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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
basement \							
0	13300000	7420	4	2	3	yes	no
no							
1	12250000	8960	4	4	4	yes	no
no							
2	12250000	9960	3	2	2	yes	no
yes							
3	12215000	7500	4	2	2	yes	no
yes							

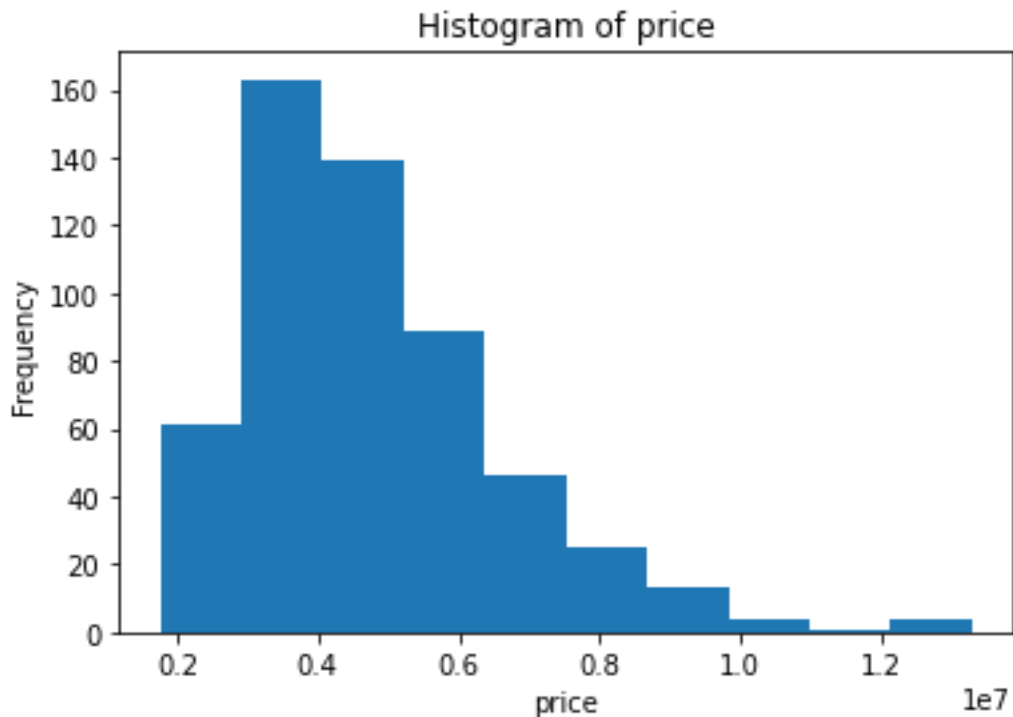
4	11410000	7420	4	1	2	yes	yes
yes							

	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.

