

House_price_prediction

December 18, 2025

```
[1]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
```

```
[2]: df1 = pd.read_csv("bengaluru_house_prices.csv")
df1.head()
```

```
[2]:
```

		area_type	availability	location	size \
0	Super built-up	Area	19-Dec	Electronic City Phase II	2 BHK
1	Plot	Area	Ready To Move	Chikka Tirupathi	4 Bedroom
2	Built-up	Area	Ready To Move	Uttarahalli	3 BHK
3	Super built-up	Area	Ready To Move	Lingadheeranahalli	3 BHK
4	Super built-up	Area	Ready To Move	Kothanur	2 BHK

	society	total_sqft	bath	balcony	price
0	Coomee	1056	2.0	1.0	39.07
1	Theanmp	2600	5.0	3.0	120.00
2	NaN	1440	2.0	3.0	62.00
3	Soiewre	1521	3.0	1.0	95.00
4	NaN	1200	2.0	1.0	51.00

```
[3]: df1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 13320 entries, 0 to 13319
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   area_type       13320 non-null  object
1   availability     13320 non-null  object
2   location        13319 non-null  object
3   size            13304 non-null  object
4   society         7818 non-null   object
5   total_sqft      13320 non-null  object
6   bath            13247 non-null  float64
7   balcony         12711 non-null  float64
8   price           13320 non-null  float64
dtypes: float64(3), object(6)
```

memory usage: 936.7+ KB

```
[4]: df1.shape
```

```
[4]: (13320, 9)
```

```
[5]: df1.groupby('area_type')['area_type'].agg('count')
```

```
[5]: area_type
Built-up Area      2418
Carpet Area         87
Plot Area          2025
Super built-up Area 8790
Name: area_type, dtype: int64
```

```
[6]: df1.value_counts('area_type')
```

```
[6]: area_type
Super built-up Area      8790
Built-up Area            2418
Plot Area                2025
Carpet Area               87
Name: count, dtype: int64
```

```
[7]: df1.nunique()
```

```
[7]: area_type      4
availability     81
location        1305
size             31
society         2688
total_sqft      2117
bath            19
balcony          4
price          1994
dtype: int64
```

```
[8]: df2= df1.drop(['area_type','society','balcony', 'availability' ],_
    ↪axis='columns')
df2.head()
```

```
[8]:
```

	location	size	total_sqft	bath	price
0	Electronic City Phase II	2 BHK	1056	2.0	39.07
1	Chikka Tirupathi	4 Bedroom	2600	5.0	120.00
2	Uttarahalli	3 BHK	1440	2.0	62.00
3	Lingadheeranahalli	3 BHK	1521	3.0	95.00
4	Kothanur	2 BHK	1200	2.0	51.00

```
[9]: df2.isnull().sum()
```

```
[9]: location      1
     size         16
     total_sqft    0
     bath         73
     price         0
     dtype: int64
```

```
[10]: df3 = df2.dropna()
      df3.isnull().sum()
```

```
[10]: location      0
     size         0
     total_sqft    0
     bath         0
     price         0
     dtype: int64
```

```
[11]: df3.shape
```

```
[11]: (13246, 5)
```

```
[12]: df3.head()
```

```
[12]:
```

	location	size	total_sqft	bath	price
0	Electronic City Phase II	2 BHK	1056	2.0	39.07
1	Chikka Tirupathi	4 Bedroom	2600	5.0	120.00
2	Uttarahalli	3 BHK	1440	2.0	62.00
3	Lingadheeranahalli	3 BHK	1521	3.0	95.00
4	Kothanur	2 BHK	1200	2.0	51.00

```
[13]: df3['size'].unique()
```

```
[13]: array(['2 BHK', '4 Bedroom', '3 BHK', '4 BHK', '6 Bedroom', '3 Bedroom',
        '1 BHK', '1 RK', '1 Bedroom', '8 Bedroom', '2 Bedroom',
        '7 Bedroom', '5 BHK', '7 BHK', '6 BHK', '5 Bedroom', '11 BHK',
        '9 BHK', '9 Bedroom', '27 BHK', '10 Bedroom', '11 Bedroom',
        '10 BHK', '19 BHK', '16 BHK', '43 Bedroom', '14 BHK', '8 BHK',
        '12 Bedroom', '13 BHK', '18 Bedroom'], dtype=object)
```

```
[14]: df3['bhk'] = df3['size'].apply(lambda x : int(x.split(' ')[0]))
```

C:\Users\dell\AppData\Local\Temp\ipykernel_14048\3847263516.py:1:

SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: <https://pandas.pydata.org/pandas->

docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
df3['bhk'] = df3['size'].apply(lambda x : int(x.split(' ')[0]))

```
[15]: df3.head()
```

```
[15]:
```

	location	size	total_sqft	bath	price	bhk
0	Electronic City Phase II	2 BHK	1056	2.0	39.07	2
1	Chikka Tirupathi	4 Bedroom	2600	5.0	120.00	4
2	Uttarahalli	3 BHK	1440	2.0	62.00	3
3	Lingadheeranahalli	3 BHK	1521	3.0	95.00	3
4	Kothanur	2 BHK	1200	2.0	51.00	2

```
[16]: df3['bhk'].unique()
```

```
[16]: array([ 2,  4,  3,  6,  1,  8,  7,  5, 11,  9, 27, 10, 19, 16, 43, 14, 12,
        13, 18], dtype=int64)
```

```
[17]: df3[df3['bhk']>20]
```

```
[17]:
```

	location	size	total_sqft	bath	price	bhk
1718	Electronic City Phase II	27 BHK	8000	27.0	230.0	27
4684	Munnekollal	43 Bedroom	2400	40.0	660.0	43

```
[18]: df3.total_sqft.unique()
```

```
[18]: array(['1056', '2600', '1440', ..., '1133 - 1384', '774', '4689'],
        dtype=object)
```

```
[19]: def is_float(x):
        try:
            float(x)
        except:
            return False
        return True
```

```
[20]: df3[~df3['total_sqft'].apply(is_float)].head(10)
```

```
[20]:
```

	location	size	total_sqft	bath	price	bhk
30	Yelahanka	4 BHK	2100 - 2850	4.0	186.000	4
122	Hebbal	4 BHK	3067 - 8156	4.0	477.000	4
137	8th Phase JP Nagar	2 BHK	1042 - 1105	2.0	54.005	2
165	Sarjapur	2 BHK	1145 - 1340	2.0	43.490	2
188	KR Puram	2 BHK	1015 - 1540	2.0	56.800	2
410	Kengeri	1 BHK	34.46Sq. Meter	1.0	18.500	1
549	Hennur Road	2 BHK	1195 - 1440	2.0	63.770	2
648	Arekere	9 Bedroom	4125Perch	9.0	265.000	9
661	Yelahanka	2 BHK	1120 - 1145	2.0	48.130	2

