

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

df = pd.read_csv('/content/House Price India.csv')
dfh = df.head()
dfl = df.tail()
dfh
```

	id	Date	number of bedrooms	number of bathrooms	living area \
0	6762810145	42491	5	2.50	3650
1	6762810635	42491	4	2.50	2920
2	6762810998	42491	5	2.75	2910
3	6762812605	42491	4	2.50	3310
4	6762812919	42491	3	2.00	2710

	lot area	number of floors	waterfront present	number of views \
0	9050	2.0	0	4
1	4000	1.5	0	0
2	9480	1.5	0	0
3	42998	2.0	0	0
4	4500	1.5	0	0

	condition of the house	...	Built Year	Renovation Year	Postal Code \
0	5	...	1921	0	122003
1	5	...	1909	0	122004
2	3	...	1939	0	122004
3	3	...	2001	0	122005
4	4	...	1929	0	122006

	Lattitude	Longitude	living_area_renov	lot_area_renov \
0	52.8645	-114.557	2880	5400
1	52.8878	-114.470	2470	4000
2	52.8852	-114.468	2940	6600
3	52.9532	-114.321	3350	42847
4	52.9047	-114.485	2060	4500

	Number of schools nearby	Distance from the airport	Price
0	2	58	2380000
1	2	51	1400000
2	1	53	1200000
3	3	76	838000
4	1	51	805000

[5 rows x 23 columns]

df

	id	Date	number of bedrooms	number of bathrooms	\
0	6762810145	42491	5	2.50	
1	6762810635	42491	4	2.50	
2	6762810998	42491	5	2.75	
3	6762812605	42491	4	2.50	
4	6762812919	42491	3	2.00	
...	...	...	...	...	
14615	6762830250	42734	2	1.50	
14616	6762830339	42734	3	2.00	
14617	6762830618	42734	2	1.00	
14618	6762830709	42734	4	1.00	
14619	6762831463	42734	3	1.00	

	living area	lot area	number of floors	waterfront present	\
0	3650	9050	2.0	0	
1	2920	4000	1.5	0	
2	2910	9480	1.5	0	
3	3310	42998	2.0	0	
4	2710	4500	1.5	0	
...	...	...	...	...	
14615	1556	20000	1.0	0	
14616	1680	7000	1.5	0	
14617	1070	6120	1.0	0	
14618	1030	6621	1.0	0	
14619	900	4770	1.0	0	

	number of views	condition of the house	...	Built Year	\
0	4	5	...	1921	
1	0	5	...	1909	
2	0	3	...	1939	
3	0	3	...	2001	
4	0	4	...	1929	
...	...	...	...	...	
14615	0	4	...	1957	
14616	0	4	...	1968	
14617	0	3	...	1962	
14618	0	4	...	1955	
14619	0	3	...	1969	

living_area_renov \	Renovation Year	Postal Code	Lattitude	Longitude
0	0	122003	52.8645	-114.557
2880				
1	0	122004	52.8878	-114.470
2470				
2	0	122004	52.8852	-114.468
2940				
3	0	122005	52.9532	-114.321
3350				
4	0	122006	52.9047	-114.485
2060				
...	...	...	...	...
...				
14615	0	122066	52.6191	-114.472
2250				
14616	0	122072	52.5075	-114.393
1540				
14617	0	122056	52.7289	-114.507
1130				
14618	0	122042	52.7157	-114.411
1420				
14619	2009	122018	52.5338	-114.552
900				

lot_area_renov	Number of schools nearby	Distance from the airport \
0	5400	2
58		
1	4000	2
51		
2	6600	1
53		
3	42847	3
76		
4	4500	1
51		
...	...	...
...		
14615	17286	3
76		
14616	7480	3
59		
14617	6120	2
64		
14618	6631	3
54		
14619	3480	2

55

	Price
0	2380000
1	1400000
2	1200000
3	838000
4	805000
...	...
14615	221700
14616	219200
14617	209000
14618	205000
14619	146000

[14620 rows x 23 columns]

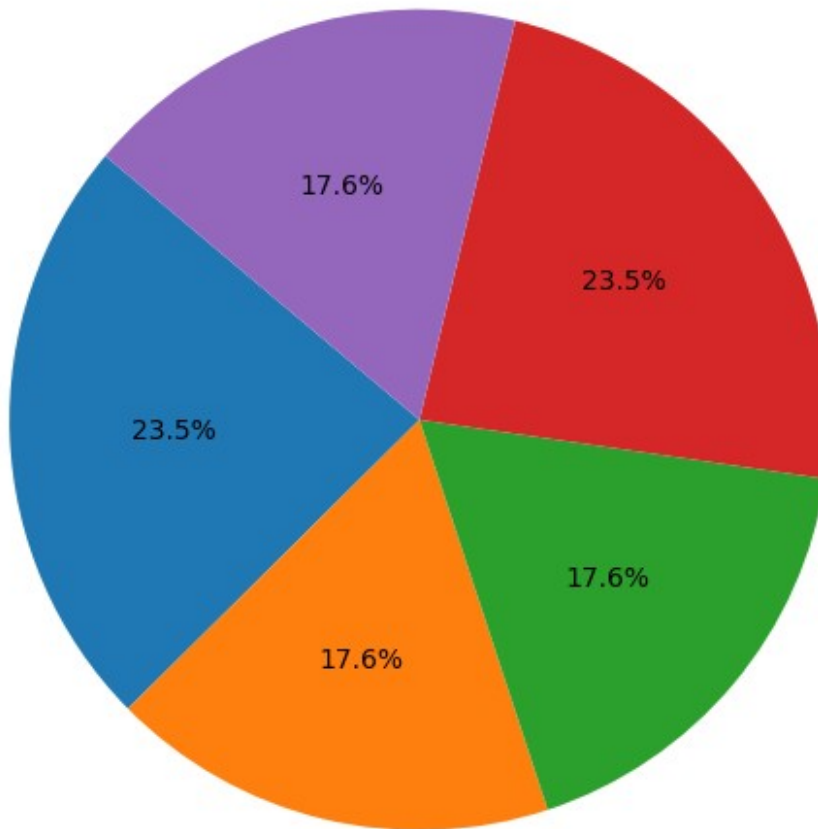
df.shape

(14620, 23)

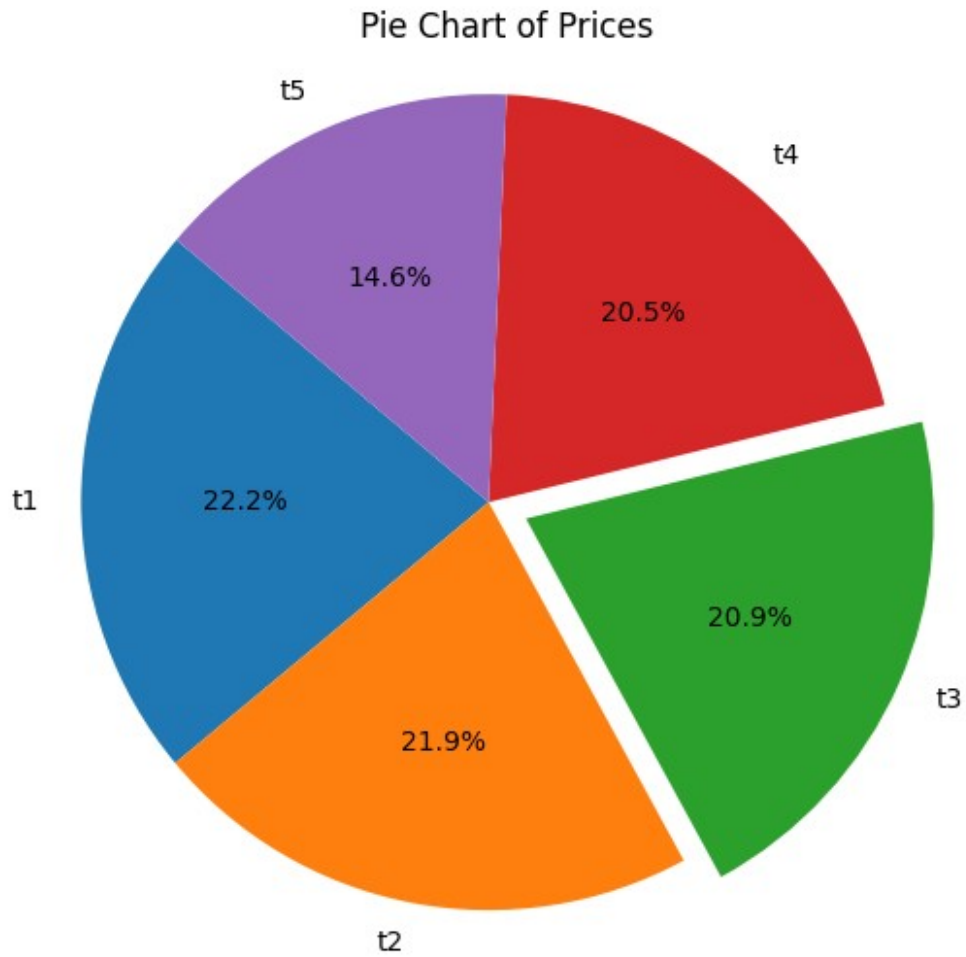
## Univariate Analysis

