

ASSIGNMENT 3

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```
In [16]: import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```

2. Load the dataset into the tool.

```
In [19]: df=pd.read_csv('Housing.csv')
```

```
In [20]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   price                 545 non-null   int64
1   area                 545 non-null   int64
2   bedrooms             545 non-null   int64
3   bathrooms            545 non-null   int64
4   stories              545 non-null   int64
5   mainroad             545 non-null   object
6   guestroom           545 non-null   object
7   basement            545 non-null   object
8   hotwaterheating     545 non-null   object
9   airconditioning     545 non-null   object
10  parking              545 non-null   int64
11  furnishingstatus    545 non-null   object
dtypes: int64(6), object(6)
memory usage: 51.2+ KB
```

```
In [21]: df.head()
```

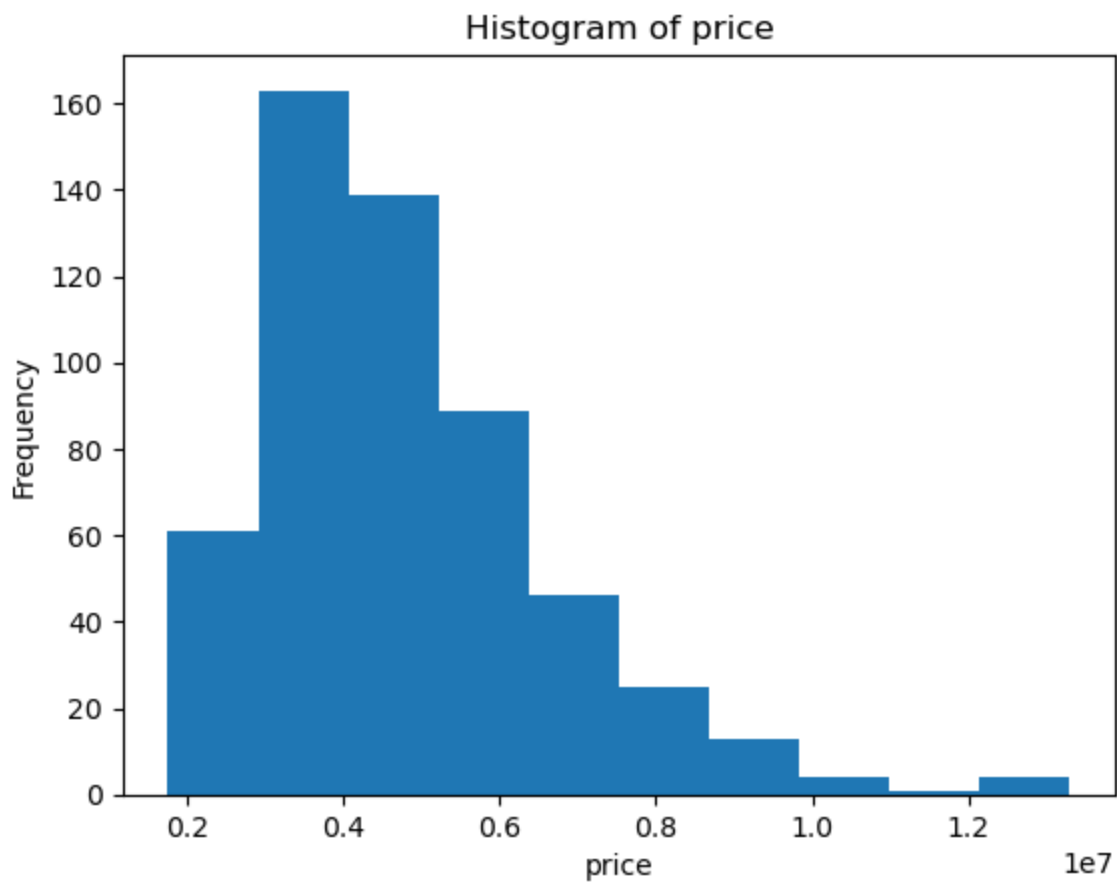
```
Out[21]:
