69.323529	Apple M1	
...
888	69.323529	13th Gen Intel Core i3 1315U	
889	71.000000	6th Gen AMD Ryzen 7 6800H	
890	89.000000	7th Gen AMD Ryzen 9 7940HS	
891	73.000000	13th Gen Intel Core i7 13700H	
892	84.000000	7th Gen AMD Ryzen 9 7940HS	

	CPU	Ram	Ram_type	ROM	ROM_type	\
0	Hexa Core, 12 Threads	8 GB	DDR4	512 GB	SSD	
1	Hexa Core (2P + 4E), 8 Threads	8 GB	DDR4	512 GB	SSD	
2	Dual Core, 4 Threads	8 GB	DDR4	512 GB	SSD	
3	12 Cores (4P + 8E), 16 Threads	16 GB	LPDDR5	512 GB	SSD	

4	Octa Core (4P + 4E)	8 GB	DDR4	256 GB	SSD
...
888	Hexa Core (2P + 4E), 8 Threads	8 GB	DDR4	512 GB	SSD
889	Octa Core, 16 Threads	16 GB	DDR	1 TB	SSD
890	Octa Core, 16 Threads	32 GB	DDR5	1 TB	SSD
891	14 Cores (6P + 8E), 20 Threads	16 GB	DDR4	512 GB	SSD
892	Octa Core, 16 Threads	16 GB	DDR4	1 TB	SSD

	GPU	display_size	resolution_width	\
0	4GB AMD Radeon RX 6500M	15.6	1920	
1	Intel UHD Graphics	15.6	1920	
2	Intel Iris Xe Graphics	14.0	1920	
3	Intel Integrated Iris Xe	14.0	2240	
4	Apple M1 Integrated Graphics	13.3	2560	
...
888	Integrated Intel UHD Graphics	15.6	1920	
889	6GB NVIDIA GeForce RTX 3060	15.6	2560	
890	8GB NVIDIA GeForce RTX 4060	14.0	2560	
891	6GB NVIDIA GeForce RTX 4050	15.6	1920	
892	6GB NVIDIA GeForce RTX 4050	15.6	1920	

	resolution_height	OS	warranty
0	1080	Windows 11 OS	1
1	1080	Windows 11 OS	1
2	1080	Windows 11 OS	1
3	1400	Windows 11 OS	1
4	1600	Windows 11 OS	1

```
# converting from TB>GB unit for Ram column
```

```
def turn_Ram_into_GB(value):
    if 'TB' in value:
        return float(value[:value.find('TB')]) * 1024
```

```
df ['Ram'] = df['Ram'].apply(turn_Ram_into_GB)
```

```
def turn_ROM_into_GB(value):
    if 'TB' in value:
        return int(value[:value.find('TB')]) * 1024
```

```
df ['ROM'] = df['ROM'].apply(turn_ROM_into_GB)
```

```
#Remove the unit (String)
```

```
df ['Ram'] = df['Ram'].str[:-2]
```

```
df ['ROM'] = df['ROM'].str[:-2]
print(df)
```

	brand	name	price	\
0	HP	Victus 15-fb0157AX Gaming Laptop	49900	
1	HP	15s-fq5007TU Laptop	39900	
2	Acer	One 14 Z8-415 Laptop	26990	
3	Lenovo	Yoga Slim 6 14IAP8 82WU0095IN Laptop	59729	
4	Apple	MacBook Air 2020 MGN03HN Laptop	69990	
...
888	Asus	Vivobook 15X 2023 K3504VAB-NJ321WS Laptop	44990	
889	Asus	TUF A15 FA577RM-HQ032WS Laptop	110000	
890	Asus	ROG Zephyrus G14 2023 GA402XV-N2034WS Gaming L...	189990	
891	Asus	TUF Gaming F15 2023 FX507VU-LP083WS Gaming Laptop	129990	
892	Asus	TUF Gaming A15 2023 FA577XU-LP041WS Gaming Laptop	131990	

	spec_rating	processor	\
0	73.000000	5th Gen AMD Ryzen 5 5600H	
1	60.000000	12th Gen Intel Core i3 1215U	

```

2      69.323529  11th Gen Intel Core i3 1115G4
3      66.000000  12th Gen Intel Core i5 1240P
4      69.323529                               Apple M1
..      ...
888     69.323529  13th Gen Intel Core i3 1315U
889     71.000000      6th Gen AMD Ryzen 7 6800H
890     89.000000      7th Gen AMD Ryzen 9 7940HS
891     73.000000  13th Gen Intel Core i7 13700H
892     84.000000      7th Gen AMD Ryzen 9 7940HS

```

```

                                CPU  Ram  Ram_type  ROM  ROM_type  \
0      Hexa Core, 12 Threads      8    DDR4    512      SSD
1      Hexa Core (2P + 4E), 8 Threads  8    DDR4    512      SSD
2      Dual Core, 4 Threads      8    DDR4    512      SSD
3      12 Cores (4P + 8E), 16 Threads 16  LPDDR5    512      SSD
4      Octa Core (4P + 4E)      8    DDR4    256      SSD
..      ...
888     Hexa Core (2P + 4E), 8 Threads  8    DDR4    512      SSD
889      Octa Core, 16 Threads     16    DDR      1      SSD
890      Octa Core, 16 Threads     32   DDR5      1      SSD
891     14 Cores (6P + 8E), 20 Threads 16   DDR4    512      SSD
892      Octa Core, 16 Threads     16   DDR4      1      SSD

```

```

                                GPU  display_size  resolution_width  \
0      4GB AMD Radeon RX 6500M      15.6      1920
1      Intel UHD Graphics          15.6      1920
2      Intel Iris Xe Graphics      14.0      1920
3      Intel Integrated Iris Xe      14.0      2240
4      Apple M1 Integrated Graphics  13.3      2560
..      ...
888     Integrated Intel UHD Graphics  15.6      1920
889      6GB NVIDIA GeForce RTX 3060  15.6      2560
890      8GB NVIDIA GeForce RTX 4060  14.0      2560
891      6GB NVIDIA GeForce RTX 4050  15.6      1920
892      6GB NVIDIA GeForce RTX 4050  15.6      1920

```

```

                                resolution_height  OS  warranty
0      1080  Windows 11 OS      1
1      1080  Windows 11 OS      1
2      1080  Windows 11 OS      1
3      1400  Windows 11 OS      1
.      ...      ..      .

```

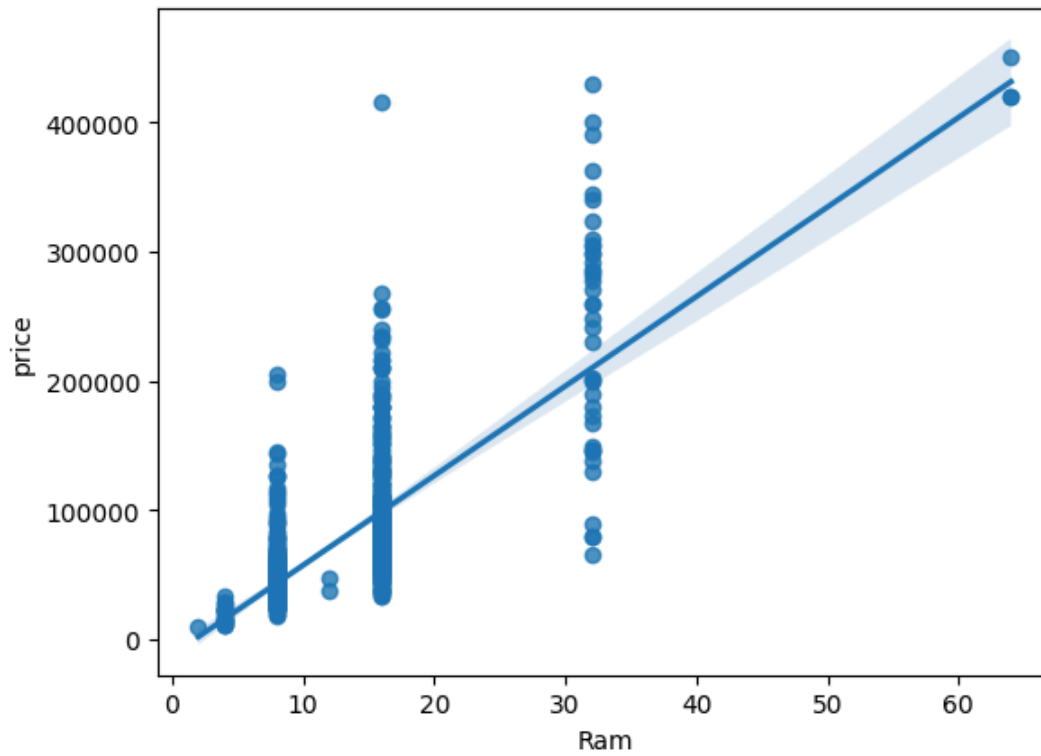
```

#conver Ram, ROM to int
df['Ram'] = df['Ram'].astype('int')
df['ROM'] = df['ROM'].astype('int')


# plots
sns.regplot(x='Ram', y='price', data=df)

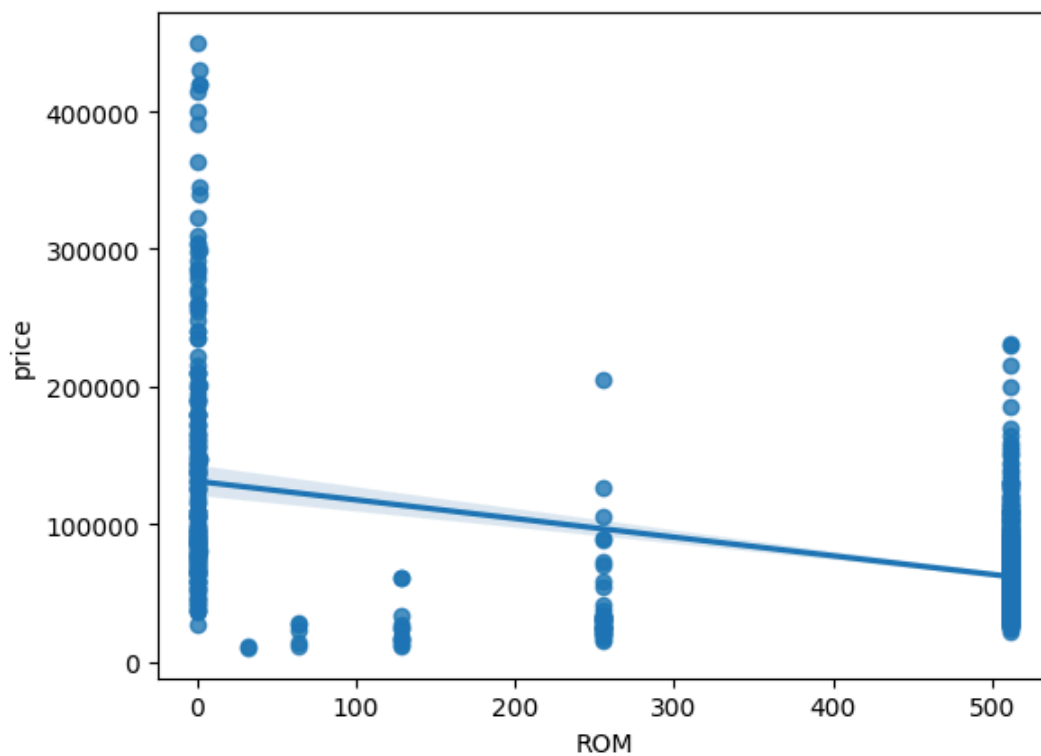
```