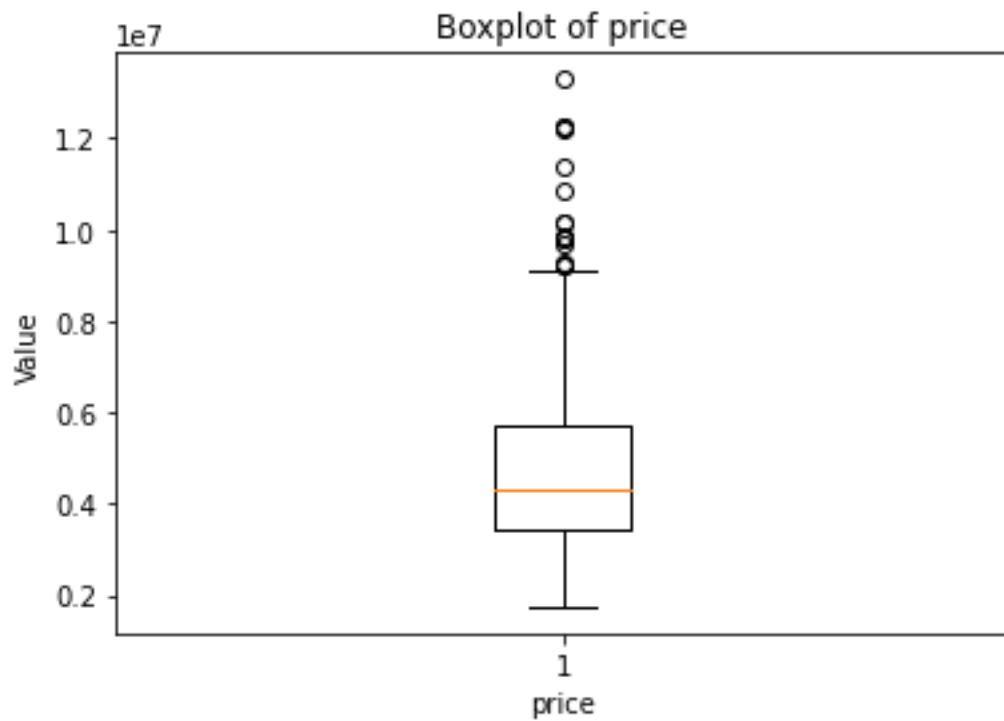
Univariate Analysis

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

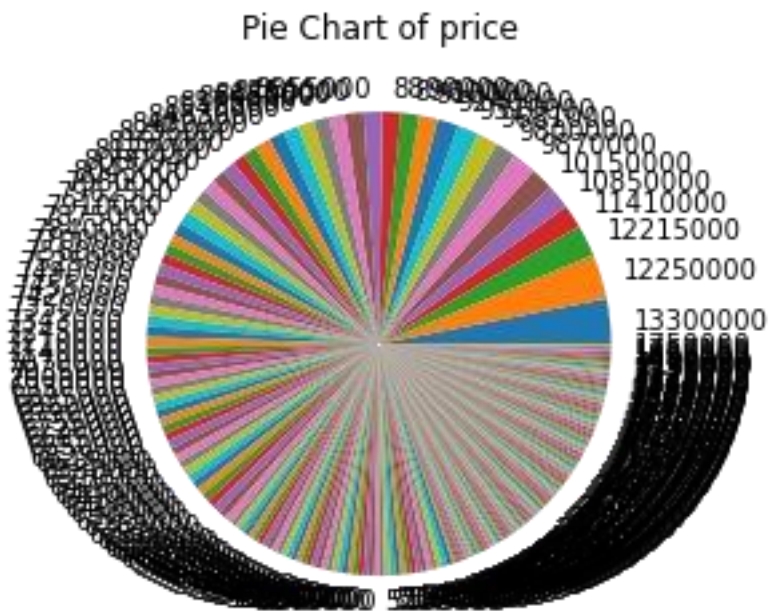


```
# Boxplot
plt.boxplot(df['price'])
plt.title('Boxplot of price')