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning
0	13300000	7420	4	2	3	yes	no	no		no
1	12250000	8960	4	4	4	yes	no	no		no
2	12250000	9960	3	2	2	yes	no	yes		no
3	12215000	7500	4	2	2	yes	no	yes		no
4	11410000	7420	4	1	2	yes	yes	yes		no

3. Perform Below Visualizations.

1. Univariate Analysis

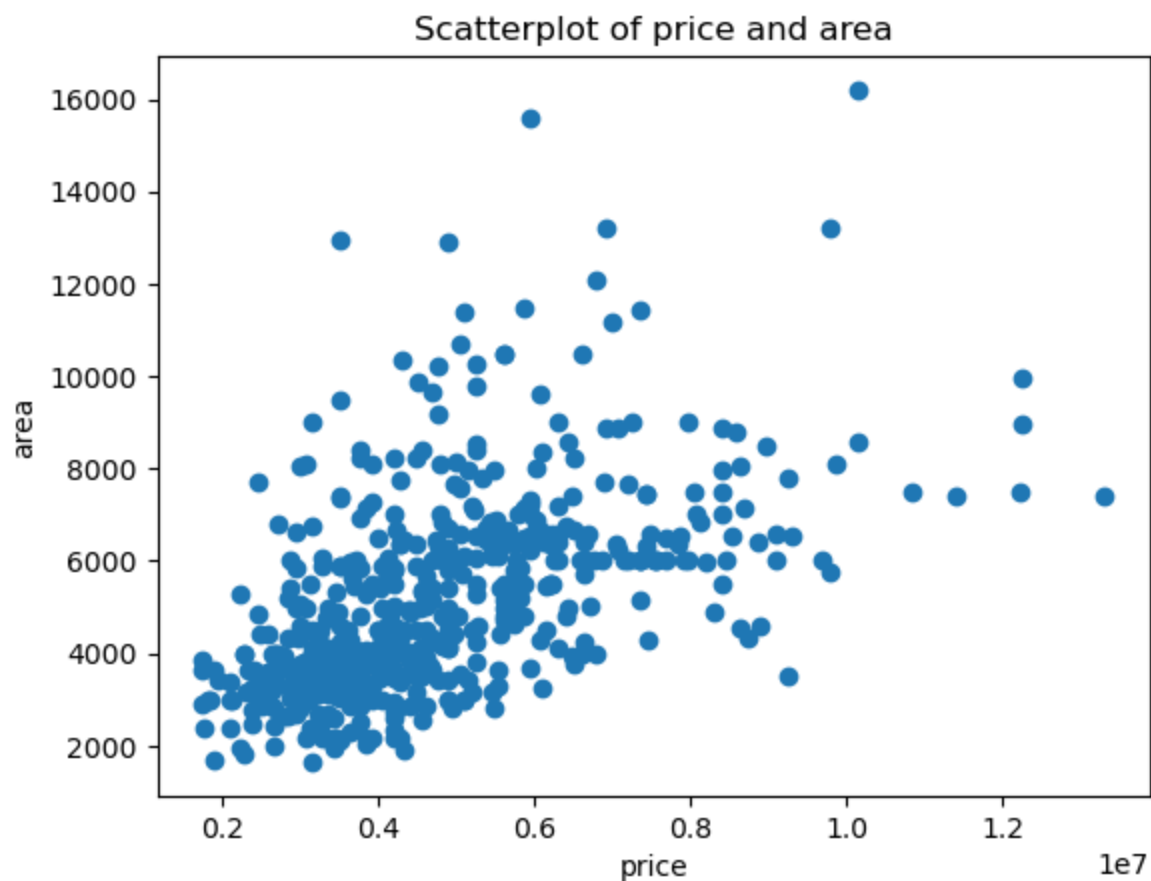
```
In [22]: #3. Perform Below Visualizations.  
#1. Univariate Analysis  
# Histogram  
plt.hist(df['price'], bins=10)  
plt.title('Histogram of price')  
plt.xlabel('price')  
plt.ylabel('Frequency')  
plt.show()
```



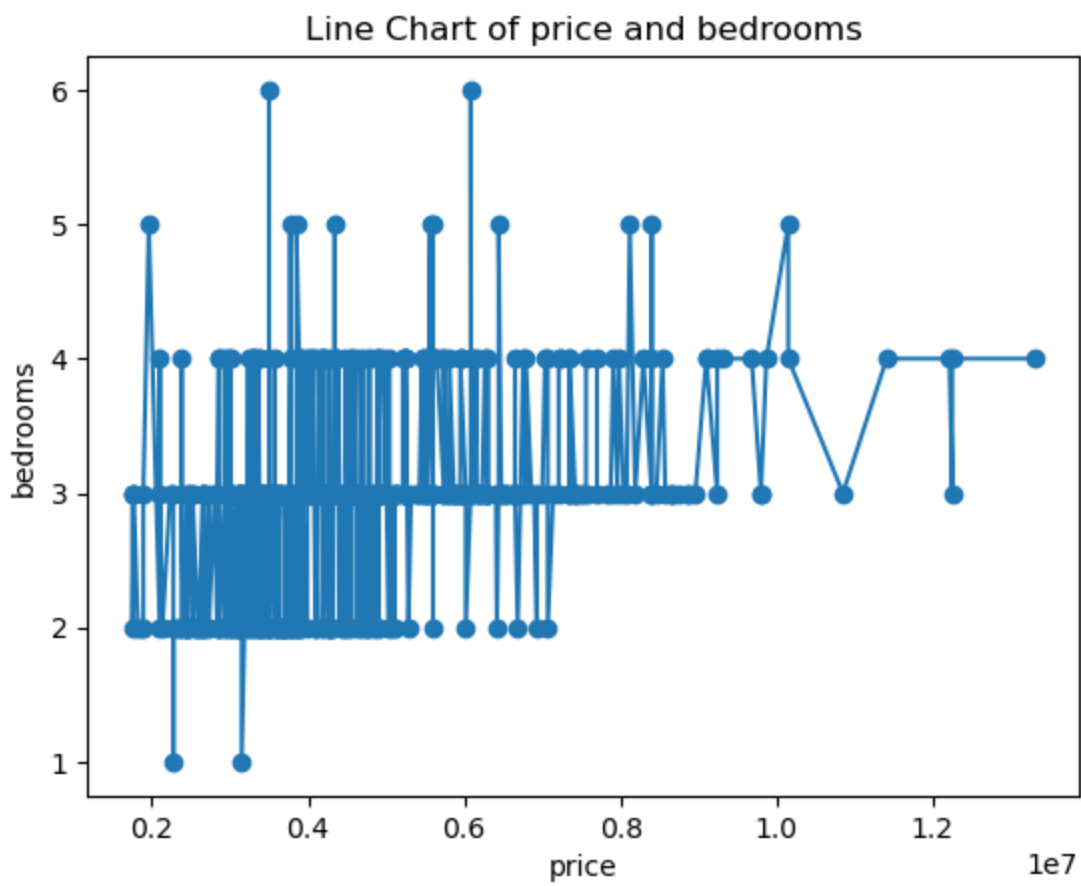
```
In [23]: # Boxplot  
plt.boxplot(df['price'])  
plt.title('Boxplot of price')  
plt.xlabel('price')  
plt.ylabel('Value')  
plt.show()
```



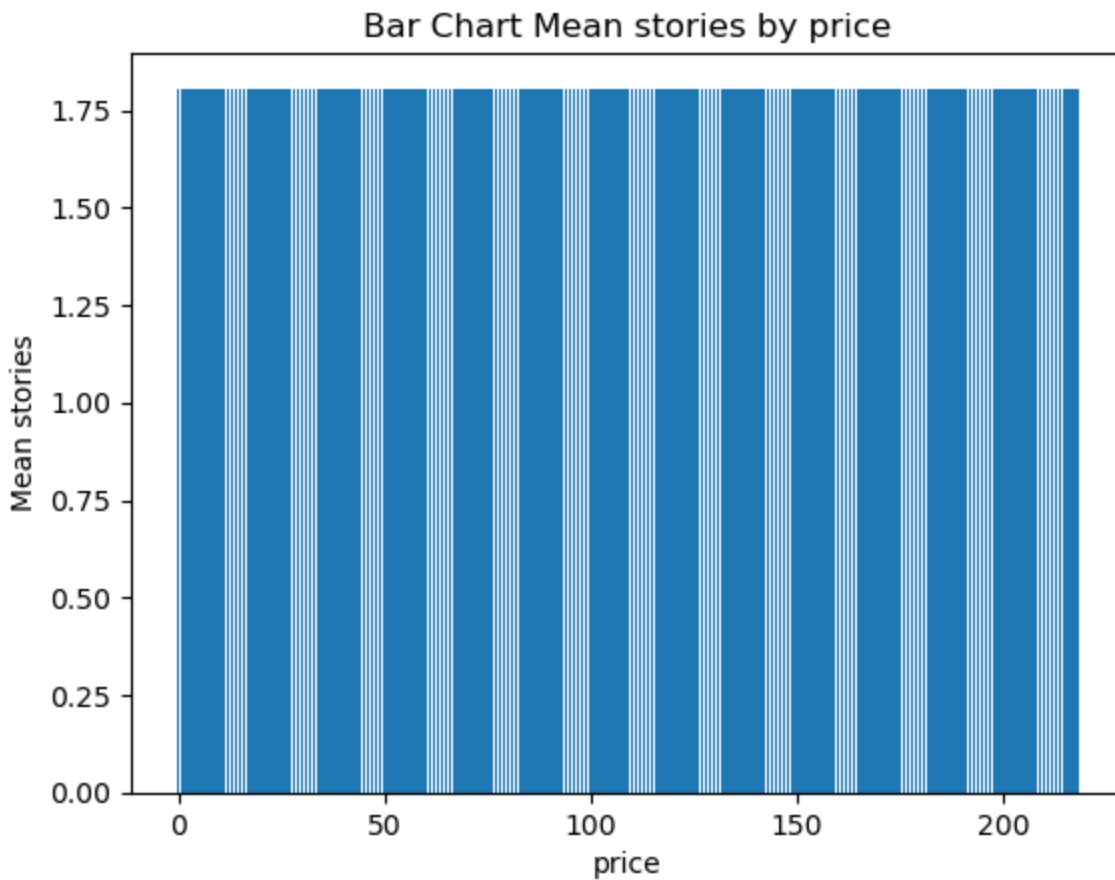
```
In [25]: # Bivariate analysis
# Scatterplot
plt.scatter(df['price'], df['area'])
plt.title('Scatterplot of price and area')
plt.xlabel('price')
plt.ylabel('area')
plt.show()
```



```
In [27]: # Line chart
plt.plot(df['price'], df['bedrooms'], 'o-')
plt.title('Line Chart of price and bedrooms')
plt.xlabel('price')
plt.ylabel('bedrooms')
plt.show()
```



```
In [34]: # Bar chart
plt.bar(df['price'].unique(), df['stories'].mean(), align='center')
plt.title('Bar Chart Mean stories by price')
plt.xlabel('price')
plt.ylabel('Mean stories')
plt.show()
```

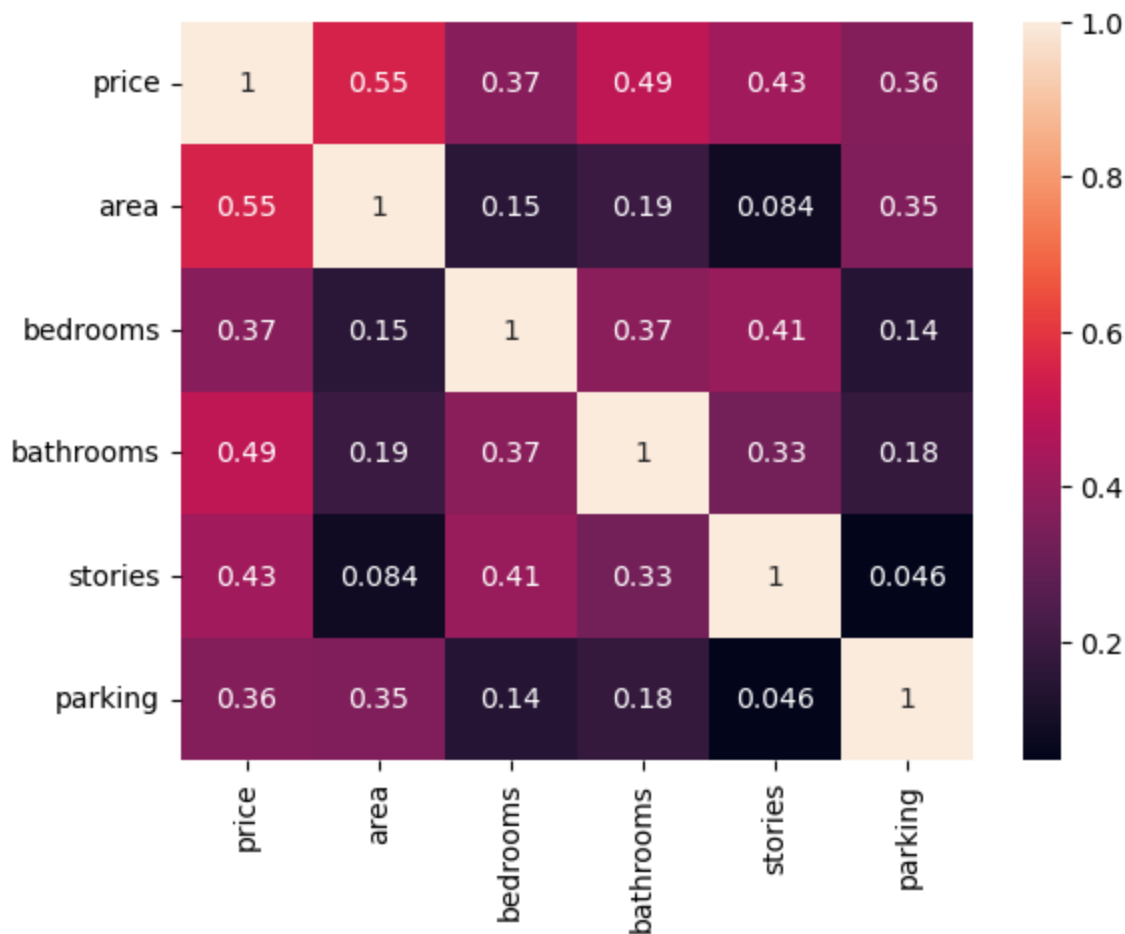


3. Multivariate analysis

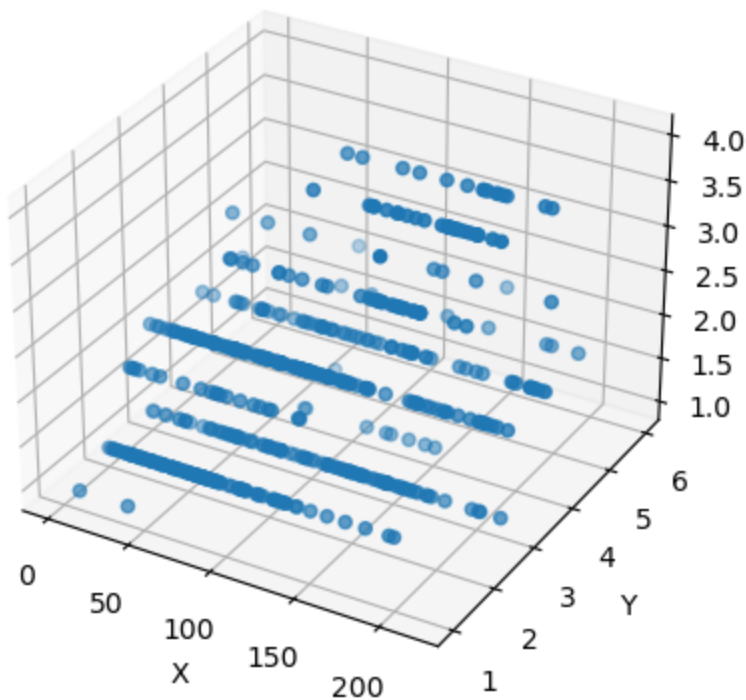
```
In [32]: # Multivariate analysis
# Heatmap
df['price'] = df['price'].astype('category').cat.codes
sns.heatmap(df.corr(), annot=True)
plt.show()
```

C:\Users\HP\AppData\Local\Temp\ipykernel_3832\46082707.py:4: 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.

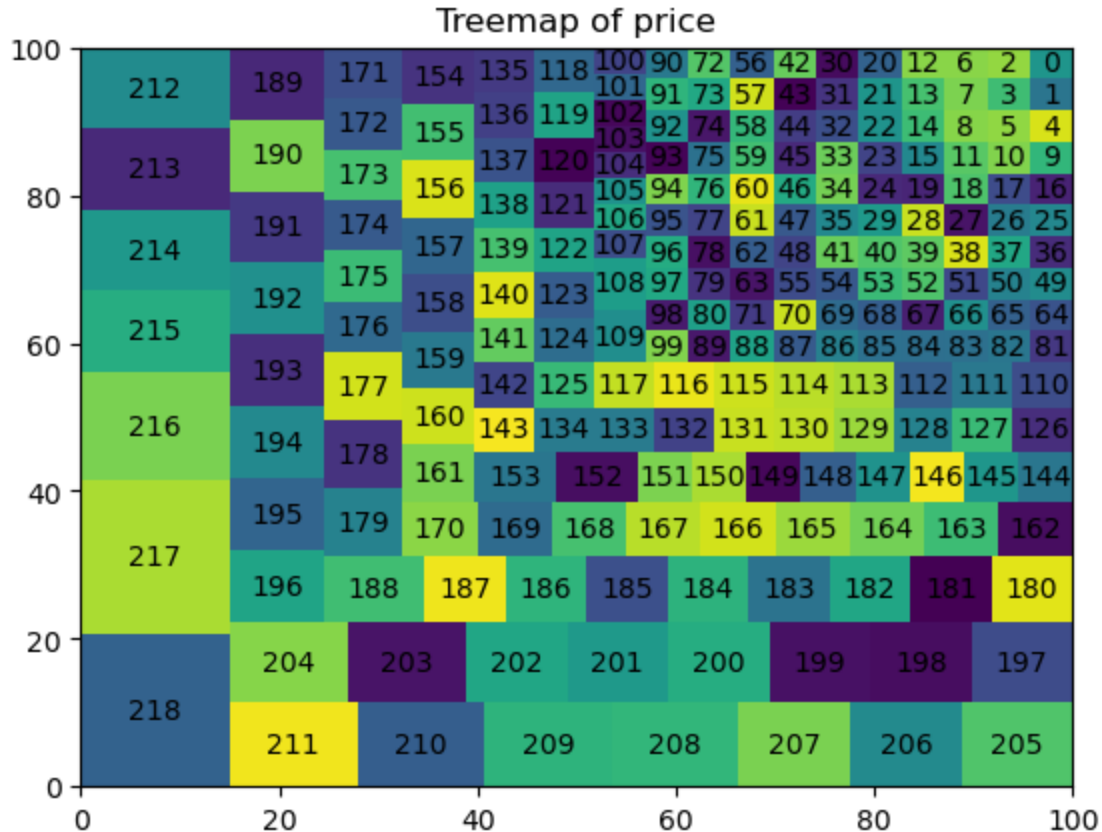
```
sns.heatmap(df.corr(), annot=True)
```



```
In [36]: # Multivariate analysis
# 3D scatterplot
from mpl_toolkits.mplot3d import Axes3D
x = df['price']
y = df['bedrooms']
z = df['stories']
fig = plt.figure()
ax = fig.add_subplot(111, projection='3d')
ax.scatter(x, y, z)
ax.set_xlabel('X')
ax.set_ylabel('Y')
ax.set_zlabel('Z')
plt.show()
```



```
In [47]: # Treemap
import squarify
plt.figure()
squarify.plot(df['price'].value_counts(), label=df['price'].unique())
plt.title('Treemap of price')
plt.show()
```



4. Perform descriptive statistics on the dataset.