 <Axes: xlabel='Ram', ylabel='price'>



```
sns.regplot(x='ROM', y='price', data=df)
```

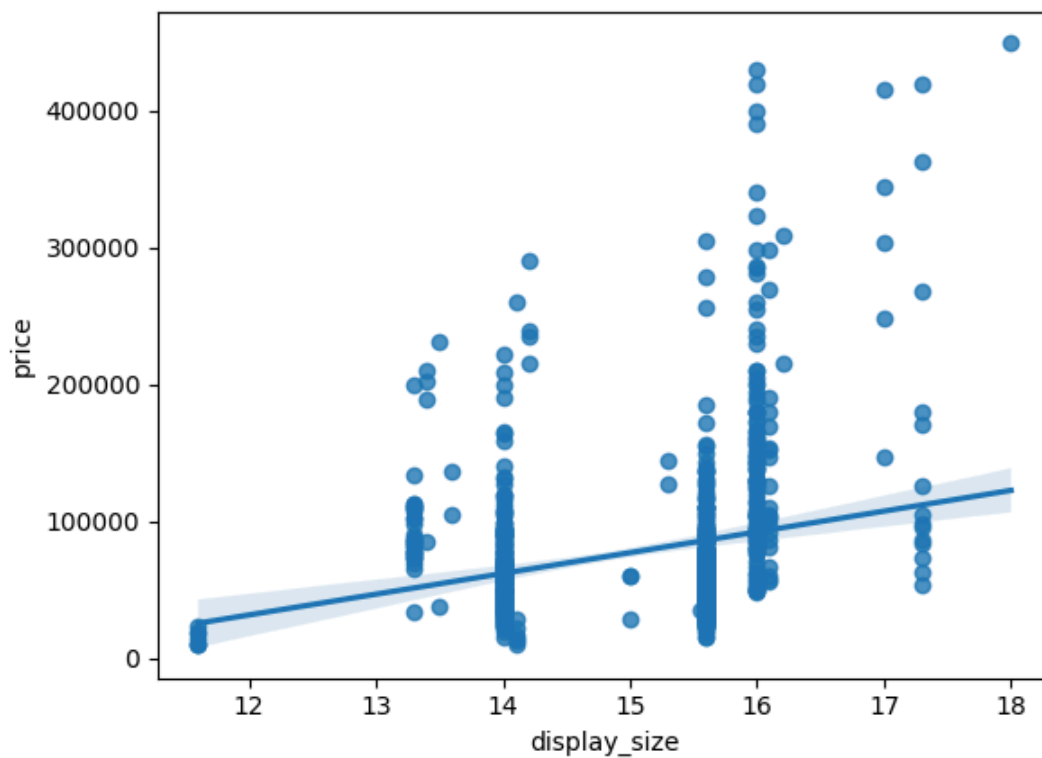
 <Axes: xlabel='ROM', ylabel='price'>



Double-click (or enter) to edit

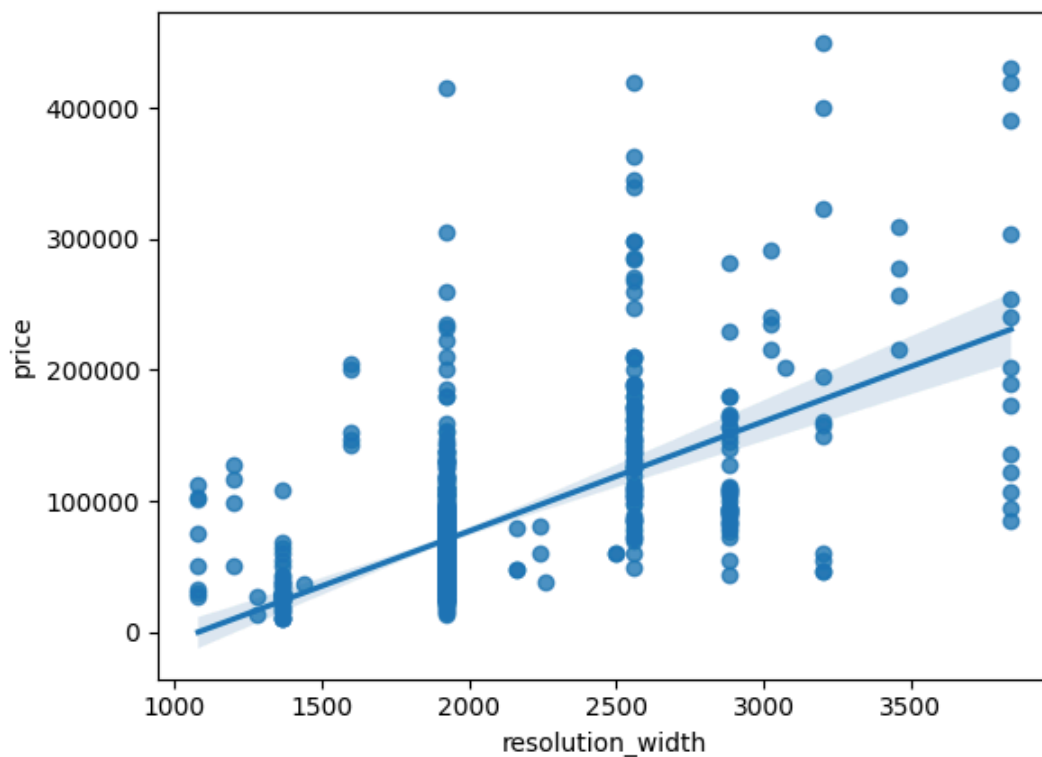
```
sns.regplot(x='display_size', y='price', data=df)
```

```
<Axes: xlabel='display_size', ylabel='price'>
```




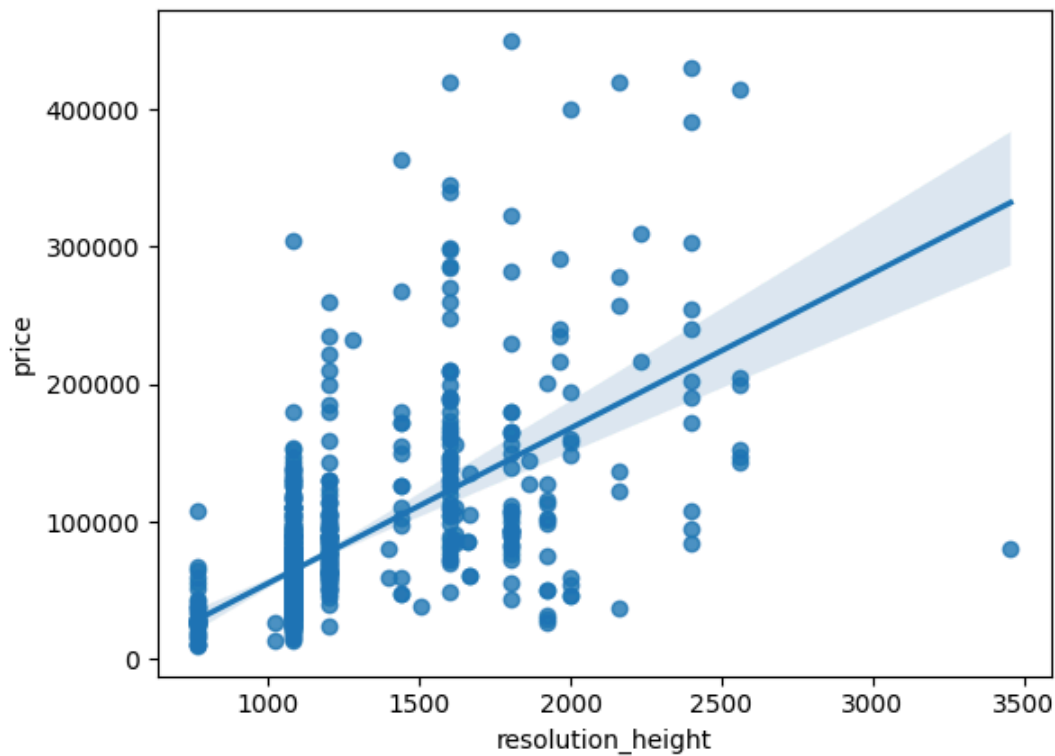
```
sns.regplot(x='resolution_width', y='price', data=df)
```

```
<Axes: xlabel='resolution_width', ylabel='price'>
```




