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1. Import the packages required

```
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
```

2. Load the dataset into the tool.

```
df=pd.read_csv('Housing.csv')
```

```
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
```

```
df.head()
```

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom
0	13300000	7420	4	2	3	yes	no
1	12250000	8960	4	4	4	yes	no

2	12250000	9960	3	2	2	yes	no
yes							
3	12215000	7500	4	2	2	yes	no
yes							
4	11410000	7420	4	1	2	yes	yes
yes							

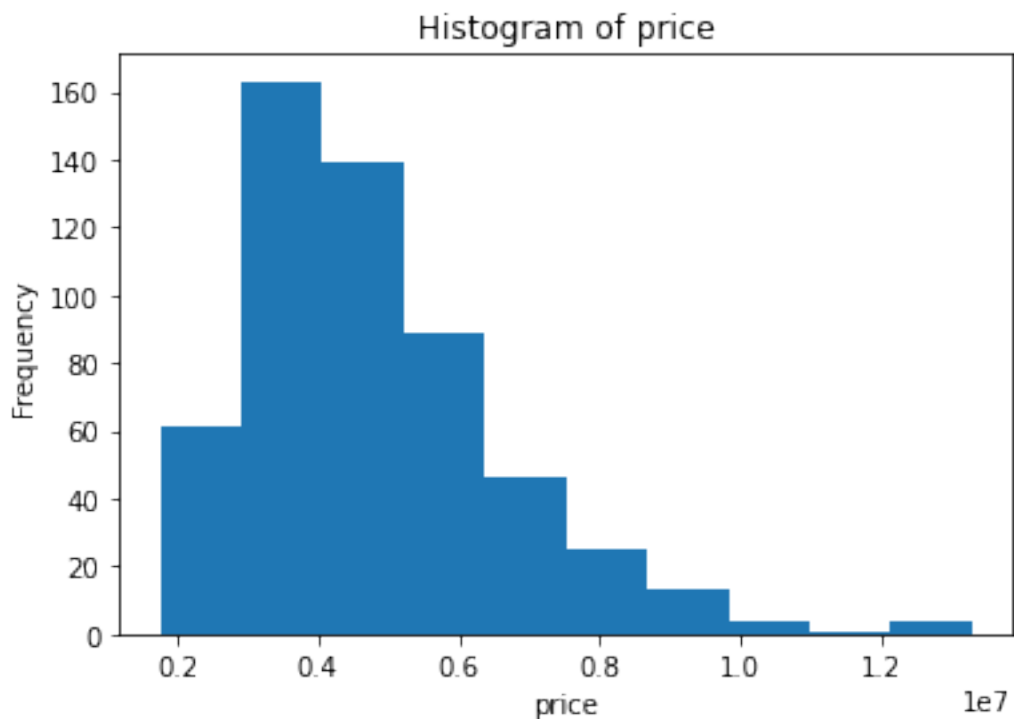
	hotwaterheating	airconditioning	parking	furnishingstatus
0	no	yes	2	furnished
1	no	yes	3	furnished
2	no	no	2	semi-furnished
3	no	yes	3	furnished
4	no	yes	2	furnished

