

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

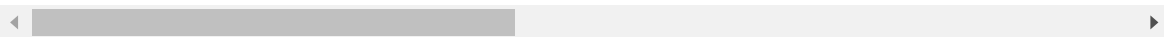
## Load Dataset

```
In [23]: df = pd.read_csv('House Price India.csv')
df.head()
```

Out[23]:

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	coi
0	6762810145	42491	5	2.50	3650	9050	2.0	0	4	
1	6762810635	42491	4	2.50	2920	4000	1.5	0	0	
2	6762810998	42491	5	2.75	2910	9480	1.5	0	0	
3	6762812605	42491	4	2.50	3310	42998	2.0	0	0	
4	6762812919	42491	3	2.00	2710	4500	1.5	0	0	

5 rows × 23 columns



In [3]: df.info()

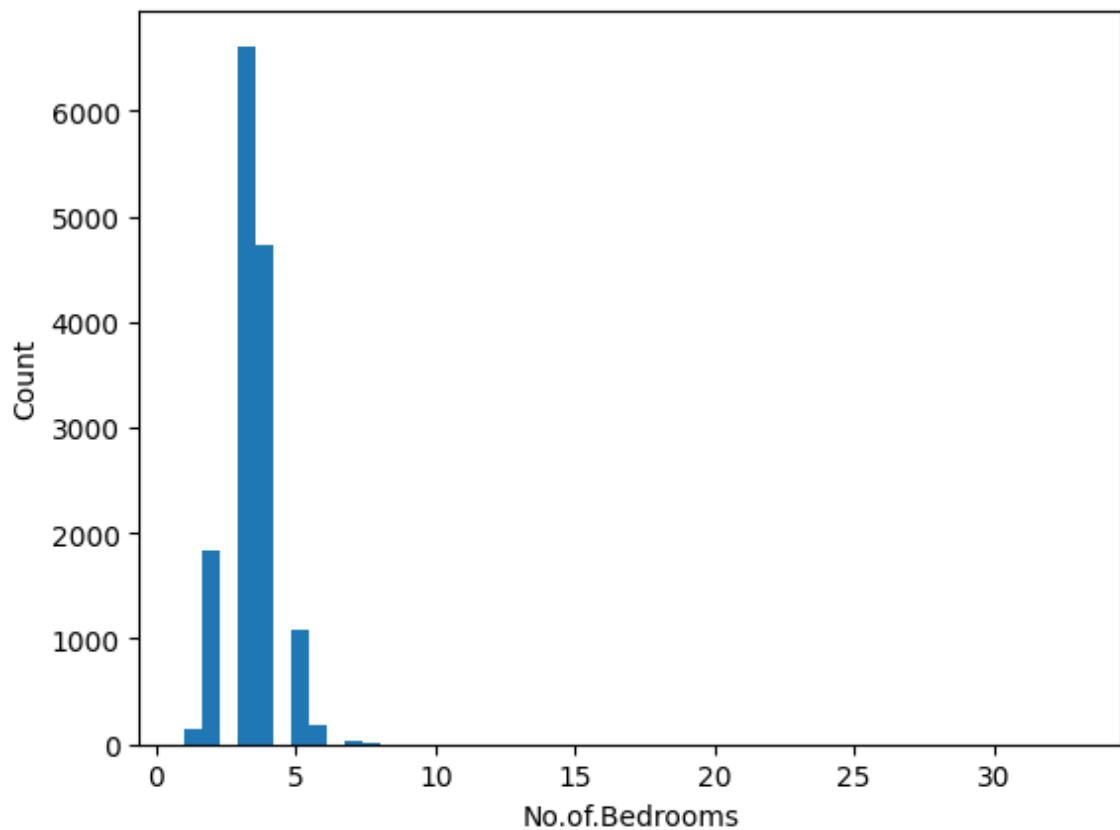
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14620 entries, 0 to 14619
Data columns (total 23 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   id                                    14620 non-null  int64
1   Date                                 14620 non-null  int64
2   number of bedrooms                  14620 non-null  int64
3   number of bathrooms                 14620 non-null  float64
4   living area                         14620 non-null  int64
5   lot area                           14620 non-null  int64
6   number of floors                    14620 non-null  float64
7   waterfront present                  14620 non-null  int64
8   number of views                     14620 non-null  int64
9   condition of the house              14620 non-null  int64
10  grade of the house                  14620 non-null  int64
11  Area of the house(excluding basement) 14620 non-null  int64
12  Area of the basement                14620 non-null  int64
13  Built Year                          14620 non-null  int64
14  Renovation Year                     14620 non-null  int64
15  Postal Code                         14620 non-null  int64
16  Lattitude                           14620 non-null  float64
17  Longitude                           14620 non-null  float64
18  living_area_renov                   14620 non-null  int64
19  lot_area_renov                      14620 non-null  int64
20  Number of schools nearby             14620 non-null  int64
21  Distance from the airport           14620 non-null  int64
22  Price                              14620 non-null  int64
dtypes: float64(4), int64(19)
memory usage: 2.6 MB
```

## Univariate Analysis

### Histogram