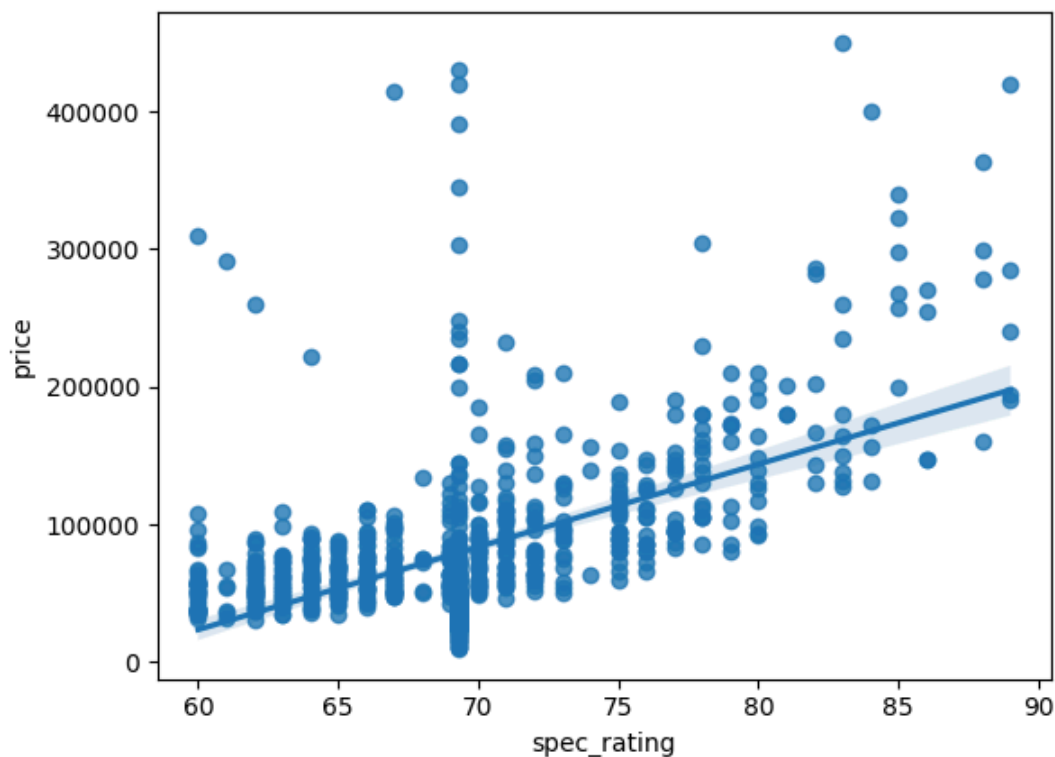
```
sns.regplot(x='resolution_height', y='price', data=df)
```


 <Axes: xlabel='resolution_height', ylabel='price'>



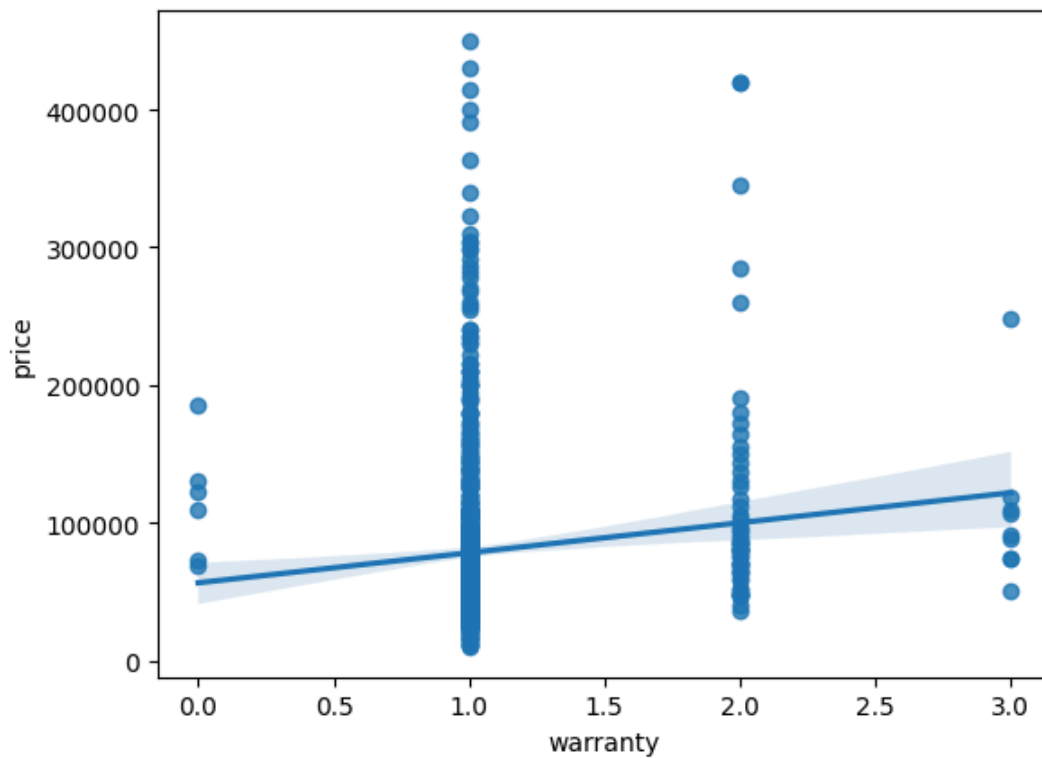
```
sns.regplot(x='spec_rating', y='price', data=df)
```

 <Axes: xlabel='spec_rating', ylabel='price'>



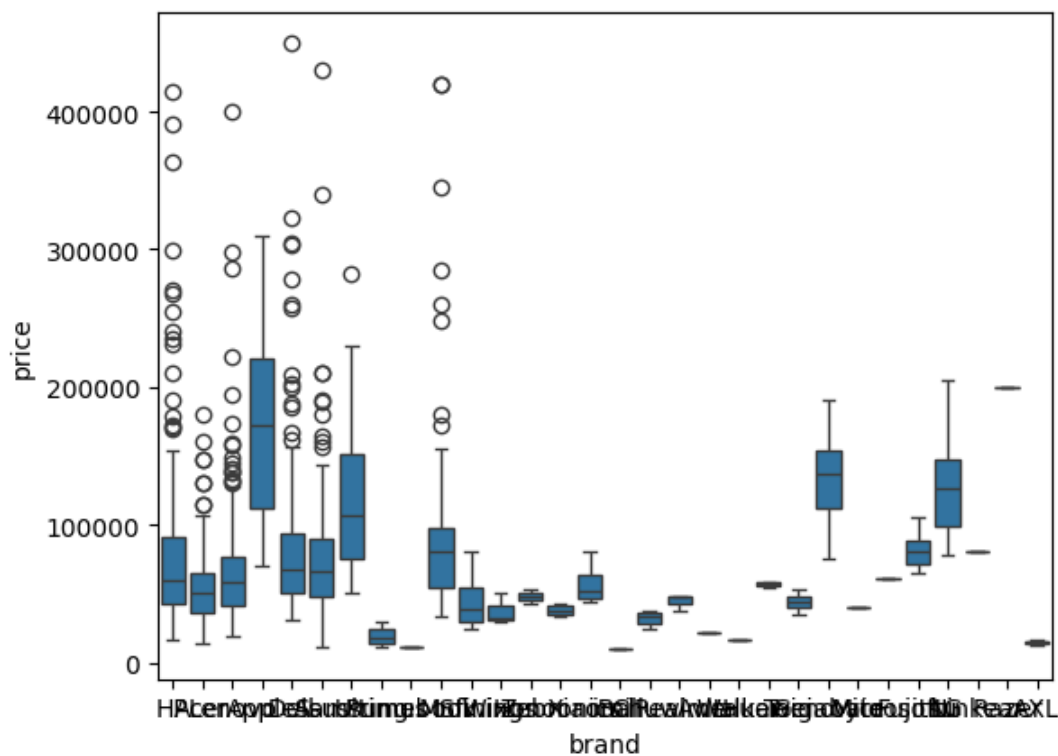
```
sns.regplot(x='warranty', y='price', data=df)
```

```
<Axes: xlabel='warranty', ylabel='price'>
```




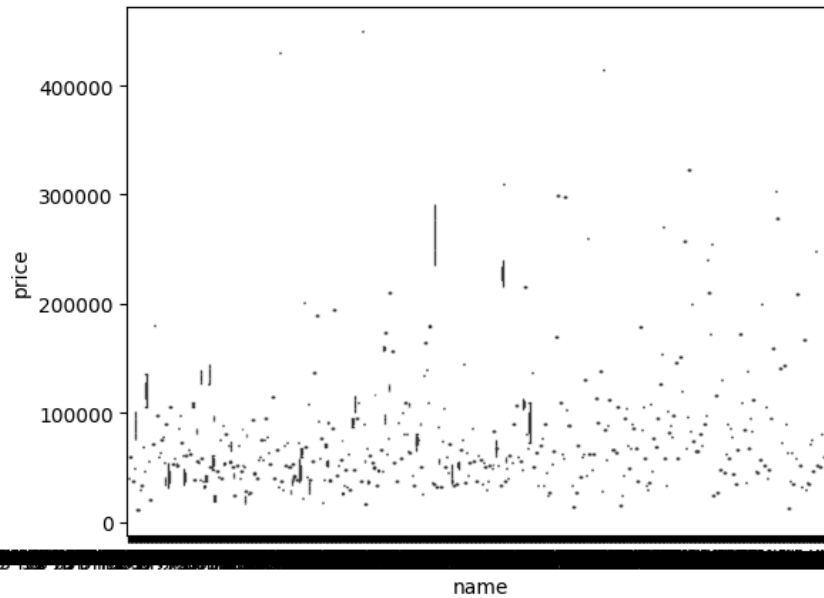
```
#boxplots
sns.boxplot(x='brand', y='price', data=df)
```

```
<Axes: xlabel='brand', ylabel='price'>
```



