

Real Estate Price Prediction

```
In [1]: import numpy as np
import pandas as pd
from matplotlib import pyplot as plt
%matplotlib inline
import matplotlib
matplotlib.rcParams['figure.figsize']=(20,10)
```

```
In [2]: df1=pd.read_csv(r"C:\Users\maram\Downloads\Bengaluru_House_Data.csv")
df1.head(2)
```

```
Out[2]:
```

	area_type	availability	location	size	society	total_sqft	bath	balcony	price
0	Super built-up Area	19-Dec	Electronic City Phase II	2 BHK	Coomee	1056	2.0	1.0	39.07
1	Plot Area	Ready To Move	Chikka Tirupathi	4 Bedroom	Theanmp	2600	5.0	3.0	120.00

```
In [3]: df1.shape
```

```
Out[3]: (13320, 9)
```

```
In [4]: df1.groupby('area_type')['area_type'].agg('count')
```

```
Out[4]:
```

area_type	
Built-up Area	2418
Carpet Area	87
Plot Area	2025
Super built-up Area	8790

Name: area_type, dtype: int64

```
In [5]: #df1['area_type'].value_counts()
```

```
In [6]: df1.columns.values
```

```
Out[6]: array(['area_type', 'availability', 'location', 'size', 'society',
        'total_sqft', 'bath', 'balcony', 'price'], dtype=object)
```

```
In [7]: df1.drop(['area_type','availability','society','balcony'],axis=1,inplace=True)
```

```
In [8]: df1.head(3)
```

```
Out[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

```
In [9]: df1.isnull().sum()
```

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

```
In [10]: df1.dropna(inplace=True)
```

```
In [11]: df1.isnull().sum()
```

```
Out[11]: location      0
        size          0
        total_sqft     0
        bath          0
        price         0
        dtype: int64
```

```
In [12]: df1['size'].unique()
```

```
Out[12]: 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)
```

```
In [13]: df1['bhk']=df1['size'].apply(lambda x :int(x.split(' ')[0]))
        # The lambda function splits each element (which is assumed to be a string) by a space
```

```
In [14]: df1['bhk'].unique()
```

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

```
In [15]: df1.head()
```

```
Out[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

```
In [16]: #df1['bhk1'] = df1['size'].str.replace(r'\D', '', regex=True)
```

```
In [17]: #df1['bhk1'].unique()
```

```
In [18]: df1[df1['bhk']>20]
```

Out[18]:

	location	size	total_sqft	bath	price	bhk
1718	2Electronic City Phase II	27 BHK	8000	27.0	230.0	27
4684	Munnekollal	43 Bedroom	2400	40.0	660.0	43

In [19]: *# you can not have 43 bedrooms while u have total sqft = 2400*

In [20]: `df1.total_sqft.unique()`

Out[20]: `array(['1056', '2600', '1440', ..., '1133 - 1384', '774', '4689'],
dtype=object)`

In [21]:

```
def is_float(x):
    try: #Im trying to convert x to float
        float(x)
    except: #if its on form of interval or any return false
        return False
    return True
```

In [22]: `df1[~df1['total_sqft'].apply(is_float)].head(10)`

Out[22]:

	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

In [23]:

```
def tokens_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
```

In [24]: `tokens_sqft_to_num('2011')`

Out[24]: 2011.0

In [25]: `tokens_sqft_to_num('235 - 243')`

Out[25]: 356.5

In [26]: `df2=df1.copy()`

In [27]: `df2['total_sqft']=df1['total_sqft'].apply(tokens_sqft_to_num)`

In [28]: `df2.head()`

Out[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
3	Lingadheeranahalli	3 BHK	1521.0	3.0	95.00	3
4	Kothanur	2 BHK	1200.0	2.0	51.00	2

Feature Engineering

In [29]: `df3=df2.copy()`

In [30]: `df3['price_per_sqft']=df3['price']*100000/df3['total_sqft']`
`df3.head(3)`

Out[30]:

	location	size	total_sqft	bath	price	bhk	price_per_sqft
0	Electronic City Phase II	2 BHK	1056.0	2.0	39.07	2	3699.810606
1	Chikka Tirupathi	4 Bedroom	2600.0	5.0	120.00	4	4615.384615
2	Uttarahalli	3 BHK	1440.0	2.0	62.00	3	4305.555556