```
In [10]: plt.hist(df['number of bedrooms'],bins=50)  
plt.xlabel("No.of.Bedrooms")  
plt.ylabel("Count")
```

```
Out[10]: Text(0, 0.5, 'Count')
```



From the above graph we can clearly see that the peak count above 6000 is at range between 0 to 5. As the no.of.bedrooms increases after 5 the count values decreases tremendously.

## Distplot

```
In [11]: sns.distplot(df['Price'],bins=30)
```

C:\Users\ELCOT\AppData\Local\Temp\ipykernel\_70496\507312228.py:1: UserWarning:

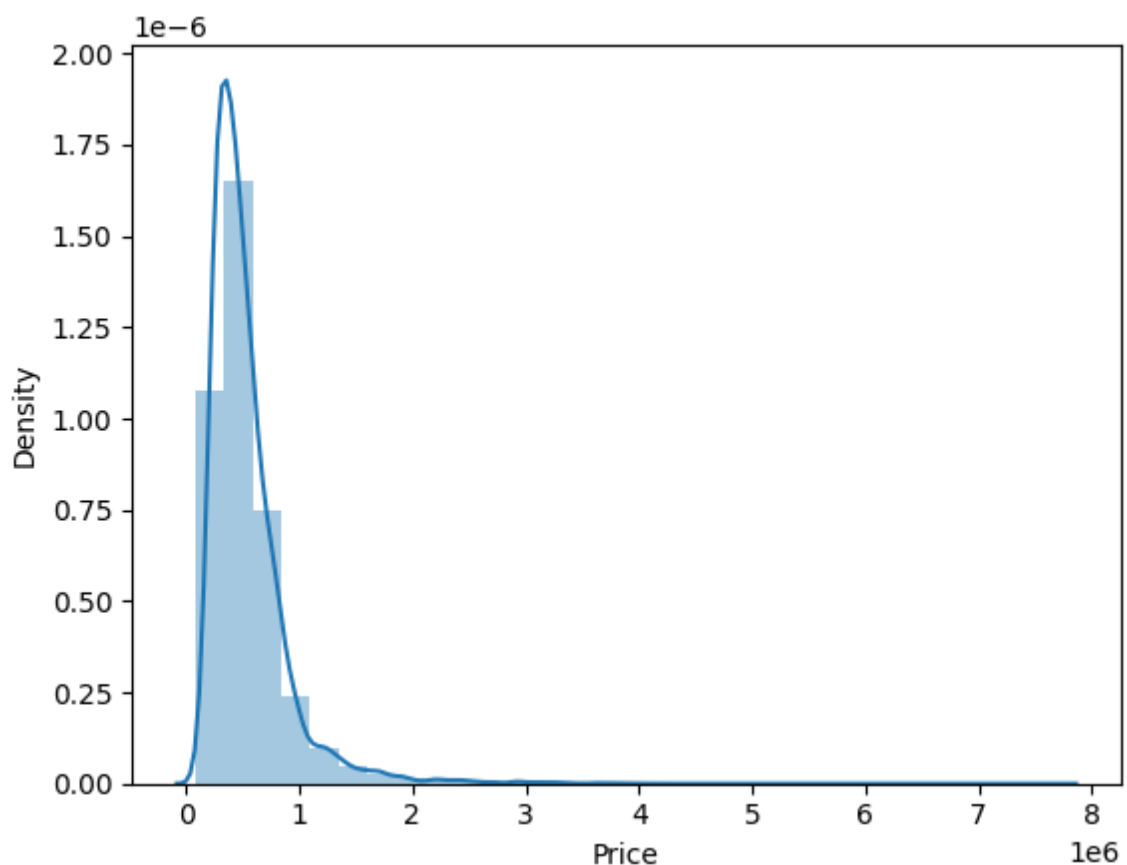
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see <https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751> (<https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751>)