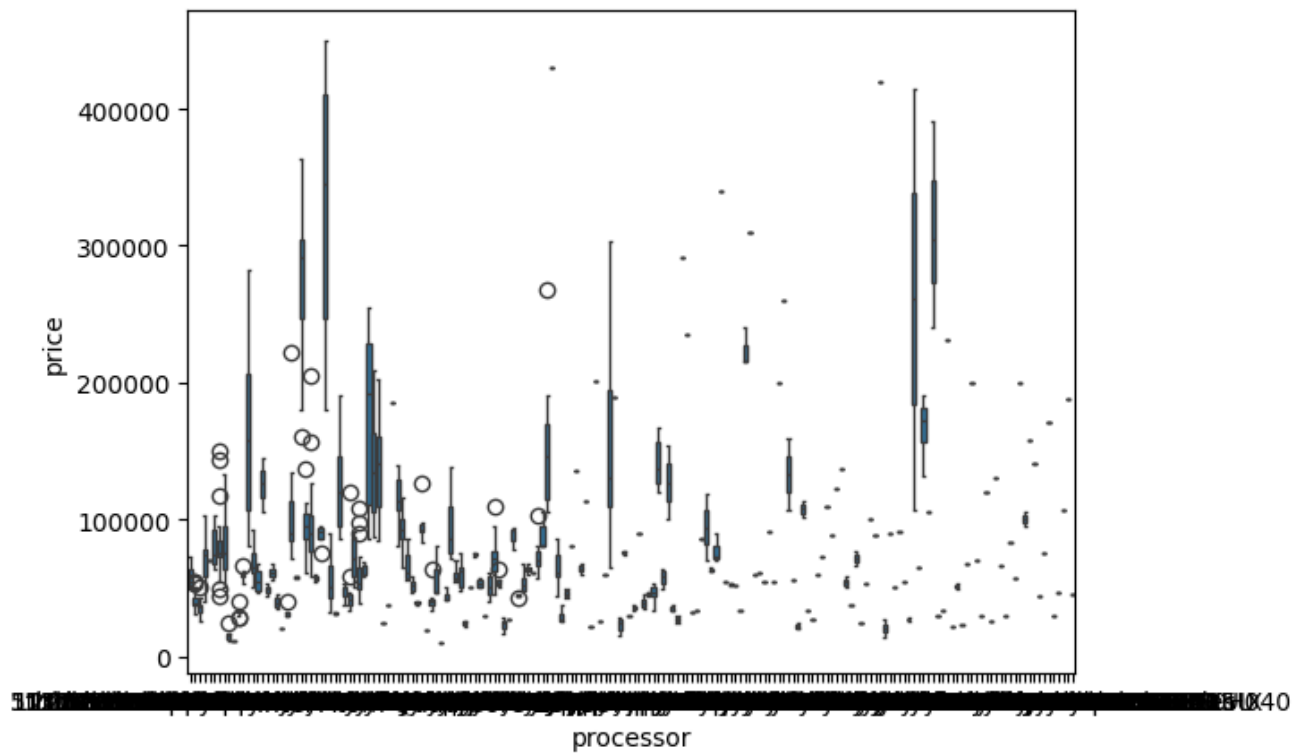
```
sns.boxplot(x='name', y='price', data=df)
```

 <Axes: xlabel='name', ylabel='price'>



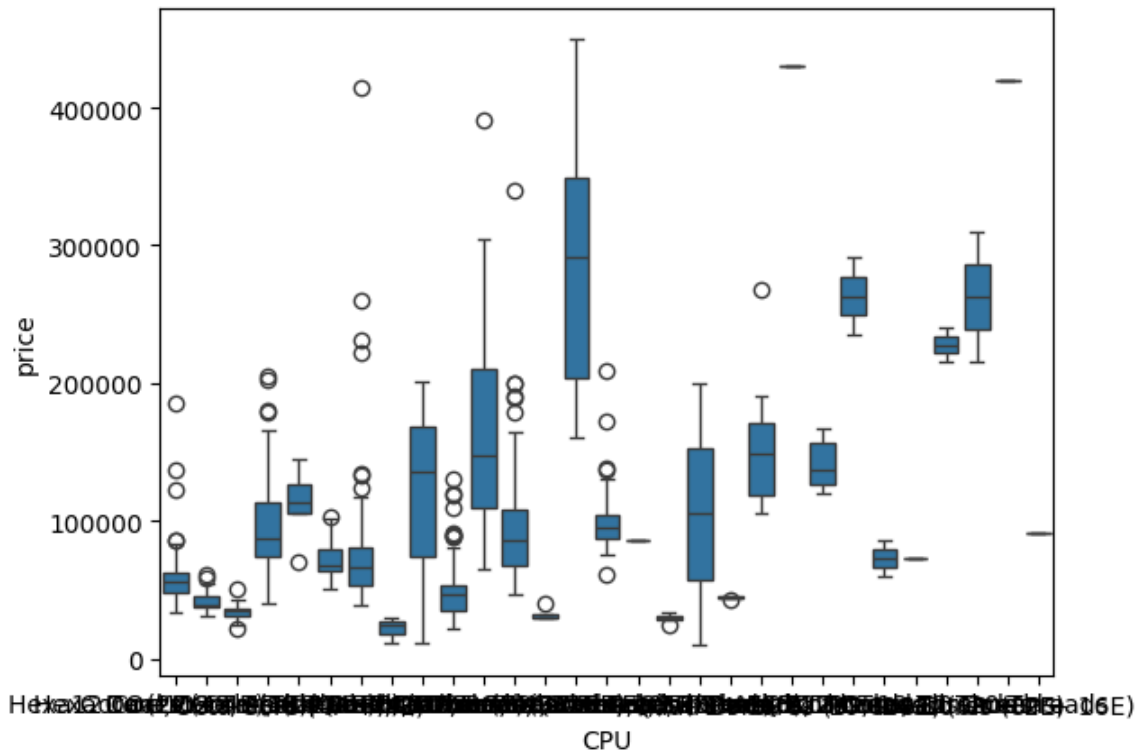
```
sns.boxplot(x='processor', y='price', data=df)
```

 <Axes: xlabel='processor', ylabel='price'>



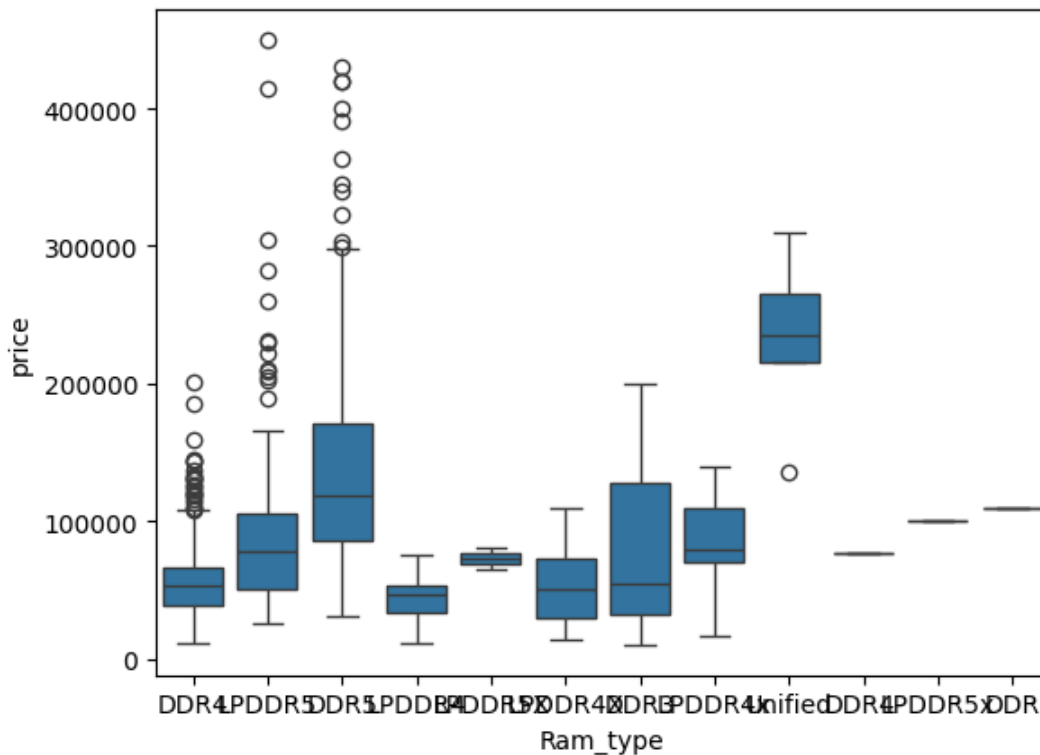
```
sns.boxplot(x='CPU', y='price', data=df)
```

```
>>> sns.boxplot(x='CPU', y='price', data=df)
```




