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

df=pd.read_csv("/content/Housing.csv")
df

\Rightarrow		price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishi
	0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	
	1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	
	2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	semi-
	3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	
	4	11410000	7420	4	1	2	yes	yes	yes	no	yes	2	no	
	540	1820000	3000	2	1	1	yes	no	yes	no	no	2	no	ur
	541	1767150	2400	3	1	1	no	no	no	no	no	0	no	semi-
	542	1750000	3620	2	1	1	yes	no	no	no	no	0	no	ur
	543	1750000	2910	3	1	1	no	no	no	no	no	0	no	
	544	1750000	3850	3	1	2	yes	no	no	no	no	0	no	ur

545 rows × 13 columns

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 545 entries, 0 to 544
Data columns (total 13 columns):

Column Non-Null Count Dtype

```
price
                      545 non-null
                                      int64
    area
                      545 non-null
                                      int64
1
                      545 non-null
                                      int64
2
    bedrooms
    bathrooms
                      545 non-null
                                      int64
3
4
    stories
                      545 non-null
                                      int64
    mainroad
                      545 non-null
                                      object
                      545 non-null
    guestroom
                                      object
6
                      545 non-null
    basement
                                      object
7
8
   hotwaterheating
                      545 non-null
                                      object
    airconditioning 545 non-null
9
                                      object
10 parking
                      545 non-null
                                      int64
11 prefarea
                      545 non-null
                                      object
12 furnishingstatus 545 non-null
                                      object
```

dtypes: int64(6), object(7)
memory usage: 55.5+ KB

df.shape

(545, 13)

df.size

7085

df.head()

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishing
0	13300000	7420	4	2	3	yes	no	no	no	yes	2	yes	fu
1	12250000	8960	4	4	4	yes	no	no	no	yes	3	no	fu
2	12250000	9960	3	2	2	yes	no	yes	no	no	2	yes	semi-fu
3	12215000	7500	4	2	2	yes	no	yes	no	yes	3	yes	fu
1	11/110000	7/120		1	2	V/AC	VAS	1/25	no	VAC	2	no	fill

df.tail()

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	parking	prefarea	furnishin
540	1820000	3000	2	1	1	yes	no	yes	no	no	2	no	unf
541	1767150	2400	3	1	1	no	no	no	no	no	0	no	semi-f
542	1750000	3620	2	1	1	yes	no	no	no	no	0	no	unf
543	1750000	2910	3	1	1	no	no	no	no	no	0	no	fı
511	1750000	225N	2	1	2	VAC	no	no	no	no	Λ	no	unfi

df.dtypes

```
price
                     int64
area
                     int64
bedrooms
                     int64
bathrooms
                     int64
stories
                     int64
mainroad
                    object
guestroom
                    object
basement
                    object
hotwaterheating
                    object
airconditioning
                    object
                    int64
parking
prefarea
                    object
furnishingstatus
                    object
dtype: object
```

df.columns

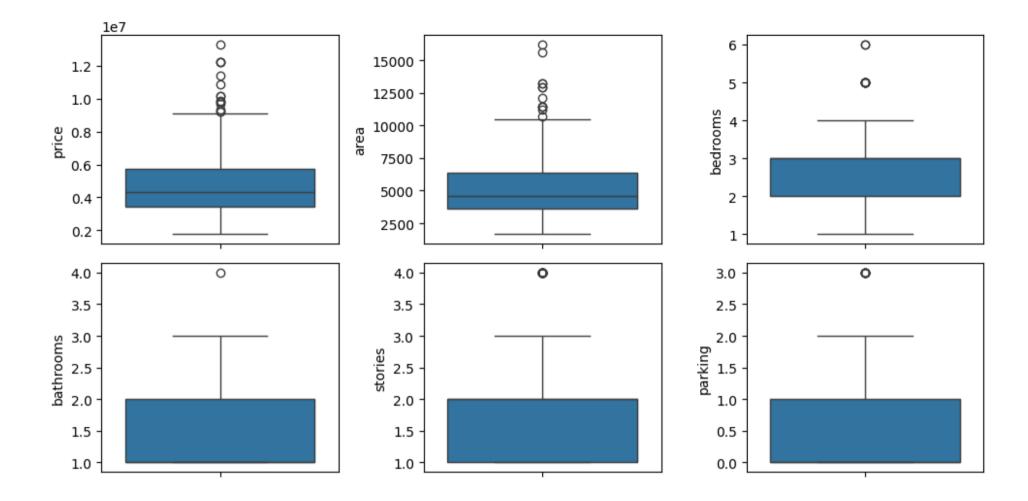
df.isna().sum()