```
sns.distplot(df['Price'],bins=30)
```

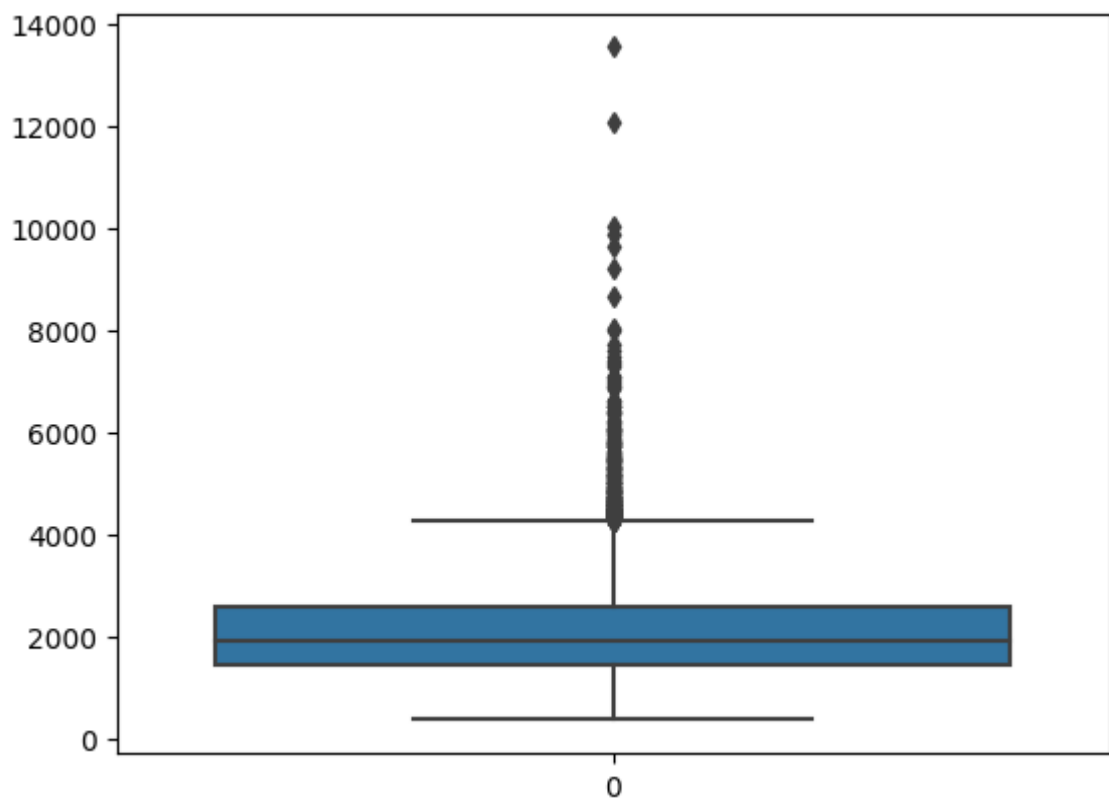
```
Out[11]: <AxesSubplot: xlabel='Price', ylabel='Density'>
```



## Boxplot

```
In [12]: sns.boxplot(df['living area'])
```

```
Out[12]: <AxesSubplot: >
```

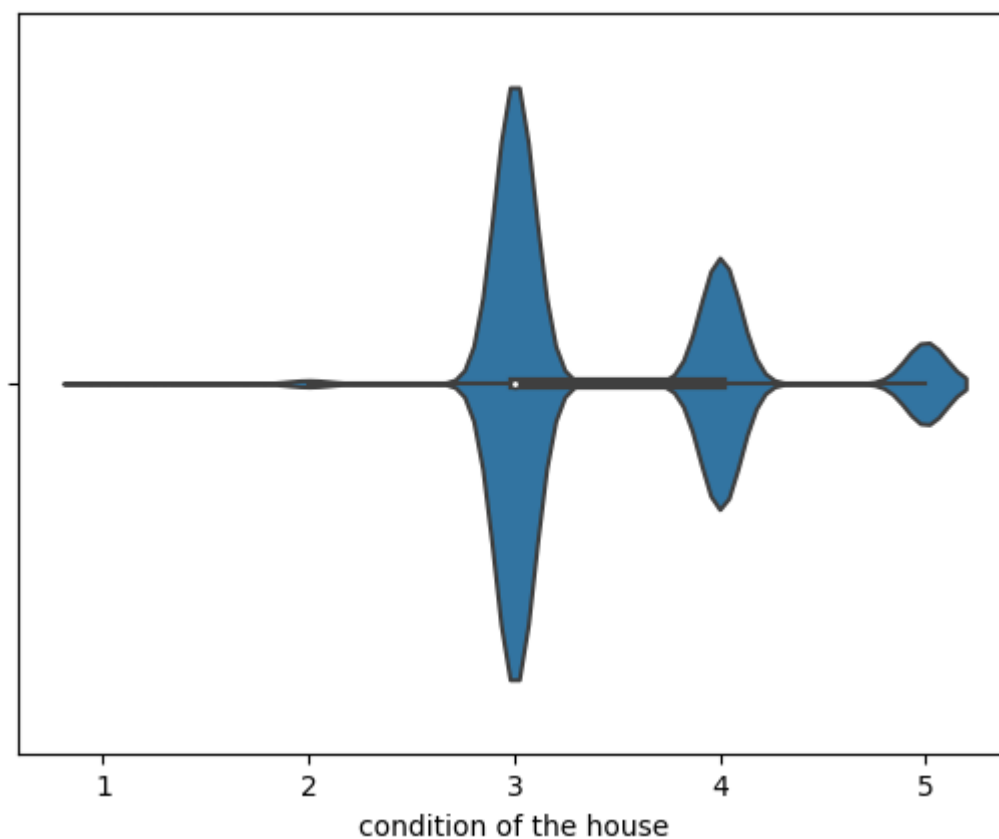


Boxplot is also used for detect the outlier in data set. It captures the summary of the data efficiently with a simple box and whiskers and allows us to compare easily across groups. Boxplot for living area and it contains many outliers and many outliers present in the features. The above one is a sample for detecting outliers.

## Violinplot