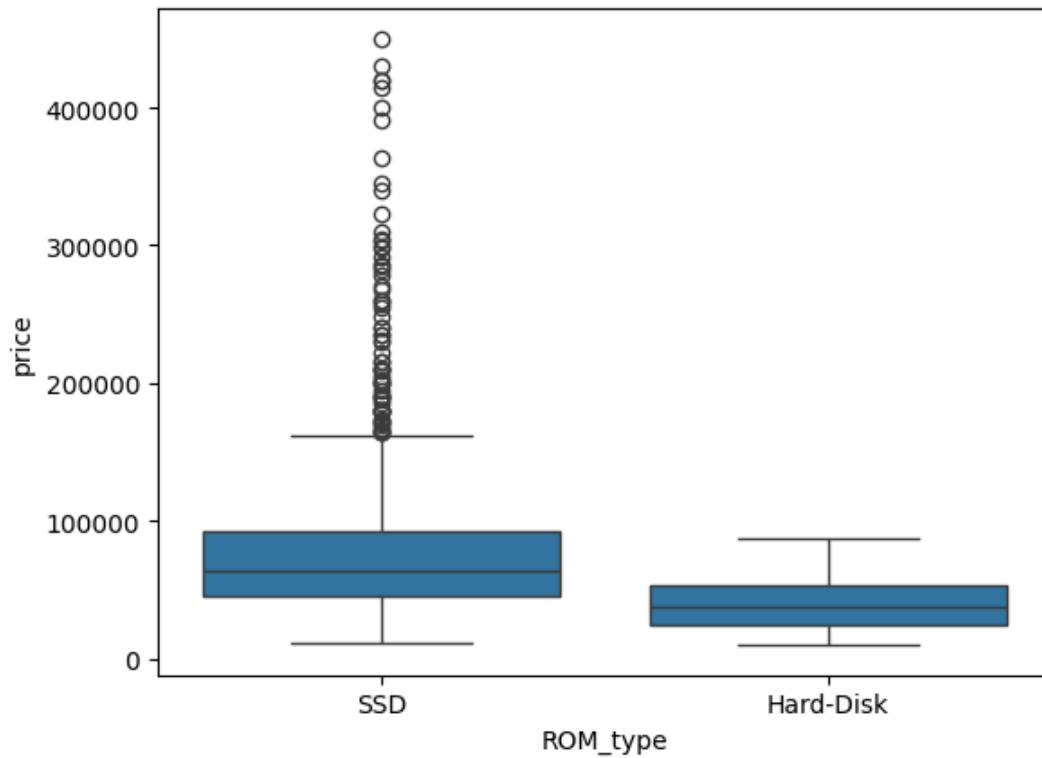
```
sns.boxplot(x='Ram_type', y='price', data=df)
```

```
>>> sns.boxplot(x='Ram_type', y='price', data=df)
```



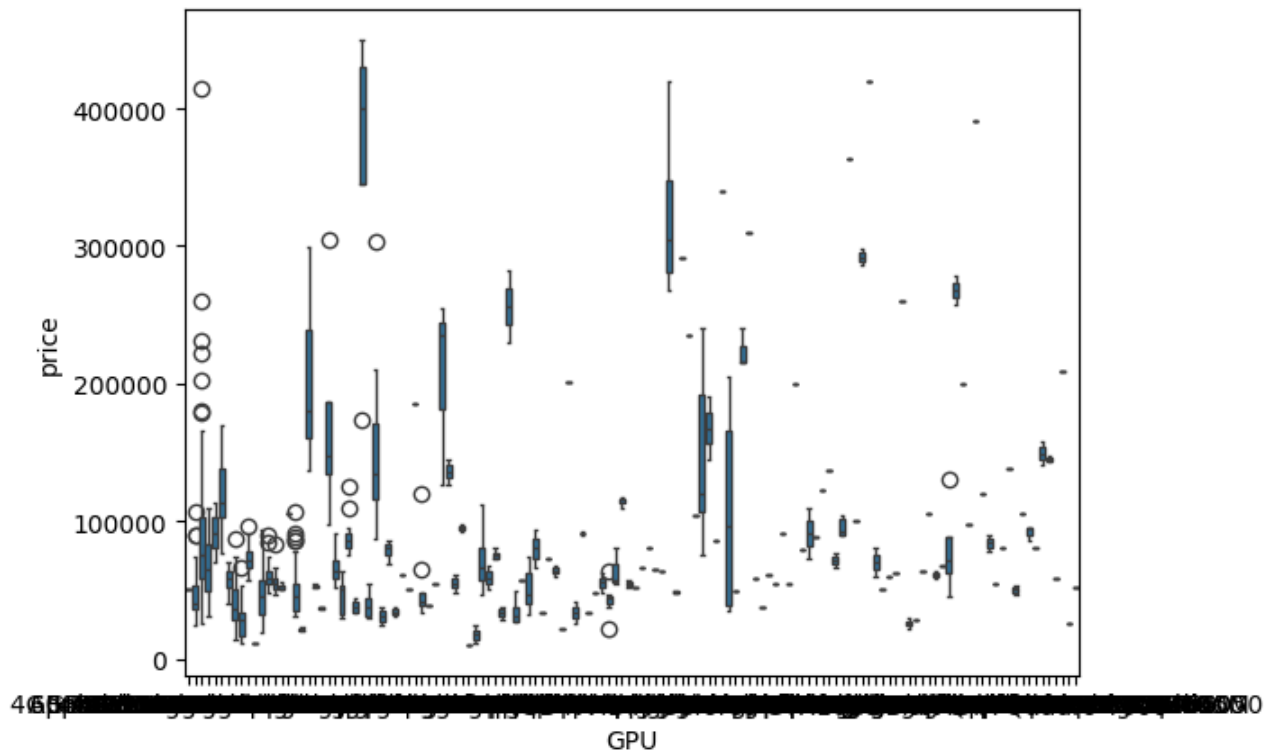
```
sns.boxplot(x='ROM_type', y='price', data=df)
```

 <Axes: xlabel='ROM_type', ylabel='price'>



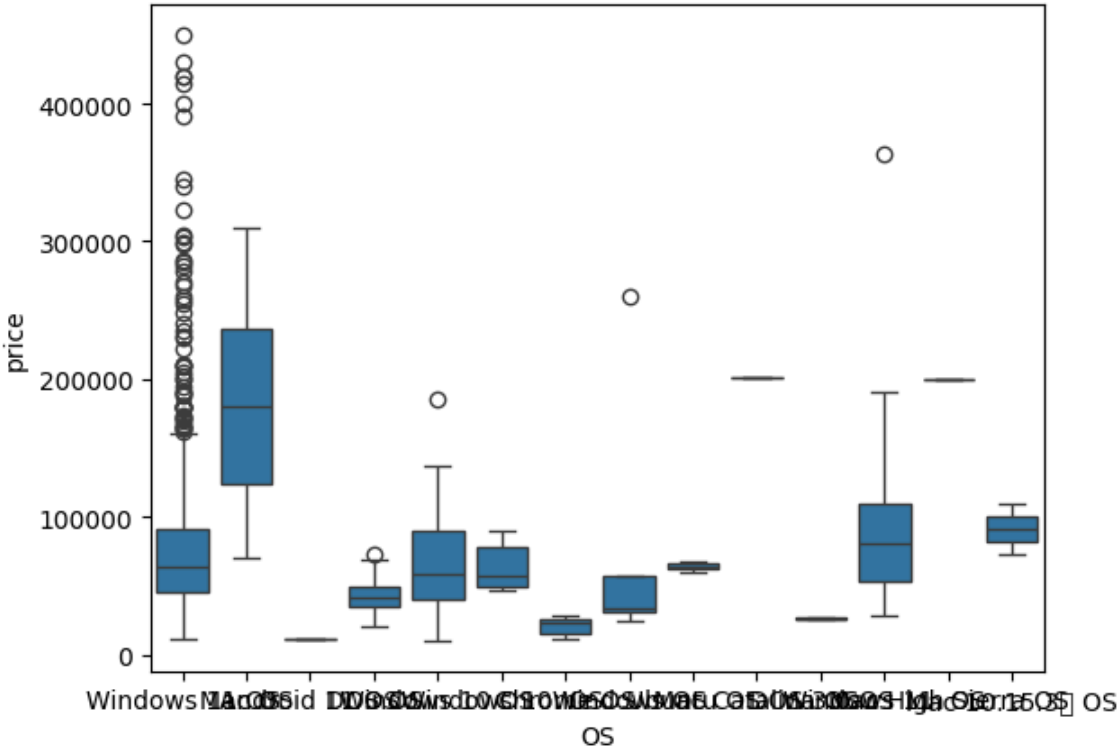
```
sns.boxplot(x='GPU', y='price', data=df)
```

 <Axes: xlabel='GPU', ylabel='price'>



```
sns.boxplot(x='OS', y='price', data=df)
```

```
<Axes: xlabel='OS', ylabel='price'>
```