```
In [49]: #4. Perform descriptive statistics on the dataset.
df.describe()
```

```
Out[49]:
```

	price	area	bedrooms	bathrooms	stories	parking
count	545.000000	545.000000	545.000000	545.000000	545.000000	545.000000
mean	95.728440	5150.541284	2.965138	1.286239	1.805505	0.693578
std	56.256108	2170.141023	0.738064	0.502470	0.867492	0.861586
min	0.000000	1650.000000	1.000000	1.000000	1.000000	0.000000
25%	51.000000	3600.000000	2.000000	1.000000	1.000000	0.000000
50%	87.000000	4600.000000	3.000000	1.000000	2.000000	0.000000
75%	137.000000	6360.000000	3.000000	2.000000	2.000000	1.000000
max	218.000000	16200.000000	6.000000	4.000000	4.000000	3.000000

```
In [50]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   price                  545 non-null    int16
1   area                   545 non-null    int64
2   bedrooms               545 non-null    int64
3   bathrooms              545 non-null    int64
4   stories                545 non-null    int64
5   mainroad               545 non-null    object
6   guestroom              545 non-null    object
7   basement               545 non-null    object
8   hotwaterheating        545 non-null    object
9   airconditioning        545 non-null    object
10  parking                545 non-null    int64
11  furnishingstatus       545 non-null    object
dtypes: int16(1), int64(5), object(6)
memory usage: 48.0+ KB
```

5. Check for Missing values and deal with them

```
In [51]: #5. Check for Missing values and deal with them
df.isnull().sum()
```

```
Out[51]: price          0
area          0
bedrooms      0
bathrooms     0
stories       0
mainroad      0
guestroom     0
basement      0
hotwaterheating 0
airconditioning 0
parking       0
furnishingstatus 0
dtype: int64
```

6 Find the outliers and replace the outliers


```
In [52]: #6. Find the outliers and replace them outliers
target_column = 'price'
Q1 = df[target_column].quantile(0.25)
Q3 = df[target_column].quantile(0.75)
IQR = Q3 - Q1
```

```
In [53]: IQR
```

```
Out[53]: 86.0
```

```
In [54]: lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR
```

```
In [55]: lower_bound
```

```
Out[55]: -78.0
```

```
In [56]: upper_bound
```

```
Out[56]: 266.0
```

```
In [57]: outliers = df[(df[target_column] < lower_bound) | (df[target_column] > upper_bound)]
```

```
In [58]: median_value = df[target_column].median()
df.loc[(df[target_column] < lower_bound) | (df[target_column] > upper_bound), target_col
```

```
In [59]: median_value
```

```
Out[59]: 87.0
```

```
In [60]: df
```

```
Out[60]:
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	aircondition
0	218	7420	4	2	3	yes	no	no	no	
1	217	8960	4	4	4	yes	no	no	no	
2	217	9960	3	2	2	yes	no	yes	no	
3	216	7500	4	2	2	yes	no	yes	no	
4	215	7420	4	1	2	yes	yes	yes	no	
...
540	2	3000	2	1	1	yes	no	yes	no	
541	1	2400	3	1	1	no	no	no	no	
542	0	3620	2	1	1	yes	no	no	no	
543	0	2910	3	1	1	no	no	no	no	
544	0	3850	3	1	2	yes	no	no	no	

545 rows × 12 columns

```
In [61]: print(df)
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	218	7420	4	2	3	yes	no	no	
1	217	8960	4	4	4	yes	no	no	
2	217	9960	3	2	2	yes	no	yes	
3	216	7500	4	2	2	yes	no	yes	
4	215	7420	4	1	2	yes	yes	yes	
..	
540	2	3000	2	1	1	yes	no	yes	
541	1	2400	3	1	1	no	no	no	
542	0	3620	2	1	1	yes	no	no	
543	0	2910	3	1	1	no	no	no	
544	0	3850	3	1	2	yes	no	no	

	hotwaterheating	airconditioning	parking	furnishingstatus
0	no	yes	2	furnished
1	no	yes	3	furnished
2	no	no	2	semi-furnished
3	no	yes	3	furnished
4	no	yes	2	furnished
..
540	no	no	2	unfurnished
541	no	no	0	semi-furnished
542	no	no	0	unfurnished
543	no	no	0	furnished
544	no	no	0	unfurnished

[545 rows x 12 columns]

7. Check for Categorical columns and perform encoding.

```
In [62]: #7. Check for Categorical columns and perform encoding.
from sklearn.preprocessing import LabelEncoder
df.dtypes
```

```
Out[62]: price                int16
area                int64
bedrooms            int64
bathrooms            int64
stories              int64
mainroad             object
guestroom            object
basement             object
hotwaterheating      object
airconditioning      object
parking              int64
furnishingstatus     object
dtype: object
```

```
In [63]: categorical_columns = df.select_dtypes(include=['object']).columns
df_encoded = pd.get_dummies(df, columns=categorical_columns)
```

```
In [64]: categorical_columns
```

```
Out[64]: Index(['mainroad', 'guestroom', 'basement', 'hotwaterheating',
               'airconditioning', 'furnishingstatus'],
              dtype='object')
```