```
In [13]: sns.violinplot(x=df['condition of the house'])
```

```
Out[13]: <AxesSubplot: xlabel='condition of the house'>
```



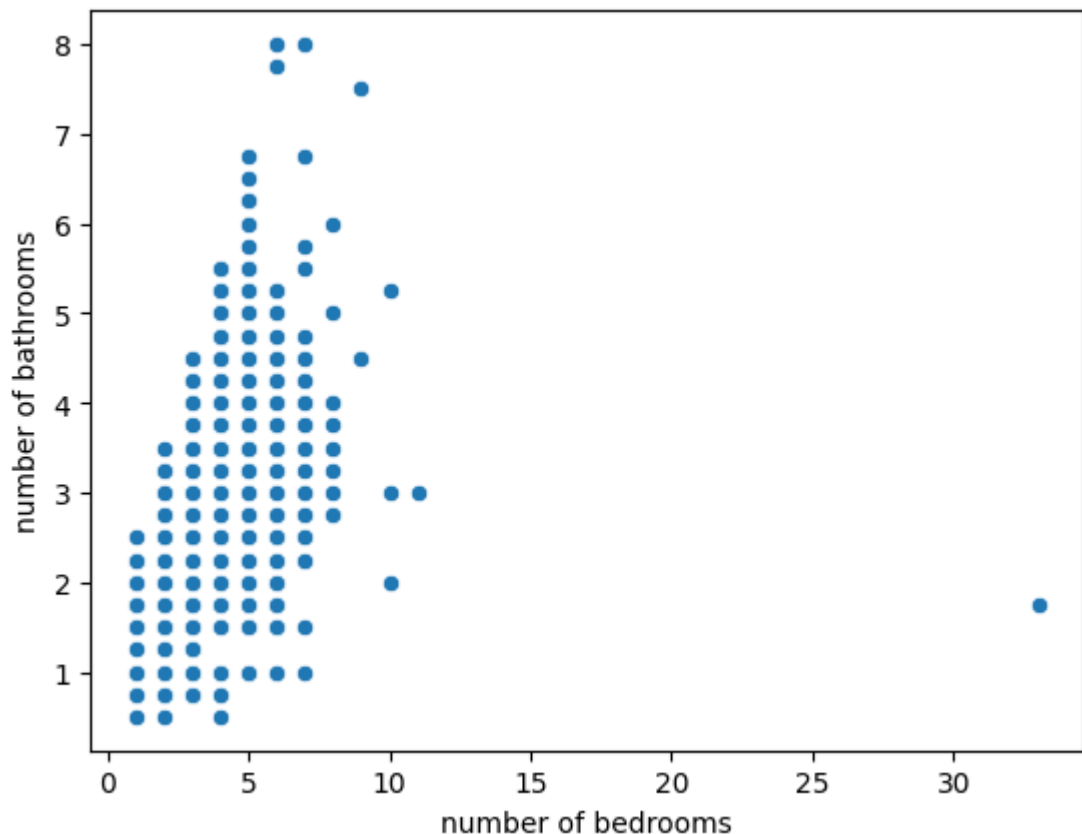
violinplot is used to visualize the distribution numerical data and it shows the full distribution of data. The mean value of the variable "condition of the house" lies in 3 and the interquartile ranges between 3 to 4. The rest thin lines represents the rest distributions, except for the points that are determined to be the outliers. The higher probability lies in 3 and lowest probability lies above 5.

## Bivariate Analysis

### Scatterplot

```
In [14]: sns.scatterplot(x=df['number of bedrooms'],y=df['number of bathrooms'])
```