price	0
area	0
bedrooms	0

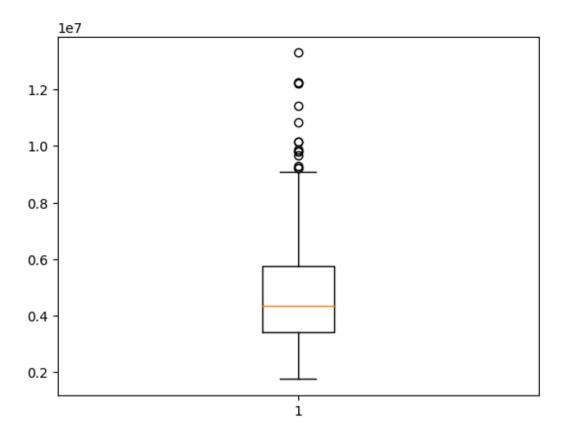
```
stories
                         0
     mainroad
                         0
                         0
     guestroom
     basement
                         0
     hotwaterheating
                         0
                         0
     airconditioning
                         0
     parking
     prefarea
                         0
     furnishingstatus
                         0
     dtype: int64
# Outlier Analysis
fig, axs = plt.subplots(2,3, figsize = (10,5))
plt1 = sns.boxplot(df['price'], ax = axs[0,0])
plt2 = sns.boxplot(df['area'], ax = axs[0,1])
plt3 = sns.boxplot(df['bedrooms'], ax = axs[0,2])
plt1 = sns.boxplot(df['bathrooms'], ax = axs[1,0])
plt2 = sns.boxplot(df['stories'], ax = axs[1,1])
plt3 = sns.boxplot(df['parking'], ax = axs[1,2])
```

bathrooms

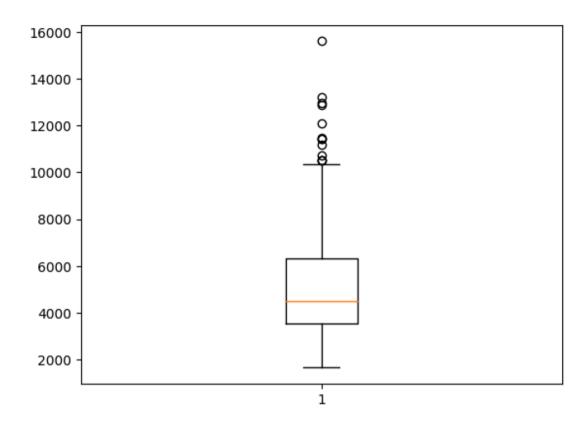
plt.tight_layout()



```
# outlier treatment for price
plt.boxplot(df.price)
Q1 = df.price.quantile(0.25)
Q3 = df.price.quantile(0.75)
IQR = Q3 - Q1
df = df[(df.price >= Q1 - 1.5*IQR) & (df.price <= Q3 + 1.5*IQR)]</pre>
```



```
# outlier treatment for area
plt.boxplot(df.area)
Q1 = df.area.quantile(0.25)
Q3 = df.area.quantile(0.75)
IQR = Q3 - Q1
df = df[(df.area >= Q1 - 1.5*IQR) & (df.area <= Q3 + 1.5*IQR)]</pre>
```



dummy=pd.get_dummies(df[['mainroad','guestroom', 'basement', 'hotwaterheating', 'airconditioning', 'prefarea', 'furnishingstatus']],drop_first=Tru
dummy

	mainroad_yes	guestroom_yes	basement_yes	hotwaterheating_yes	airconditioning_yes	prefarea_yes	furnishingstatus_semi- furnished	furnishing
15	1	0	1	0	0	0	1	
16	1	1	1	0	1	1	0	
17	1	0	0	0	1	0	0	
18	1	1	0	0	1	0	0	
19	1	0	0	0	1	1	1	
540	1	0	1	0	0	0	0	
541	0	0	0	0	0	0	1	
542	1	0	0	0	0	0	0	
543	0	0	0	0	0	0	0	
544	1	0	0	0	0	0	0	
517 ro	ws × 8 columns							