672 Bettahalsoor 4 Bedroom 3090 - 5002 4.0 445.000 4

```
[21]: def convert_sqft_to_num(x):
      tokens = x.split('-')
      if len(tokens)==2:
          return (float(tokens[0])+float(tokens[1]))/2
      try:
          return float(x)
      except:
          return None
```

```
[22]: convert_sqft_to_num('2166')
```

```
[22]: 2166.0
```

```
[23]: convert_sqft_to_num('2100 - 2850')
```

```
[23]: 2475.0
```

```
[24]: convert_sqft_to_num('4125Perch')
```

```
[25]: #applying this function into the total_sqft column to get the average
```

```
[26]: df4 = df3.copy()
      df4['total_sqft'] = df4['total_sqft'].apply(convert_sqft_to_num)
      df4.head(3)
```

```
[26]:
```

	location	size	total_sqft	bath	price	bhk
0	Electronic City Phase II	2 BHK	1056.0	2.0	39.07	2
1	Chikka Tirupathi	4 Bedroom	2600.0	5.0	120.00	4
2	Uttarahalli	3 BHK	1440.0	2.0	62.00	3

```
[27]: df4.loc[30]
```

```
[27]: location      Yelahanka
      size        4 BHK
      total_sqft  2475.0
      bath        4.0
      price       186.0
      bhk         4
      Name: 30, dtype: object
```

```
[28]: df4.head(3)
```

```
[28]:
```

	location	size	total_sqft	bath	price	bhk
0	Electronic City Phase II	2 BHK	1056.0	2.0	39.07	2
1	Chikka Tirupathi	4 Bedroom	2600.0	5.0	120.00	4
2	Uttarahalli	3 BHK	1440.0	2.0	62.00	3

```
[29]: #we cleaned the dataset
```

```
[30]: #now want to apply feature engineering
```

```
[31]: df5 = df4.copy()

df5['price_per_sqft']=df5['price']*100000/df5['total_sqft']
df5.head()
```

```
[31]:
```

	location	size	total_sqft	bath	price	bhk	\
0	Electronic City Phase II	2 BHK	1056.0	2.0	39.07	2	
1	Chikka Tirupathi	4 Bedroom	2600.0	5.0	120.00	4	
2	Uttarahalli	3 BHK	1440.0	2.0	62.00	3	
3	Lingadheeranahalli	3 BHK	1521.0	3.0	95.00	3	
4	Kothanur	2 BHK	1200.0	2.0	51.00	2	

	price_per_sqft
0	3699.810606
1	4615.384615
2	4305.555556
3	6245.890861
4	4250.000000

```
[32]: #check locations
```

```
[33]: df5['location'].value_counts()
```

```
[33]: location
Whitefield          534
Sarjapur Road       392
Electronic City     302
Kanakpura Road      266
Thanisandra         233
...
Vidyapeeta          1
Maruthi Extension   1
Okalipura            1
Old Town             1
Abshot Layout        1
Name: count, Length: 1304, dtype: int64
```

```
[34]: df5['location'].nunique()
```

```
[34]: 1304
```

```
[35]: df5.shape
```

```
[35]: (13246, 7)
```

```
[36]: # too many locations->cannot one-hot encoding
```

```
[37]: df5.location=df5.location.apply(lambda x: x.strip())
location_stats = df5.groupby('location')['location'].agg('count').
    ↪sort_values(ascending=False)
location_stats
```

```
[37]: location
Whitefield          535
Sarjapur Road       392
Electronic City     304
Kanakapura Road     266
Thanisandra         236
...
1 Giri Nagar        1
Kanakapura Road,    1
Kanakapura main Road 1
Karnataka Shabarimala 1
whitefiled          1
Name: location, Length: 1293, dtype: int64
```

```
[38]: len(location_stats[location_stats<=10])
```

```
[38]: 1052
```

```
[39]: location_stats_less_than_10 = location_stats[location_stats<=10]
location_stats_less_than_10
```

```
[39]: location
Basapura           10
1st Block Koramangala 10
Gunjur Palya       10
Kalkere            10
Sector 1 HSR Layout 10
..
1 Giri Nagar        1
Kanakapura Road,    1
Kanakapura main Road 1
Karnataka Shabarimala 1
whitefiled          1
Name: location, Length: 1052, dtype: int64
```

```
[40]: len(df5['location'].unique())
```

```
[40]: 1293
```

```
[41]: df5['location'] = df5.location.apply(lambda x: 'other' if x in location_stats_less_than_10 else x)
len(df5['location'].unique())
```

```
[41]: 242
```

```
[42]: df5.head()
```

```
[42]:
```

	location	size	total_sqft	bath	price	bhk	\
0	Electronic City Phase II	2 BHK	1056.0	2.0	39.07	2	
1	Chikka Tirupathi	4 Bedroom	2600.0	5.0	120.00	4	
2	Uttarahalli	3 BHK	1440.0	2.0	62.00	3	
3	Lingadheeranahalli	3 BHK	1521.0	3.0	95.00	3	
4	Kothanur	2 BHK	1200.0	2.0	51.00	2	

	price_per_sqft
0	3699.810606
1	4615.384615
2	4305.555556
3	6245.890861
4	4250.000000

```
[43]: # outlier removal
```

```
[44]: df5.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Index: 13246 entries, 0 to 13319
Data columns (total 7 columns):
#   Column                Non-Null Count  Dtype
---  -
0   location              13246 non-null  object
1   size                  13246 non-null  object
2   total_sqft            13200 non-null  float64
3   bath                  13246 non-null  float64
4   price                  13246 non-null  float64
5   bhk                    13246 non-null  int64
6   price_per_sqft        13200 non-null  float64
dtypes: float64(4), int64(1), object(2)
memory usage: 1.3+ MB
```

```
[45]: df5[df5['total_sqft']/df5['bhk']<300].head()
```

```
[45]:
```

	location	size	total_sqft	bath	price	bhk	\
9	other	6 Bedroom	1020.0	6.0	370.0	6	
45	HSR Layout	8 Bedroom	600.0	9.0	200.0	8	
58	Murugeshpalya	6 Bedroom	1407.0	4.0	150.0	6	
68	Devarachikkanahalli	8 Bedroom	1350.0	7.0	85.0	8	


```
70          other  3 Bedroom      500.0   3.0  100.0    3
```