3. Perform Below Visualizations.

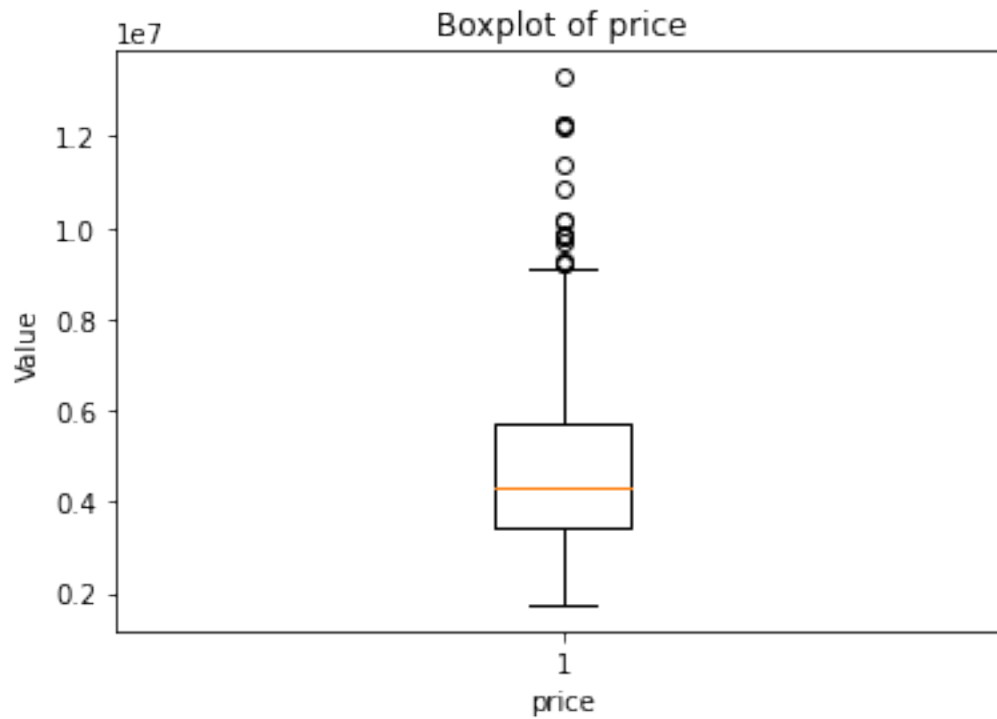
Univariate Analysis

Histogram

```
plt.hist(df['price'], bins=10)
plt.title('Histogram of price')
plt.xlabel('price')
plt.ylabel('Frequency')
plt.show()
```



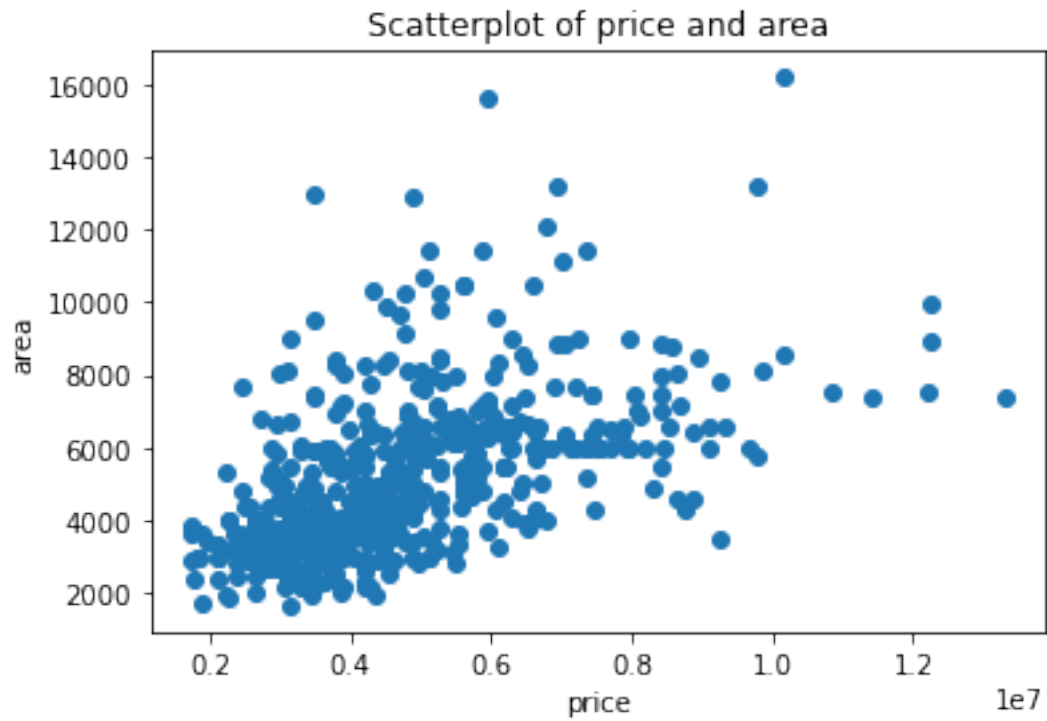
```
# Boxplot
plt.boxplot(df['price'])
plt.title('Boxplot of price')
plt.xlabel('price')
plt.ylabel('Value')
plt.show()
```