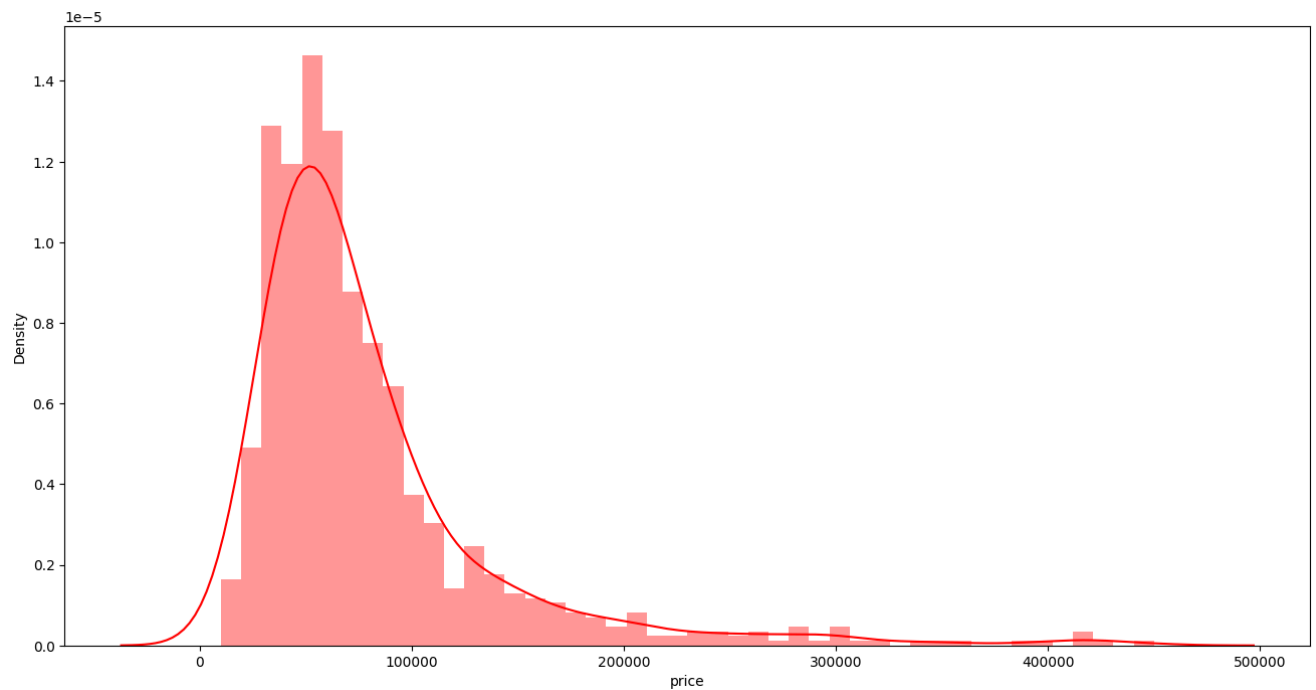


```
df.info()
```

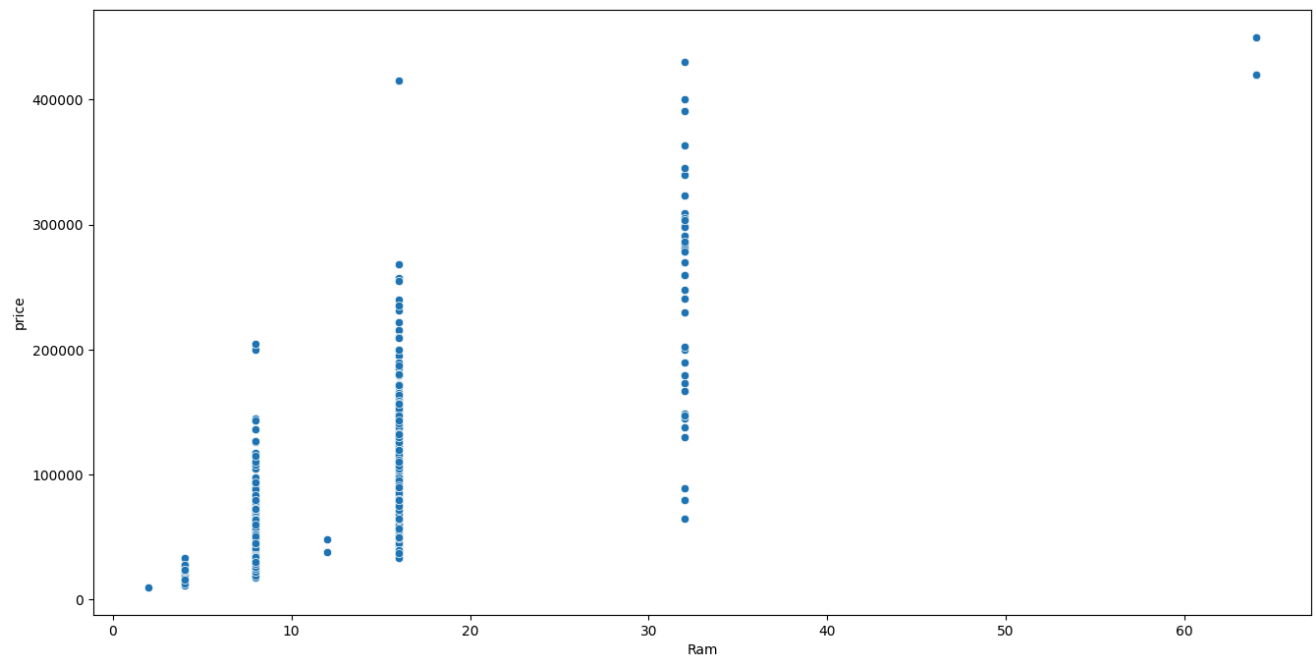
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 893 entries, 0 to 892
Data columns (total 16 columns):
#   Column                Non-Null Count  Dtype
---  -
0   brand                 893 non-null    object
1   name                  893 non-null    object
2   price                 893 non-null    int64
3   spec_rating           893 non-null    float64
4   processor             893 non-null    object
5   CPU                   893 non-null    object
6   Ram                   893 non-null    int64
7   Ram_type              893 non-null    object
8   ROM                   893 non-null    int64
9   ROM_type              893 non-null    object
10  GPU                   893 non-null    object
11  display_size          893 non-null    float64
12  resolution_width      893 non-null    int64
13  resolution_height     893 non-null    int64
14  OS                     893 non-null    object
15  warranty              893 non-null    int64
dtypes: float64(2), int64(6), object(8)
memory usage: 111.8+ KB
```

```
#Exploratory Data Analysis
```

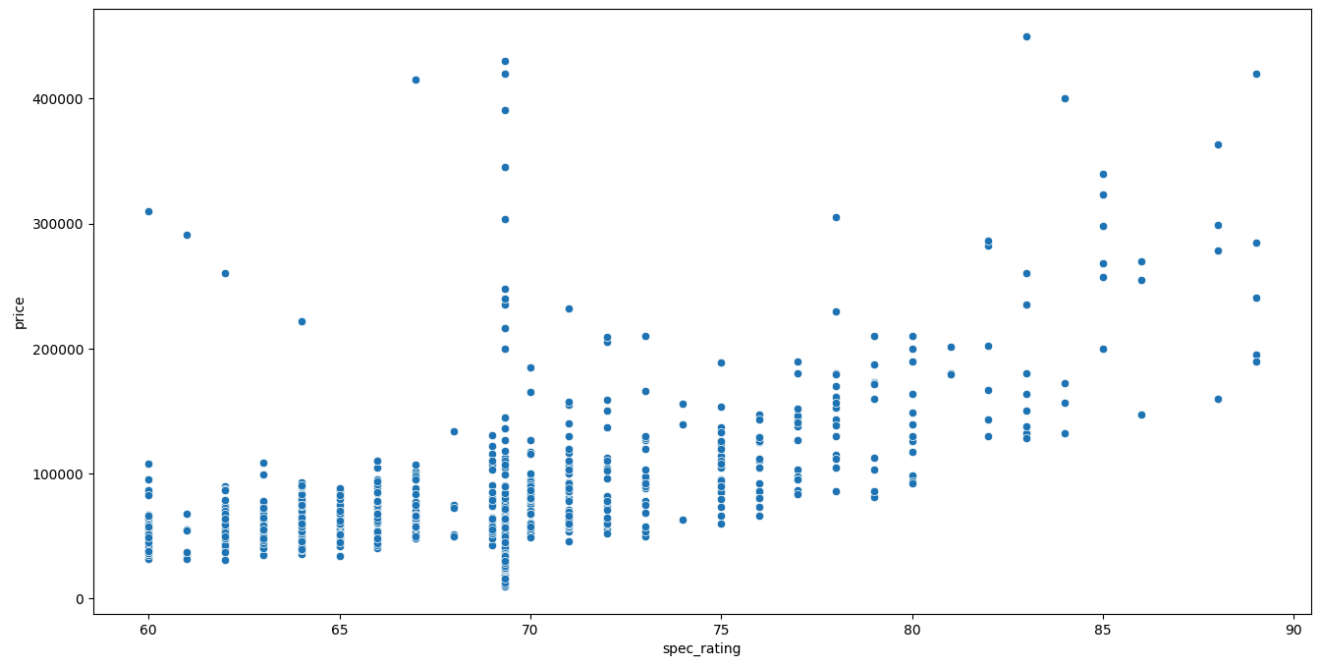
```
#Price Distribution plot
plt.figure(figsize=(16,8))
sns.distplot(df['price'],color='red')
plt.show()
```



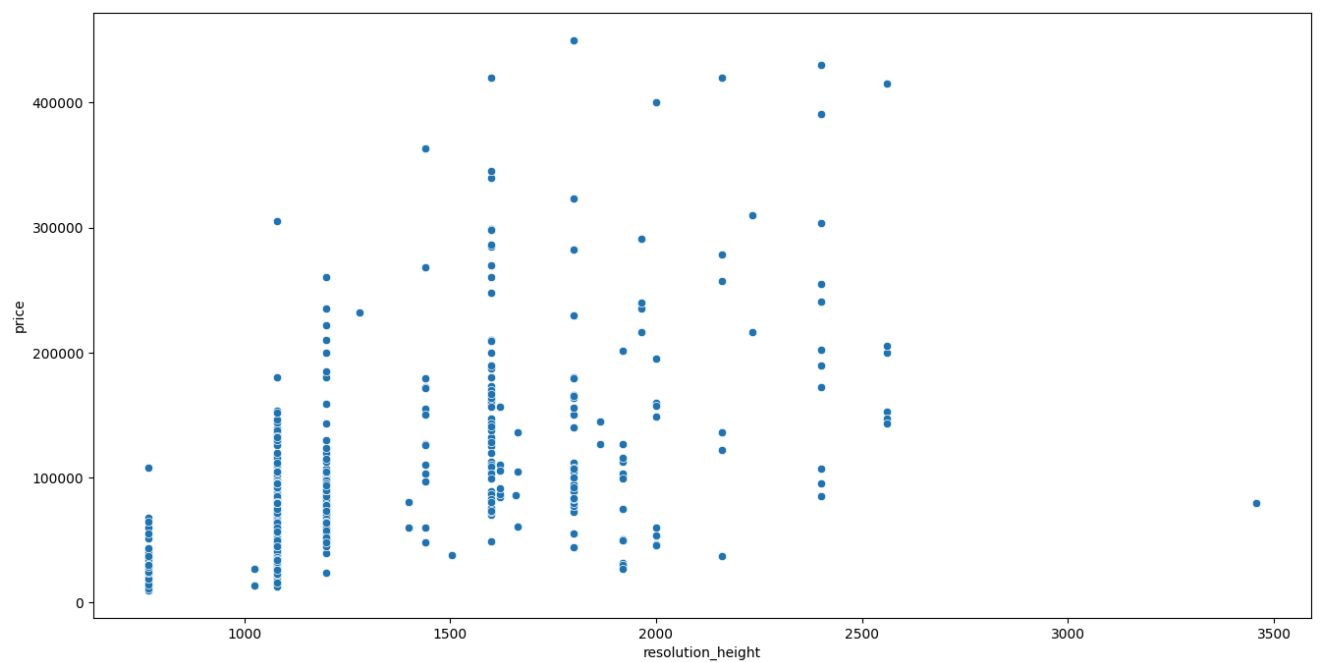
```
#Scatterplots
plt.figure(figsize=(16,8))
sns.scatterplot(x = df['Ram'],y = df['price'])
plt.show()
```



```
plt.figure(figsize=(16,8))
sns.scatterplot(x = df['spec_rating'],y = df['price'])
plt.show()
```



```
plt.figure(figsize=(16,8))  
sns.scatterplot(x = df['resolution_height'],y = df['price'])  
plt.show()
```




```
#Transformation-converting categorical data into numerical data
```

```
labelencoder = LabelEncoder()
df.brand = labelencoder.fit_transform(df.brand)
df.name = labelencoder.fit_transform(df.name)
df.processor = labelencoder.fit_transform(df.processor)
df.CPU = labelencoder.fit_transform(df.CPU)
df.Ram_type = labelencoder.fit_transform(df.Ram_type)
df.ROM_type = labelencoder.fit_transform(df.ROM_type)
df.GPU = labelencoder.fit_transform(df.GPU)
df.OS = labelencoder.fit_transform(df.OS)
print(df)
```

	brand	name	price	spec_rating	processor	CPU	Ram	Ram_type	ROM	\
0	9	647	49900	73.000000	105	19	8	2	512	
1	9	38	39900	60.000000	34	18	8	2	512	
2	1	440	26990	69.323529	15	17	8	2	512	
3	14	784	59729	66.000000	39	6	16	8	512	
4	2	378	69990	69.323529	160	21	8	2	256	
...	
888	3	697	44990	69.323529	80	18	8	2	512	
889	3	528	110000	71.000000	122	24	16	0	1	
890	3	506	189990	89.000000	143	24	32	4	1	
891	3	534	129990	73.000000	73	8	16	2	512	
892	3	530	131990	84.000000	143	24	16	2	1	

	ROM_type	GPU	display_size	resolution_width	resolution_height	OS	\
0	1	19	15.6	1920	1080	12	
1	1	126	15.6	1920	1080	12	
2	1	123	14.0	1920	1080	12	
3	1	111	14.0	2240	1400	12	
4	1	91	13.3	2560	1600	7	
...	
888	1	95	15.6	1920	1080	12	
889	1	52	15.6	2560	1440	11	
890	1	63	14.0	2560	1600	12	
891	1	53	15.6	1920	1080	12	
892	1	53	15.6	1920	1080	12	

	warranty
0	1
1	1
2	1
3	1
4	1
...	...
888	1
889	1
890	1
891	1
892	1

```
[893 rows x 16 columns]
```

```
#Coefficient of correlations
```

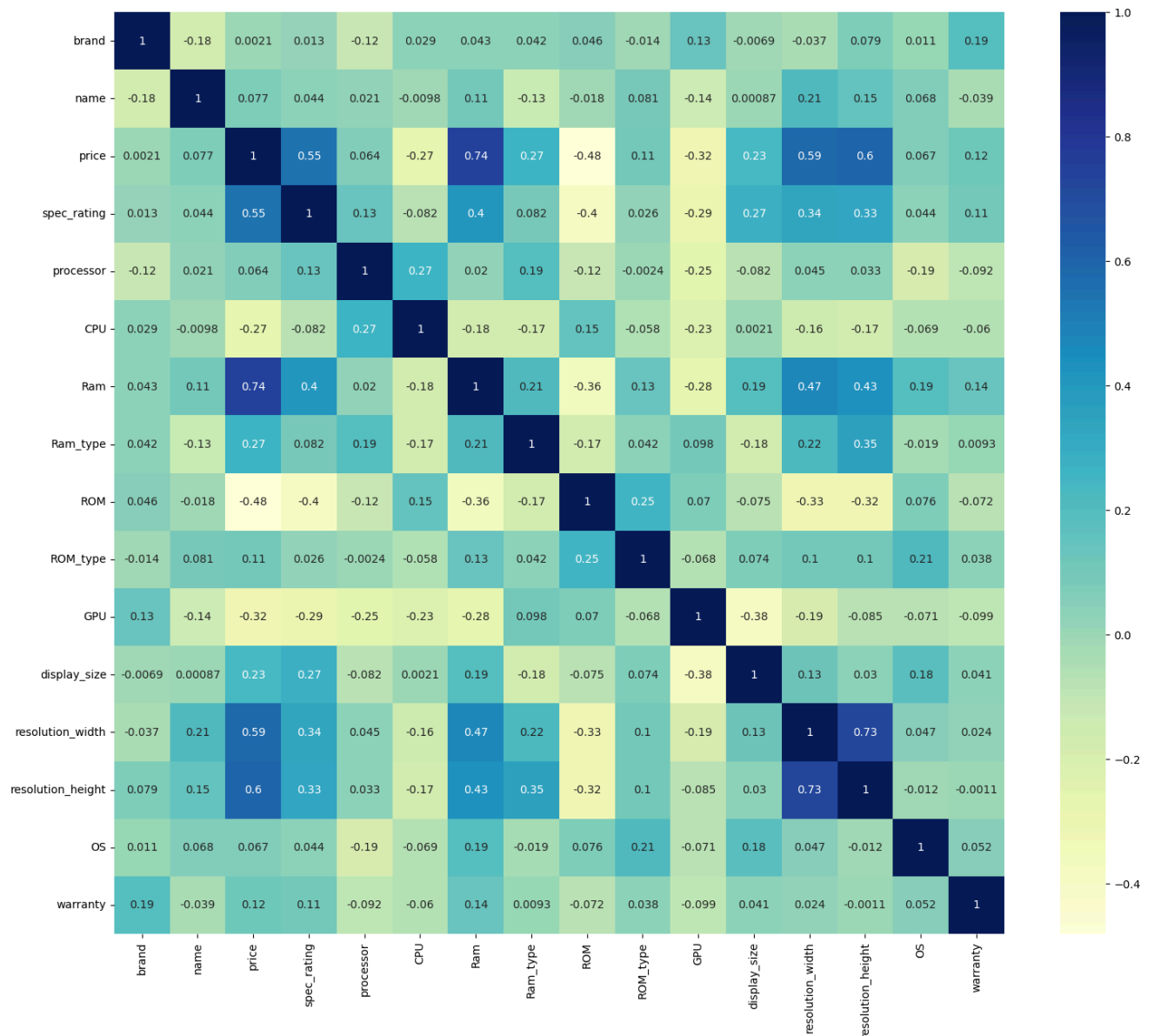
```
target_correlations = df.corr()['price'].apply(abs).sort_values()
target_correlations
```

brand	0.002070
processor	0.064416
OS	0.066910
name	0.077481
ROM_type	0.105690

```
warranty      0.117101
display_size  0.233815
Ram_type      0.267315
CPU           0.273260
GPU           0.324344
ROM           0.481178
spec_rating   0.546391
resolution_width 0.586042
resolution_height 0.604748
Ram           0.736924
price         1.000000
Name: price, dtype: float64
```

```
plt.figure(figsize=(18, 15))
sns.heatmap(df.corr(), annot=True, cmap='YlGnBu')
```

↔ <Axes: >



```
#select the features which impact the target variable(price)
selected_features =target_correlations[-12:].index
selected_features=list(selected_features)
selected_features
```