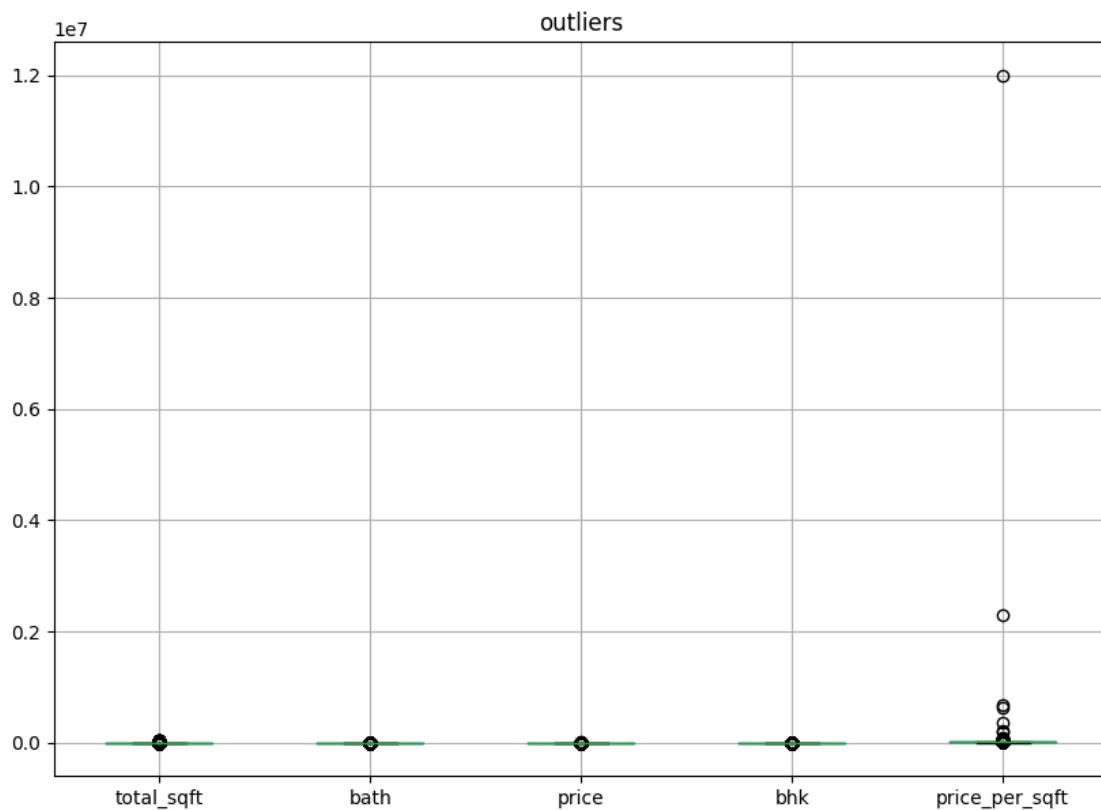
```
    price_per_sqft
9      36274.509804
45     33333.333333
58     10660.980810
68       6296.296296
70      20000.000000
```

```
[46]: df5.shape
```

```
[46]: (13246, 7)
```

```
[47]: import matplotlib.pyplot as plt

df5.boxplot(figsize=(10,7), vert = True)
plt.title("outliers")
plt.show()
```



```
[48]: df6 = df5[~(df5['total_sqft']/df5['bhk']<300)]
df6.shape
```

```
[48]: (12502, 7)
```

```
[49]: df6['price_per_sqft'].describe()
```

```
[49]: count      12456.000000
      mean       6308.502826
      std       4168.127339
      min        267.829813
      25%       4210.526316
      50%       5294.117647
      75%       6916.666667
      max      176470.588235
      Name: price_per_sqft, dtype: float64
```

```
[50]: df6.describe()
```

```
[50]:
```

	total_sqft	bath	price	bhk	price_per_sqft
count	12456.000000	12502.000000	12502.000000	12502.000000	12456.000000
mean	1590.189927	2.564790	111.311915	2.650696	6308.502826
std	1260.404795	1.084946	152.089966	0.981698	4168.127339
min	300.000000	1.000000	9.000000	1.000000	267.829813
25%	1115.000000	2.000000	49.000000	2.000000	4210.526316
50%	1300.000000	2.000000	70.000000	3.000000	5294.117647
75%	1700.000000	3.000000	115.000000	3.000000	6916.666667
max	52272.000000	16.000000	3600.000000	16.000000	176470.588235

```
[51]: def remove_pps_outliers (df):
      df_out = pd.DataFrame()
      for key, subdf in df.groupby ('location'):
          m= np.mean(subdf.price_per_sqft)
          st = np.std(subdf.price_per_sqft)
          reduced_df = subdf[(subdf.price_per_sqft>(m-st)) & (subdf.
      ↪price_per_sqft<=(m+st))]
          df_out = pd.concat([ df_out, reduced_df], ignore_index=True)
      return df_out

      df7 = remove_pps_outliers(df6)
      df7.shape
```

```
[51]: (10241, 7)
```