```
#Pie plot of the df(Head)
plt.figure(figsize=(6, 6)) # Set the figure size
plt.pie(dfh['number of floors'],autopct='%1.1f%%', startangle=140)
plt.title('Pie Chart of number of floors')
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a
circle.
plt.show()
```

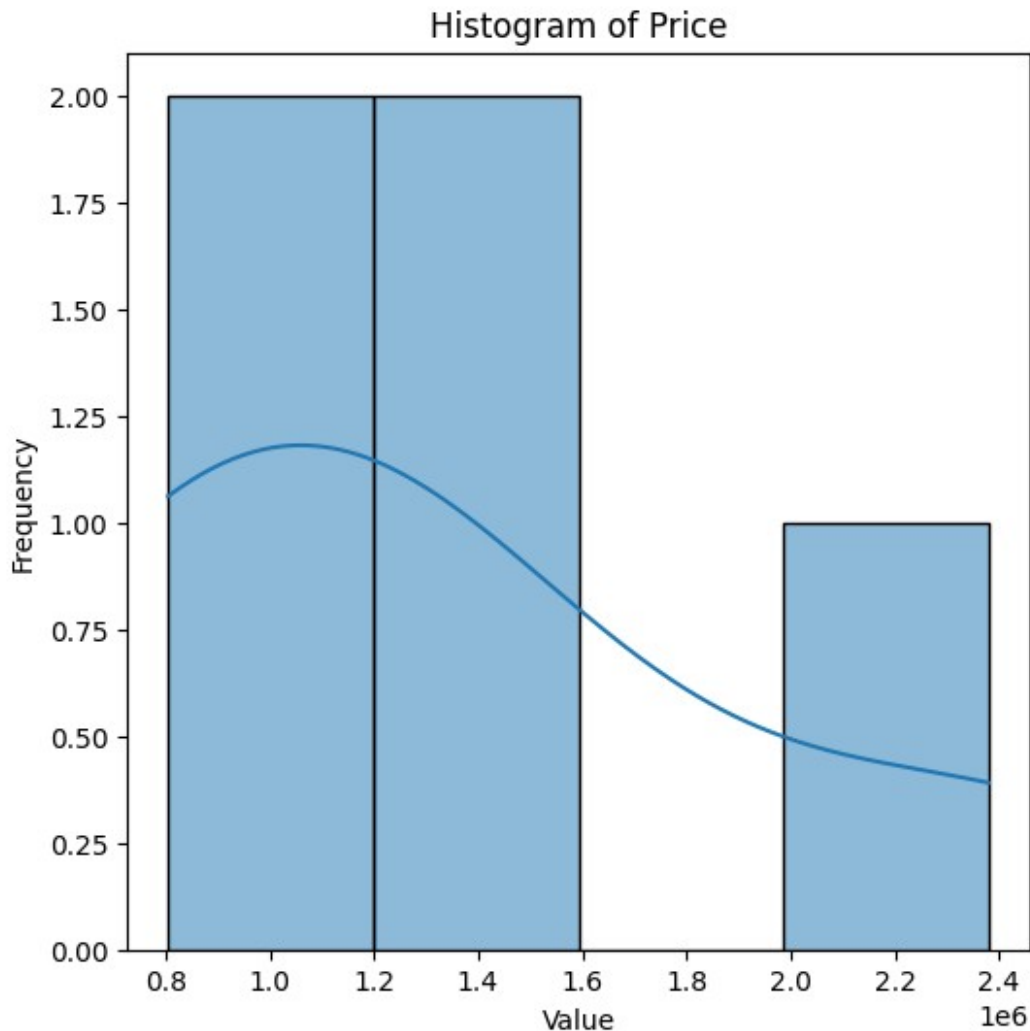
Pie Chart of number of floors



```
#pie plot of df(tail)
plt.figure(figsize=(6, 6)) # Set the figure size
plt.pie(dfl['Price'],
[0,0,0.1,0,0],labels=['t1','t2','t3','t4','t5'],autopct='%1.1f%%',
startangle=140)
plt.title('Pie Chart of Prices')
plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a
circle.
plt.show()
```



```
#histogram plot of df(head)
plt.figure(figsize=(6, 6))
sns.histplot(dfh['Price'], kde=True)
plt.title('Histogram of Price')
plt.xlabel('Value')
plt.ylabel('Frequency')
plt.show()
```