```
Out[14]: <AxesSubplot: xlabel='number of bedrooms', ylabel='number of bathrooms'>
```

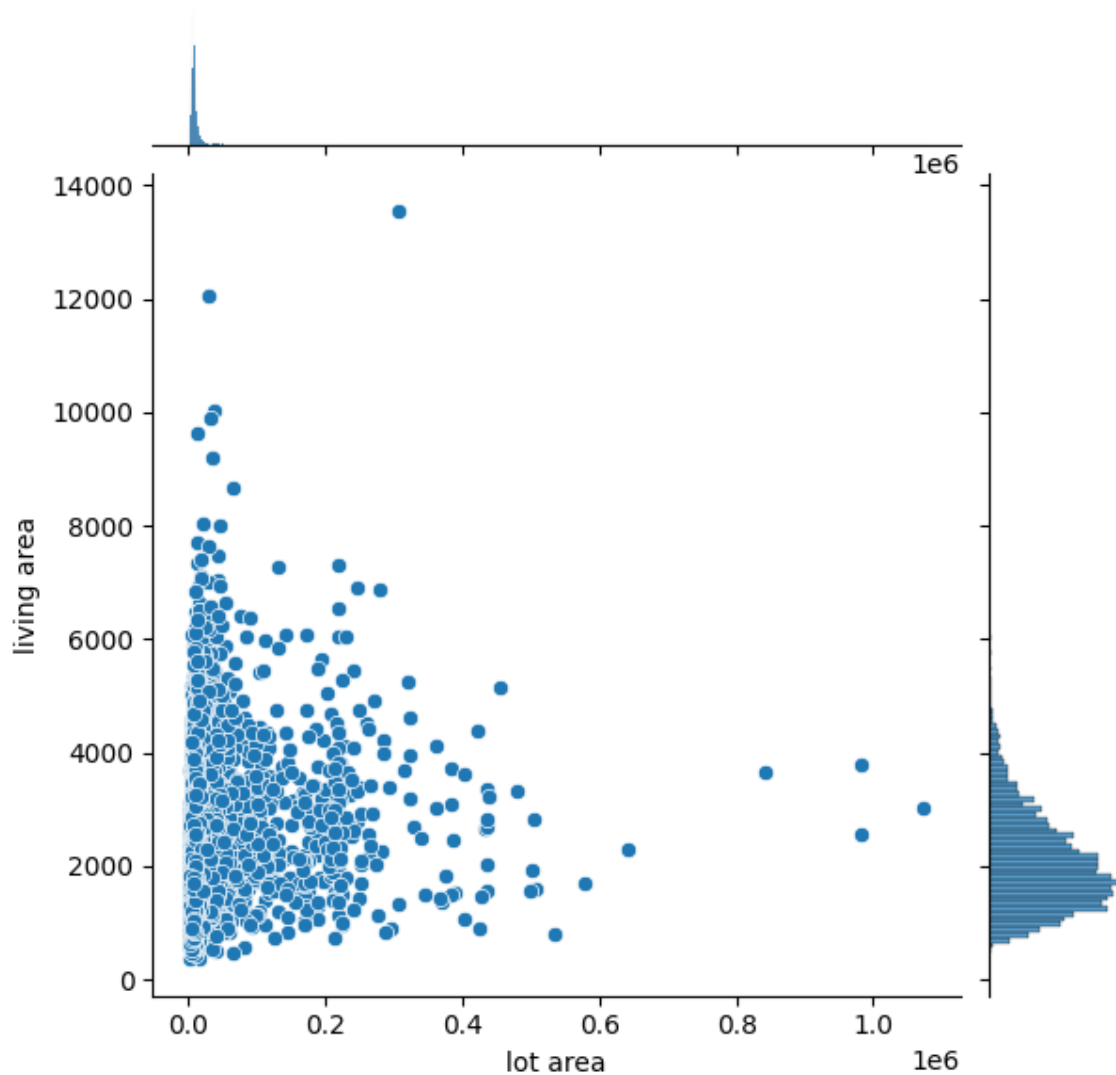


he scatterplot is used to show distributions between two variables. For no.of.bathrooms and no.of.bedrooms as far as the bathroom increases the bedroom number increases. And there are some outliers present in them.

## Jointplot