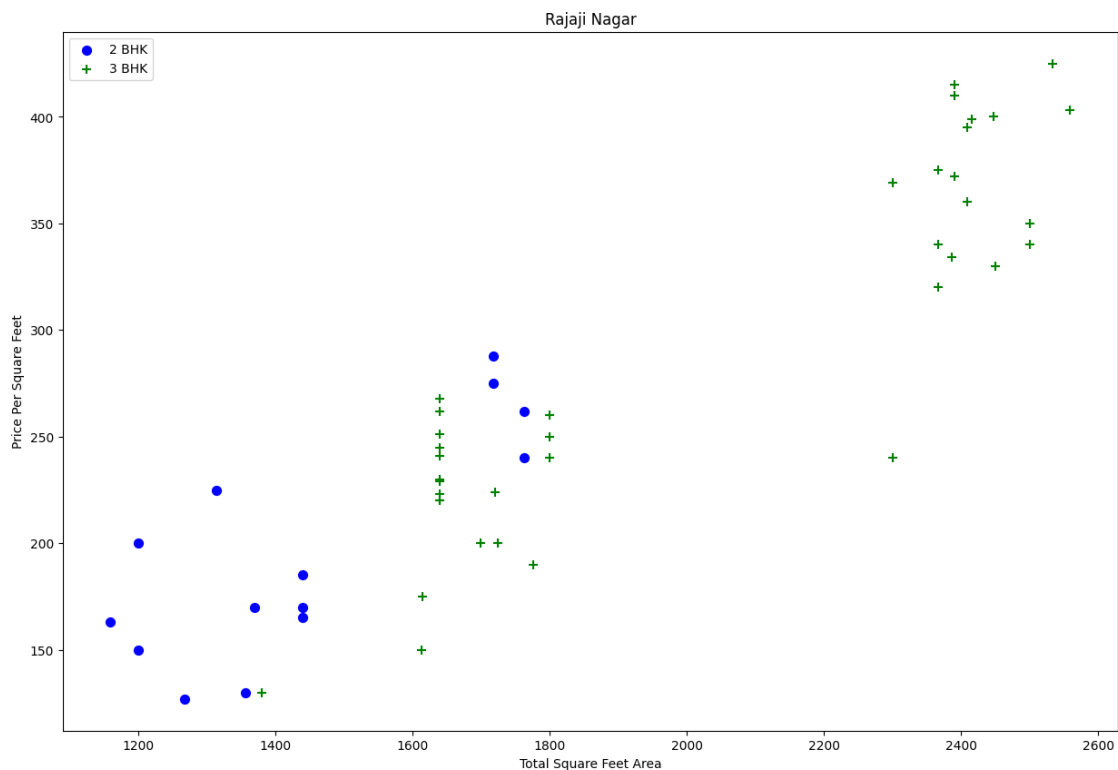
```
[52]: import matplotlib
      import matplotlib.pyplot as plt
      def plot_scatter_chart (df,location):
          bhk2 = df[(df.location==location)&(df.bhk==2)]
          bhk3 = df[(df.location==location)&(df.bhk==3)]
          matplotlib.rcParams['figure.figsize']=(15,10)
```

```

plt.scatter(bhk2.total_sqft, bhk2.price, color='blue', label='2 BHK', s=50)
plt.scatter(bhk3.total_sqft, bhk3.price, marker='+', color='green', label='3
↳BHK', s=50)
plt.xlabel("Total Square Feet Area")
plt.ylabel("Price Per Square Feet")
plt.title(location)
plt.legend()

plot_scatter_chart(df7, "Rajaji Nagar")
plt.show()

```



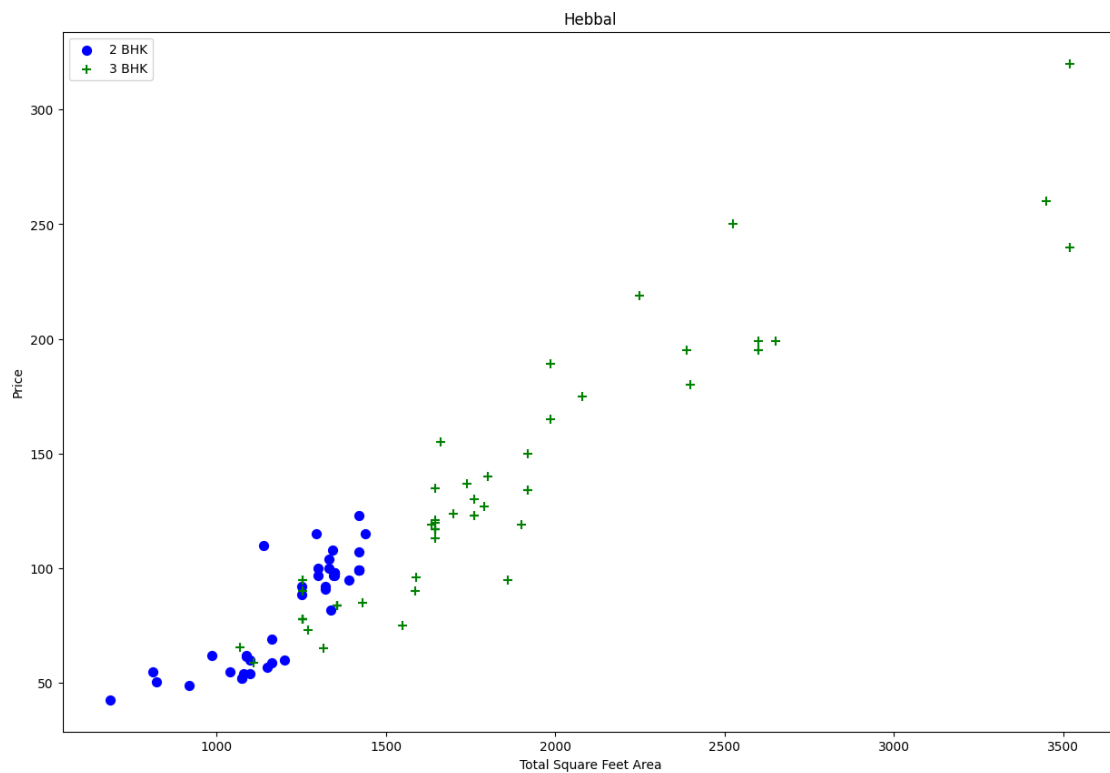
```

[53]: import matplotlib
import matplotlib.pyplot as plt
def plot_scatter_chart (df,location):
    bhk2 = df[(df.location==location)&(df.bhk==2)]
    bhk3 = df[(df.location==location)&(df.bhk==3)]
    matplotlib.rcParams['figure.figsize']=(15,10)
    plt.scatter(bhk2.total_sqft, bhk2.price, color='blue', label='2 BHK', s=50)
    plt.scatter(bhk3.total_sqft, bhk3.price, marker='+', color='green', label='3
↳BHK', s=50)
    plt.xlabel("Total Square Feet Area")