In [31]: `len(df3.location.unique())`

Out[31]: 1304

Dimensionality reduction

In [32]: `df3.location=df3.location.apply(lambda x:x.strip())`

In [33]: `location_stats=df3.groupby('location')['location'].agg('count').sort_values(ascending=False)`
`location_stats.head(10)`

```
Out[33]: location
Whitefield      535
Sarjapur Road   392
Electronic City 304
Kanakpura Road  266
Thanisandra     236
Yelahanka       210
Uttarahalli     186
Hebbal          176
Marathahalli    175
Raja Rajeshwari Nagar 171
Name: location, dtype: int64
```

```
In [34]: len(location_stats[location_stats<=10])
```

```
Out[34]: 1052
```

```
In [35]: location_stats_less_than_10=location_stats[location_stats<=10]
location_stats_less_than_10
```

```
Out[35]: 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
```

```
In [36]: len(df3.location.unique())
```

```
Out[36]: 1293
```

```
In [37]: df3.location=df3.location.apply(lambda x: 'other' if x in location_stats_less_than_10 else x)
len(df3.location.unique())
```

```
Out[37]: 242
```

Outlier detuction/Removal

```
In [38]: df3.columns.values
```

```
Out[38]: array(['location', 'size', 'total_sqft', 'bath', 'price', 'bhk',
        'price_per_sqft'], dtype=object)
```

```
In [39]: df3[df3.total_sqft/df3.bhk<300].head()
```

Out[39]:

	location	size	total_sqft	bath	price	bhk	price_per_sqft
9	other	6 Bedroom	1020.0	6.0	370.0	6	36274.509804
45	HSR Layout	8 Bedroom	600.0	9.0	200.0	8	33333.333333
58	Murugeshpalya	6 Bedroom	1407.0	4.0	150.0	6	10660.980810
68	Devarachikkanahalli	8 Bedroom	1350.0	7.0	85.0	8	6296.296296
70	other	3 Bedroom	500.0	3.0	100.0	3	20000.000000

In [40]: df3.shape

Out[40]: (13246, 7)

In [41]: df4=df3[~(df3.total_sqft/df3.bhk<300)]

In [42]: df4.shape

Out[42]: (12502, 7)

In [43]: df4

Out[43]:

	location	size	total_sqft	bath	price	bhk	price_per_sqft
0	Electronic City Phase II	2 BHK	1056.0	2.0	39.07	2	3699.810606
1	Chikka Tirupathi	4 Bedroom	2600.0	5.0	120.00	4	4615.384615
2	Uttarahalli	3 BHK	1440.0	2.0	62.00	3	4305.555556
3	Lingadheeranahalli	3 BHK	1521.0	3.0	95.00	3	6245.890861
4	Kothanur	2 BHK	1200.0	2.0	51.00	2	4250.000000
...
13315	Whitefield	5 Bedroom	3453.0	4.0	231.00	5	6689.834926
13316	other	4 BHK	3600.0	5.0	400.00	4	11111.111111
13317	Raja Rajeshwari Nagar	2 BHK	1141.0	2.0	60.00	2	5258.545136
13318	Padmanabhanagar	4 BHK	4689.0	4.0	488.00	4	10407.336319
13319	Doddathoguru	1 BHK	550.0	1.0	17.00	1	3090.909091

12502 rows × 7 columns

In [44]: df4.price_per_sqft.describe()

```
Out[44]: count      12456.000000
         mean       6290.299983
         std        4175.619211
         min        267.829813
         25%        4186.725844
         50%        5281.690141
         75%        6904.652015
         max       176470.588235
         Name: price_per_sqft, dtype: float64
```

Check above data points. We have 6 bhk apartment with 1020 sqft. Another one is 8 bhk and total sqft is 600. These are clear data errors that can be removed safely

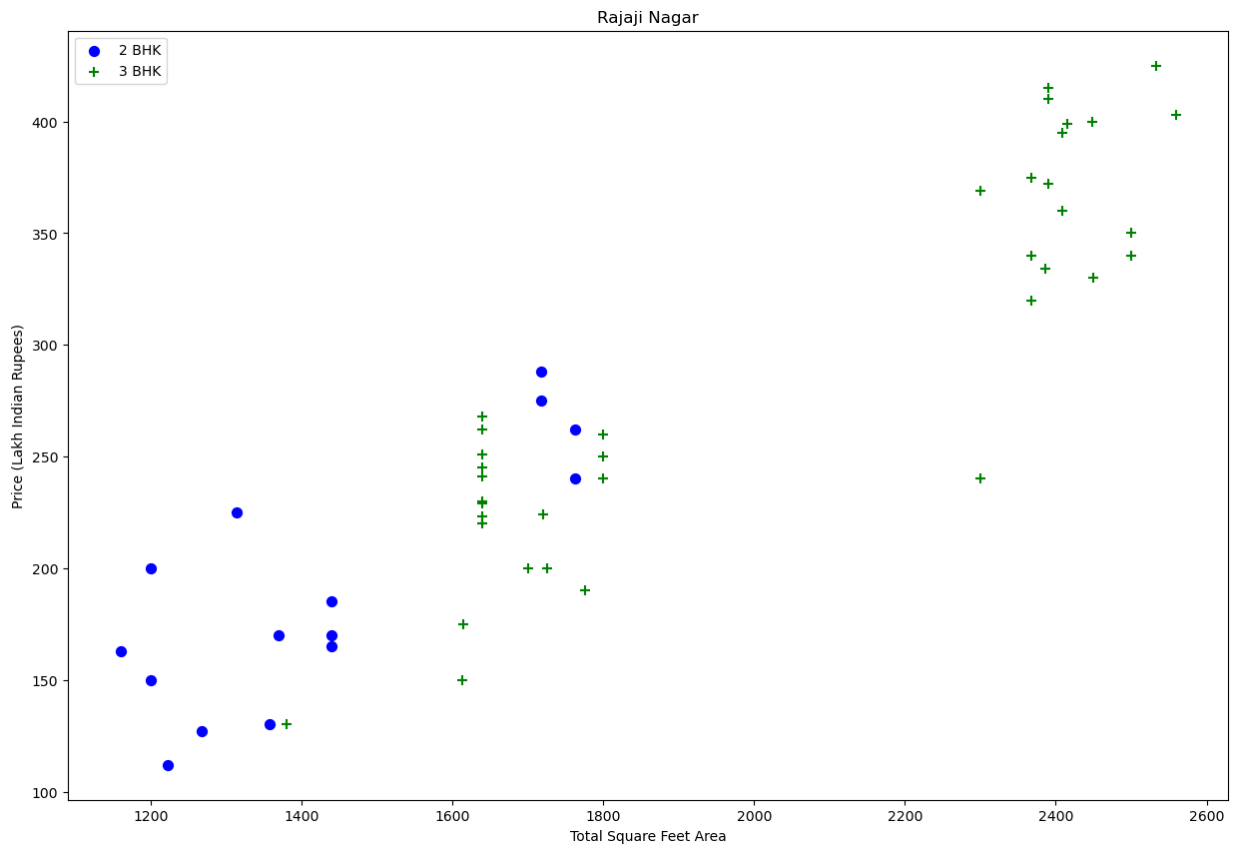
```
In [45]: def remove_pps_outliers(df):
         df_out = pd.DataFrame()
         for key, subdf in df.groupby('location'):
             m = np.mean(subdf.price_per_sqft)
             st = np.std(subdf.price_per_sqft)
             reduced_df = subdf[(subdf.price_per_sqft>(m-st)) & (subdf.price_per_sqft<=(m+st))]
             df_out = pd.concat([df_out,reduced_df],ignore_index=True)
         return df_out
         df5 = remove_pps_outliers(df4)
         df5.shape
```

```
Out[45]: (10231, 7)
```

Let's check if for a given location how does the 2 BHK and 3 BHK property prices look like

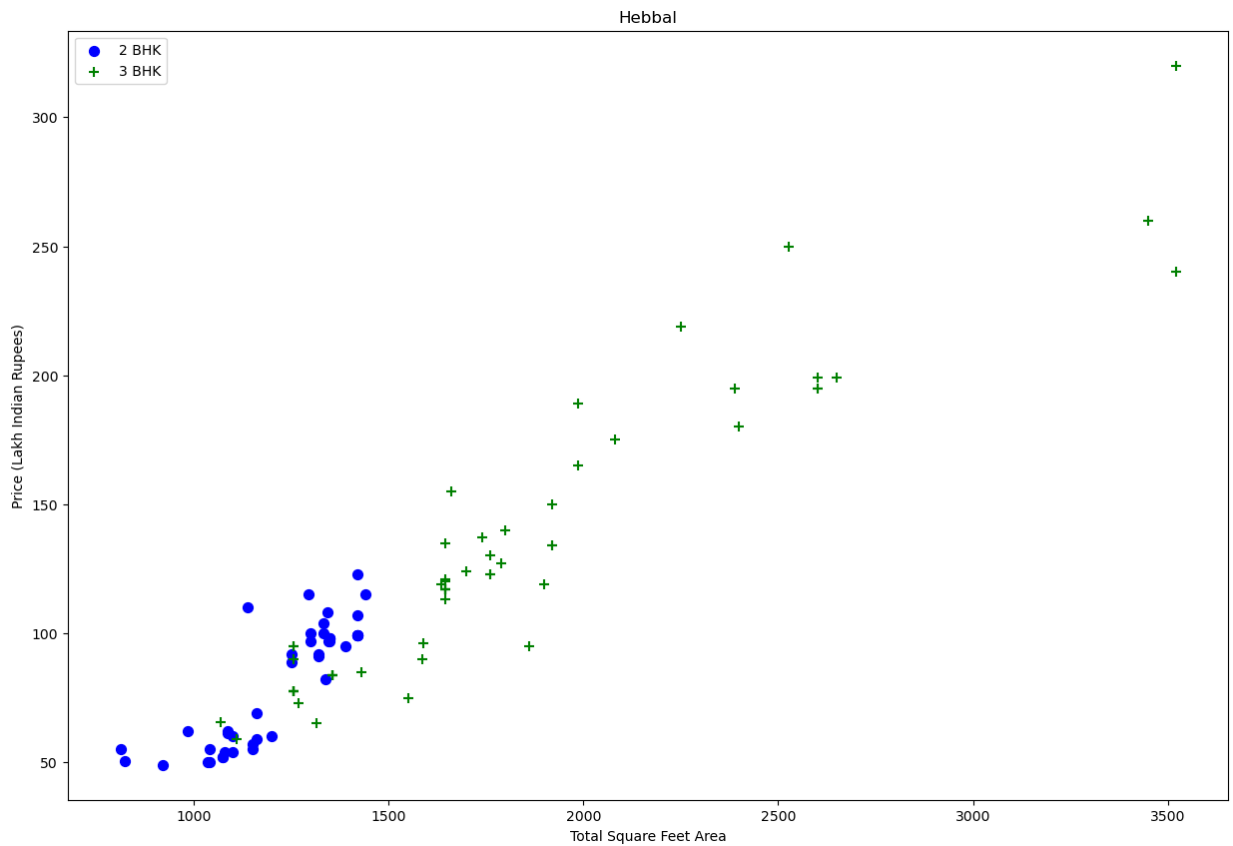
```
In [46]: def plot_scatter_chart(df,location):
         bhk2 = df[(df.location==location) & (df.bhk==2)]
         bhk3 = df[(df.location==location) & (df.bhk==3)]
         matplotlib.rcParams['figure.figsize'] = (15,10)
         plt.scatter(bhk2.total_sqft,bhk2.price,color='blue',label='2 BHK', s=50)
         plt.scatter(bhk3.total_sqft,bhk3.price,marker='+', color='green',label='3 BHK', s=50)
         plt.xlabel("Total Square Feet Area")
         plt.ylabel("Price (Lakh Indian Rupees)")
         plt.title(location)
         plt.legend()

         plot_scatter_chart(df5,"Rajaji Nagar")
```



```
In [47]: 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 (Lakh Indian Rupees)")
    plt.title(location)
    plt.legend()

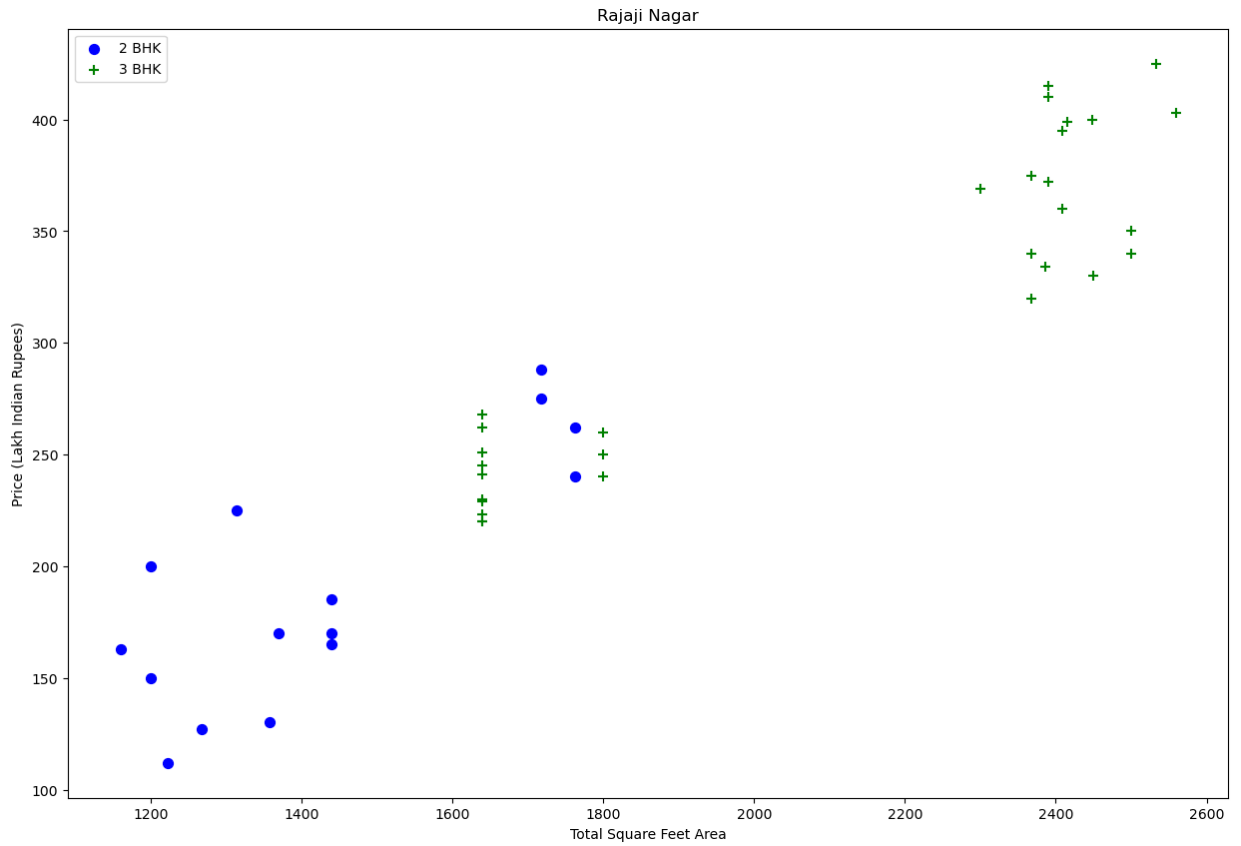
    plot_scatter_chart(df5,"Hebbal")
```