```
sns.distplot(df['living area'])
```

```
<ipython-input-8-f6cb9bd3998b>:1: UserWarning:
```

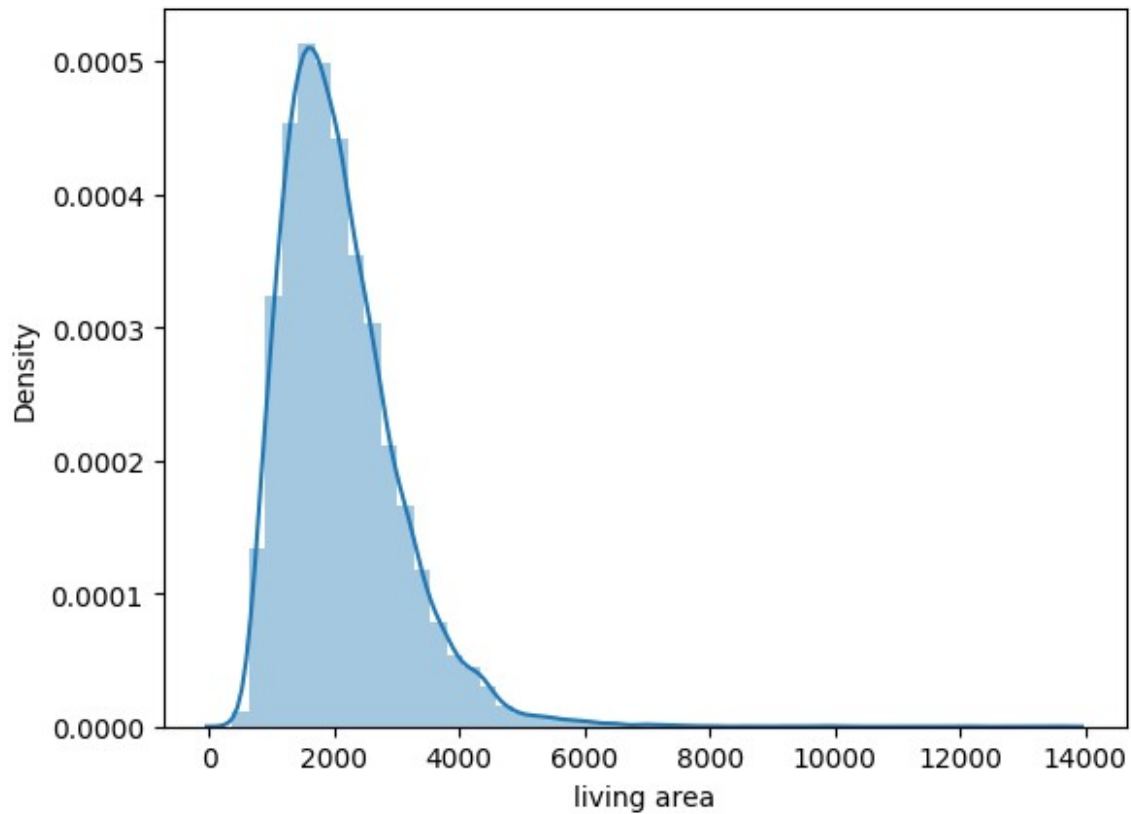
```
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.
```

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>

```
sns.distplot(df['living area'])
```

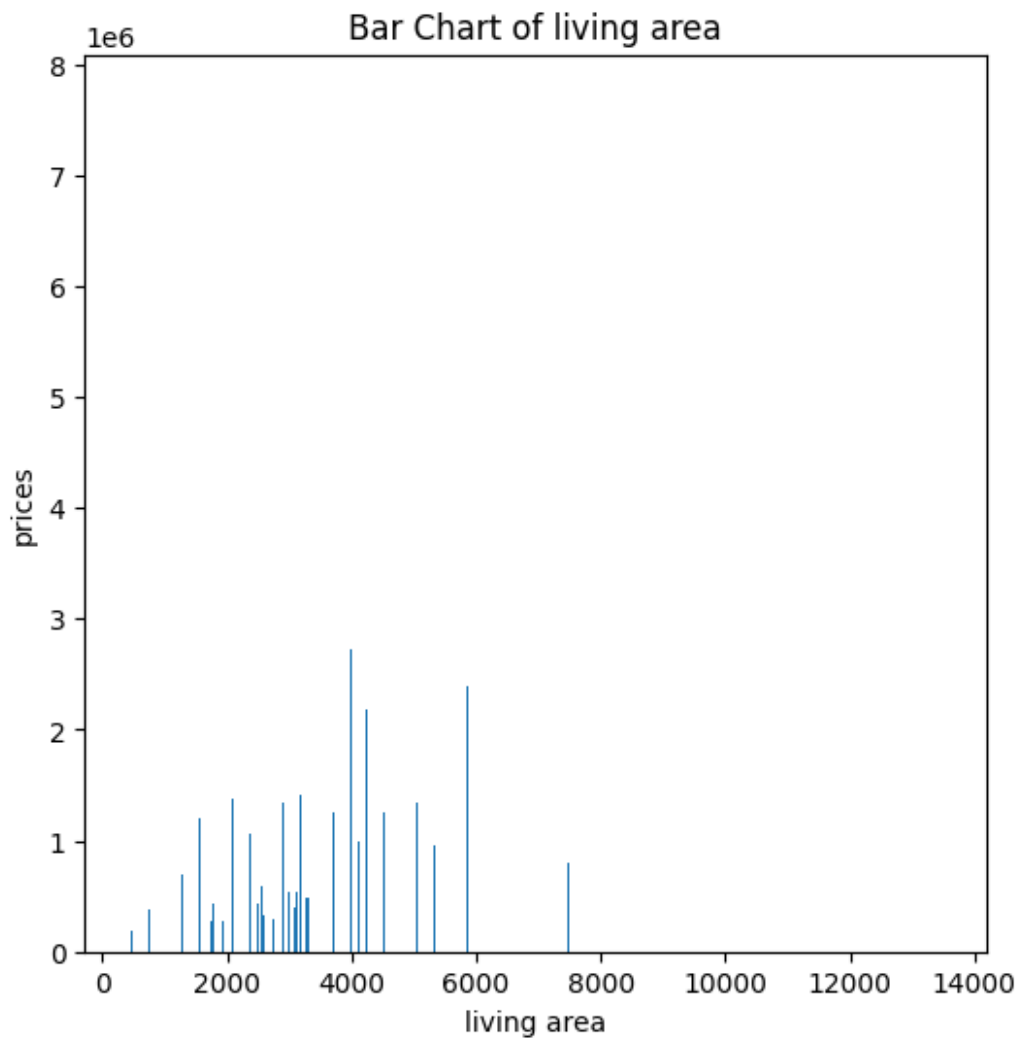
```
<Axes: xlabel='living area', ylabel='Density'>
```



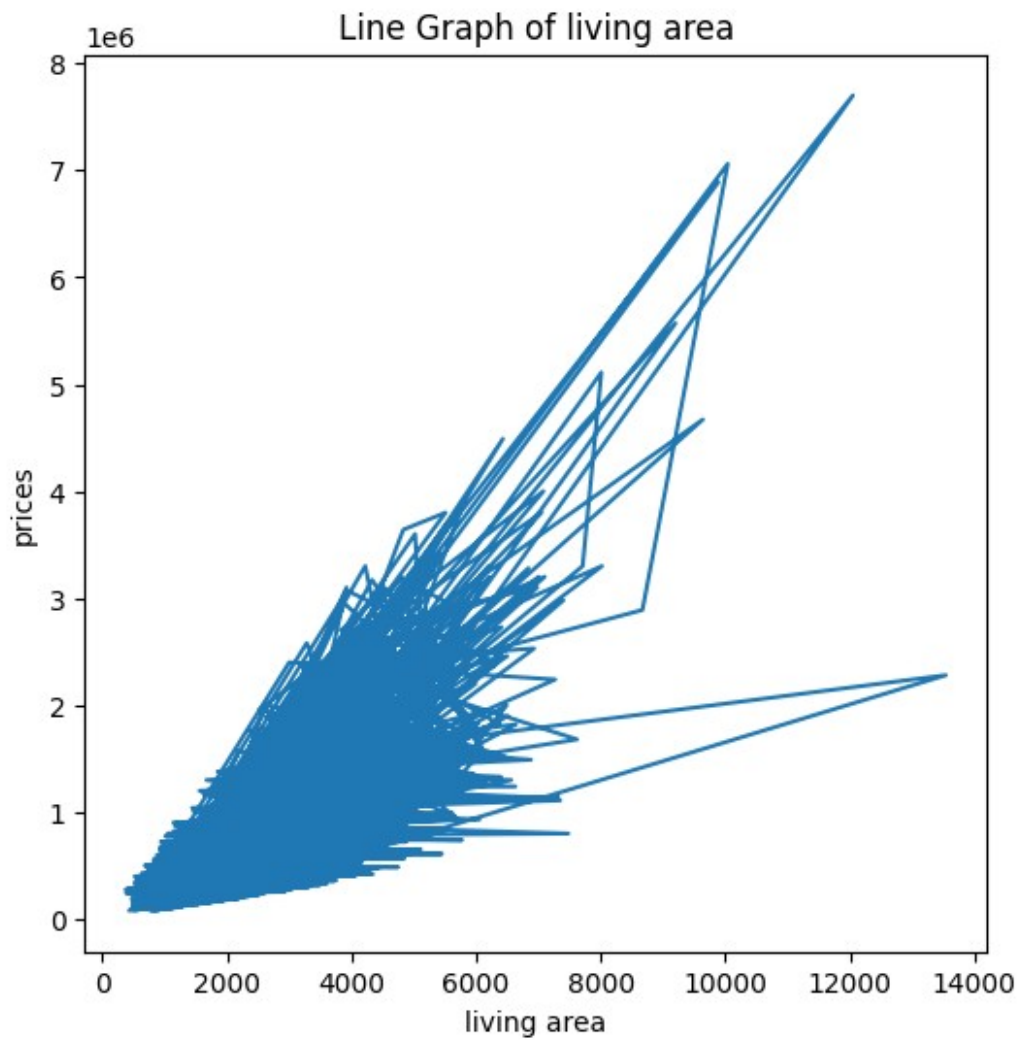
### Bivariate Analysis

```
#Barplot of df [Comparison between 'living area' feature and 'Price']  
plt.figure(figsize=(6, 6))  
plt.bar(df['living area'], df['Price'])  
plt.title('Bar Chart of living area')  
plt.xlabel('living area')  
plt.ylabel('prices')  
plt.show()
```

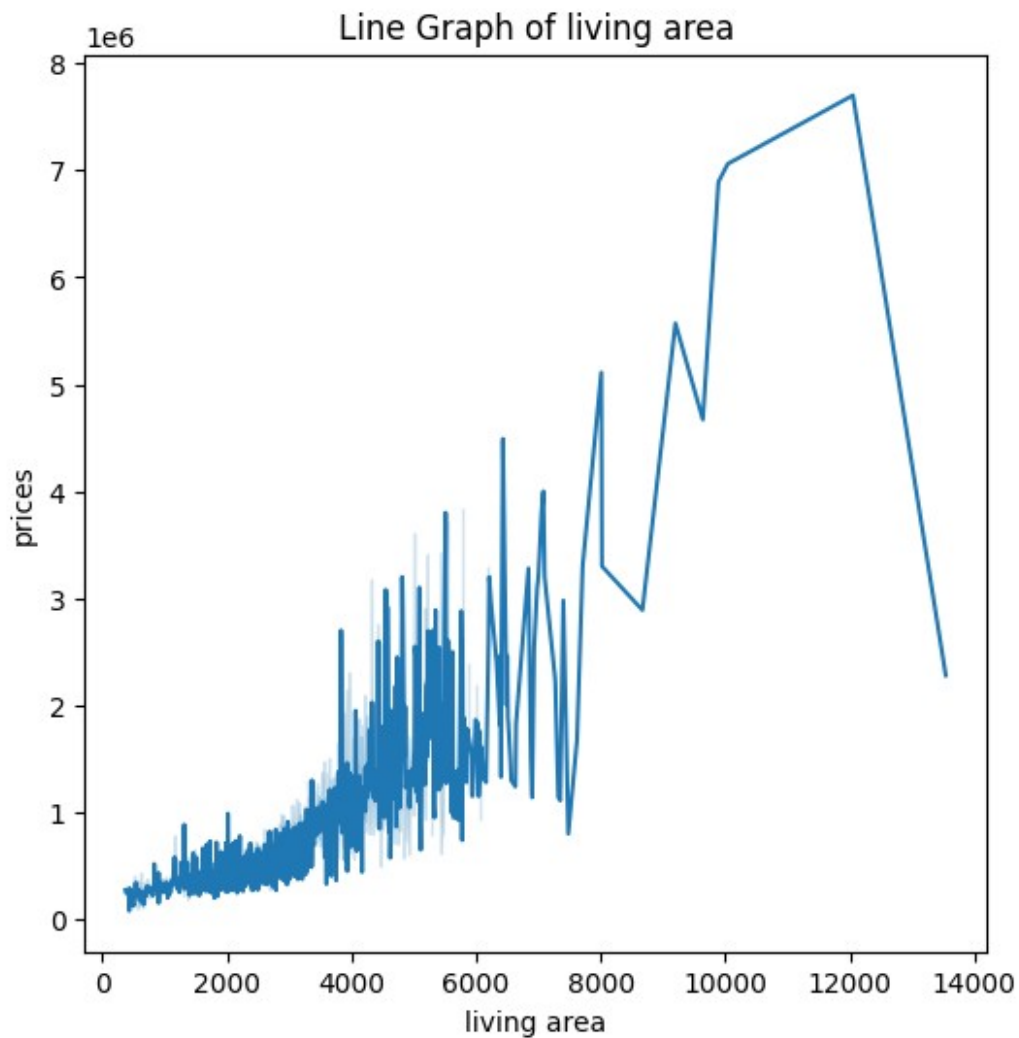




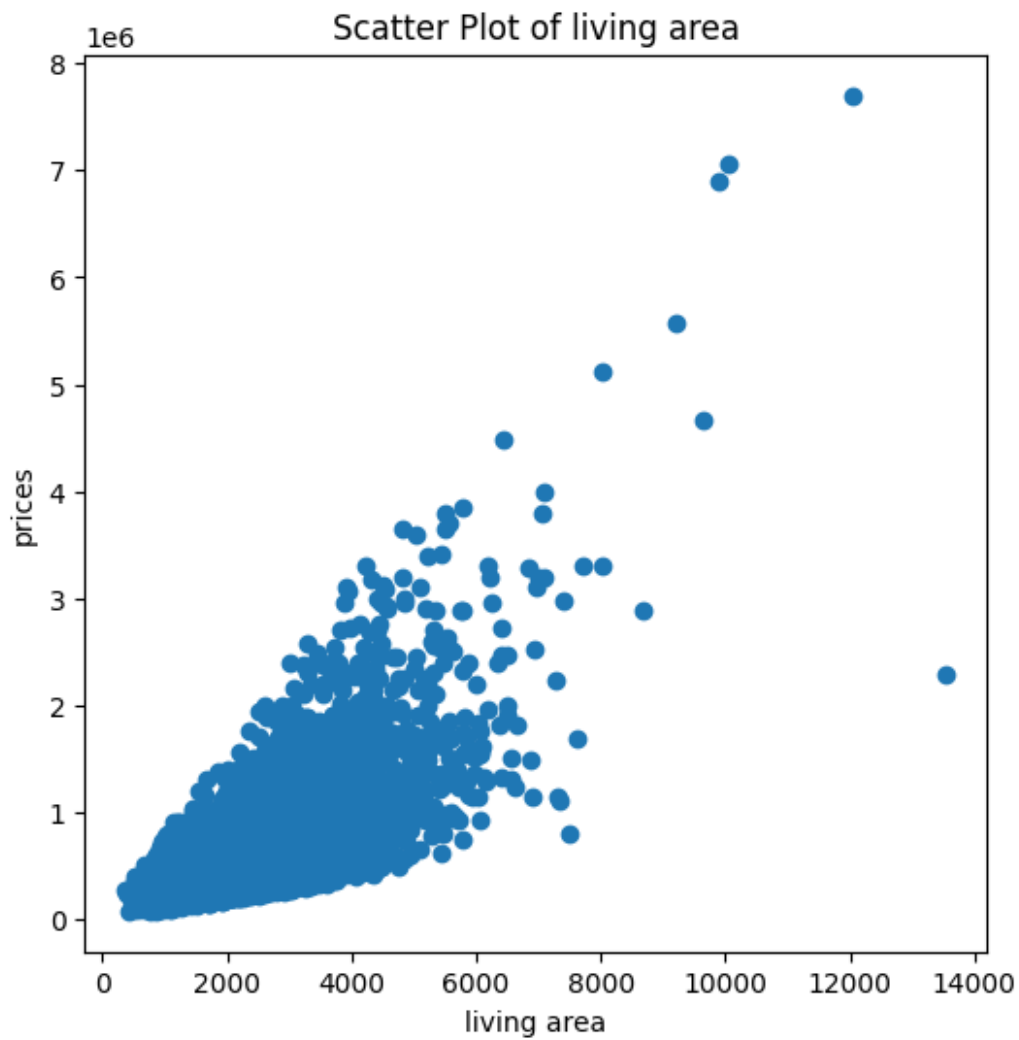
```
#linegraph of df [Comparison between 'living area' feature and  
'Price']  
plt.figure(figsize=(6, 6))  
plt.plot(df['living area'], df['Price'])  
plt.title('Line Graph of living area')  
plt.xlabel('living area')  
plt.ylabel('prices')  
plt.show()
```



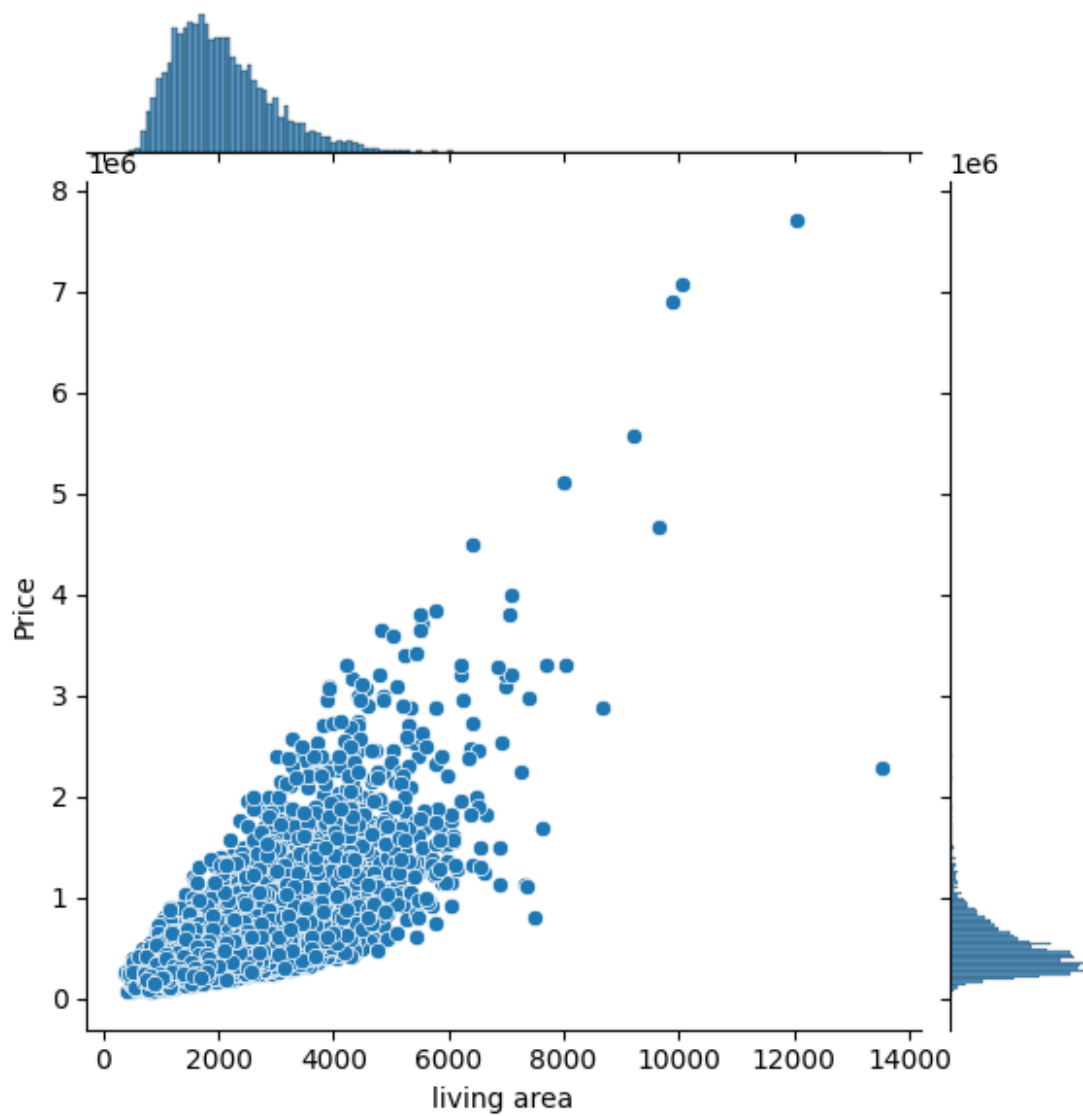
```
#lineplot of df [Comparison between 'living area' feature and 'Price']  
plt.figure(figsize=(6, 6))  
sns.lineplot(x= df['living area'], y= df['Price'])  
plt.title('Line Graph of living area')  
plt.xlabel('living area')  
plt.ylabel('prices')  
plt.show()
```



```
#Satterplot of df [Comparison between 'living area' feature and  
'Price']  
plt.figure(figsize=(6, 6))  
plt.scatter(df['living area'], df['Price'])  
plt.title('Scatter Plot of living area')  
plt.xlabel('living area')  
plt.ylabel('prices')  
plt.show()
```

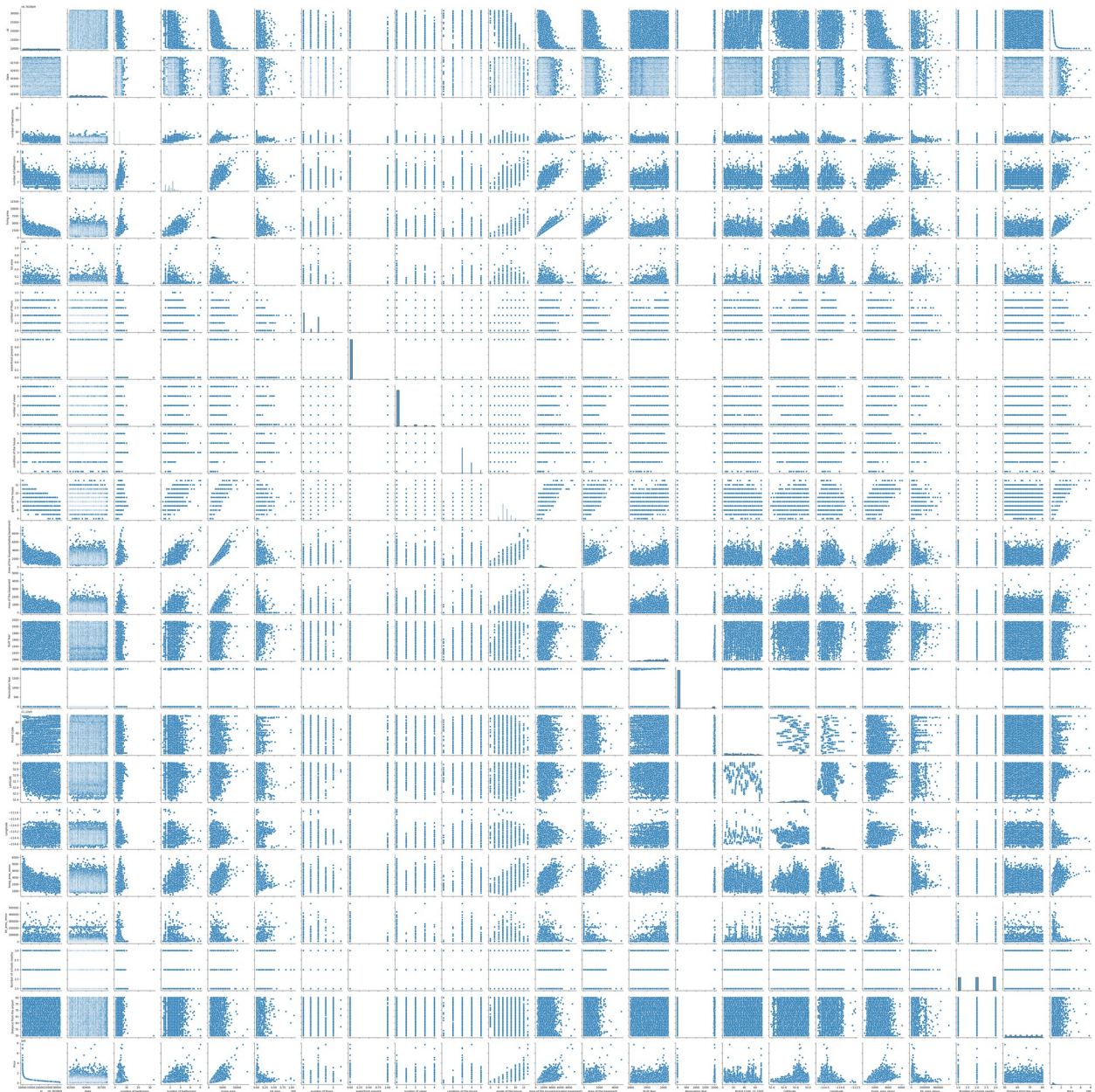


```
# JointPlot [Comparison between 'living area' feature and 'Price']  
sns.jointplot(x='living area',y= 'Price',data =df)  
<seaborn.axisgrid.JointGrid at 0x7e171bd75d80>
```



### Multivariate Analysis

```
#Pair Plot  
sns.pairplot(df)  
  
<seaborn.axisgrid.PairGrid at 0x7e170055edd0>
```



```
df.shape
```

```
(14620, 23)
```

```
df.corr()
```

	id	Date	number of
bedrooms \			
id	1.000000	0.045966	-
0.329034			
Date	0.045966	1.000000	-
0.015663			
number of bedrooms	-0.329034	-0.015663	

1.000000			
number of bathrooms	-0.516909	-0.026485	
0.509784			
living area	-0.648127	-0.021958	
0.570526			
lot area	-0.100269	0.004392	
0.034416			
number of floors	-0.312305	-0.010335	
0.177294			
waterfront present	-0.112937	0.012006	-
0.006257			
number of views	-0.293004	-0.004782	
0.078665			
condition of the house	-0.045061	-0.027402	
0.026597			
grade of the house	-0.673448	-0.033097	
0.352945			
Area of the house(excluding basement)	-0.565116	-0.015994	
0.473599			
Area of the basement	-0.290806	-0.015711	
0.300332			
Built Year	-0.068645	-0.005869	
0.152954			
Renovation Year	-0.109155	-0.011636	
0.016132			
Postal Code	0.294709	0.018243	-
0.044156			
Lattitude	-0.479334	-0.023327	-
0.013163			
Longitude	-0.070841	-0.018231	
0.135712			
living_area_renov	-0.599900	-0.032495	
0.389855			
lot_area_renov	-0.089604	-0.000050	
0.029400			
Number of schools nearby	-0.004821	-0.004071	
0.003397			
Distance from the airport	-0.004542	0.011457	-
0.006157			
Price	-0.773114	-0.027919	
0.308460			

	number of bathrooms	living
area \		
id	-0.516909	-
0.648127		
Date	-0.026485	-
0.021958		
number of bedrooms	0.509784	

0.570526		
number of bathrooms	1.000000	
0.753517		
living area	0.753517	
1.000000		
lot area	0.080806	
0.174420		
number of floors	0.502924	
0.354743		
waterfront present	0.060104	
0.105837		
number of views	0.183789	
0.287728		
condition of the house	-0.128232	-
0.063358		
grade of the house	0.663054	
0.761835		
Area of the house(excluding basement)	0.684391	
0.875793		
Area of the basement	0.287190	
0.441491		
Built Year	0.498127	
0.309602		
Renovation Year	0.049669	
0.059400		
Postal Code	-0.105546	-
0.080303		
Lattitude	0.031156	
0.054518		
Longitude	0.223904	
0.240208		
living_area_renov	0.570530	
0.757571		
lot_area_renov	0.078627	
0.180312		
Number of schools nearby	0.002180	
0.002370		
Distance from the airport	0.009206	
0.002511		
Price	0.531735	
0.712169		
	lot area	number of floors \
id	-0.100269	-0.312305
Date	0.004392	-0.010335
number of bedrooms	0.034416	0.177294
number of bathrooms	0.080806	0.502924
living area	0.174420	0.354743
lot area	1.000000	-0.004138



number of floors	-0.004138	1.000000
waterfront present	0.026282	0.016316
number of views	0.078308	0.020153
condition of the house	-0.008548	-0.269928
grade of the house	0.110546	0.463082
Area of the house(excluding basement)	0.183553	0.525643
Area of the basement	0.019755	-0.242976
Built Year	0.051615	0.481565
Renovation Year	0.006848	0.006705
Postal Code	0.070131	-0.129788
Lattitude	-0.090983	0.050731
Longitude	0.221432	0.127550
living_area_renov	0.149744	0.285093
lot_area_renov	0.706812	-0.010120
Number of schools nearby	-0.012671	-0.007579
Distance from the airport	0.003291	0.016567
Price	0.081992	0.262732

	waterfront present	number of
views \		
id	-0.112937	-
0.293004		
Date	0.012006	-
0.004782		
number of bedrooms	-0.006257	
0.078665		
number of bathrooms	0.060104	
0.183789		
living area	0.105837	
0.287728		
lot area	0.026282	
0.078308		
number of floors	0.016316	
0.020153		
waterfront present	1.000000	
0.400206		
number of views	0.400206	
1.000000		
condition of the house	0.018644	
0.052533		
grade of the house	0.079831	
0.254532		
Area of the house(excluding basement)	0.071865	
0.162672		
Area of the basement	0.085441	
0.293062		
Built Year	-0.024226	-
0.055357		
Renovation Year	0.085865	

0.102944		
Postal Code	0.038318	
0.039268		
Lattitude	-0.021795	-
0.004555		
Longitude	-0.047791	-
0.079706		
living_area_renov	0.085743	
0.281452		
lot_area_renov	0.032055	
0.072300		
Number of schools nearby	0.001563	
0.008004		
Distance from the airport	0.001448	-
0.001657		
Price	0.263687	
0.395973		

	condition of the house	...	\
id	-0.045061	...	
Date	-0.027402	...	
number of bedrooms	0.026597	...	
number of bathrooms	-0.128232	...	
living area	-0.063358	...	
lot area	-0.008548	...	
number of floors	-0.269928	...	
waterfront present	0.018644	...	
number of views	0.052533	...	
condition of the house	1.000000	...	
grade of the house	-0.152530	...	
Area of the house(excluding basement)	-0.167695	...	