```

```
plt.ylabel("Price")
plt.title(location)
plt.legend()
```

```
plot_scatter_chart(df7, "Hebbal")
plt.show()
```



```
[54]: #remove those 2 BHK apartments whose price_per_sqft is less than mean_
      ↪ price_per_sqft of i BHK apartment
```

```
[55]: def remove_bhk_outliers(df):
      exclude_indices = np.array([])
      for location, location_df in df.groupby('location'):
          bhk_stats = {}
          for bhk, bhk_df in location_df.groupby('bhk'):
              bhk_stats[bhk] = {
                  'mean': np.mean(bhk_df.price_per_sqft ),
                  'std' : np.std(bhk_df.price_per_sqft),
                  'count' : bhk_df.shape[0]
              }
          for bhk, bhk_df in location_df.groupby('bhk'):
```

```

        stats = bhk_stats.get(bhk-1)
        if stats and stats['count']>5:
            exclude_indices= np.append(exclude_indices,bhk_df[bhk_df.
↪price_per_sqft<(stats['mean']))].index.values )
        return df.drop(exclude_indices, axis='index')

df8 = remove_bhk_outliers(df7)
df8.shape

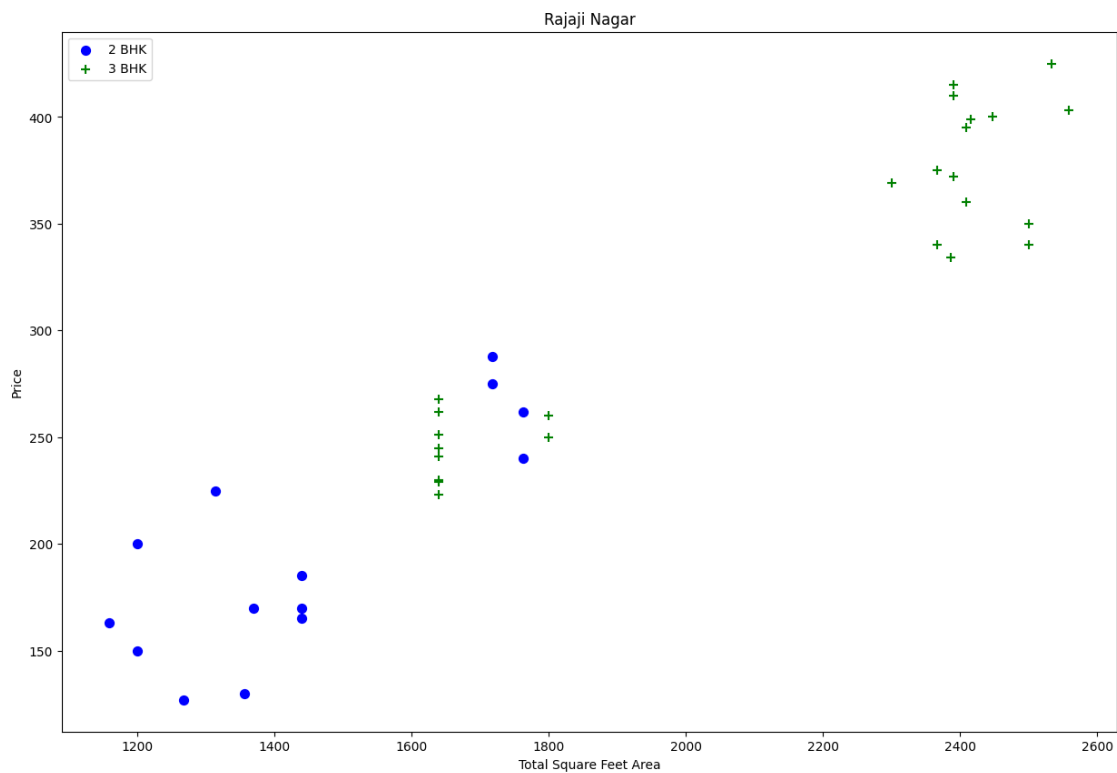
```

[55]: (7329, 7)

```

[56]: plot_scatter_chart(df8, "Rajaji Nagar")
plt.show()

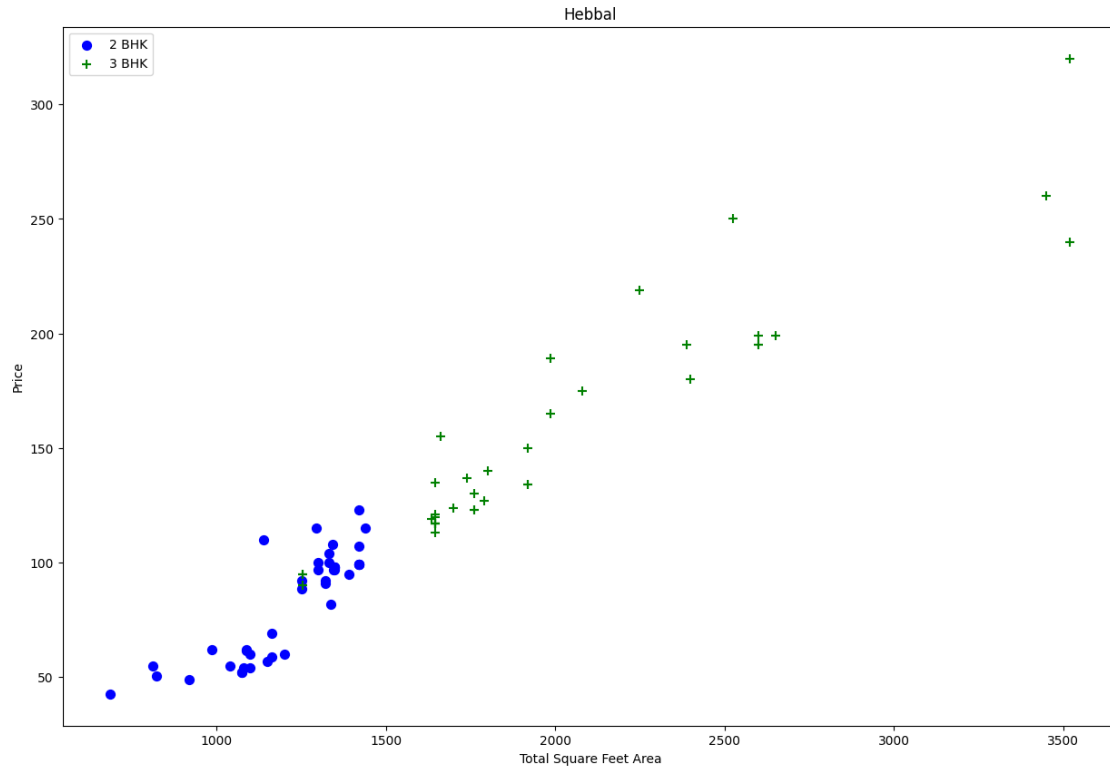
```



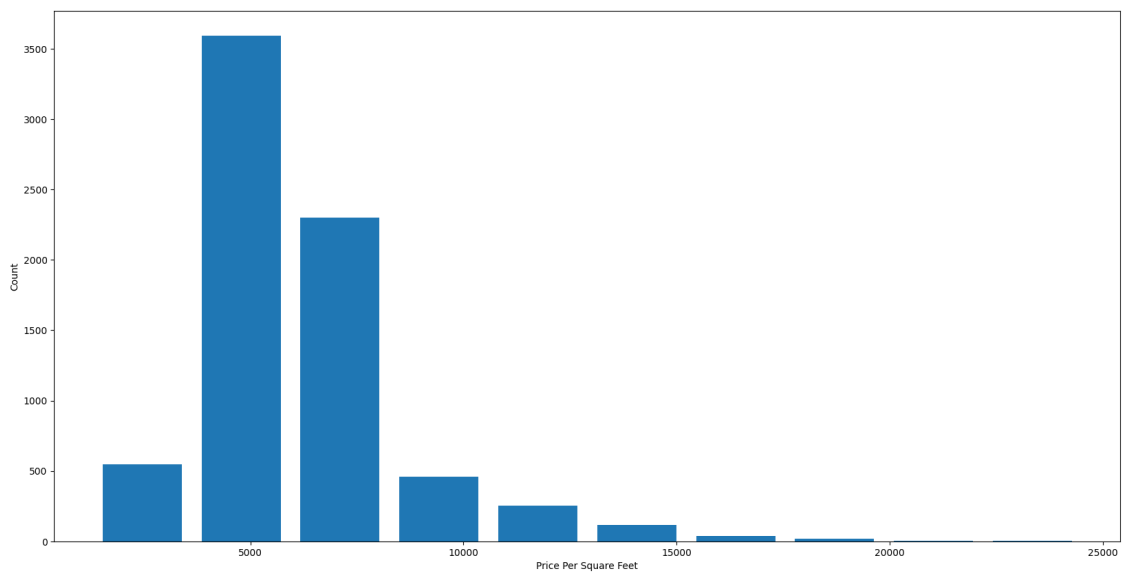
```

[57]: plot_scatter_chart(df8, "Hebbal")
plt.show()

```



```
[58]: matplotlib.rcParams["figure.figsize"]=(20,10)
plt.hist(df8.price_per_sqft, rwidth=0.8)
plt.xlabel("Price Per Square Feet")
plt.ylabel("Count")
plt.show()
```



```
[59]: #dataset has normal distribution
```

```
[60]: df8.bath.unique()
```

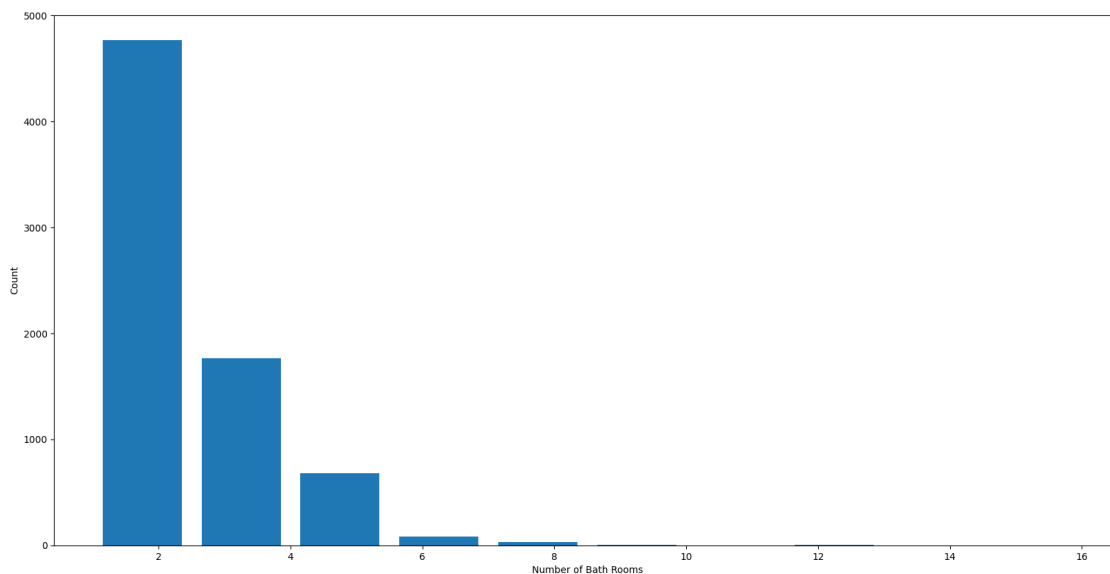
```
[60]: array([ 4.,  3.,  2.,  5.,  8.,  1.,  6.,  7.,  9., 12., 16., 13.]
```

```
[61]: df8[df8['bath']>10]
```

```
[61]:
```

	location	size	total_sqft	bath	price	bhk	price_per_sqft
5277	Neeladri Nagar	10 BHK	4000.0	12.0	160.0	10	4000.000000
8486	other	10 BHK	12000.0	12.0	525.0	10	4375.000000
8575	other	16 BHK	10000.0	16.0	550.0	16	5500.000000
9308	other	11 BHK	6000.0	12.0	150.0	11	2500.000000
9639	other	13 BHK	5425.0	13.0	275.0	13	5069.124424

```
[62]: matplotlib.rcParams["figure.figsize"]=(20,10)
plt.hist(df8.bath, rwidth=0.8)
plt.xlabel("Number of Bath Rooms")
plt.ylabel("Count")
plt.show()
```



```
[63]: df8[df8['bath']>df8.bhk+2]
```

```
[63]:
```

	location	size	total_sqft	bath	price	bhk	price_per_sqft
1626	Chikkabanavar	4 Bedroom	2460.0	7.0	80.0	4	3252.032520
5238	Nagasandra	4 Bedroom	7000.0	8.0	450.0	4	6428.571429

6711	Thanisandra	3 BHK	1806.0	6.0	116.0	3	6423.034330
8411	other	6 BHK	11338.0	9.0	1000.0	6	8819.897689

```
[64]: df9 = df8[df8['bath']<df8.bhk+2]
df9.shape
```

```
[64]: (7251, 7)
```

```
[65]: df10 = df9.drop(["size", "price_per_sqft"], axis='columns')
df10.head()
```

```
[65]:
```

		location	total_sqft	bath	price	bhk
0	1st Block	Jayanagar	2850.0	4.0	428.0	4
1	1st Block	Jayanagar	1630.0	3.0	194.0	3
2	1st Block	Jayanagar	1875.0	2.0	235.0	3
3	1st Block	Jayanagar	1200.0	2.0	130.0	3
4	1st Block	Jayanagar	1235.0	2.0	148.0	2

```
[66]: # create machine learning model
```

```
[67]: # to turn this categorical column into numbers we using one hot encoding
```

```
[73]: dummies = pd.get_dummies(df10['location'])
dummies.head()
```

```
[73]:
```

	1st Block	Jayanagar	1st Phase	JP Nagar	2nd Phase	Judicial Layout	\
0		True		False		False	
1		True		False		False	
2		True		False		False	
3		True		False		False	
4		True		False		False	

	2nd Stage	Nagarbhavi	5th Block	Hbr Layout	5th Phase	JP Nagar	\
0		False		False		False	
1		False		False		False	
2		False		False		False	
3		False		False		False	
4		False		False		False	

	6th Phase	JP Nagar	7th Phase	JP Nagar	8th Phase	JP Nagar	\
0		False		False		False	
1		False		False		False	
2		False		False		False	
3		False		False		False	
4		False		False		False	

	9th Phase	JP Nagar	...	Vishveshwarya Layout	Vishwapriya Layout	\
0		False	...	False		False

1	False	...	False	False
2	False	...	False	False
3	False	...	False	False
4	False	...	False	False

	Vittasandra	Whitefield	Yelachenahalli	Yelahanka	Yelahanka New Town	\
0	False	False	False	False	False	
1	False	False	False	False	False	
2	False	False	False	False	False	
3	False	False	False	False	False	
4	False	False	False	False	False	

	Yelenahalli	Yeshwanthpur	other
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False

[5 rows x 242 columns]

```
[77]: df11 = pd.concat([df10, dummies],axis = 'columns')
df11.head()
```

```
[77]:
```

	location	total_sqft	bath	price	bhk	1st Block Jayanagar	\
0	1st Block Jayanagar	2850.0	4.0	428.0	4	True	
1	1st Block Jayanagar	1630.0	3.0	194.0	3	True	
2	1st Block Jayanagar	1875.0	2.0	235.0	3	True	
3	1st Block Jayanagar	1200.0	2.0	130.0	3	True	
4	1st Block Jayanagar	1235.0	2.0	148.0	2	True	

	1st Phase JP Nagar	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	5th Block Hbr Layout	...	Vishveshwarya Layout	Vishwapriya Layout	\
0	False	...	False	False	
1	False	...	False	False	
2	False	...	False	False	
3	False	...	False	False	
4	False	...	False	False	

	Vittasandra	Whitefield	Yelachenahalli	Yelahanka	Yelahanka New Town	\
0	False	False	False	False	False	

1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False

	Yelenahalli	Yeshwanthpur	other
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False

[5 rows x 247 columns]

```
[81]: df11= df11.drop('other', axis=1)
df11.head()
```

```
[81]:
```

	location	total_sqft	bath	price	bhk	1st Block Jayanagar	\
0	1st Block Jayanagar	2850.0	4.0	428.0	4	True	
1	1st Block Jayanagar	1630.0	3.0	194.0	3	True	
2	1st Block Jayanagar	1875.0	2.0	235.0	3	True	
3	1st Block Jayanagar	1200.0	2.0	130.0	3	True	
4	1st Block Jayanagar	1235.0	2.0	148.0	2	True	

	1st Phase JP Nagar	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	5th Block Hbr Layout	...	Vijayanagar	Vishveshwarya Layout	\
0	False	...	False	False	
1	False	...	False	False	
2	False	...	False	False	
3	False	...	False	False	
4	False	...	False	False	

	Vishwapriya Layout	Vittasandra	Whitefield	Yelachenahalli	Yelahanka	\
0	False	False	False	False	False	
1	False	False	False	False	False	
2	False	False	False	False	False	
3	False	False	False	False	False	
4	False	False	False	False	False	

	Yelahanka New Town	Yelenahalli	Yeshwanthpur
0	False	False	False

1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False

[5 rows x 246 columns]

```
[83]: df12 = df11.drop('location',axis=1)
df12.head()
```

```
[83]: total_sqft  bath  price  bhk  1st Block Jayanagar  1st Phase JP Nagar \
0      2850.0    4.0  428.0    4                True                False
1      1630.0    3.0  194.0    3                True                False
2      1875.0    2.0  235.0    3                True                False
3      1200.0    2.0  130.0    3                True                False
4      1235.0    2.0  148.0    2                True                False
```

	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	5th Block Hbr Layout	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	5th Phase JP Nagar	...	Vijayanagar	Vishveshwarya Layout	\
0	False	...	False	False	
1	False	...	False	False	
2	False	...	False	False	
3	False	...	False	False	
4	False	...	False	False	

	Vishwapriya Layout	Vittasandra	Whitefield	Yelachenahalli	Yelahanka	\
0	False	False	False	False	False	
1	False	False	False	False	False	
2	False	False	False	False	False	
3	False	False	False	False	False	
4	False	False	False	False	False	

	Yelahanka New Town	Yelenahalli	Yeshwanthpur
0	False	False	False
1	False	False	False
2	False	False	False
3	False	False	False
4	False	False	False

[5 rows x 245 columns]

```
[85]: df12.shape
```

```
[85]: (7251, 245)
```

```
[87]: # now creating the independant variable
```

```
[89]: X = df12.drop('price', axis=1)
X.head()
```

```
[89]:
```

	total_sqft	bath	bhk	1st Block Jayanagar	1st Phase JP Nagar	\
0	2850.0	4.0	4	True	False	
1	1630.0	3.0	3	True	False	
2	1875.0	2.0	3	True	False	
3	1200.0	2.0	3	True	False	
4	1235.0	2.0	2	True	False	

	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	5th Block Hbr Layout	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	5th Phase JP Nagar	6th Phase JP Nagar	... Vijayanagar	\
0	False	False	False	
1	False	False	False	
2	False	False	False	
3	False	False	False	
4	False	False	False	

	Vishveshwarya Layout	Vishwapriya Layout	Vittasandra	Whitefield	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	

	Yelachenahalli	Yelahanka	Yelahanka New Town	Yelenahalli	Yeshwanthpur
0	False	False	False	False	False
1	False	False	False	False	False
2	False	False	False	False	False
3	False	False	False	False	False
4	False	False	False	False	False

```
[5 rows x 244 columns]
```

```
[91]: y = df12['price']
      y.head()
```

```
[91]: 0    428.0
      1    194.0
      2    235.0
      3    130.0
      4    148.0
      Name: price, dtype: float64
```

```
[95]: from sklearn.model_selection import train_test_split
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2,
      ↪ random_state=10)
```

```
[101]: from sklearn.linear_model import LinearRegression
      lr_clf = LinearRegression()
      lr_clf.fit(X_train, y_train)
      lr_clf.score(X_test, y_test)
```

```
[101]: 0.8452277697874369
```

```
[105]: from sklearn.model_selection import ShuffleSplit
      from sklearn.model_selection import cross_val_score

      cv = ShuffleSplit(n_splits=5, test_size=0.2, random_state=0)

      cross_val_score(LinearRegression(), X, y, cv=cv)
```

```
[105]: array([0.82430186, 0.77166234, 0.85089567, 0.80837764, 0.83653286])
```

```
[107]: # we getting score more than 80 percent, now we will do hyper parameter tuning
```

```
[128]: # from sklearn.model_selection import GridSearchCV
      # from sklearn.linear_model import Lasso
      # from sklearn.tree import DecisionTreeRegressor

      # def find_best_model_using_grid_search_cv(X, y):
      #     algos = {
      #         'linear_regression': {
      #             'model': LinearRegression(),
      #             'params': {
      #                 'normalize': [True, False]
      #             }
      #         },
      #         'lasso': {
      #             'model': Lasso(),
      #             'params': {
      #                 'alpha': [1, 2],
```

```

#         'selection' : ['random', 'cyclic']
#     }
# },
#     'decision tree' :{
#         'model' : DecisionTreeRegressor(),
#         'params' :{
#             'criterion' : ['mse', 'friedman_mse'],
#             'splitter' : ['best', 'random']
#         }
#     }
# }

# scores = []
# cv = ShuffleSplit(n_splits=5, test_size=0.2, random_state=0)
# for algo_name, config in algos.items():
#     gs = GridSearchCV(config['model'], config['params'],
# ↪cv=cv, return_train_score=False)
#     gs.fit(X,y)
#     scores.append({
#         'model' : algo_name,
#         'best_score':gs.best_score_,
#         'best_params':gs.best_params_
#     })
# return pd.DataFrame(scores, columns=['model', 'best_score', 'best_params'])

# find_best_model_using_grid_search_cv(X,y)

```

```

[130]: from sklearn.model_selection import GridSearchCV, ShuffleSplit
from sklearn.linear_model import LinearRegression, Lasso
from sklearn.tree import DecisionTreeRegressor
import pandas as pd

def find_best_model_using_grid_search_cv(X, y):
    algos = {
        'linear_regression': {
            'model': LinearRegression(),
            'params': {
                'fit_intercept': [True, False]    # normalize removed in sklearn
            }
        },
        'lasso': {
            'model': Lasso(),
            'params': {
                'alpha': [1, 2],
                'selection': ['random', 'cyclic']
            }
        }
    },

```

```

        'decision_tree': {
            'model': DecisionTreeRegressor(),
            'params': {
                'criterion': ['squared_error', 'friedman_mse'],
                'splitter': ['best', 'random']
            }
        }
    }

    scores = []
    cv = ShuffleSplit(n_splits=5, test_size=0.2, random_state=0)

    for algo_name, config in algos.items():
        gs = GridSearchCV(config['model'], config['params'], cv=cv,
        ↪return_train_score=False)
        gs.fit(X, y)
        scores.append({
            'model': algo_name,
            'best_score': gs.best_score_,
            'best_params': gs.best_params_
        })

    return pd.DataFrame(scores, columns=['model', 'best_score', 'best_params'])
find_best_model_using_grid_search_cv(X,y)

```

```

[130]:
      model  best_score \
0  linear_regression    0.819001
1           lasso      0.687429
2  decision_tree      0.728380

      best_params
0  {'fit_intercept': False}
1  {'alpha': 1, 'selection': 'cyclic'}
2  {'criterion': 'friedman_mse', 'splitter': 'best'}

```

```

[131]: #so best one is linear regression

```

```

[136]: X.columns

```

```

[136]: Index(['total_sqft', 'bath', 'bhk', '1st Block Jayanagar',
            '1st Phase JP Nagar', '2nd Phase Judicial Layout',
            '2nd Stage Nagarbhavi', '5th Block Hbr Layout', '5th Phase JP Nagar',
            '6th Phase JP Nagar',
            ...,
            'Vijayanagar', 'Vishveshwarya Layout', 'Vishwapriya Layout',
            'Vittasandra', 'Whitefield', 'Yelachenahalli', 'Yelahanka',
            'Yelahanka New Town', 'Yelenahalli', 'Yeshwanthpur'],

```

```
dtype='object', length=244)
```

```
[138]: np.where(X.columns=="2nd Phase Judicial Layout")[0][0]
```

```
[138]: 5
```

```
[140]: def predict_price(location, sqft, bath, bhk):  
    loc_index = np.where(X.columns == location)[0][0]  
  
    x = np.zeros(len(X.columns))  
    x[0] = sqft  
    x[1] = bath  
    x[2] = bhk  
    if loc_index>=0:  
        x[loc_index] = 1  
  
    return lr_clf.predict([x])[0]
```

```
[142]: predict_price("1st Phase JP Nagar", 1000,2,2)
```

```
C:\Users\dell\anaconda3\envs\tf2\lib\site-  
packages\sklearn\utils\validation.py:2749: UserWarning: X does not have valid  
feature names, but LinearRegression was fitted with feature names  
warnings.warn(
```

```
[142]: 83.49904677201745
```

```
[144]: #we get 83 lacks
```

```
[150]: predict_price("1st Phase JP Nagar", 1000,3,3)
```

```
C:\Users\dell\anaconda3\envs\tf2\lib\site-  
packages\sklearn\utils\validation.py:2749: UserWarning: X does not have valid  
feature names, but LinearRegression was fitted with feature names  
warnings.warn(
```

```
[150]: 86.80519395228475
```

```
[162]: predict_price("Indira Nagar", 1000,3,4)
```

```
C:\Users\dell\anaconda3\envs\tf2\lib\site-  
packages\sklearn\utils\validation.py:2749: UserWarning: X does not have valid  
feature names, but LinearRegression was fitted with feature names  
warnings.warn(
```

```
[162]: 182.81142425609204
```

```
[164]: df1.head()
```



```
[164]:
```

		area_type	availability	location	size \
0	Super built-up	Area	19-Dec	Electronic City Phase II	2 BHK
1	Plot	Area	Ready To Move	Chikka Tirupathi	4 Bedroom
2	Built-up	Area	Ready To Move	Uttarahalli	3 BHK
3	Super built-up	Area	Ready To Move	Lingadheeranahalli	3 BHK
4	Super built-up	Area	Ready To Move	Kothanur	2 BHK

	society	total_sqft	bath	balcony	price
0	Coomee	1056	2.0	1.0	39.07
1	Theanmp	2600	5.0	3.0	120.00
2	NaN	1440	2.0	3.0	62.00
3	Soiewre	1521	3.0	1.0	95.00
4	NaN	1200	2.0	1.0	51.00

```
[166]: #now export this model into a pickle file
```

```
[170]: import pickle
with open('bangaluru_home_price_model.pickle', 'wb') as f:
    pickle.dump(lr_clf, f)
```

```
[172]: import json
columns={
    'data_columns' : [col.lower() for col in X.columns]
}
with open("columns.json", "w") as f:
    f.write(json.dumps(columns))
```

```
[ ]:
```