plt.xlabel('price')
```

```
plt.ylabel('Value')
plt.show()
```

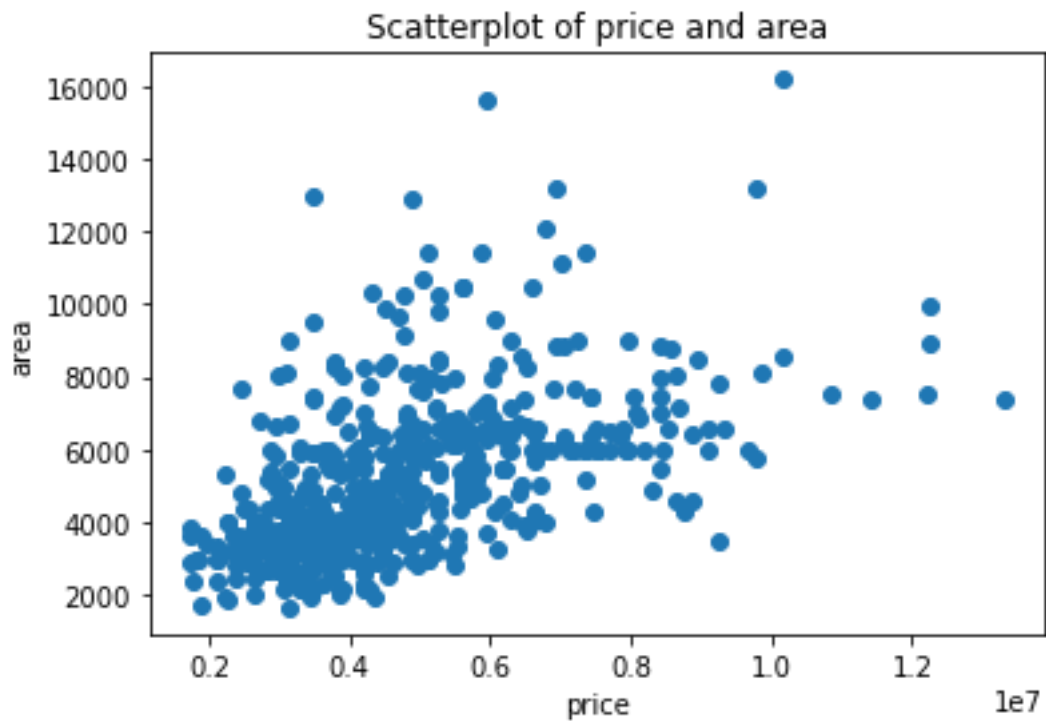


```
#Pie Chart
plt.pie(df['price'].value_counts(), labels=df['price'].unique())
plt.title('Pie Chart of price') plt.show()
```

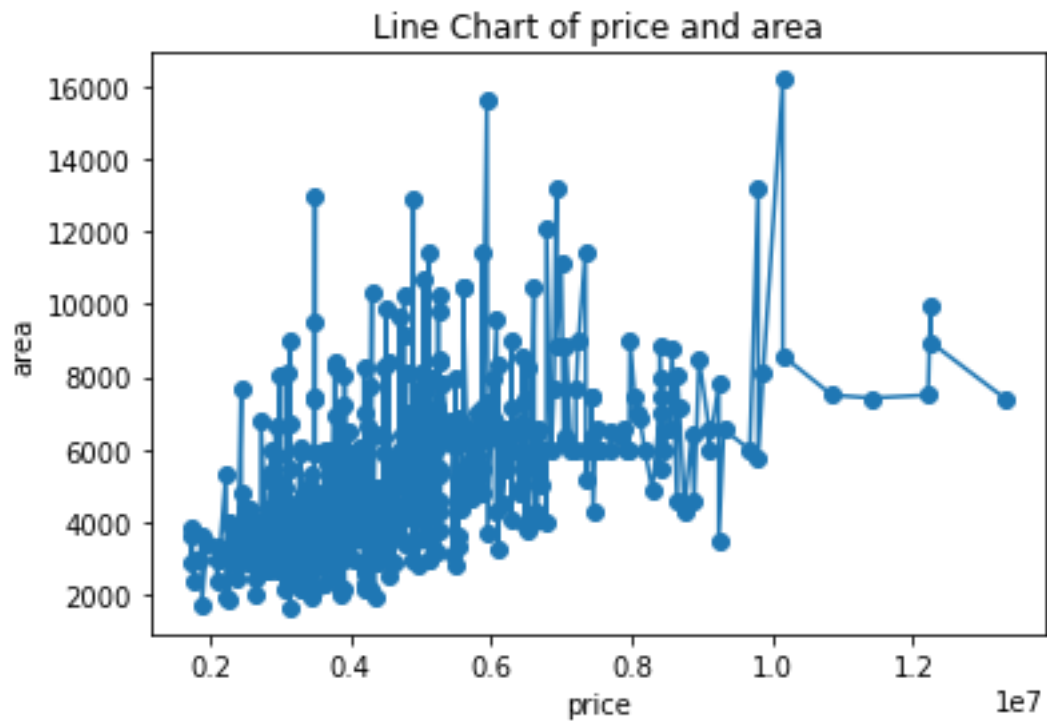


Bivariate analysis

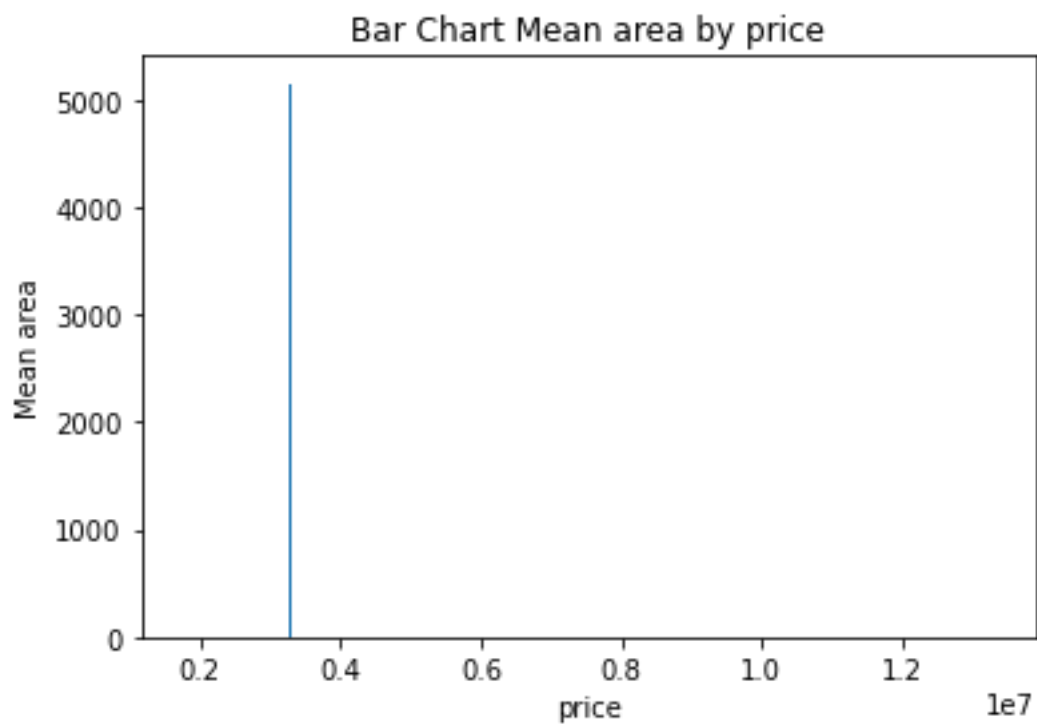
```
# Bivariate analysis # Scatterplot
plt.scatter(df['price'], df['area'])
plt.title('Scatterplot of price and area')
plt.xlabel('price') plt.ylabel('area')
plt.show()
```



```
# 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()
```

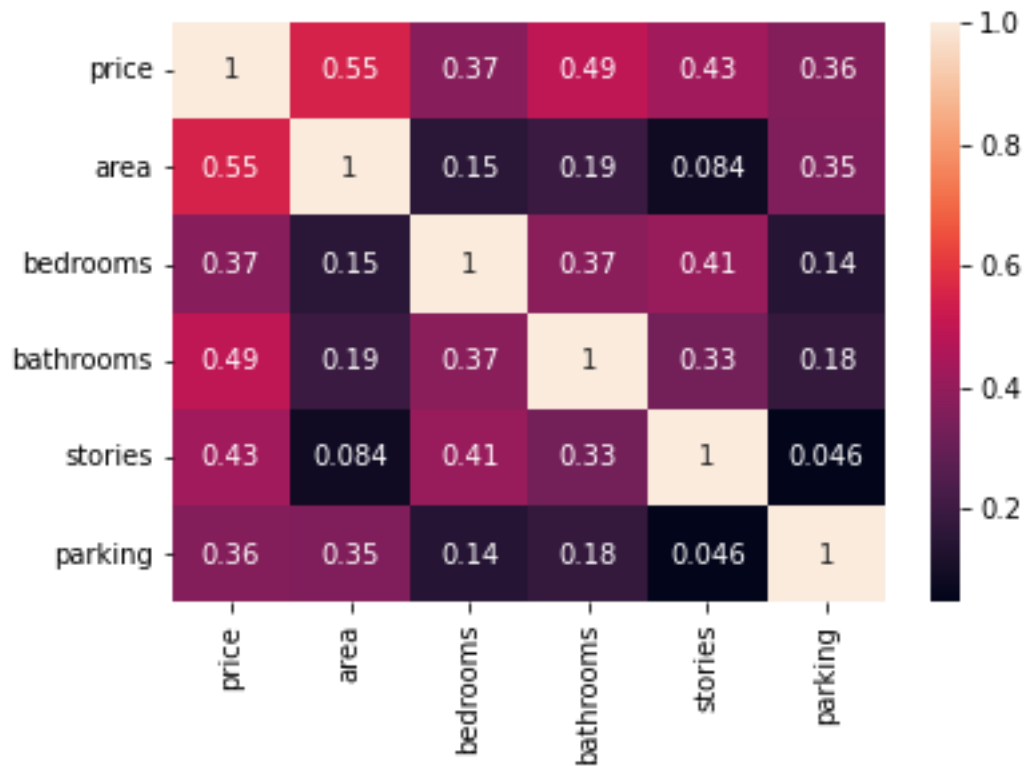


Multivariate analysis

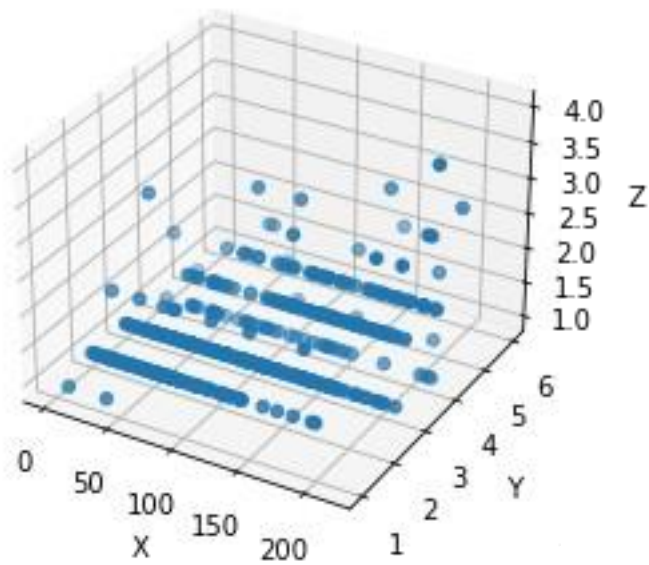
```
# Multivariate analysis
```

```
# Heatmap
```

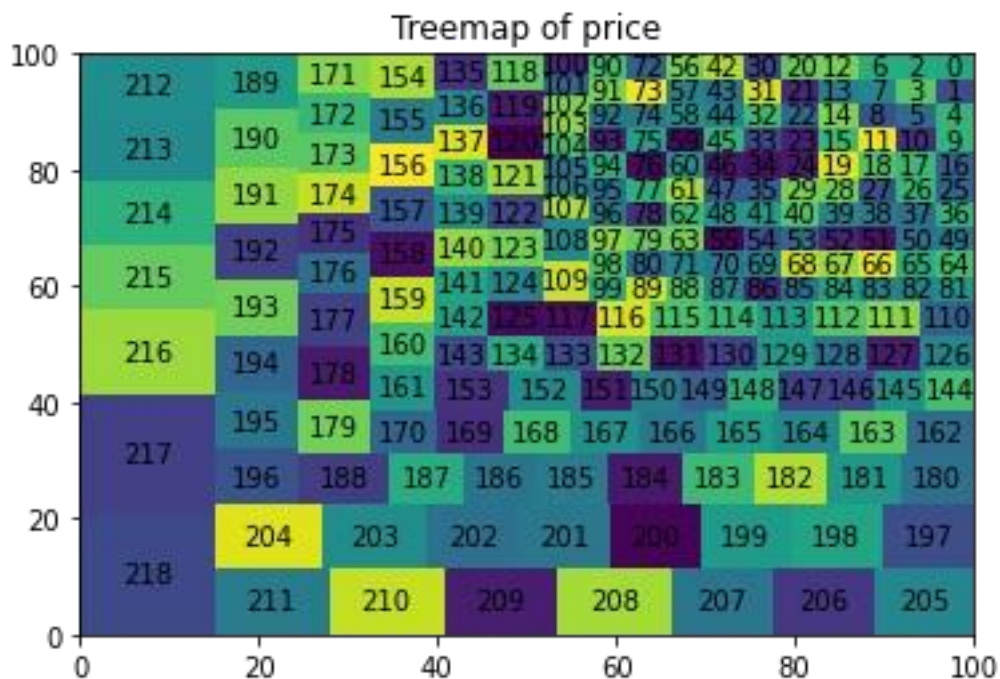
```
df['price'] = df['price'].astype('category').cat.codes  
sns.heatmap(df.corr(), annot=True) plt.show()
```



```
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

```

count    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
furnishingsstatus 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

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
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    semifurnished
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    semifurnished
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    semifurnished
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    semifurnished
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	hotwaterheating	airconditioning	parking
furnishingstatus				
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
```