```
['ROM_type',
 'warranty',
 'display_size',
 'Ram_type',
 'CPU',
 'GPU',
 'ROM',
 'spec_rating',
 'resolution_width',
 'resolution_height',
 'Ram',
 'price']
```

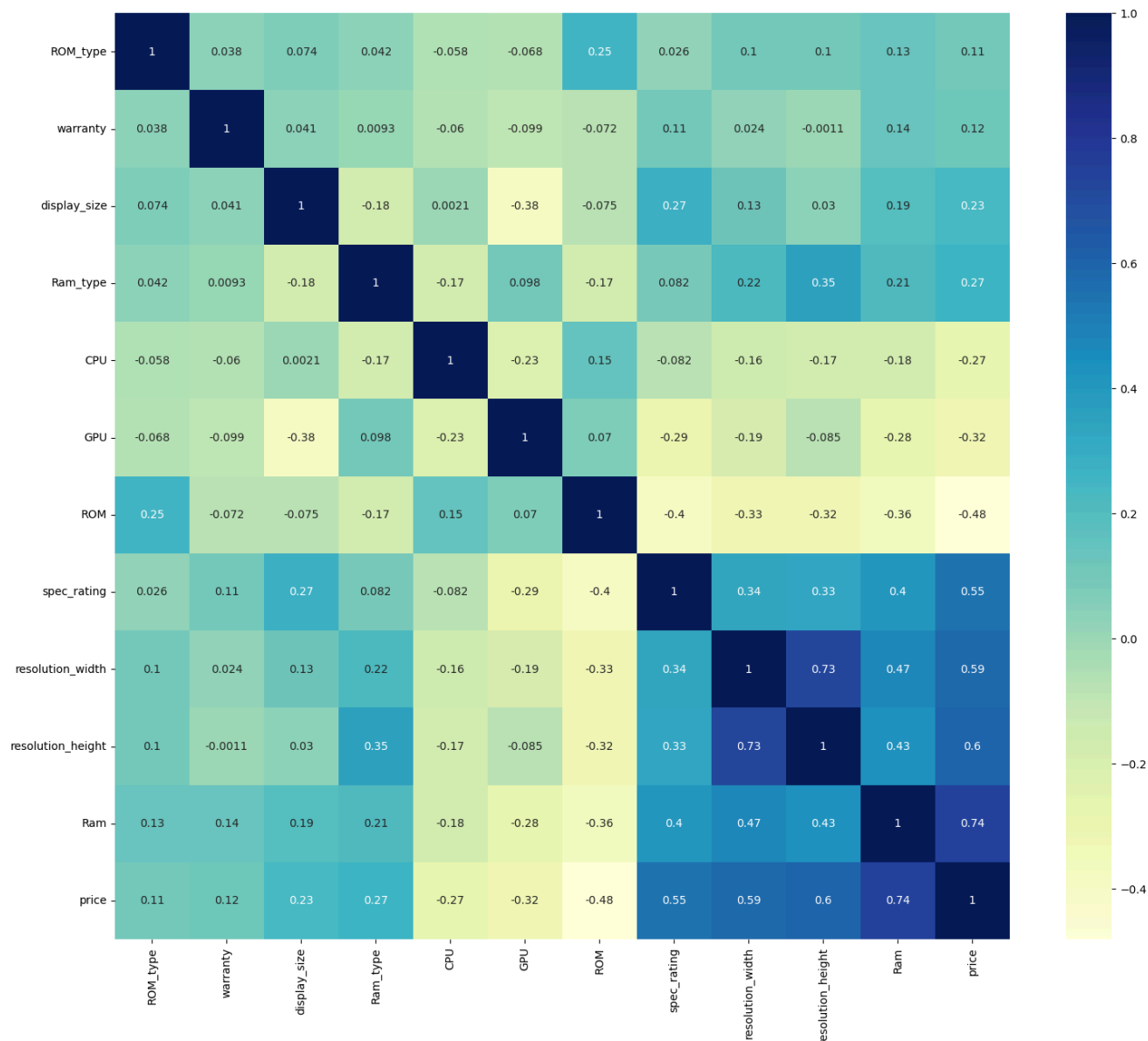
```
#check the corr of the selected features
limited_df =df[selected_features]
limited_df
```

```
ROM_type warranty display_size Ram_type CPU GPU ROM spec_rating resolution_w.
```

0	1	1	15.6	2	19	19	512	73.000000
1	1	1	15.6	2	18	126	512	60.000000
2	1	1	14.0	2	17	123	512	69.323529
3	1	1	14.0	8	6	111	512	66.000000
4	1	1	13.3	2	21	91	256	69.323529
...
888	1	1	15.6	2	18	95	512	69.323529
889	1	1	15.6	0	24	52	1	71.000000
890	1	1	14.0	4	24	63	1	89.000000
891	1	1	15.6	2	8	53	512	73.000000
892	1	1	15.6	2	24	53	1	84.000000

893 rows x 12 columns

```
plt.figure(figsize=(18, 15))
sns.heatmap(limited_df.corr(), annot=True, cmap='YlGnBu')
```

 <Axes: >


The more blue areas founded, the more correlated to the price

```
#Scaling(normalization)
```

```
X, y =limited_df.drop('price', axis=1), limited_df['price']
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.1)
```

```
scaler = StandardScaler()
```

```
X_train_scaled = scaler.fit_transform(X_train)
```

```
X_test_scaled = scaler.transform(X_test)
```

```
#Fit and Evaluate Model
```

```
#Multiple_linear_Regression(MLR)
model = LinearRegression()
model_mlr = model.fit(X_train,y_train)
#Predict price using test
y_pred_MLR = model_mlr.predict(X_test)
```

```
#Calculating the Mean Square Error for MLR
mse_MLR = mean_squared_error(y_test,y_pred_MLR)
print('The mean square error for Multiple Linear Regression: ', mse_MLR)
```

```
➞ The mean square error for Multiple Linear Regression: 794801305.136414
```

```
#Calculating the Mean Absolute Error for MLR
mae_MLR= mean_absolute_error(y_test, y_pred_MLR)
print('The mean absolute error for Multiple Linear Regression: ', mae_MLR)
```

```
➞ The mean absolute error for Multiple Linear Regression: 19139.406643778206
```

```
#Random Forest Regressor
forest = RandomForestRegressor()

forest.fit(X_train_scaled, y_train)
forest.score(X_test_scaled, y_test)
```

```
➞ 0.7125265891251428
```

```
#prediction of laptop price using testing
y_pred_RF = forest.predict(X_test_scaled)
```

```
#Random Forest Evaluation
#Calculating the Mean Square Error for Random Forest Model
mse_RF = mean_squared_error(y_test, y_pred_RF)
print('The mean square error of price and predicted value is: ', mse_RF)
```

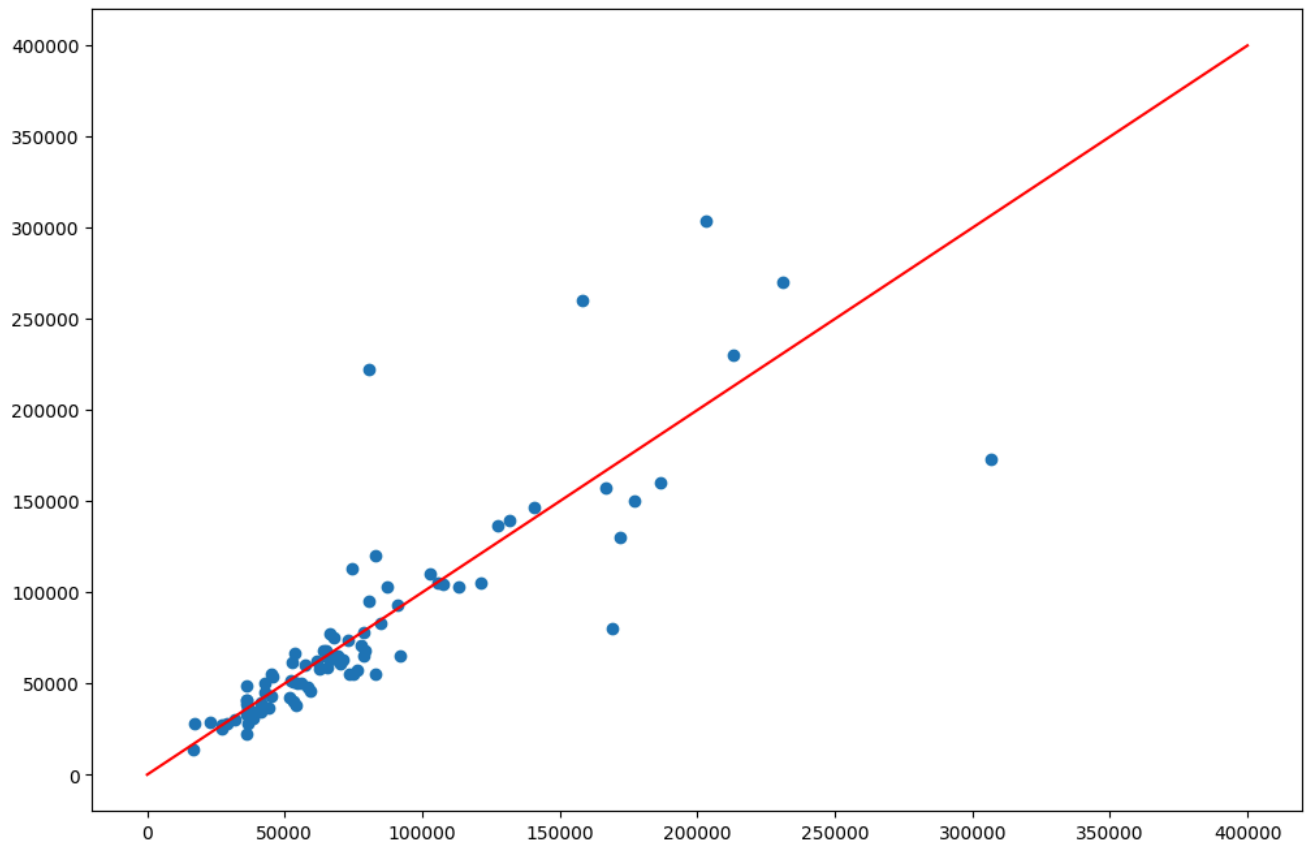
```
➞ The mean square error of price and predicted value is: 901324956.1502393
```

```
#Calculating the absolute Square Error for Random Forest Model
mae_RF= mean_absolute_error(y_test, y_pred_RF)
print('The mean absolute error of price and predicted value is: ', mae_RF)
```

```
➞ The mean absolute error of price and predicted value is: 15194.167113442421
```

```
#line and scatter plot
plt.figure(figsize=(12,8))
plt.scatter(y_pred_RF, y_test)
plt.plot(range(0, 400000), range(0, 400000), c='red')
```

↗ [matplotlib.lines.Line2D at 0x7ba3c4d9a3e0]



```
#LASSO Model
LassoModel = Lasso()
model_lm = LassoModel.fit(X_train,y_train)

y_pred_lasso = model_lm.predict(X_test)
```

```
#LASSO Evaluation
#Mean Absolute Error for LASSO Model
mae_lasso= mean_absolute_error(y_test,y_pred_lasso)
print('The mean absolute error of price and predicted value is: ', mae_lasso)
```

↗ The mean absolute error of price and predicted value is: 19140.350232765788

```
#Mean squared Error for LASSO Model
mse_lasso = mean_squared_error(y_test, y_pred_lasso)
print('The mean square error of price and predicted value is: ', mse_lasso)
```

↗ The mean square error of price and predicted value is: 794845643.8141383

```
scores = [('MLR', mae_MLR),
          ('Random Forest', mae_RF),
          ('LASSO', mae_lasso)
        ]
```

```
mae = pd.DataFrame(data = scores, columns=['Model', 'MAE Score'])  
mae
```