```
In [66]: print(df_encoded)
```

	price	area	bedrooms	bathrooms	stories	parking	mainroad_no	\
0	218	7420	4	2	3	2	0	
1	217	8960	4	4	4	3	0	
2	217	9960	3	2	2	2	0	
3	216	7500	4	2	2	3	0	
4	215	7420	4	1	2	2	0	
..	
540	2	3000	2	1	1	2	0	
541	1	2400	3	1	1	0	1	
542	0	3620	2	1	1	0	0	
543	0	2910	3	1	1	0	1	
544	0	3850	3	1	2	0	0	

	mainroad_yes	guestroom_no	guestroom_yes	basement_no	basement_yes	\
0	1	1	0	1	0	
1	1	1	0	1	0	
2	1	1	0	0	1	
3	1	1	0	0	1	
4	1	0	1	0	1	
..	
540	1	1	0	0	1	
541	0	1	0	1	0	
542	1	1	0	1	0	
543	0	1	0	1	0	
544	1	1	0	1	0	

	hotwaterheating_no	hotwaterheating_yes	airconditioning_no	\
0	1	0	0	
1	1	0	0	
2	1	0	1	
3	1	0	0	
4	1	0	0	
..	
540	1	0	1	
541	1	0	1	
542	1	0	1	
543	1	0	1	
544	1	0	1	

	airconditioning_yes	furnishingstatus_furnished	\
0	1	1	
1	1	1	
2	0	0	
3	1	1	
4	1	1	
..	
540	0	0	
541	0	0	
542	0	0	
543	0	1	
544	0	0	

	furnishingstatus_semi-furnished	furnishingstatus_unfurnished
0	0	0
1	0	0
2	1	0
3	0	0
4	0	0
..
540	0	1
541	1	0
542	0	1
543	0	0
544	0	1

[545 rows x 19 columns]

8. Split the data into dependent and independent variables.

In [65]: *#8. Split the data into dependent and independent variables.*

```
dependent_variable = 'price'  
independent_variables = df.drop(dependent_variable, axis=1)  
dependent_variable = df[dependent_variable]
```

In [67]: `print(dependent_variable)`

```
0      218  
1      217  
2      217  
3      216  
4      215  
...  
540     2  
541     1  
542     0  
543     0  
544     0  
Name: price, Length: 545, dtype: int16
```

In [68]: `independent_variables`

Out[68]:

	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	price
0	7420	4	2	3	yes	no	no	no	yes	
1	8960	4	4	4	yes	no	no	no	yes	
2	9960	3	2	2	yes	no	yes	no	no	
3	7500	4	2	2	yes	no	yes	no	yes	
4	7420	4	1	2	yes	yes	yes	no	yes	
...
540	3000	2	1	1	yes	no	yes	no	no	
541	2400	3	1	1	no	no	no	no	no	
542	3620	2	1	1	yes	no	no	no	no	
543	2910	3	1	1	no	no	no	no	no	
544	3850	3	1	2	yes	no	no	no	no	

545 rows × 11 columns

In [69]: `print(independent_variables)`

	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	\
0	7420	4	2	3	yes	no	no	
1	8960	4	4	4	yes	no	no	
2	9960	3	2	2	yes	no	yes	
3	7500	4	2	2	yes	no	yes	
4	7420	4	1	2	yes	yes	yes	
..	
540	3000	2	1	1	yes	no	yes	
541	2400	3	1	1	no	no	no	
542	3620	2	1	1	yes	no	no	
543	2910	3	1	1	no	no	no	
544	3850	3	1	2	yes	no	no	

	hotwaterheating	airconditioning	parking	furnishingstatus
0	no	yes	2	furnished
1	no	yes	3	furnished
2	no	no	2	semi-furnished
3	no	yes	3	furnished
4	no	yes	2	furnished
..
540	no	no	2	unfurnished
541	no	no	0	semi-furnished
542	no	no	0	unfurnished
543	no	no	0	furnished
544	no	no	0	unfurnished

[545 rows x 11 columns]

9. Scale the independent variables

```
In [70]: #9. Scale the independent variables
from sklearn.preprocessing import StandardScaler
columns_to_scale = ['price', 'bedrooms', 'bathrooms', 'area', 'stories', 'parking']
scaler = StandardScaler()
df[columns_to_scale] = scaler.fit_transform(df[columns_to_scale])
```

```
In [71]: df
```

```
Out[71]:
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating
0	2.175477	1.046726	1.403419	1.421812	1.378217	yes	no	no	no
1	2.157685	1.757010	1.403419	5.405809	2.532024	yes	no	no	no
2	2.157685	2.218232	0.047278	1.421812	0.224410	yes	no	yes	no
3	2.139893	1.083624	1.403419	1.421812	0.224410	yes	no	yes	no
4	2.122101	1.046726	1.403419	-0.570187	0.224410	yes	yes	yes	no
...
540	-1.667633	-0.991879	-1.308863	-0.570187	-0.929397	yes	no	yes	no
541	-1.685425	-1.268613	0.047278	-0.570187	-0.929397	no	no	no	no
542	-1.703217	-0.705921	-1.308863	-0.570187	-0.929397	yes	no	no	no
543	-1.703217	-1.033389	0.047278	-0.570187	-0.929397	no	no	no	no
544	-1.703217	-0.599839	0.047278	-0.570187	0.224410	yes	no	no	no

545 rows × 12 columns

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom \
0	2.175477	1.046726	1.403419	1.421812	1.378217	yes	no
1	2.157685	1.757010	1.403419	5.405809	2.532024	yes	no
2	2.157685	2.218232	0.047278	1.421812	0.224410	yes	no
3	2.139893	1.083624	1.403419	1.421812	0.224410	yes	no
4	2.122101	1.046726	1.403419	-0.570187	0.224410	yes	yes
..
540	-1.667633	-0.991879	-1.308863	-0.570187	-0.929397	yes	no
541	-1.685425	-1.268613	0.047278	-0.570187	-0.929397	no	no
542	-1.703217	-0.705921	-1.308863	-0.570187	-0.929397	yes	no
543	-1.703217	-1.033389	0.047278	-0.570187	-0.929397	no	no
544	-1.703217	-0.599839	0.047278	-0.570187	0.224410	yes	no

	basement	hotwaterheating	airconditioning	parking	furnishingstatus	
0	no		no	yes	1.517692	furnished
1	no		no	yes	2.679409	furnished
2	yes		no	no	1.517692	semi-furnished
3	yes		no	yes	2.679409	furnished
4	yes		no	yes	1.517692	furnished
..
540	yes		no	no	1.517692	unfurnished
541	no		no	no	-0.805741	semi-furnished
542	no		no	no	-0.805741	unfurnished
543	no		no	no	-0.805741	furnished
544	no		no	no	-0.805741	unfurnished