df1=pd.concat([df,dummy],axis=1)
df1

	price	area	bedrooms	bathrooms	stories	mainroad	guestroom	basement	hotwaterheating	airconditioning	• • •	prefarea	furnishir
15	9100000	6000	4	1	2	yes	no	yes	no	no		no	semi-
16	9100000	6600	4	2	2	yes	yes	yes	no	yes		yes	un
17	8960000	8500	3	2	4	yes	no	no	no	yes		no	
18	8890000	4600	3	2	2	yes	yes	no	no	yes		no	
19	8855000	6420	3	2	2	yes	no	no	no	yes		yes	semi-
540	1820000	3000	2	1	1	yes	no	yes	no	no		no	un
541	1767150	2400	3	1	1	no	no	no	no	no		no	semi-
542	1750000	3620	2	1	1	yes	no	no	no	no		no	un
543	1750000	2910	3	1	1	no	no	no	no	no		no	
544	1750000	3850	3	1	2	yes	no	no	no	no		no	un
517 rc	ws × 21 co	lumns											

df1=df1.drop(['mainroad','guestroom', 'basement', 'hotwaterheating', 'airconditioning', 'prefarea', 'furnishingstatus'],axis=1)
df1

	price	area	bedrooms	bathrooms	stories	parking	mainroad_yes	guestroom_yes	basement_yes	hotwaterheating_yes	airconditioning
15	9100000	6000	4	1	2	2	1	0	1	0	
16	9100000	6600	4	2	2	1	1	1	1	0	
17	8960000	8500	3	2	4	2	1	0	0	0	
18	8890000	4600	3	2	2	2	1	1	0	0	
19	8855000	6420	3	2	2	1	1	0	0	0	
540	1820000	3000	2	1	1	2	1	0	1	0	
541	1767150	2400	3	1	1	0	0	0	0	0	
542	1750000	3620	2	1	1	0	1	0	0	0	
543	1750000	2910	3	1	1	0	0	0	0	0	
544	1750000	3850	3	1	2	0	1	0	0	0	
517 rd	ows × 14 co	lumns									

df1.dtypes

price	int64
area	int64
bedrooms	int64
bathrooms	int64
stories	int64
parking	int64
mainroad_yes	uint8
<pre>guestroom_yes</pre>	uint8
basement_yes	uint8
hotwaterheating_yes	uint8
airconditioning_yes	uint8
prefarea_yes	uint8
furnishingstatus_semi-furnished	uint8
furnishingstatus_unfurnished	uint8
dtype: object	

	area	bedrooms	bathrooms	stories	parking	mainroad_yes	<pre>guestroom_yes</pre>	basement_yes	hotwaterheating_yes	airconditioning_yes	pre [.]
15	6000	4	1	2	2	1	0	1	0	0	
16	6600	4	2	2	1	1	1	1	0	1	
17	8500	3	2	4	2	1	0	0	0	1	
18	4600	3	2	2	2	1	1	0	0	1	
19	6420	3	2	2	1	1	0	0	0	1	
540	3000	2	1	1	2	1	0	1	0	0	
541	2400	3	1	1	0	0	0	0	0	0	
542	3620	2	1	1	0	1	0	0	0	0	
543	2910	3	1	1	0	0	0	0	0	0	
544	3850	3	1	2	0	1	0	0	0	0	
517 rc	ws × 13	3 columns									

```
y=df1['price']
y
```

- 15 9100000
- 16 9100000
- 17 8960000
- 18 8890000
- 19 8855000
 - • •
- 540 1820000
- 541 1767150
- 542 1750000

543 1750000544 1750000

Name: price, Length: 517, dtype: int64

from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,random_state=42)
x_train

	area	bedrooms	bathrooms	stories	parking	mainroad_yes	<pre>guestroom_yes</pre>	basement_yes	hotwaterheating_yes	airconditioning_yes	pre
474	4352	4	1	2	1	0	0	0	0	0	
543	2910	3	1	1	0	0	0	0	0	0	
178	7085	3	1	1	2	1	1	1	0	0	
40	6550	3	1	2	0	1	0	1	0	1	
292	2747	4	2	2	0	0	0	0	0	0	
91	6750	2	1	1	2	1	1	1	0	0	
127	6500	3	2	3	0	1	0	0	0	1	
297	3640	3	2	2	0	1	0	1	0	0	
463	3090	2	1	1	0	1	1	1	0	0	
122	6254	4	2	1	1	1	0	1	0	0	
397 rc	wc v 1	3 columns									

387 rows × 13 columns

x_test

	area	bedrooms	bathrooms	stories	parking	mainroad_yes	guestroom_yes	basement_yes	hotwaterheating_yes	airconditioning_yes	pre [.]
331	7260	3	2	1	3	1	1	1	0	0	
529	3970	3	1	2	0	1	0	1	0	0	
469	4600	2	1	1	0	1	0	0	0	0	
177	6050	3	1	1	0	1	0	1	0	0	
531	5300	3	1	1	0	0	0	0	0	1	
34	6840	5	1	2	1	1	1	1	0	1	
533	2400	3	1	2	0	1	0	0	0	0	
367	3630	2	1	1	0	1	0	0	0	1	
75	4260	4	2	2	0	1	0	0	1	0	
305	3450	3	1	2	1	1	0	0	0	0	
130 rc	ows × 1	3 columns									

y_train

... 6419000

Name: price, Length: 387, dtype: int64

```
y_test
     331
            3920000
     529
            2275000
     469
            3010000
     177
            5243000
     531
            2233000
             . . .
     34
            8120000
     533
            2100000
     367
            3675000
     75
            6650000
     305
            4193000
     Name: price, Length: 130, dtype: int64
from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
scaler.fit(x_train)
      ▼ StandardScaler
     StandardScaler()
x train=scaler.transform(x train)
x_test=scaler.transform(x_test)
from sklearn.linear model import LinearRegression
model=LinearRegression()
model.fit(x_train,y_train)
      ▼ LinearRegression
      LinearRegression()
y_pred=model.predict(x_test)
y_pred
     array([6194128.36716976, 3708265.44191532, 3278149.9224309,
            4548078.41690674, 4259819.15023141, 5810311.38242633,
```

```
3787425.1319972 , 3413010.1523162 , 3755775.93391517,
4308741.98014449, 2920161.18725972, 3203662.95236367,
4852718.18418045, 6060809.50830749, 7711000.8138782,
2985882.87344751, 6558821.03792345, 6128719.13277128,
4988962.57949334, 6012760.81566001, 5997937.55702228,
4517016.06371847, 4210161.59609396, 5521345.42322105,
5160197.9893913 , 5390074.26625152, 5550131.72250513,
2897627.51447483, 3036172.05696851, 4246203.94237666,
3266504.15143808, 3487337.13290691, 6555242.31247906,
6892836.17158825, 6723888.72882142, 3084706.13577415,
2833049.15569326, 4597490.90334397, 6595774.39402495,
4054665.22140898, 4476068.24776444, 6365274.30704325,
4620492.65670607, 4885377.57224581, 3446690.53874322,
4447366.88987447, 6128523.66140305, 2821127.06307179,
3046377.33981311, 4235631.92768213, 5552709.92599634,
3878546.01165676, 3524362.2641639, 6568288.13869669,
2213813.13678861, 4723186.21058468, 7699380.82380036,
3769851.55913972, 3098211.11904949, 6355107.31699222,
3417394.58478077, 4299535.09859223, 4952581.22704303,
7080913.47008744, 2786505.41463526, 4011266.73569139,
4750033.60728005, 4356620.86041333, 5177302.48653666,
5017742.3562727 , 7836496.16482978 , 4082054.68213363 ,
7917196.34863343, 6121406.39308625, 4232057.88027172,
6531246.85456437, 5662052.55530476, 3172778.45635302,
2716108.66946111, 6271387.75990502, 3691672.36011089,
3866678.15124159, 3045492.62251385, 5149444.44770837,
6322764.39096781, 3220110.68063684, 6132270.00914305,
2872645.53594336, 5263726.37449914, 3861082.92903906,
5938087.41805516, 3280265.16009549, 4391503.64052012,
3006740.71812295, 7044401.39977643, 6127948.66909685,
6895101.54872653, 3847734.37060897, 2533911.97021292,
4013972.73482058, 4132614.2879905 , 2816944.10965609,
2411029.42705321, 3216207.58766904, 4850094.9135177,
3955683.79323657, 4669511.64853254, 2749985.57062585,
3809401.4221589 , 2832450.09306894, 3824125.26011381,
6407063.14926889, 3018581.00128911, 3075062.46707257,
2223492.25140711, 2837097.60219231, 5429050.7923989 ,
6085843.0679925 , 3017563.16014767 , 6144820.02631941 ,
3349863.94664135, 7155487.91074025, 4205498.83943637,
5118089.32774551, 3753641.65838789, 6388013.69254177,
2986328.61547977, 3609704.51943376, 5129175.16540427,
3723911.21061878])
```

```
print(model.coef_)
     [ 421274.20553618
                         69847.80099196 335519.49728808 382920.36146923
       193767.8716274
                        137064,9099503
                                         149710.01107001 162730.10264607