```
In [15]: sns.jointplot(data = df,x = 'lot area',y = 'living area')
```

```
Out[15]: <seaborn.axisgrid.JointGrid at 0x21867173ca0>
```



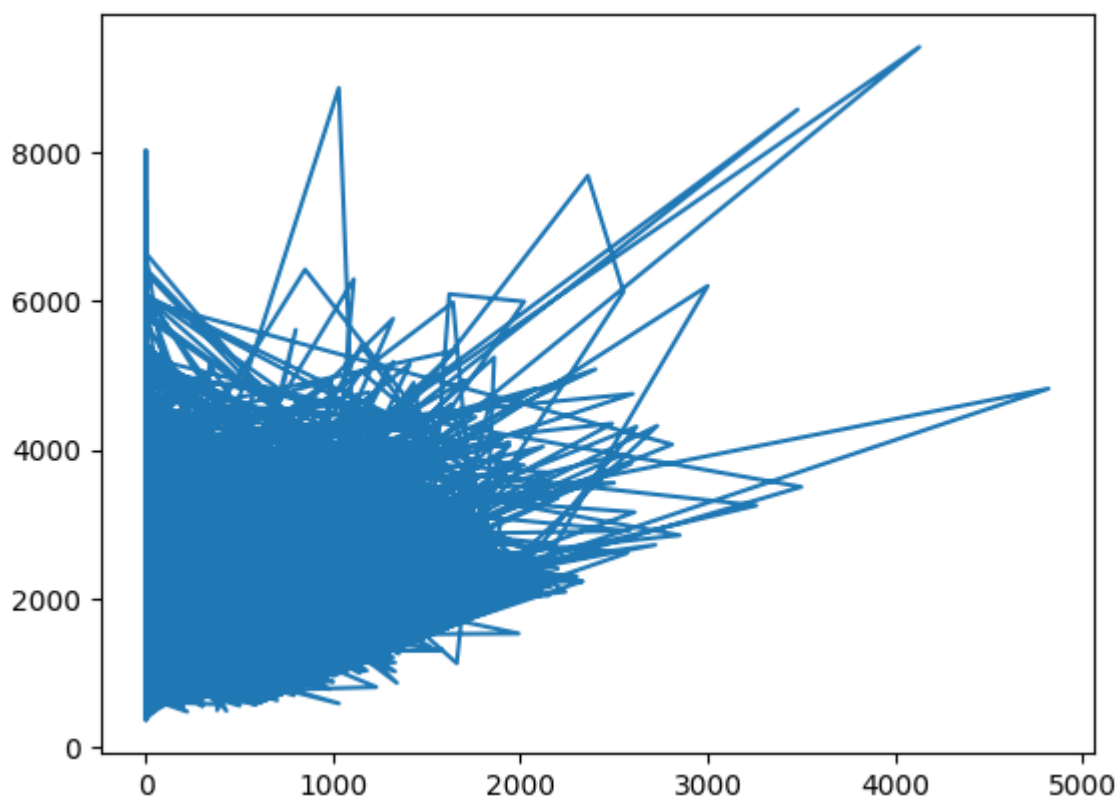


The relation between living area vs lot area and univariate of these has been shown. As far as the living area increases the lot area increases slighter and present many outliers between them. Univariate distribution of lot area remains same with slight increase in area but for living area the peak value is achieved at 2000 by gradual increase in it and then decreases until at a range of 5000.

## Line plot

```
In [16]: plt.plot(df['Area of the basement'],df['Area of the house(excluding basement
```

```
Out[16]: [<matplotlib.lines.Line2D at 0x218699cf2e0>]
```

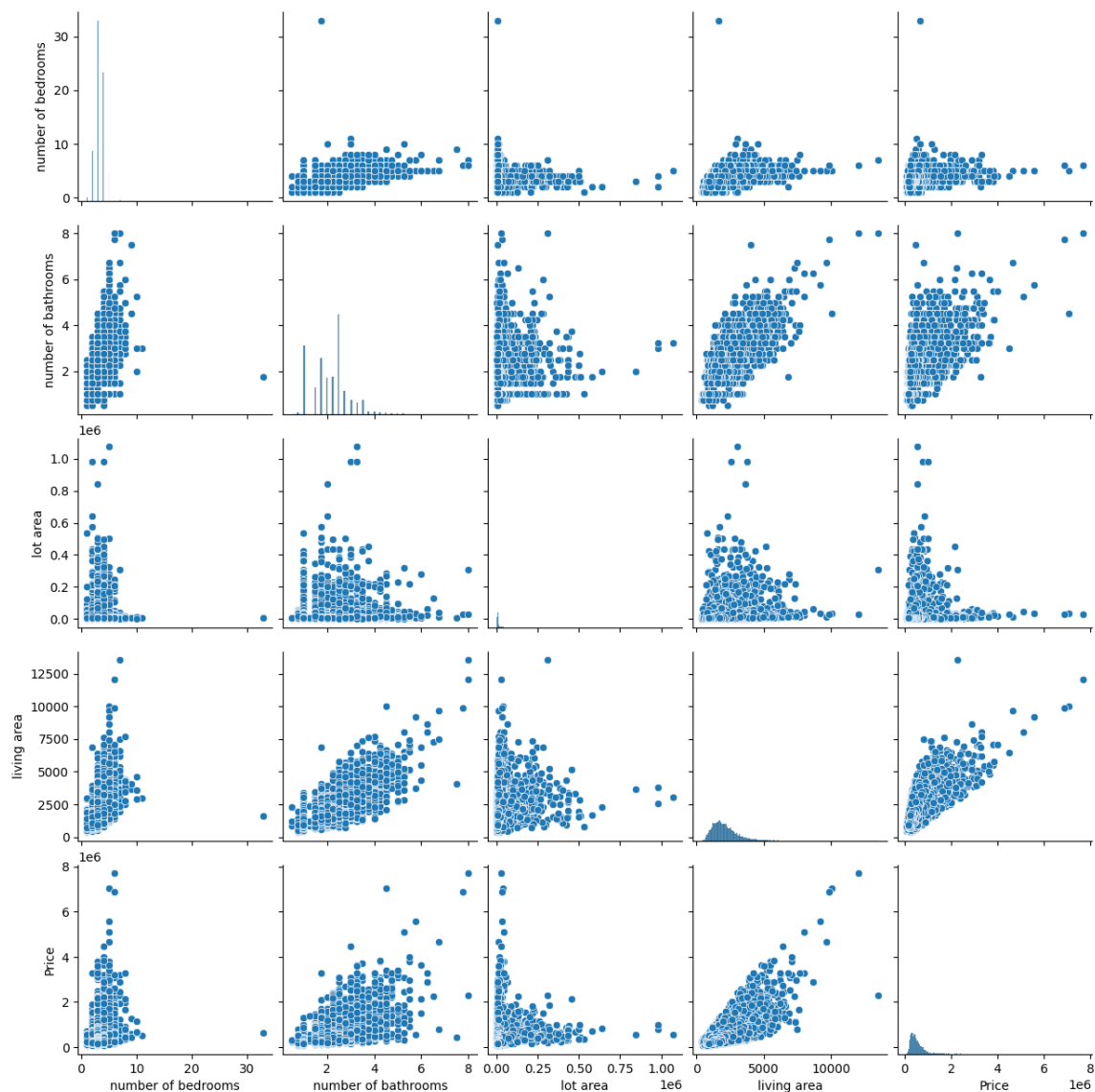


# Multivariate Analysis

## Pair plot