[545 rows x 12 columns]

10.Split the data into training and testing

```
In [73]: #10.Split the data into training and testing
from sklearn.model_selection import train_test_split
X = df.drop('price', axis=1)
y = df['price']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=4
```

In [74]: X_train

```
Out[74]:
```

	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	aircondition
167	-0.253922	-1.308863	1.421812	-0.929397	yes	no	no	no	
368	0.225750	-1.308863	-0.570187	-0.929397	no	no	no	no	
301	-0.752043	0.047278	-0.570187	0.224410	yes	no	no	no	
527	-1.528742	-1.308863	-0.570187	-0.929397	no	no	yes	no	
382	-0.922695	0.047278	-0.570187	0.224410	yes	no	yes	no	
...
71	0.391790	1.403419	1.421812	2.532024	yes	no	no	no	
106	0.138117	1.403419	1.421812	-0.929397	yes	no	yes	no	
270	-0.300045	0.047278	1.421812	1.378217	yes	no	no	yes	
435	-0.512207	-1.308863	-0.570187	-0.929397	yes	no	no	no	
102	0.161178	0.047278	1.421812	2.532024	yes	yes	no	no	

408 rows × 11 columns

Out[75]:

	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	aircondition
316	0.345668	1.403419	1.421812	0.224410	no	no	yes	no	
77	0.622401	0.047278	1.421812	1.378217	yes	no	no	no	
360	-0.512207	-1.308863	-0.570187	-0.929397	yes	no	no	no	
90	-0.069433	0.047278	-0.570187	0.224410	yes	no	no	no	
493	-0.549105	0.047278	-0.570187	-0.929397	yes	no	no	no	
...
172	1.498725	0.047278	-0.570187	0.224410	yes	yes	yes	no	
124	0.633932	0.047278	1.421812	2.532024	yes	no	no	no	
388	-0.692084	0.047278	-0.570187	0.224410	yes	no	no	no	
521	-0.699002	-1.308863	-0.570187	-0.929397	no	no	no	no	
503	-0.530656	0.047278	-0.570187	-0.929397	yes	no	no	no	

137 rows × 11 columns

In [76]:

y_train

Out[76]:

167 0.520805
368 -0.635687
301 -0.262051
527 -1.525296
382 -0.706855
...
71 1.285868
106 0.983401
270 -0.155298
435 -0.920362
102 1.001194
Name: price, Length: 408, dtype: float64

In [77]:

y_test

Out[77]:

316 -0.386596
77 1.232492
360 -0.600102
90 1.125739
493 -1.276205
...
172 0.503013
124 0.876648
388 -0.742440
521 -1.454127
503 -1.329582
Name: price, Length: 137, dtype: float64

In [78]:

from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()

In [79]:

df['mainroad']=le.fit_transform(df['mainroad'])
df['guestroom']=le.fit_transform(df['guestroom'])
df['basement']=le.fit_transform(df['basement'])
df['hotwaterheating']=le.fit_transform(df['hotwaterheating'])
df['airconditioning']=le.fit_transform(df['airconditioning'])
df['furnishingstatus']=le.fit_transform(df['furnishingstatus'])

Out[80]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	aircc
0	2.175477	1.046726	1.403419	1.421812	1.378217	1	0	0	0	
1	2.157685	1.757010	1.403419	5.405809	2.532024	1	0	0	0	
2	2.157685	2.218232	0.047278	1.421812	0.224410	1	0	1	0	
3	2.139893	1.083624	1.403419	1.421812	0.224410	1	0	1	0	
4	2.122101	1.046726	1.403419	-0.570187	0.224410	1	1	1	0	

11. Build the Model

In [81]:

```
#11. Build the Model
from sklearn.linear_model import LinearRegression
model=LinearRegression()
X_train, X_test, y_train, y_test = train_test_split(df, df['price'], test_size=0.25)
```

In [83]:

```
model.fit(X_train,y_train)
```

Out[83]:

▼ LinearRegression

LinearRegression()

12. Train the model

In [84]:

```
#12. Train the model
X_train
```

Out[84]:

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	aircc
147	0.698727	0.161178	0.047278	1.421812	0.224410	1	0	0	0	
284	-0.226467	1.208154	-1.308863	-0.570187	-0.929397	1	0	0	0	
396	-0.742440	-0.696696	-1.308863	-0.570187	-0.929397	1	0	0	0	
88	1.143531	1.042114	0.047278	-0.570187	-0.929397	1	1	1	0	
138	0.734311	-0.069433	0.047278	-0.570187	1.378217	1	0	0	0	
...
242	0.022624	-0.696696	0.047278	-0.570187	0.224410	1	0	0	0	
343	-0.493349	-0.493758	-1.308863	-0.570187	-0.929397	1	0	0	0	
59	1.445998	0.391790	0.047278	1.421812	2.532024	1	1	0	0	
317	-0.386596	-0.073123	0.047278	1.421812	0.224410	1	0	0	0	
185	0.414052	-0.991879	0.047278	-0.570187	0.224410	1	0	1	0	

408 rows × 12 columns

In [85]:

```
y_train
```



```
Out[85]: 147    0.698727
284    -0.226467
396    -0.742440
88      1.143531
138     0.734311
...
242     0.022624
343    -0.493349
59      1.445998
317    -0.386596
185     0.414052
Name: price, Length: 408, dtype: float64
```

13. Test the model

```
In [86]: #13. Test the model
score = model.score(X_test, y_test)
```

```
In [87]: X_test
```

```
Out[87]:
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating
539	-1.649841	-0.996491	-1.308863	-0.570187	-0.929397	0	0	0	0
328	-0.439973	-0.300045	0.047278	1.421812	0.224410	0	0	1	0
226	0.111585	0.008975	0.047278	-0.570187	2.532024	1	0	0	0
342	-0.493349	0.923119	0.047278	-0.570187	0.224410	1	0	0	0
490	-1.258413	-0.369228	0.047278	-0.570187	0.224410	0	0	0	1
...
110	0.930025	0.668524	0.047278	-0.570187	-0.929397	1	1	1	0
155	0.663142	0.437912	0.047278	1.421812	-0.929397	1	0	1	0
209	0.253922	0.723870	0.047278	-0.570187	-0.929397	1	0	0	0
7	2.086516	5.096263	2.759560	3.413810	0.224410	1	0	0	0
500	-1.329582	-1.084123	0.047278	-0.570187	-0.929397	1	0	0	0

137 rows × 12 columns

```
In [88]: y_test
```

```
Out[88]: 539    -1.649841
328    -0.439973
226     0.111585
342    -0.493349
490    -1.258413
...
110     0.930025
155     0.663142
209     0.253922
7       2.086516
500    -1.329582
Name: price, Length: 137, dtype: float64
```

```
In [89]: score
```

```
Out[89]: 1.0
```

```
In [90]: predictions = model.predict(X_test)
```

```
In [91]: predictions
```

```
Out[91]: array([-1.64984094, -0.43997282,  0.11158471, -0.49334935, -1.25841302,
        -0.97373817, -0.74243985, -0.26205104,  1.12573887,  1.25028412,
         0.25392213,  1.55275115,  1.60612768, -0.51114153,  1.33924501,
        -0.457765   , -0.79581638,  1.81963382, -1.32958173,  0.6097657   ,
        -0.95594599, -0.97373817, -0.92036163, -0.457765   , -0.19088232,
         0.14716906, -0.457765   , -0.38659628, -0.84919292,  0.44963609,
        -0.6178946   , -0.10192143,  0.37846738,  0.85885619,  1.87301035,
        -0.26205104, -0.56451807,  0.37846738, -1.13386777, -0.03075272,
         0.2005456   , -1.34737391,  1.57054332,  0.50301263, -1.40075045,
        -0.26205104, -1.6854253   , -0.31542757,  0.48522045,  1.07236233,
        -1.70321748,  2.10430867,  0.09379253, -0.74243985,  0.07600035,
        -1.66763312, -1.4363348   , -1.2762052   , -1.20503648,  0.18275342,
        -0.54672589,  0.3606752   ,  1.46379025,  0.94781709, -0.26205104,
         1.37482936, -0.92036163, -0.40438846, -0.65347896, -1.29399738,
        -0.83140074, -1.22282866,  0.50301263,  0.44963609, -0.24425886,
        -0.74243985, -0.2086745   ,  0.02262382, -1.20503648, -0.79581638,
        -0.77802421, -1.22282866,  1.94417907, -0.22646668,  0.07600035,
         1.64171204,  1.97976342, -0.74243985,  1.5883355   , -0.56451807,
        -0.74243985, -0.36880411,  0.82327184,  1.76625728, -1.40075045,
        -0.97373817,  2.17547738, -0.81360856,  0.00483164,  1.23249194,
         0.25392213,  1.21469976, -1.06269906, -0.65347896,  1.23249194,
        -1.08049124,  2.06872432,  1.41041372,  0.32509085, -0.457765   ,
        -0.65347896,  0.69872659, -1.06269906,  2.03313996, -0.99153035,
        -0.49334935,  2.05093214, -0.60010242,  1.49937461, -0.72464767,
         1.05457015, -0.13750579, -0.92036163, -0.79581638, -1.57867223,
        -1.70321748,  0.41405174, -0.74243985, -1.09828341, -1.20503648,
         0.2005456   ,  1.23249194,  0.93002491,  0.66314223,  0.25392213,
         2.08651649, -1.32958173])
```

14. Measure the performance using Metrics

```
In [92]: #14. Measure the performance using Metrics
from sklearn.metrics import mean_squared_error, r2_score, mean_absolute_error
y_pred = model.predict(X_test)
```

```
In [93]: error=y_test-y_pred
```

```
In [94]: error
```

```
Out[94]: 539    -6.661338e-16
        328     1.165734e-15
        226     9.575674e-16
        342     6.661338e-16
        490    -3.108624e-15
        ...
        110    -6.661338e-16
        155     5.662137e-15
        209    -7.993606e-15
         7     -1.909584e-14
        500    -3.774758e-15
        Name: price, Length: 137, dtype: float64
```

```
In [95]: se=error*error
```

```
In [96]: se
```

```
Out[96]: 539      4.437343e-31
          328      1.358936e-30
          226      9.169352e-31
          342      4.437343e-31
          490      9.663546e-30
          ...
          110      4.437343e-31
          155      3.205980e-29
          209      6.389773e-29
           7       3.646510e-28
          500      1.424880e-29
          Name: price, Length: 137, dtype: float64
```

```
In [97]: mse=np.mean(se)
```

```
In [98]: mse
```

```
Out[98]: 2.464189690914093e-29
```

```
In [99]: mse2=mean_squared_error(y_test,y_pred)
```

```
In [100]: mse2
```

```
Out[100]: 2.464189690914093e-29
```

```
In [101]: mae=mean_absolute_error(y_test,y_pred)
```

```
In [102]: mae
```

```
Out[102]: 3.9692625706925375e-15
```

```
In [103]: rmse=np.sqrt(mse2)
```

```
In [104]: rmse
```

```
Out[104]: 4.964060526337378e-15
```

```
In [105]: r2=r2_score(y_test,y_pred)
```

```
In [106]: r2
```

```
Out[106]: 1.0
```

```
In [ ]:
```