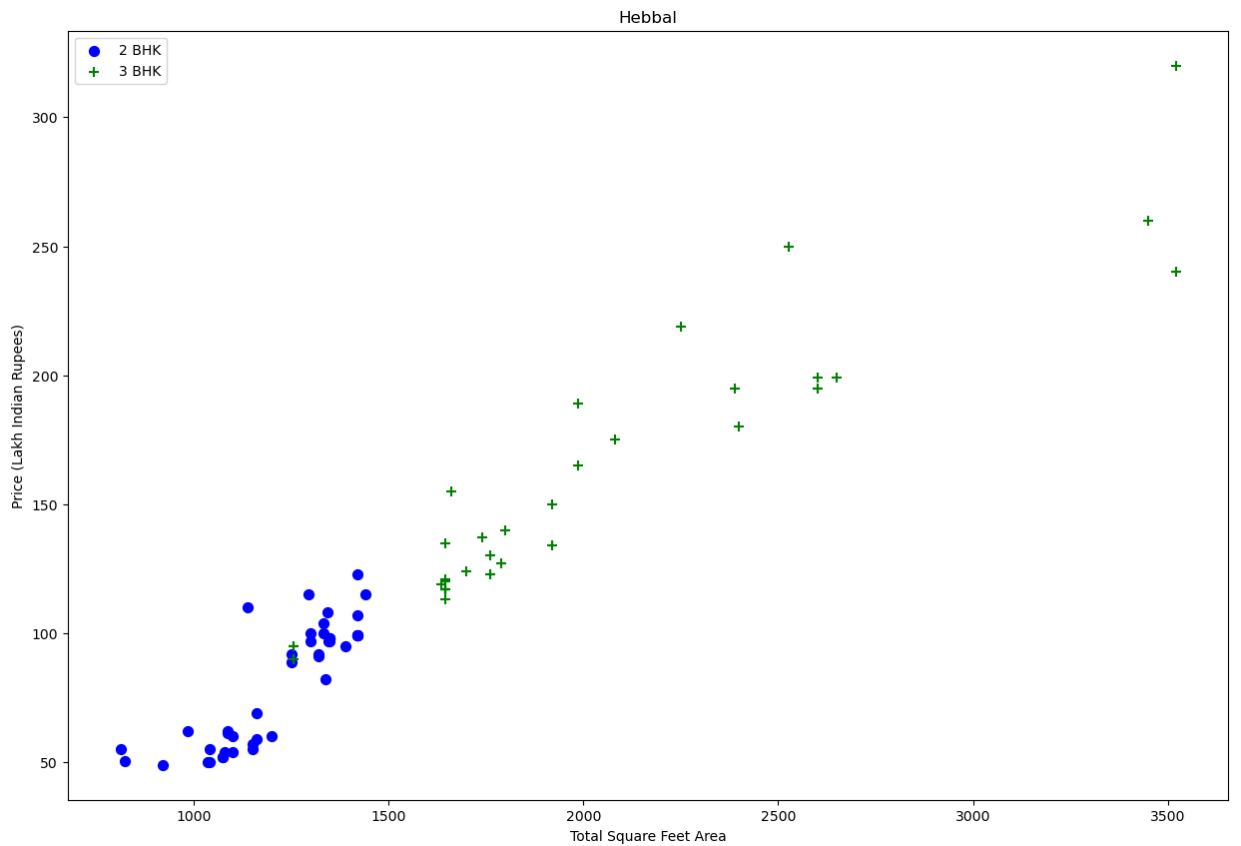
```
In [48]: def remove_bhk_outliers(df):
exclude_indices = np.array([])
for location, location_df in df.groupby('location'):
    bhk_stats = {}
    for bhk, bhk_df in location_df.groupby('bhk'):
        bhk_stats[bhk] = {
            'mean': np.mean(bhk_df.price_per_sqft),
            'std': np.std(bhk_df.price_per_sqft),
            '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['std'] * 3].index)
    return df.drop(exclude_indices, axis='index')
df6 = remove_bhk_outliers(df5)
# df8 = df7.copy()
df6.shape
```

Out[48]: (7326, 7)

```
In [49]: plot_scatter_chart(df6, "Rajaji Nagar")
```



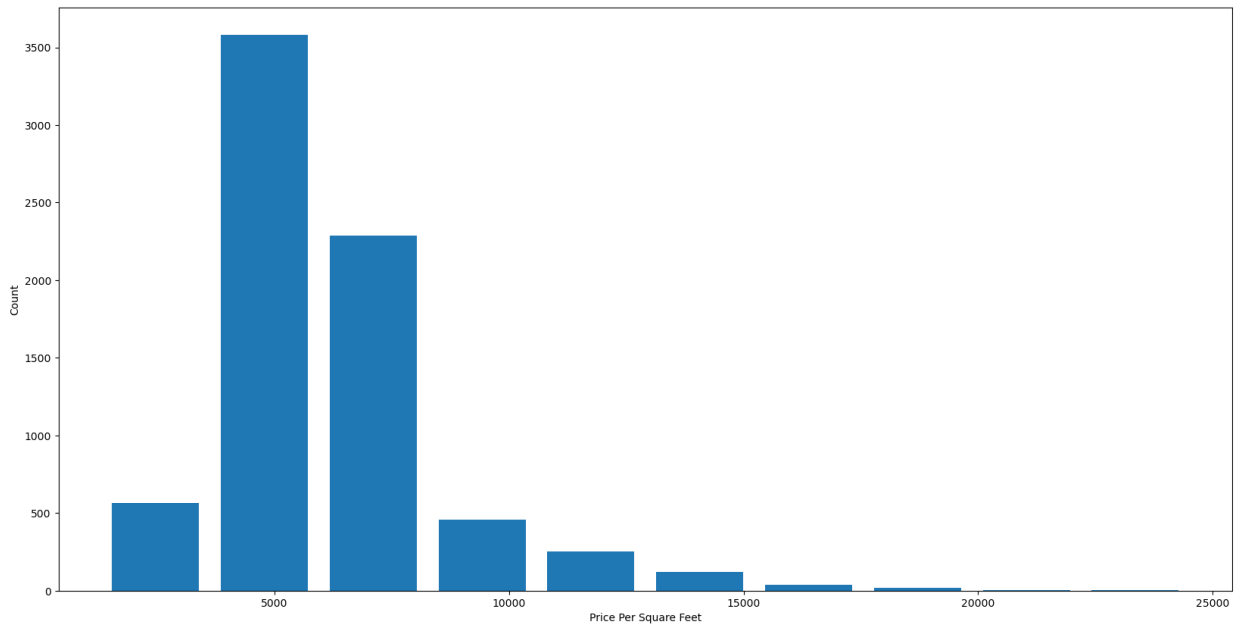
```
In [50]: plot_scatter_chart(df6, "Hebbal")
```



```
In [51]: import matplotlib
matplotlib.rcParams["figure.figsize"] = (20,10)
plt.hist(df6.price_per_sqft,rwidth=0.8)
```

```
plt.xlabel("Price Per Square Feet")  
plt.ylabel("Count")
```

Out[51]: Text(0, 0.5, 'Count')

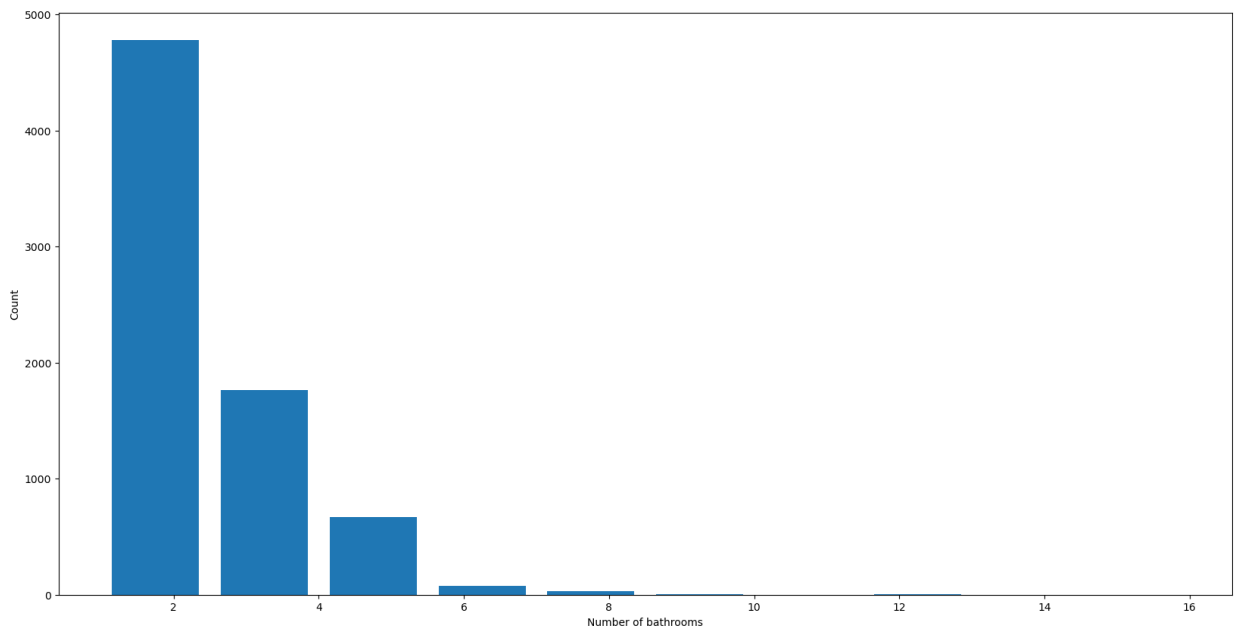


In [52]: `df6.bath.unique()`

Out[52]: array([4., 3., 2., 5., 8., 1., 6., 7., 9., 12., 16., 13.])

```
plt.hist(df6.bath,rwidth=0.8)  
plt.xlabel("Number of bathrooms")  
plt.ylabel("Count")
```

Out[53]: Text(0, 0.5, 'Count')



In [54]: `df6[df6.bath>10]`

Out[54]:

	location	size	total_sqft	bath	price	bhk	price_per_sqft
5273	Neeladri Nagar	10 BHK	4000.0	12.0	160.0	10	4000.000000
8476	other	10 BHK	12000.0	12.0	525.0	10	4375.000000
8565	other	16 BHK	10000.0	16.0	550.0	16	5500.000000
9298	other	11 BHK	6000.0	12.0	150.0	11	2500.000000
9629	other	13 BHK	5425.0	13.0	275.0	13	5069.124424

It is unusual to have 2 more bathrooms than number of bedrooms in a home

In [55]:

```
df6[df6.bath>df6.bhk+2]
```

Out[55]:

	location	size	total_sqft	bath	price	bhk	price_per_sqft
1625	Chikkabanavar	4 Bedroom	2460.0	7.0	80.0	4	3252.032520
5234	Nagasandra	4 Bedroom	7000.0	8.0	450.0	4	6428.571429
6705	Thanisandra	3 BHK	1806.0	6.0	116.0	3	6423.034330
8401	other	6 BHK	11338.0	9.0	1000.0	6	8819.897689

Again the business manager has a conversation with you (i.e. a data scientist) that if you have 4 bedroom home and even if you have bathroom in all 4 rooms plus one guest bathroom, you will have total bath = total bed + 1 max. Anything above that is an outlier or a data error and can be removed

In [56]:

```
df7 = df6[df6.bath<df6.bhk+2]
df7.shape
```

Out[56]:

```
(7252, 7)
```

Hot encoding

In [57]:

```
df8 = df7.drop(['size', 'price_per_sqft'], axis='columns')
df8.head(3)
```

Out[57]:

	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

In [58]:

```
dummies = pd.get_dummies(df8.location)
dummies.head(3)
```

Out[58]:

	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	7th Phase JP Nagar	8th Phase JP Nagar	9th Phase JP Nagar	...	Vishves
0	1	0	0	0	0	0	0	0	0	0	...	
1	1	0	0	0	0	0	0	0	0	0	...	
2	1	0	0	0	0	0	0	0	0	0	...	

3 rows × 242 columns

```
In [59]: df11 = pd.concat([df8,dummies.drop('other',axis='columns')],axis='columns')
df11.head()
```

Out[59]:

	location	total_sqft	bath	price	bhk	1st Block Jayanagar	1st Phase JP Nagar	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	5th Block Hbr Layout	...	Vijaya
0	1st Block Jayanagar	2850.0	4.0	428.0	4	1	0	0	0	0	...	
1	1st Block Jayanagar	1630.0	3.0	194.0	3	1	0	0	0	0	...	
2	1st Block Jayanagar	1875.0	2.0	235.0	3	1	0	0	0	0	...	
3	1st Block Jayanagar	1200.0	2.0	130.0	3	1	0	0	0	0	...	
4	1st Block Jayanagar	1235.0	2.0	148.0	2	1	0	0	0	0	...	

5 rows × 246 columns

```
In [60]: df12 = df11.drop('location',axis='columns')
df12.head(2)
```

Out[60]:

	total_sqft	bath	price	bhk	1st Block Jayanagar	1st Phase JP Nagar	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	5th Block Hbr Layout	5th Phase JP Nagar	...	Vijayana
0	2850.0	4.0	428.0	4	1	0	0	0	0	0	...	
1	1630.0	3.0	194.0	3	1	0	0	0	0	0	...	

2 rows × 245 columns

Build a Model Now...

In [61]: `df12.shape`

Out[61]: (7252, 245)

In [62]: `X = df12.drop(['price'],axis='columns')`
`X.head(3)`

Out[62]:

	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
0	2850.0	4.0	4	1	0	0	0	0	0	0	...	
1	1630.0	3.0	3	1	0	0	0	0	0	0	...	
2	1875.0	2.0	3	1	0	0	0	0	0	0	...	

3 rows × 244 columns

In [63]: `y = df12.price`
`y.head(3)`

Out[63]:

```
0    428.0
1    194.0
2    235.0
Name: price, dtype: float64
```

In [64]: `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)`

In [65]: `from sklearn.linear_model import LinearRegression`
`lr_clf = LinearRegression()`
`lr_clf.fit(X_train,y_train)`
`lr_clf.score(X_test,y_test)`

Out[65]: 0.8515492485425586

Use K Fold cross validation to measure accuracy of our LinearRegression model

In [66]: `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)`

Out[66]: array([0.83419256, 0.84066871, 0.85637292, 0.84326517, 0.84613103])

In [67]: `from sklearn.model_selection import GridSearchCV`
`from sklearn.linear_model import Lasso`
`from sklearn.tree import DecisionTreeRegressor`

```

def find_best_model_using_gridsearchcv(X, y):
    algos = {
        'linear_regression': {
            'model': LinearRegression(),
            'params': {
                'fit_intercept': [True, False],
                'n_jobs': [-1] # If you want to specify other parameters for LinearRe
            }
        },
        '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_gridsearchcv(X, y)

```

```

C:\ProgramData\anaconda3\lib\site-packages\sklearn\model_selection\_validation.py:37
8: FitFailedWarning:
10 fits failed out of a total of 20.
The score on these train-test partitions for these parameters will be set to nan.
If these failures are not expected, you can try to debug them by setting error_score
='raise'.

Below are more details about the failures:
-----
10 fits failed with the following error:
Traceback (most recent call last):
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\model_selection\_validation
n.py", line 686, in _fit_and_score
    estimator.fit(X_train, y_train, **fit_params)
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py", line 12
47, in fit
    super().fit(
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\tree\_classes.py", line 17
7, in fit
    self._validate_params()
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py", line 581, in _va
lvalidate_params
    validate_parameter_constraints(
  File "C:\ProgramData\anaconda3\lib\site-packages\sklearn\utils\_param_validation.p
y", line 97, in validate_parameter_constraints
    raise InvalidParameterError(
sklearn.utils._param_validation.InvalidParameterError: The 'criterion' parameter of D
ecisionTreeRegressor must be a str among {'poisson', 'squared_error', 'absolute_erro
r', 'friedman_mse'}. Got 'mse' instead.

    warnings.warn(some_fits_failed_message, FitFailedWarning)
C:\ProgramData\anaconda3\lib\site-packages\sklearn\model_selection\_search.py:952: Us
erWarning: One or more of the test scores are non-finite: [      nan      nan 0.68
251641 0.62267566]
    warnings.warn(

```

```

Out[67]:

```

	model	best_score	best_params
0	linear_regression	0.844581	{'fit_intercept': False, 'n_jobs': -1}
1	lasso	0.717742	{'alpha': 1, 'selection': 'random'}
2	decision_tree	0.682516	{'criterion': 'friedman_mse', 'splitter': 'best'}

```

In [70]: 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]

```

```

In [71]: predict_price('1st Phase JP Nagar',1000, 2, 2)

```



```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names
warnings.warn(
```

```
Out[71]: 84.22924041793821
```

```
In [72]: predict_price('1st Phase JP Nagar',1000, 3, 3)
```

```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names
warnings.warn(
```

```
Out[72]: 87.41872656162407
```

```
In [73]: predict_price('Indira Nagar',1000, 2, 2)
```

```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names
warnings.warn(
```

```
Out[73]: 181.82117905122465
```

```
In [74]: predict_price('Indira Nagar',1000, 3, 3)
```

```
C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names
warnings.warn(
```

```
Out[74]: 185.01066519491047
```

Export the tested model to a pickle file

```
In [75]: import pickle
with open('bangalore_home_prices_model.pickle','wb') as f:
    pickle.dump(lr_clf,f)
```

Export location and column information to a file that will be useful later on in our prediction application

```
In [76]: import json
columns = {
    'data_columns' : [col.lower() for col in X.columns]
}
with open("columns.json","w") as f:
    f.write(json.dumps(columns))
```