```
In [17]: X = df[['number of bedrooms', 'number of bathrooms', 'lot area', 'living area',  
sns.pairplot(X, dropna=True)
```

```
Out[17]: <seaborn.axisgrid.PairGrid at 0x218672039a0>
```

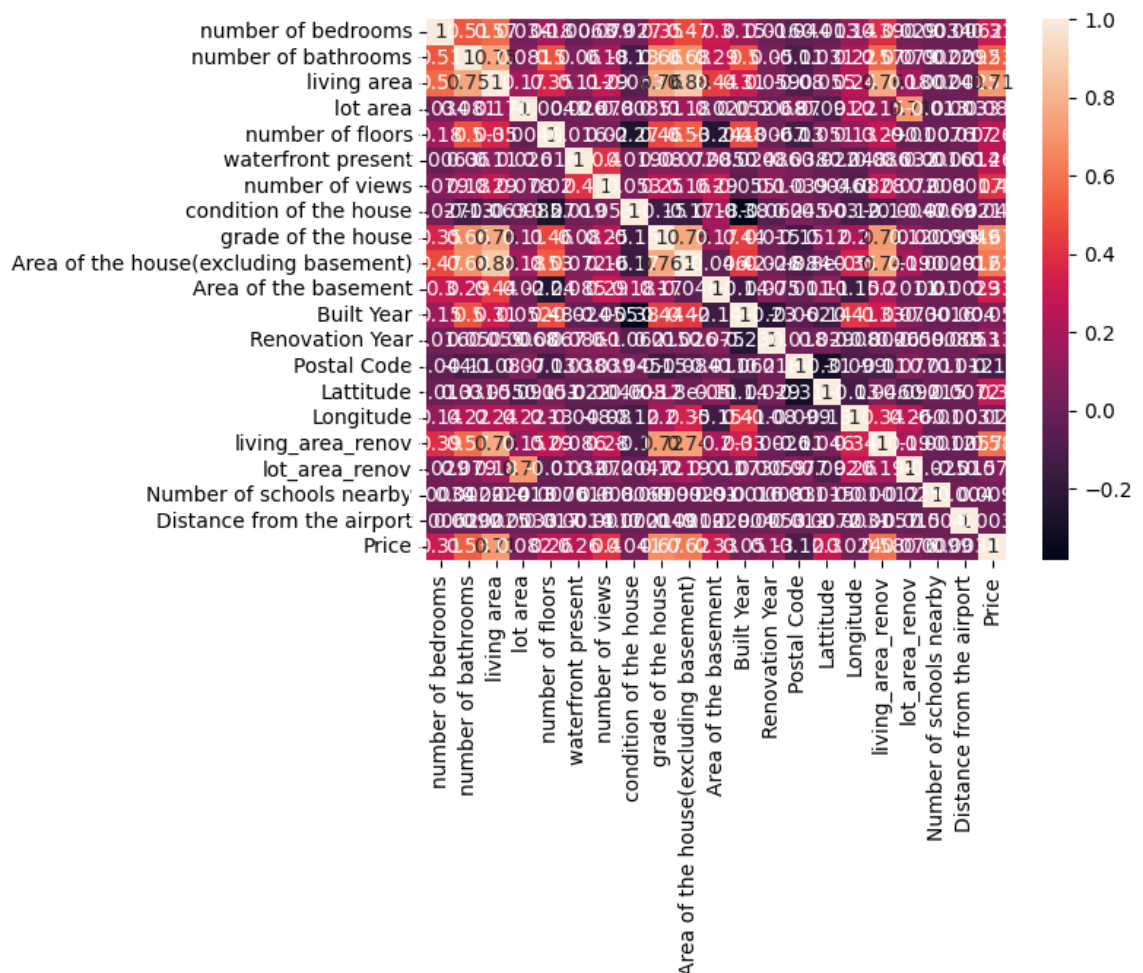


From pairplot we can clearly see that some variable are linear to some variable and logistic to some variables. Most of the variables are linear to other variables. But in all variables outliers present in it.

## Heat map