       135340.16389145 400561.34939058 215809.60601501 -24587.66366152
      -141010.24401814]
list(zip(x,model.coef_))
     [('area', 421274.2055361774),
      ('bedrooms', 69847.80099196325),
      ('bathrooms', 335519.49728808226),
      ('stories', 382920.3614692315),
      ('parking', 193767.8716274008),
      ('mainroad_yes', 137064.9099503031),
      ('guestroom ves', 149710.0110700127),
      ('basement yes', 162730.1026460734),
      ('hotwaterheating yes', 135340.16389144532),
      ('airconditioning yes', 400561.3493905797),
      ('prefarea yes', 215809.6060150095),
      ('furnishingstatus semi-furnished', -24587.663661522216),
      ('furnishingstatus unfurnished', -141010.24401813897)]
print(model.intercept )
     4543252.687338501
from sklearn.metrics import mean_absolute_error
print(mean absolute error(y pred,y test))
     720789,0592011426
from sklearn.metrics import mean absolute percentage error
print(mean absolute percentage error(y pred,y test))
     0.16370154111169832
from sklearn.metrics import mean squared error
print(mean squared error(y pred,y test))
```

```
mse=mean squared error(y pred,y test)
rmse=np.sqrt(mse)
rmse
     954319.4162162034
from sklearn.metrics import r2 score
print(r2_score(y_pred,y_test))
     0.5509313402870675
from sklearn.preprocessing import PolynomialFeatures
poly=PolynomialFeatures()
x_poly=poly.fit_transform(x)
x poly
     array([[1.00e+00, 6.00e+03, 4.00e+00, ..., 1.00e+00, 0.00e+00, 0.00e+00],
            [1.00e+00, 6.60e+03, 4.00e+00, ..., 0.00e+00, 0.00e+00, 1.00e+00],
            [1.00e+00, 8.50e+03, 3.00e+00, ..., 0.00e+00, 0.00e+00, 0.00e+00],
            . . . ,
            [1.00e+00, 3.62e+03, 2.00e+00, ..., 0.00e+00, 0.00e+00, 1.00e+00],
            [1.00e+00, 2.91e+03, 3.00e+00, ..., 0.00e+00, 0.00e+00, 0.00e+00],
            [1.00e+00, 3.85e+03, 3.00e+00, ..., 0.00e+00, 0.00e+00, 1.00e+00]])
model1=LinearRegression()
model1.fit(x poly,y)
      ▼ LinearRegression
     LinearRegression()
y_poly=model1.predict(x_poly)
y poly
     array([5212043.46412995, 7958165.1018231, 7871580.2449039,
            7254161.92279949, 7591210.03811599, 7078274.73053734,
```

```
8075534.41363492, 6655796.43025099, 6973508.06943973,
7262500.3545027 , 8069633.12802633 , 7855181.68434715 ,
4823953.33207059, 8650256.71149069, 7277923.55665832,
8139829.82662969, 6964903.78311496, 6632808.08992082,
6839153.47904213, 6877783.19297101, 7823314.49834665,
6833998.69879143, 7785759.35242216, 6866669.9702782,
7699649.32672868, 6023188.08170411, 7982217.77247437,
8083176.22763896, 6632035.97472083, 7500377.79090599,
6305921.11819862, 7583339.71726626, 7254015.97946179,
4930099.28507092, 7391294.33181166, 7338377.71343498,
6672820.15471011, 7789788.60925507, 7791968.21726042,
7301348.08615963, 4891330.08256173, 7690534.56505232,
7383270.96174543, 7587398.06276466, 5484425.30690609,
5312547.5785801 , 7023776.84006026, 7615121.50763191,
6906086.64329245, 5548009.18901891, 5648920.44382125,
6289908.78131224, 6464183.63583391, 6789743.80944074,
6827193.47085024, 7438613.33300808, 5910193.89731809,
7145906.56940692, 7162600.98000578, 6400918.22094498,
6378719.77617056, 7838001.31583158, 5741736.33549489,
6950932.29553033, 5804808.17068816, 6430512.58618441,
5933036.2423356 , 4158224.76905461, 6260996.29997177,
7013493.09967076, 4582568.07685421, 5124579.29991309,
6550539.14171991, 5903852.54361692, 6426370.15397618,
6521709.55469852, 5919871.3437011, 6394136.67183848,
7032248.59083216, 5076496.72890513, 6814594.74973433,
5545985.40325069, 7620116.30129828, 6375712.03807781,
6171528.12790365, 5603280.92523459, 6661743.12870595,
5935167.50114324, 4984790.17990351, 6017645.93728299,
5651496.28190588, 5686605.09776249, 4705409.87273714,
5645720.92202947, 4786111.0959348 , 6289913.82476288,
5697354.008966 , 5406148.30384827, 5793518.30975962,
5905771.85642414, 5628542.46522015, 5599986.00726839,
```