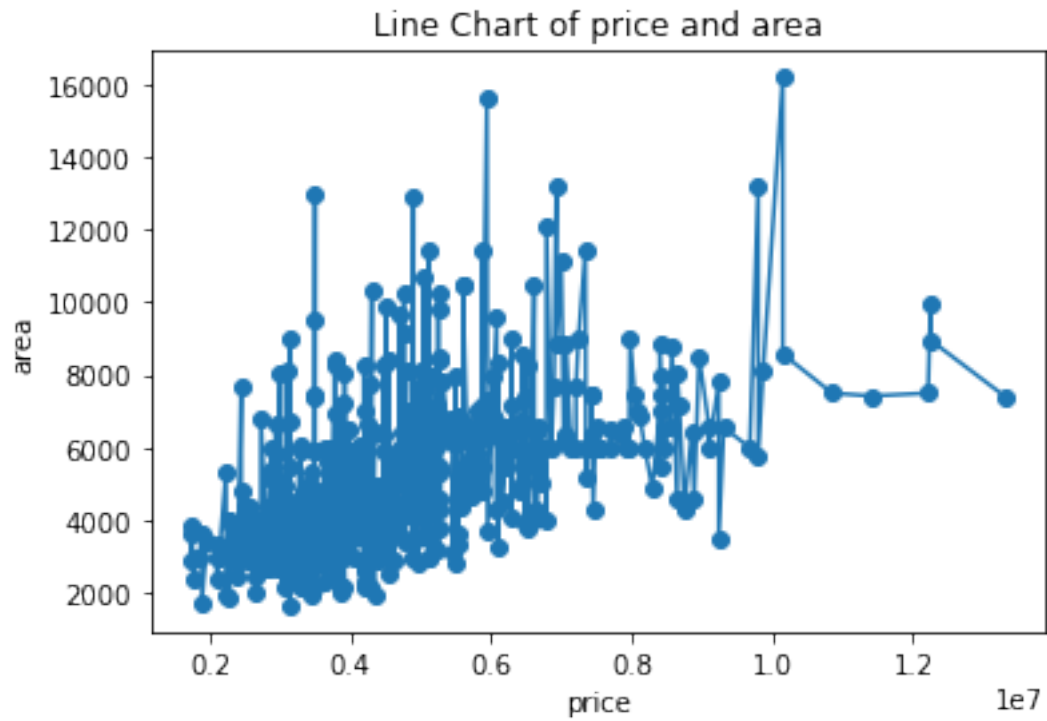
```
#Pie Chart
plt.pie(df['price'].value_counts(), labels=df['price'].unique())
plt.title('Pie Chart of price')
plt.show()
```

Bivariate analysis

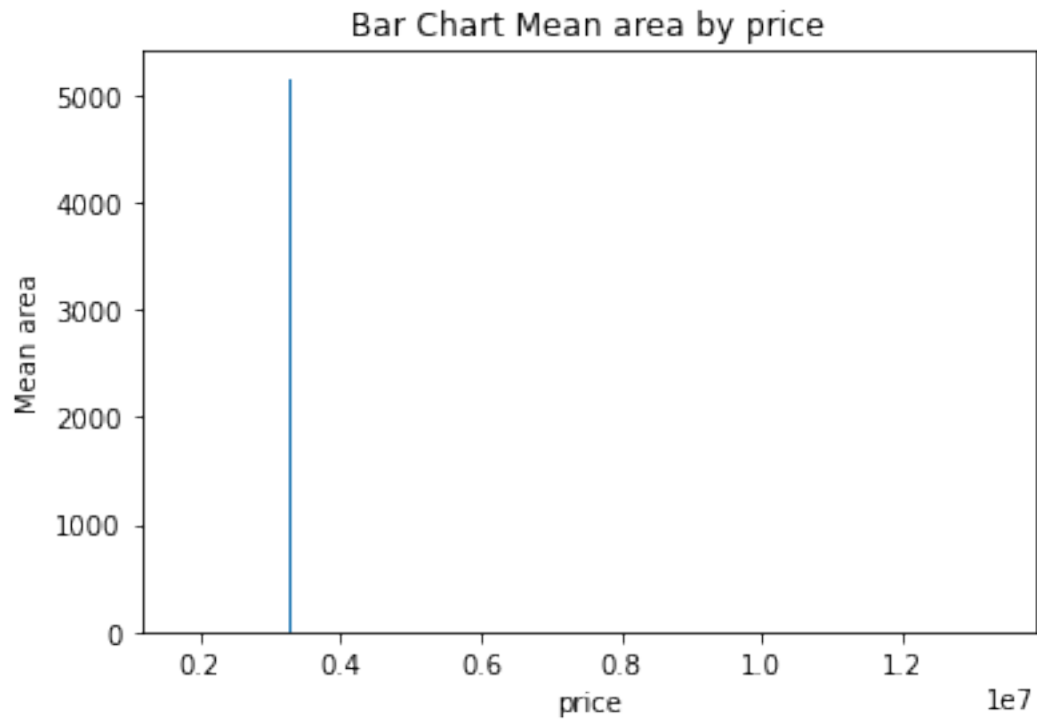
```
plt.scatter(df['price'], df['area'])
plt.title('Scatterplot of price and area')
plt.xlabel('price')
plt.ylabel('area')
plt.show()
```



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



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



Multivariate analysis

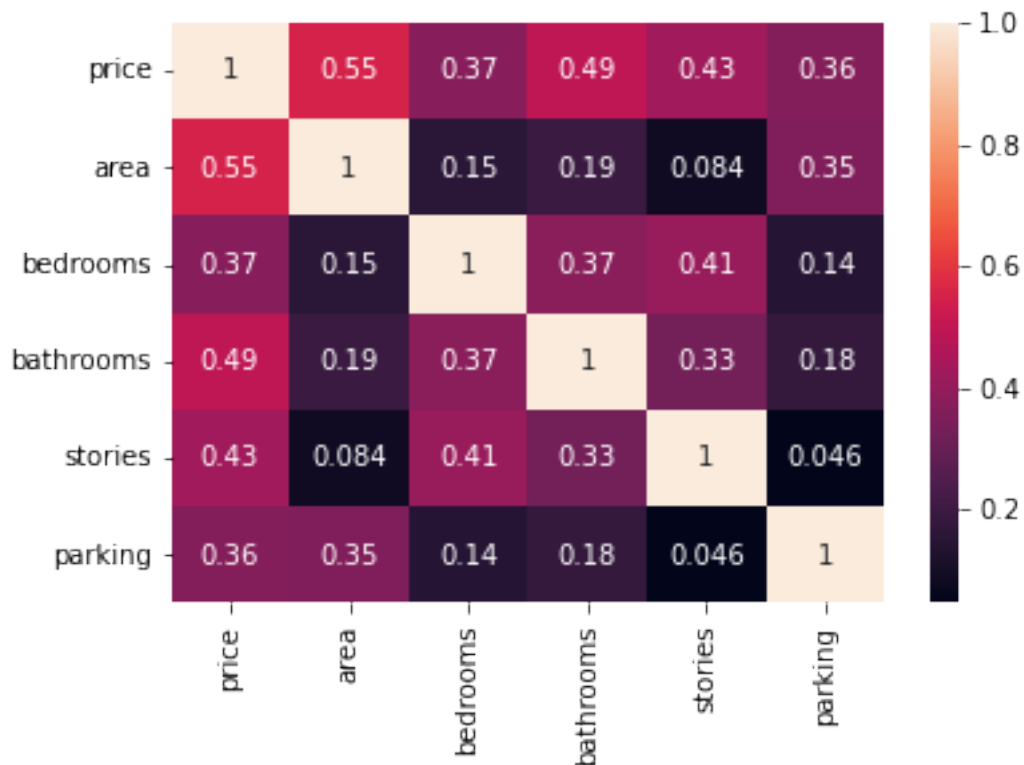
Multivariate analysis

Heatmap

```
df['price'] = df['price'].astype('category').cat.codes
```

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

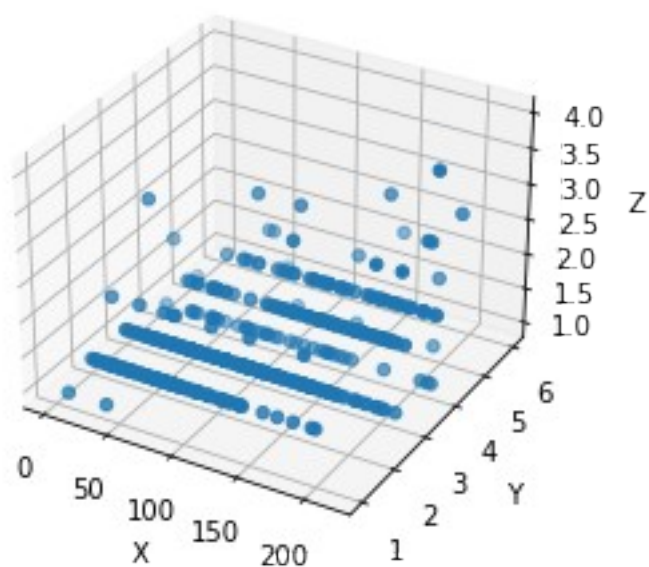
```
plt.show()
```



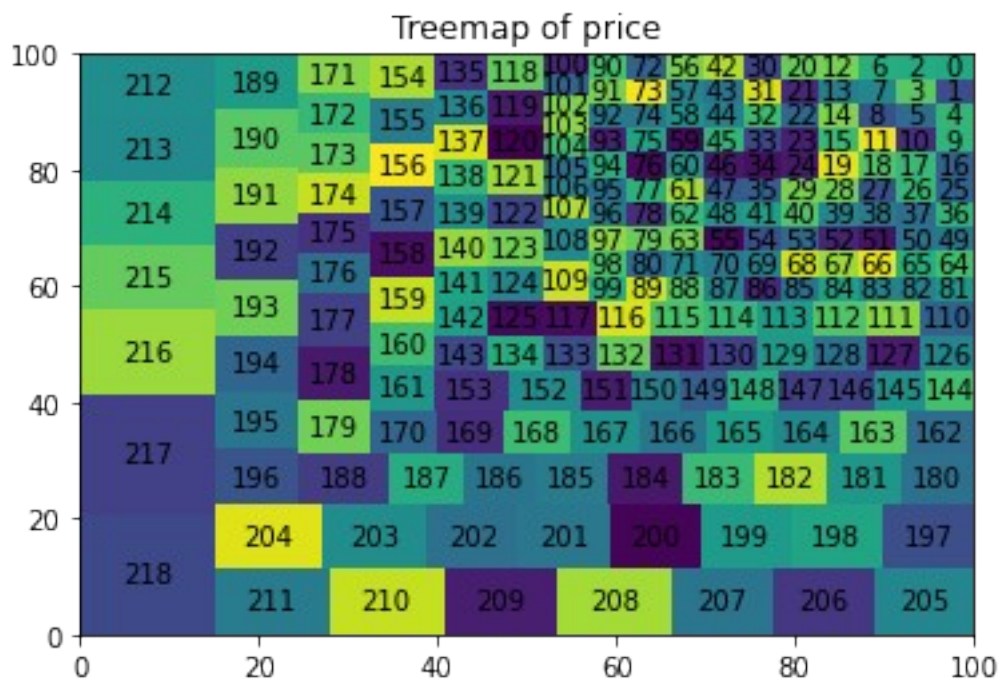
```

from mpl_toolkits.mplot3d import Axes3D
x = df['price']
y = df['bedrooms']
z = df['bathrooms']
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()

```

```
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.

#4. Perform descriptive statistics on the dataset.

```
df.describe()
```

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

	parking
count	545.000000
mean	0.693578
std	0.861586
min	0.000000
25%	0.000000
50%	0.000000
75%	1.000000
max	3.000000

```
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

```
df.isnull().sum()
```

```

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

```

target_column = 'price'
Q1 = df[target_column].quantile(0.25)
Q3 = df[target_column].quantile(0.75)
IQR = Q3 - Q1

IQR

86.0

lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

lower_bound

-78.0

upper_bound

266.0

outliers = df[(df[target_column] < lower_bound) | (df[target_column] >
upper_bound)]

median_value = df[target_column].median()
df.loc[(df[target_column] < lower_bound) | (df[target_column] >
upper_bound), target_column] = median_value

median_value

87.0

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]

```
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.

#7. Check for Categorical columns and perform encoding.

```

from sklearn.preprocessing import LabelEncoder
df.dtypes

```

```

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

```

```

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

```

```
categorical_columns
```

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

```
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.

#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]
```

```
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
```

```
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]
```



```
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

#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])

df
```

	price	area	bedrooms	bathrooms	stories	mainroad
0	2.175477	1.046726	1.403419	1.421812	1.378217	yes
1	2.157685	1.757010	1.403419	5.405809	2.532024	yes
2	2.157685	2.218232	0.047278	1.421812	0.224410	yes
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]

```
print(df)
```

```

      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

#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=42)
```

X_train

	area	bedrooms	bathrooms	stories	mainroad	guestroom
basement \						
167	-0.253922	-1.308863	1.421812	-0.929397	1	0
0						
368	0.225750	-1.308863	-0.570187	-0.929397	0	0
0						
301	-0.752043	0.047278	-0.570187	0.224410	1	0
0						
527	-1.528742	-1.308863	-0.570187	-0.929397	0	0
1						
382	-0.922695	0.047278	-0.570187	0.224410	1	0
1						
..
...						
71	0.391790	1.403419	1.421812	2.532024	1	0
0						
106	0.138117	1.403419	1.421812	-0.929397	1	0
1						
270	-0.300045	0.047278	1.421812	1.378217	1	0
0						
435	-0.512207	-1.308863	-0.570187	-0.929397	1	0
0						
102	0.161178	0.047278	1.421812	2.532024	1	1
0						
	hotwaterheating	airconditioning	parking	furnishingstatus		
167	0	1	1.517692	1		
368	0	0	-0.805741	1		
301	0	0	-0.805741	1		
527	0	0	-0.805741	1		
382	0	0	-0.805741	0		
..		
71	0	1	-0.805741	2		
106	0	1	-0.805741	1		
270	1	0	0.355976	0		
435	0	0	-0.805741	2		
102	0	1	0.355976	1		

[408 rows x 11 columns]

X_test

	area	bedrooms	bathrooms	stories	mainroad	guestroom
basement \						
316	0.345668	1.403419	1.421812	0.224410	0	0
1						
77	0.622401	0.047278	1.421812	1.378217	1	0
0						
360	-0.512207	-1.308863	-0.570187	-0.929397	1	0
0						
90	-0.069433	0.047278	-0.570187	0.224410	1	0
0						
493	-0.549105	0.047278	-0.570187	-0.929397	1	0
0						
..
...						
172	1.498725	0.047278	-0.570187	0.224410	1	1
1						
124	0.633932	0.047278	1.421812	2.532024	1	0
0						
388	-0.692084	0.047278	-0.570187	0.224410	1	0
0						
521	-0.699002	-1.308863	-0.570187	-0.929397	0	0
0						
503	-0.530656	0.047278	-0.570187	-0.929397	1	0
0						

	hotwaterheating	airconditioning	parking	furnishingstatus
316	0	0	0.355976	2
77	0	1	-0.805741	0
360	0	0	-0.805741	1
90	0	1	-0.805741	1
493	0	0	-0.805741	0
..
172	0	1	1.517692	2
124	0	0	0.355976	0
388	0	0	-0.805741	2
521	0	0	-0.805741	2
503	0	0	-0.805741	1

[137 rows x 11 columns]

y_train

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

```

y_test

```

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

```

```

from sklearn.preprocessing import LabelEncoder
le=LabelEncoder()

```

```

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'])

```

df.head()

	price	area	bedrooms	bathrooms	stories	mainroad
guestroom \						
0	2.175477	1.046726	1.403419	1.421812	1.378217	1
0						
1	2.157685	1.757010	1.403419	5.405809	2.532024	1
0						
2	2.157685	2.218232	0.047278	1.421812	0.224410	1
0						
3	2.139893	1.083624	1.403419	1.421812	0.224410	1
0						
4	2.122101	1.046726	1.403419	-0.570187	0.224410	1
1						
basement						
hotwaterheating						
airconditioning						
parking						
furnishingstatus						
0	0		0		1	1.517692
0						
1	0		0		1	2.679409

```

0
2      1      0      0  1.517692
1
3      1      0      1  2.679409
0
4      1      0      1  1.517692
0