```
In [18]: df.drop(columns=['id', 'Date'], inplace=True)
sns.heatmap(df.corr(), annot=True)
```

```
Out[18]: <AxesSubplot: >
```



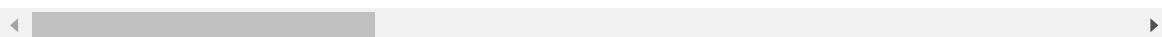
## Statistical analysis

In [5]: `df.describe()`

Out[5]:

	id	Date	number of bedrooms	number of bathrooms	living area	lot area
<b>count</b>	1.462000e+04	14620.000000	14620.000000	14620.000000	14620.000000	1.462000e+04
<b>mean</b>	6.762821e+09	42604.538646	3.379343	2.129583	2098.262996	1.509328e+04
<b>std</b>	6.237575e+03	67.347991	0.938719	0.769934	928.275721	3.791962e+04
<b>min</b>	6.762810e+09	42491.000000	1.000000	0.500000	370.000000	5.200000e+02
<b>25%</b>	6.762815e+09	42546.000000	3.000000	1.750000	1440.000000	5.010750e+03
<b>50%</b>	6.762821e+09	42600.000000	3.000000	2.250000	1930.000000	7.620000e+03
<b>75%</b>	6.762826e+09	42662.000000	4.000000	2.500000	2570.000000	1.080000e+04
<b>max</b>	6.762832e+09	42734.000000	33.000000	8.000000	13540.000000	1.074218e+06

8 rows × 7 columns



**Descriptive statistics to summarize the data by computing mean, median, mode, standard deviation and likewise other informations of data.**

## Handling missing values

In [21]: `df.isnull().any()`

Out[21]:

number of bedrooms	False
number of bathrooms	False
living area	False
lot area	False
number of floors	False
waterfront present	False
number of views	False
condition of the house	False
grade of the house	False
Area of the house(excluding basement)	False
Area of the basement	False
Built Year	False
Renovation Year	False
Postal Code	False
Lattitude	False
Longitude	False
living_area_renov	False
lot_area_renov	False
Number of schools nearby	False
Distance from the airport	False
Price	False
dtype: bool	

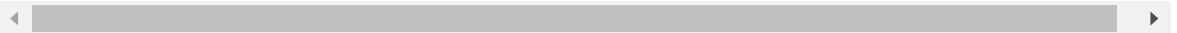
In [22]: df.info()

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14620 entries, 0 to 14619
Data columns (total 21 columns):
 #   Column                                          Non-Null Count  Dtype
---  -
 0   number of bedrooms                          14620 non-null  int64
 1   number of bathrooms                         14620 non-null  float64
 2   living area                                  14620 non-null  int64
 3   lot area                                    14620 non-null  int64
 4   number of floors                            14620 non-null  float64
 5   waterfront present                          14620 non-null  int64
 6   number of views                             14620 non-null  int64
 7   condition of the house                     14620 non-null  int64
 8   grade of the house                         14620 non-null  int64
 9   Area of the house(excluding basement)      14620 non-null  int64
10   Area of the basement                       14620 non-null  int64
11   Built Year                                 14620 non-null  int64
12   Renovation Year                           14620 non-null  int64
13   Postal Code                               14620 non-null  int64
14   Lattitude                                  14620 non-null  float64
15   Longitude                                  14620 non-null  float64
16   living_area_renov                          14620 non-null  int64
17   lot_area_renov                            14620 non-null  int64
18   Number of schools nearby                   14620 non-null  int64
19   Distance from the airport                  14620 non-null  int64
20   Price                                      14620 non-null  int64
dtypes: float64(4), int64(17)
memory usage: 2.3 MB

```

***The above information shows that the none of the columns contains any null value in it. We don't need to perform any specific operations to handle the missing values.***



In [ ]: