

house price pred

April 6, 2023

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

```
[2]: import warnings
warnings.filterwarnings("ignore")
```

```
[3]: df=pd.read_csv("Bengaluru_House_Data.csv")
df.head()
```

```
[3]:
```

		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

```
[4]: df.shape
```

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

1 Data Cleaning

```
[5]: df["area_type"].value_counts()
```

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

```
[6]: df_copy1=df.copy()
```

```
[7]: df_copy1.
      ↪drop(["availability","society","area_type","balcony"],axis=1,inplace=True)
      df_copy1.head()
```

```
[7]:
```

	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

```
[8]: df_copy1.isnull().sum()
```

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

```
[9]: median_bath=df_copy1["bath"].median()
      df_copy1.fillna(median_bath,inplace=True)
```

```
[10]: df_copy1["bath"].isnull().sum()
```

```
[10]: 0
```

```
[11]: df_copy1["size"].unique()
```

```
[11]: 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', 2.0, '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)
```

```
[12]: df_copy1.dropna(inplace=True)
```

```
[13]: df_copy1.isnull().sum()
```

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

```
[14]: # df_copy1["size_temp"]=df_copy1["size"]
      # df_copy1["size"]=df_copy1["size"].str.split(" ").str[0]
      # df_copy1["size_temp"]=df_copy1["size_temp"].str.split(" ").str[1]
      # df_copy1["size_temp"]=df_copy1["size_temp"].replace("Bedroom", "1")
      # df_copy1["size_temp"]=df_copy1["size_temp"].replace("BHK", "3")
      # df_copy1["size_temp"]=df_copy1["size_temp"].replace("RK", "2")
      # # df_copy1["size_temp"]=df_copy1["size_temp"].replace("np.nan", "0")
      # # df_copy1["size_temp"]=df_copy1["size_temp"].astype(int)
      # # df_copy1["size"].astype(int)
      # # df_copy1["size"]=df_copy1["size"]+df_copy1["size_temp"]
      # # df_copy1["size"].head()
```

```
[15]: df_copy1["BHK"]=df_copy1["size"].apply(lambda x:int(x) if type(x)== float else_
      ↪int(x.split(" ")[0]))
      df_copy1["BHK"]=df_copy1["BHK"].astype(int)
```

```
[16]: df_copy1.head()
```

```
[16]:
```

	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

```
[17]: df_copy1.BHK.unique()
```

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

```
[18]: # there is/are house with 43 bedrooms which don't feel practical
      df_copy1[df_copy1.BHK>20]
```

```
[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

```
[19]: df_copy1["total_sqft"].unique()
```

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

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

```
[21]: (~df_copy1["total_sqft"].apply(is_float)).sum()
```

```
[21]: 13320
```

```
[22]: def conv_sqft_to_num(n):
      tokens=n.split("-") # for value is in the form of 2000-2200
      if len(tokens)==2:
          return (float(tokens[0]) + float(tokens[1]))/2
      try:
          n=float(n)
          return n
      except:
          return None
```

```
[23]: a=conv_sqft_to_num("90")
      print(a)
```

```
90.0
```

```
[24]: df_copy1["total_sqft"]=df_copy1["total_sqft"].apply(conv_sqft_to_num)
```

```
[25]: df_copy1["total_sqft"].unique()
```

```
[25]: array([1056. , 2600. , 1440. , ..., 1258.5,  774. , 4689. ])
```

```
[26]: df_copy1["total_sqft"].mean()
```

```
[26]: 1559.626693912912
```

```
[27]: df_copy1.head()
```

```
[27]:
```

	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

```
[28]: df_copy1["price_per_sqft"]=(df_copy1["price"]*100000)/df_copy1["total_sqft"]
```

```
[29]: len(df_copy1["location"].unique())
```

```
[29]: 1306
```

```
[30]: (df_copy1["location"].apply(is_float)).sum()
```

```
[30]: 0
```

```
[31]: df_copy1.location=df_copy1.location.apply(lambda x: str(x).strip())
```

```
[32]: df_copy1["location"].info()
```

```
<class 'pandas.core.series.Series'>  
RangeIndex: 13320 entries, 0 to 13319  
Series name: location  
Non-Null Count  Dtype  
-----  
13320 non-null  object  
dtypes: object(1)  
memory usage: 104.2+ KB
```

```
[33]: df_copy1.shape
```

```
[33]: (13320, 7)
```

```
[34]: location_stats=df_copy1["location"].value_counts().sort_values(ascending=False)
```

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

```
[35]: 1054
```

```
[36]: location_stats_lessthan_10=location_stats[location_stats<=10]
```

```
[37]: len(df_copy1["location"].unique())
```

```
[37]: 1295
```

```
[38]: df_copy1["location"]=df_copy1["location"].apply(lambda x:"others" if x in_  
↪location_stats_lessthan_10 else x)
```

```
[39]: len(df_copy1["location"].unique())
```

```
[39]: 242
```

```
[40]: df_copy1.shape
```

```
[40]: (13320, 7)
```

```
[41]: df_copy1[df_copy1["total_sqft"]/df_copy1["BHK"]<300]
```

```
[41]:
```

	location	size	total_sqft	bath	price	BHK	\
9	others	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	others	3 Bedroom	500.0	3.0	100.0	3	
...	
13277	others	7 Bedroom	1400.0	7.0	218.0	7	
13279	others	6 Bedroom	1200.0	5.0	130.0	6	
13281	Margondanahalli	5 Bedroom	1375.0	5.0	125.0	5	
13303	Vidyaranyapura	5 Bedroom	774.0	5.0	70.0	5	
13311	Ramamurthy Nagar	7 Bedroom	1500.0	9.0	250.0	7	

	price_per_sqft
9	36274.509804
45	33333.333333
58	10660.980810
68	6296.296296
70	20000.000000
...	...
13277	15571.428571
13279	10833.333333
13281	9090.909091
13303	9043.927649
13311	16666.666667


```
[744 rows x 7 columns]
```

```
[42]: df_copy2=df_copy1[(df_copy1["total_sqft"]/df_copy1["BHK"])>=300]
df_copy2.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]: df_copy2.shape
```

```
[43]: (12530, 7)
```

```
[44]: df_copy2["price_per_sqft"].describe()
```

```
[44]: count      12530.000000
mean         6303.979357
std          4162.237981
min           267.829813
25%          4210.526316
50%          5294.117647
75%          6916.666667
max         176470.588235
Name: price_per_sqft, dtype: float64
```

```
[45]: # def remove_outliers(df):
#     for key,sub_df in df.groupby("location"):
#         m=np.mean(sub_df.price_per_sqft)
#         sd=np.std(sub_df.price_per_sqft)
#         df_out=sub_df[(sub_df.price_per_sqft>(m-sd)) & (sub_df.
#             ↪price_per_sqft<=(m+sd))]
#         return df_out
```

```
[46]: # df_copy3=remove_outliers(df_copy2)
```

```
[47]: # df_copy3.shape
```

```
[48]: IQR=6916.666667-4210.526316
lower_range=4210.526316-(1.5*IQR)
upper_range=6916.666667+(1.5*IQR)
print(lower_range)
print(upper_range)
```

```
151.3157895000004
10975.8771935
```

```
[49]: df_copy2=df_copy2[(df_copy2["price_per_sqft"]>=lower_range) &
    ↪(df_copy2["price_per_sqft"]<=upper_range)]
df_copy2.shape
```

```
[49]: (11523, 7)
```

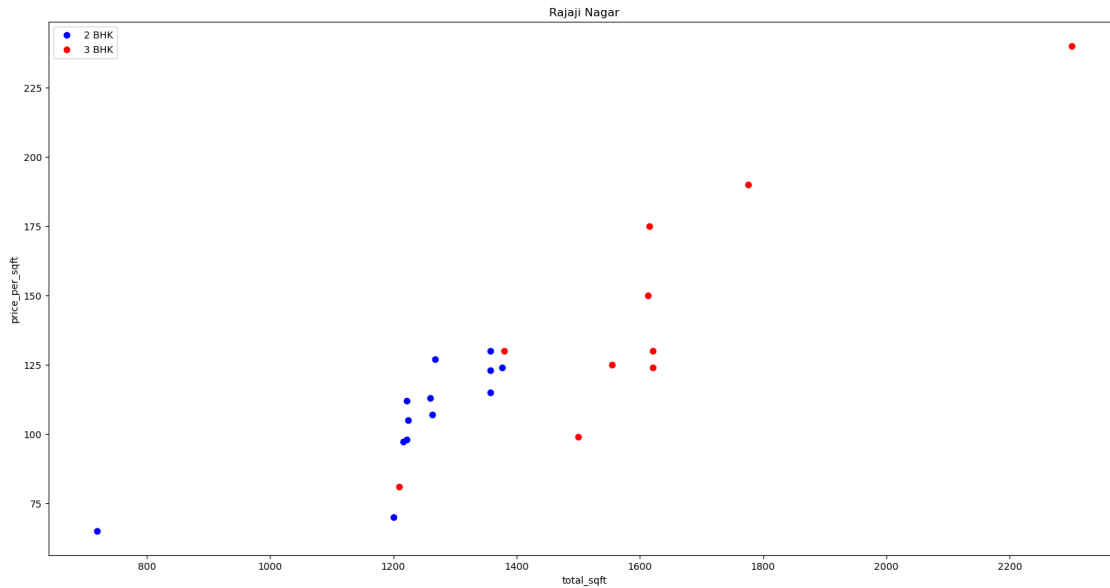
```
[50]: def plot_scatter_chart(df,location):
      bhk_2=df[(df.location==location) & (df.BHK==2)]
```

```

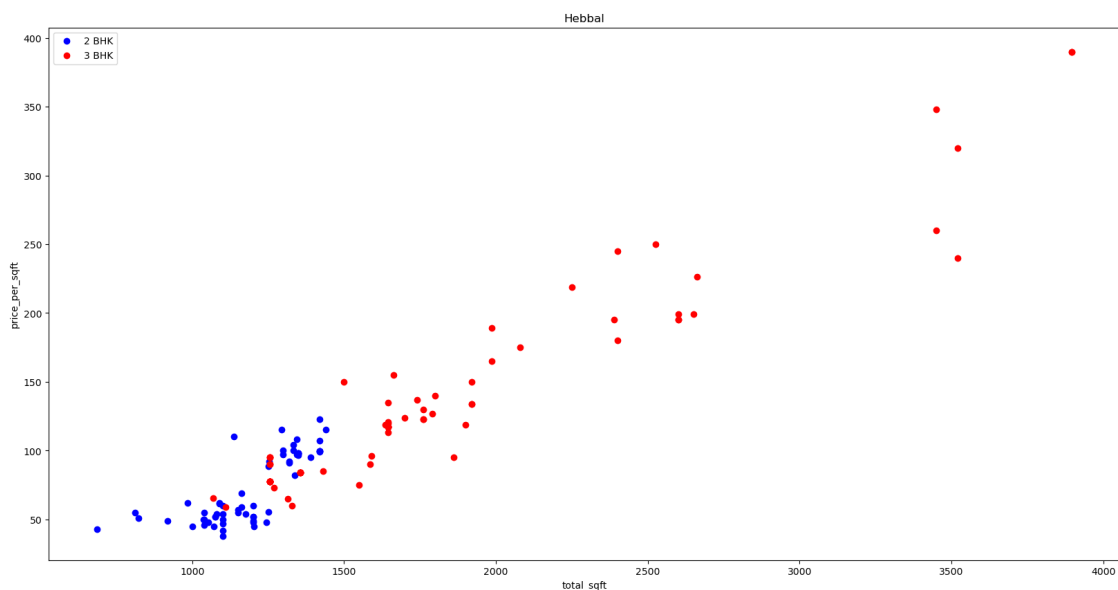
bhk_3=df[(df.location==location) & (df.BHK==3)]
plt.scatter(bhk_2.total_sqft,bhk_2.price,color="blue",label="2 BHK")
plt.scatter(bhk_3.total_sqft,bhk_3.price,color="red",label="3 BHK")
plt.xlabel("total_sqft")
plt.ylabel("price_per_sqft")
plt.title(location)
plt.legend()

```

```
[51]: plot_scatter_chart(df_copy2,"Rajaji Nagar")
```



```
[52]: plot_scatter_chart(df_copy2,"Hebbal")
```




```
[53]: # for same location there are some house with high price with less bedrooms.
```

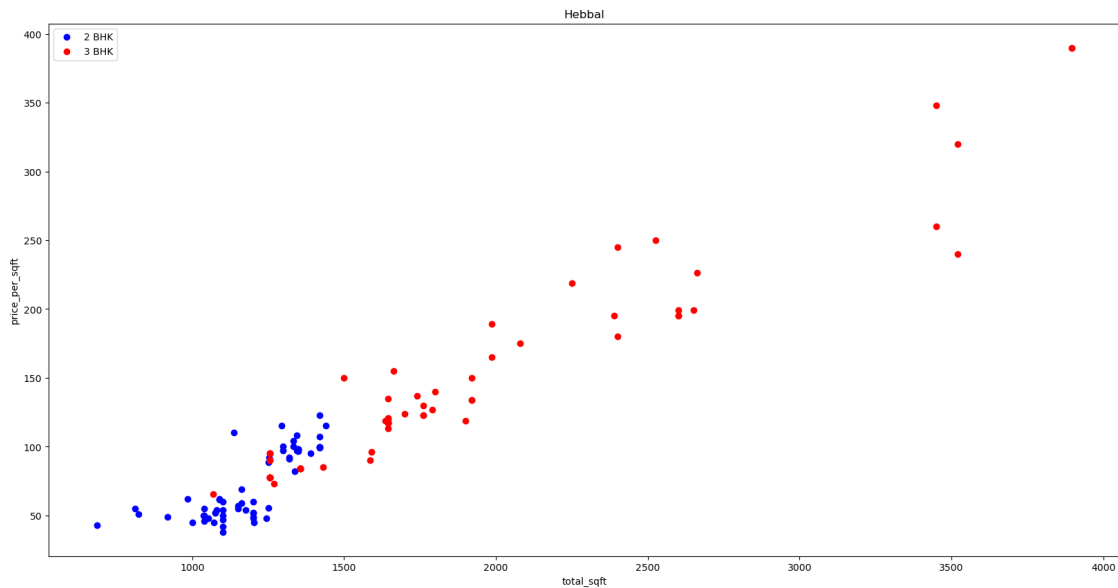
```
def remove_bhk_outliers(df):
    exclude_indices=np.array([])
    for location,location_df in df.groupby("location"):
        bhk_stats={}
        for bhk,bhk_df in location_df.groupby("BHK"):
            bhk_stats[bhk]={
                "mean":np.mean(df.price_per_sqft),
                "std":np.std(df.price_per_sqft),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

```
[54]: df_copy3=remove_bhk_outliers(df_copy2)
```

```
[55]: df_copy3.shape
```

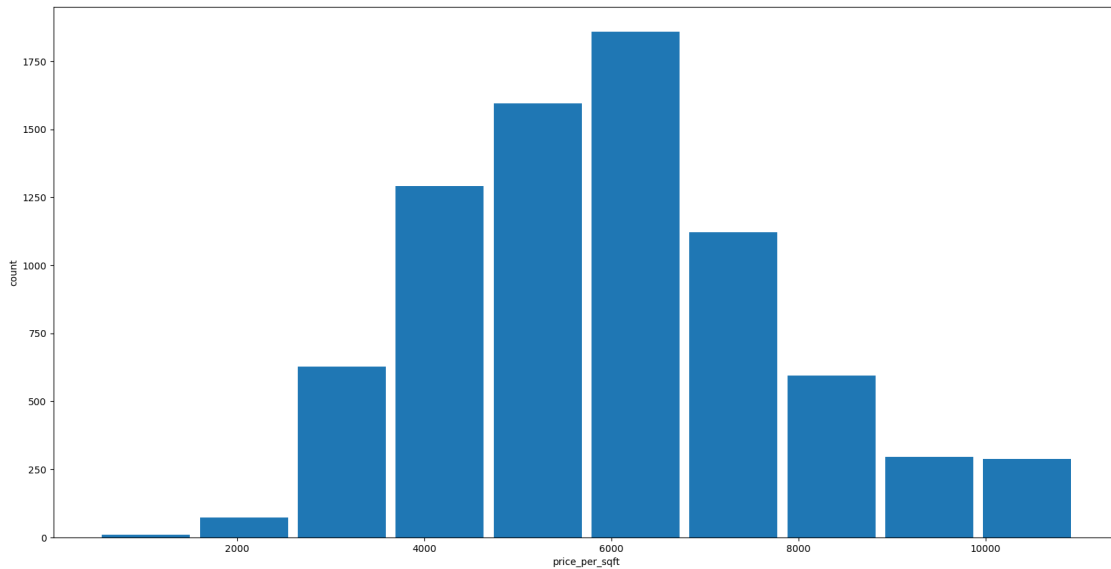
```
[55]: (7753, 7)
```

```
[56]: plot_scatter_chart(df_copy3,"Hebbal")
```



```
[57]: plt.hist(df_copy3.price_per_sqft,rwidth=0.9)
plt.xlabel("price_per_sqft")
plt.ylabel("count")
```

```
[57]: Text(0, 0.5, 'count')
```



```
[58]: df_copy3.bath.unique()
```

```
[58]: array([ 2.,  3.,  4.,  5.,  1.,  8.,  6.,  7.,  9., 12., 16., 10., 13.])
```

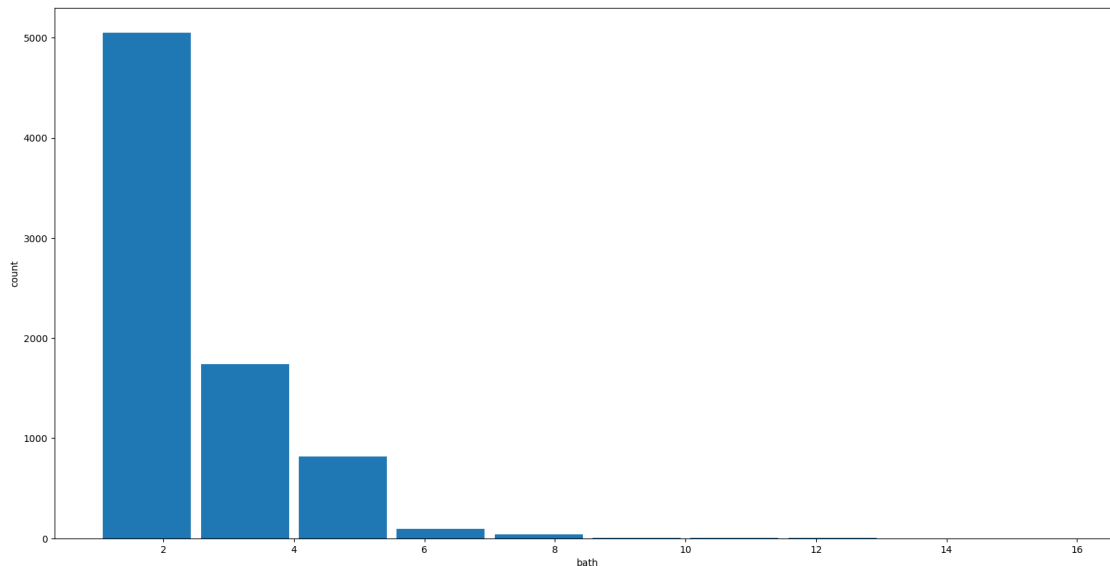
```
[59]: df_copy3[df_copy3.bath>10]
```

```
[59]:
```

	location	size	total_sqft	bath	price	BHK	price_per_sqft
3096	others	10 BHK	12000.0	12.0	525.0	10	4375.000000
3609	others	16 BHK	10000.0	16.0	550.0	16	5500.000000
7979	others	11 BHK	6000.0	12.0	150.0	11	2500.000000
8636	Neeladri Nagar	10 BHK	4000.0	12.0	160.0	10	4000.000000
9935	others	13 BHK	5425.0	13.0	275.0	13	5069.124424

```
[60]: plt.hist(df_copy3.bath,rwidth=0.9)
plt.xlabel("bath")
plt.ylabel("count")
```

```
[60]: Text(0, 0.5, 'count')
```



```
[61]: df_copy3=df_copy3[df_copy3.bath<df_copy3.BHK+2]
```

```
[62]: df_copy3.shape
```

```
[62]: (7672, 7)
```

```
[63]: df_copy4=df_copy3.drop(["size","price_per_sqft"],axis=1)
```

```
[64]: df_copy4.head()
```

```
[64]:
```

	location	total_sqft	bath	price	BHK
0	Electronic City Phase II	1056.0	2.0	39.07	2
3	Lingadheeranahalli	1521.0	3.0	95.00	3
4	Kothanur	1200.0	2.0	51.00	2
6	Old Airport Road	2732.0	4.0	204.00	4
11	Whitefield	2785.0	5.0	295.00	4

2 Encoding

```
[65]: dummies=pd.get_dummies(df_copy4.location)
dummies.head(3)
```

```
[65]:
```

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

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

	6th Phase JP Nagar	7th Phase JP Nagar	8th Phase JP Nagar	\
0	0	0	0	
3	0	0	0	
4	0	0	0	

	9th Phase JP Nagar	...	Vishveshwarya Layout	Vishwapriya Layout	\
0	0	...	0	0	
3	0	...	0	0	
4	0	...	0	0	

	Vittasandra	Whitefield	Yelachenahalli	Yelahanka	Yelahanka New Town	\
0	0	0	0	0	0	
3	0	0	0	0	0	
4	0	0	0	0	0	

	Yelenahalli	Yeshwanthpur	others
0	0	0	0
3	0	0	0
4	0	0	0

[3 rows x 241 columns]

```
[66]: df_copy5=pd.concat([df_copy4,dummies.drop("others",axis=1)],axis=1)
df_copy5.head(3)
```

```
[66]:      location  total_sqft  bath  price  BHK  \
0  Electronic City Phase II    1056.0   2.0  39.07   2
3      Lingadheeranahalli    1521.0   3.0  95.00   3
4      Kothanur            1200.0   2.0  51.00   2
```

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

	2nd Stage Nagarbhavi	5th Block Hbr Layout	...	Vijayanagar	\
0	0	0	...	0	
3	0	0	...	0	
4	0	0	...	0	

	Vishveshwarya Layout	Vishwapriya Layout	Vittasandra	Whitefield	\
0	0	0	0	0	

3	0	0	0	0
4	0	0	0	0
	Yelachenahalli	Yelahanka	Yelahanka New Town	Yelenahalli
0	0	0	0	0
3	0	0	0	0
4	0	0	0	0

[3 rows x 245 columns]

```
[67]: df_copy5=df_copy5.drop("location",axis=1)
df_copy5.shape
```

[67]: (7672, 244)

3 Model bulding

```
[68]: x=df_copy5.drop("price",axis=1)
x.head()
```

```
[68]:
```

	total_sqft	bath	BHK	1st Block Jayanagar	1st Phase JP Nagar	\
0	1056.0	2.0	2	0	0	
3	1521.0	3.0	3	0	0	
4	1200.0	2.0	2	0	0	
6	2732.0	4.0	4	0	0	
11	2785.0	5.0	4	0	0	

	2nd Phase Judicial Layout	2nd Stage Nagarbhavi	5th Block Hbr Layout	\
0	0	0	0	
3	0	0	0	
4	0	0	0	
6	0	0	0	
11	0	0	0	

	5th Phase JP Nagar	6th Phase JP Nagar	... Vijayanagar	\
0	0	0	...	0
3	0	0	...	0
4	0	0	...	0
6	0	0	...	0
11	0	0	...	0

	Vishveshwarya Layout	Vishwapriya Layout	Vittasandra	Whitefield	\
0	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	
6	0	0	0	0	

	0	0	0	1
	Yelachenahalli	Yelahanka	Yelahanka New Town	Yelenahalli
0	0	0	0	0
3	0	0	0	0
4	0	0	0	0
6	0	0	0	0
11	0	0	0	0

[5 rows x 243 columns]

```
[69]: y=df_copy5.price
      y.head()
```

```
[69]: 0      39.07
      3      95.00
      4      51.00
      6     204.00
      11     295.00
      Name: price, dtype: float64
```

```
[70]: from sklearn.model_selection import train_test_split
      x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.
      ↪22,random_state=29)
```

```
[71]: from sklearn.linear_model import LinearRegression
      lr_clf=LinearRegression()
      lr_clf.fit(x_train,y_train)
      lr_clf.score(x_test,y_test)
```

```
[71]: 0.8430104651996677
```

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

      cv=ShuffleSplit(n_splits=5,test_size=0.22,random_state=12)

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

```
[72]: array([ 8.36901632e-01,  8.63577341e-01, -2.30433018e+14, -8.26731318e+14,
      8.36353524e-01])
```

```
[73]: 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={
        "liner_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=0.22,random_state=12,)
    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)

```

```

[73]:
      model  best_score  \
0  liner_regression    0.851008
1           lasso      0.830676
2  decision tree      0.826229

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

```

```
[74]: x.columns
```

```
[74]: 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=243)
```

```
[75]: np.where(x.columns=="2nd Stage Nagarbhavi")[0][0]
```

```
[75]: 6
```

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

```
[77]: predict_price("1st Block Jayanagar",1500,3,3)
```

```
[77]: 115.96596278763438
```

```
[78]: predict_price("1st Block Jayanagar",1500,4,4)
```

```
[78]: 127.78170187851941
```

```
[79]: predict_price("Indira Nagar",1500,4,4)
```

```
[79]: 153.1193750858707
```

```
[80]: predict_price("Indira Nagar",1500,3,3)
```

```
[80]: 141.30363599498568
```

```
[81]: predict_price("Vijayanagar",2000,4,4)
```

```
[81]: 146.25340547707984
```



```
[82]: predict_price("Vijayanagar",2000,3,4)
```

```
[82]: 141.02842337329878
```

```
[83]: import pickle  
with open("banglore_home_prices_model.pickle","wb") as f:  
    pickle.dump(lr_clf,f)
```

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

```
[ ]:
```