```

11. Build the Model

#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)

model.fit(X_train,y_train)

LinearRegression()

```

12. Train the model

#12. Train the model

```

X_train

      price      area  bedrooms  bathrooms  stories  mainroad
guestroom \
473 -1.151660  1.337297 -1.308863  -0.570187 -0.929397      1
0
206  0.253922  0.299545 -1.308863  -0.570187 -0.929397      1
1
285 -0.244259  0.691585  0.047278  -0.570187  0.224410      1
1
212  0.236130 -0.798165  1.403419   1.421812  0.224410      1
0
100  1.018986  0.668524  0.047278   1.421812 -0.929397      1
0
..      ...      ...      ...      ...      ...      ...
...
450 -0.991530 -0.784329  0.047278  -0.570187  0.224410      1
0
42   1.606128  0.613177  0.047278   1.421812  2.532024      1
0
342 -0.493349  0.923119  0.047278  -0.570187  0.224410      1
0
527 -1.525296 -1.528742 -1.308863  -0.570187 -0.929397      0
0
469 -1.133868 -0.253922 -1.308863  -0.570187 -0.929397      1
0

```

	basement	hotwaterheating	airconditioning	parking
furnishingstatus				
473	0	0	0	-0.805741
2				
206	1	0	1	-0.805741
1				
285	0	0	0	-0.805741
1				
212	1	0	1	1.517692
1				
100	1	0	1	-0.805741
2				
..
...				
450	1	0	0	-0.805741
1				
42	0	0	1	1.517692
2				
342	0	0	1	-0.805741
0				
527	1	0	0	-0.805741
1				
469	0	0	0	-0.805741
0				

[408 rows x 12 columns]

y_train

473	-1.151660
206	0.253922
285	-0.244259
212	0.236130
100	1.018986

	...
450	-0.991530
42	1.606128
342	-0.493349
527	-1.525296
469	-1.133868

Name: price, Length: 408, dtype: float64

13. Test the model

#13. Test the model

score = model.score(X_test, y_test)

X_test

	price	area	bedrooms	bathrooms	stories	mainroad
guestroom \						

171	0.503013	2.360750	0.047278	-0.570187	-0.929397	1
0						
247	0.022624	1.498725	1.403419	-0.570187	2.532024	1
0						
333	-0.457765	-0.991879	0.047278	-0.570187	0.224410	1
0						
357	-0.564518	0.820727	1.403419	-0.570187	0.224410	0
0						
105	0.983401	-0.300045	0.047278	-0.570187	2.532024	1
0						
..
...						
82	1.196908	2.467293	0.047278	1.421812	-0.929397	1
0						
494	-1.293997	0.760768	-1.308863	-0.570187	-0.929397	1
0						
125	0.858856	4.819529	0.047278	-0.570187	-0.929397	1
0						
377	-0.653479	-1.061062	0.047278	1.421812	0.224410	0
0						
393	-0.742440	1.048571	0.047278	-0.570187	-0.929397	0
0						

	basement furnishingstatus	hotwaterheating	airconditioning	parking
171	0	0	0	0.355976
1				
247	0	0	0	2.679409
2				
333	0	0	0	-0.805741
1				
357	0	0	0	0.355976
0				
105	0	0	1	-0.805741
2				
..
...				
82	1	0	1	0.355976
0				
494	0	0	0	-0.805741
2				
125	0	0	1	1.517692
1				
377	1	0	0	-0.805741
2				
393	0	0	0	-0.805741
2				

[137 rows x 12 columns]

y_test

```
171    0.503013
247    0.022624
333   -0.457765
357   -0.564518
105    0.983401
```

```
...
82     1.196908
494   -1.293997
125    0.858856
377   -0.653479
393   -0.742440
```

Name: price, Length: 137, dtype: float64

score

1.0

predictions = model.predict(X_test)

predictions

```
array([ 0.50301263,  0.02262382, -0.457765   , -0.56451807,
 0.98340144,
        1.60612768,  1.30366065, -1.57867223,  1.92638689, -
0.47555718,
        0.94781709, -0.84919292, -0.08412925,  0.2005456   , -
0.26205104,
        0.50301263,  1.44599808, -0.74243985, -0.92036163, -
1.32958173,
        0.93002491, -0.58231025, -0.63568678, -0.0485449   , -
1.32958173,
        -1.13386777, -0.84919292, -0.43997282, -0.6178946   ,
2.15768521,
        0.25392213,  1.51716679, -1.25841302,  0.25392213, -
0.03075272,
        1.00119362, -0.51114153, -0.99153035,  1.23249194,
1.80184164,
        -0.42218064, -0.03075272, -1.20503648, -0.10192143, -
0.99153035,
        -0.6178946   , -0.10192143,  1.23249194, -0.0485449   , -
0.38659628,
        -1.20503648, -0.17309015,  1.76625728, -0.56451807, -
0.19088232,
        0.16496124, -0.70685549, -1.25841302, -0.97373817,
1.94417907,
        -0.19088232, -0.99153035, -0.65347896,  1.87301035, -
0.22646668,
        1.30366065,  0.37846738,  2.03313996,  1.74846511,
0.50301263,
```

```

-1.40075045, 0.43184392, -0.24425886, 1.90859471, -
0.60010242,
1.89080253, -1.63204876, -1.11607559, 0.11158471, -
0.13750579,
-0.31542757, 0.69872659, 0.25392213, 0.41405174, -
0.26205104,
-0.54672589, -1.63204876, 0.18275342, 0.50301263,
0.69872659,
1.10794669, -0.52893371, 0.53859699, 1.14353105,
0.50301263,
2.12210085, -0.49334935, -0.54672589, 1.07236233, -
0.10192143,
0.00483164, -0.74243985, 1.5883355 , -1.64984094, -
0.83140074,
-0.457765 , 1.73067293, -1.20503648, 0.91223273,
0.69872659,
1.16132322, 2.10430867, 0.02262382, -0.0485449 ,
0.64535005,
0.04041599, -1.32958173, -0.74243985, 1.837426 , -
1.09828341,
-1.06269906, 0.84106402, -1.13386777, 1.07236233, -
0.60010242,
1.07236233, -1.20503648, 1.23249194, 1.96197125,
0.00483164,
-1.32958173, 1.51716679, 1.19690758, -1.29399738,
0.85885619,
-0.65347896, -0.74243985])

```

14. Measure the performance using Metrics

#14. Measure the performance using Metrics

```

from sklearn.metrics import mean_squared_error, r2_score,
mean_absolute_error
y_pred = model.predict(X_test)

```

```

error=y_test-y_pred

```

```

error

```

```

171    4.440892e-16
247    3.469447e-17
333   -5.551115e-17
357   -4.440892e-16
105    1.110223e-15

```

```

...
82     4.440892e-16
494   -2.220446e-16
125    7.771561e-16
377   -8.881784e-16

```

```
393    -4.440892e-16
Name: price, Length: 137, dtype: float64
```

```
se=error*error
```

```
se
```

```
171    1.972152e-31
247    1.203706e-33
333    3.081488e-33
357    1.972152e-31
105    1.232595e-30
```

```
      ...
82    1.972152e-31
494    4.930381e-32
125    6.039716e-31
377    7.888609e-31
393    1.972152e-31
```

```
Name: price, Length: 137, dtype: float64
```

```
mse=np.mean(se)
```

```
mse
```

```
2.7925977603982354e-31
```

```
mse2=mean_squared_error(y_test,y_pred)
```

```
mse2
```

```
2.7925977603982354e-31
```

```
mae=mean_absolute_error(y_test,y_pred)
```

```
mae
```

```
4.4222786422255463e-16
```

```
rmse=np.sqrt(mse2)
```

```
rmse
```

```
5.284503534295569e-16
```

```
r2=r2_score(y_test,y_pred)
```

```
r2
```

```
1.0
```