Area of the basement	0.180609	...	
Built Year	-0.381718	...	
Renovation Year	-0.062126	...	
Postal Code	0.045334	...	
Lattitude	-0.002998	...	
Longitude	-0.121189	...	
living_area_renov	-0.099743	...	
lot_area_renov	-0.004748	...	
Number of schools nearby	-0.006939	...	
Distance from the airport	-0.002136	...	
Price	0.041376	...	

	Built Year	Renovation Year	\
id	-0.068645	-0.109155	
Date	-0.005869	-0.011636	
number of bedrooms	0.152954	0.016132	
number of bathrooms	0.498127	0.049669	
living area	0.309602	0.059400	

lot area	0.051615	0.006848
number of floors	0.481565	0.006705
waterfront present	-0.024226	0.085865
number of views	-0.055357	0.102944
condition of the house	-0.381718	-0.062126
grade of the house	0.440358	0.014501
Area of the house(excluding basement)	0.419369	0.025727
Area of the basement	-0.138843	0.075104
Built Year	1.000000	-0.233683
Renovation Year	-0.233683	1.000000
Postal Code	-0.062349	0.018006
Lattitude	-0.143153	0.028908
Longitude	0.414591	-0.080050
living_area_renov	0.328625	-0.002601
lot_area_renov	0.072874	0.005869
Number of schools nearby	-0.001631	-0.000826
Distance from the airport	-0.003968	0.005342
Price	0.050307	0.133173

	Postal Code	Lattitude
Longitude \		
id	0.294709	-0.479334 -
0.070841		
Date	0.018243	-0.023327 -
0.018231		
number of bedrooms	-0.044156	-0.013163
0.135712		
number of bathrooms	-0.105546	0.031156
0.223904		
living area	-0.080303	0.054518
0.240208		
lot area	0.070131	-0.090983
0.221432		
number of floors	-0.129788	0.050731
0.127550		
waterfront present	0.038318	-0.021795 -
0.047791		
number of views	0.039268	-0.004555 -
0.079706		
condition of the house	0.045334	-0.002998 -
0.121189		
grade of the house	-0.146342	0.115256
0.203754		
Area of the house(excluding basement)	-0.083730	-0.000088
0.345899		
Area of the basement	-0.010542	0.112989 -
0.145879		
Built Year	-0.062349	-0.143153
0.414591		

Renovation Year	0.018006	0.028908	-
0.080050			
Postal Code	1.000000	-0.310172	-
0.099003			
Lattitude	-0.310172	1.000000	-
0.131472			
Longitude	-0.099003	-0.131472	
1.000000			
living_area_renov	-0.108454	0.046148	
0.341221			
lot_area_renov	0.077483	-0.091622	
0.258066			
Number of schools nearby	0.010605	0.014949	-
0.010163			
Distance from the airport	0.011528	0.007193	-
0.003100			
Price	-0.115908	0.297490	
0.024414			
	living_area_renov		
lot_area_renov \			
id	-0.599900		-
0.089604			
Date	-0.032495		-
0.000050			
number of bedrooms	0.389855		
0.029400			
number of bathrooms	0.570530		
0.078627			
living area	0.757571		
0.180312			
lot area	0.149744		
0.706812			
number of floors	0.285093		-
0.010120			
waterfront present	0.085743		
0.032055			
number of views	0.281452		
0.072300			
condition of the house	-0.099743		-
0.004748			
grade of the house	0.720019		
0.116725			
Area of the house(excluding basement)	0.737744		
0.194670			
Area of the basement	0.196403		
0.011283			
Built Year	0.328625		
0.072874			

Renovation Year	-0.002601	
0.005869		
Postal Code	-0.108454	
0.077483		
Lattitude	0.046148	-
0.091622		
Longitude	0.341221	
0.258066		
living_area_renov	1.000000	
0.189225		
lot_area_renov	0.189225	
1.000000		
Number of schools nearby	-0.001203	-
0.025014		
Distance from the airport	-0.005673	-
0.014587		
Price	0.584924	
0.075535		

	Number of schools nearby \	
id	-0.004821	
Date	-0.004071	
number of bedrooms	0.003397	
number of bathrooms	0.002180	
living area	0.002370	
lot area	-0.012671	
number of floors	-0.007579	
waterfront present	0.001563	
number of views	0.008004	
condition of the house	-0.006939	
grade of the house	0.000986	
Area of the house(excluding basement)	-0.002894	
Area of the basement	0.010284	
Built Year	-0.001631	
Renovation Year	-0.000826	
Postal Code	0.010605	
Lattitude	0.014949	
Longitude	-0.010163	
living_area_renov	-0.001203	
lot_area_renov	-0.025014	
Number of schools nearby	1.000000	
Distance from the airport	0.004035	
Price	0.009890	

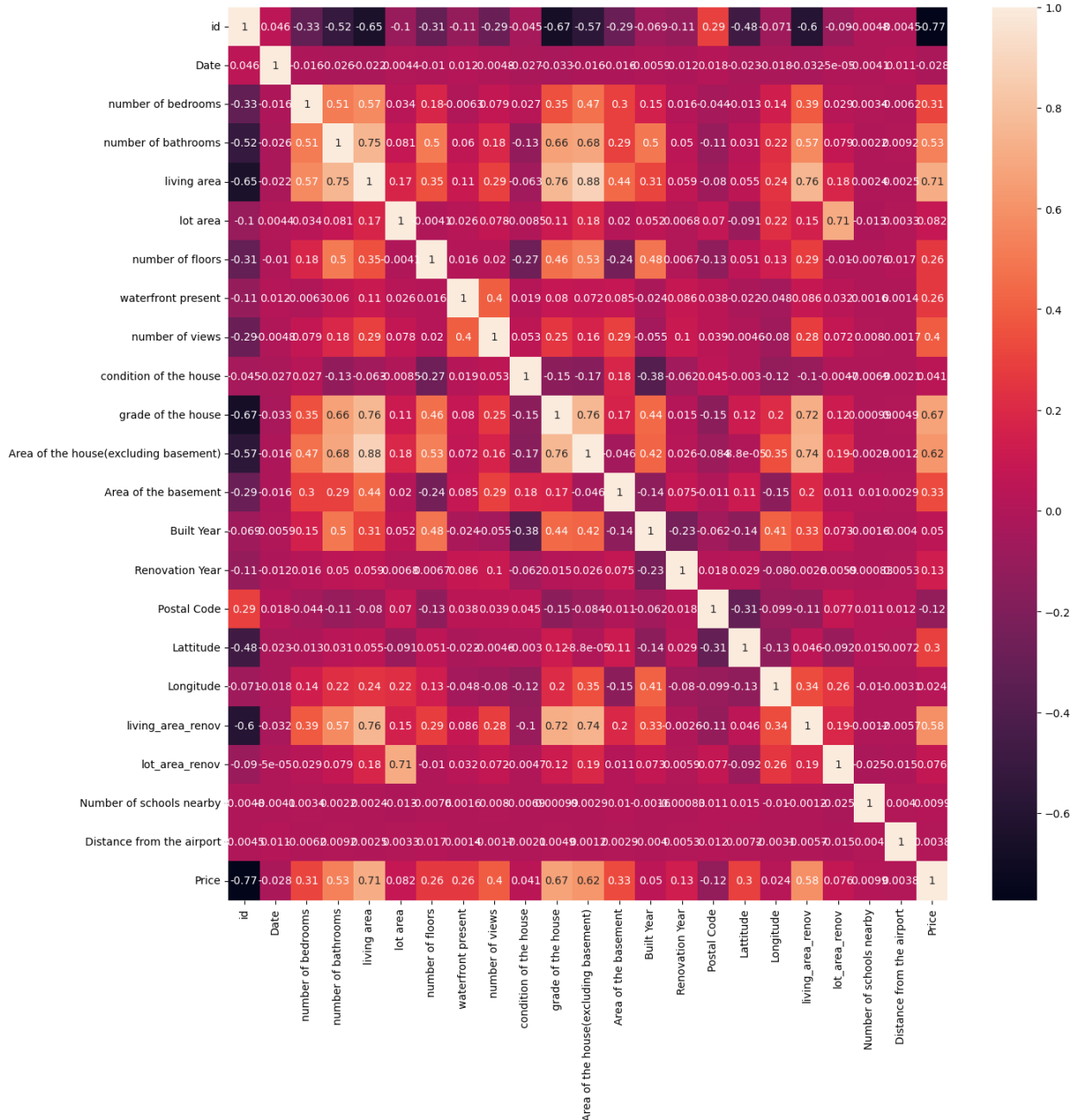
	Distance from the airport	
Price		
id	-0.004542	-
0.773114		
Date	0.011457	-

0.027919	
number of bedrooms	-0.006157
0.308460	
number of bathrooms	0.009206
0.531735	
living area	0.002511
0.712169	
lot area	0.003291
0.081992	
number of floors	0.016567
0.262732	
waterfront present	0.001448
0.263687	
number of views	-0.001657
0.395973	
condition of the house	-0.002136
0.041376	
grade of the house	0.004940
0.671814	
Area of the house(excluding basement)	0.001222
0.615220	
Area of the basement	0.002926
0.330202	
Built Year	-0.003968
0.050307	
Renovation Year	0.005342
0.133173	
Postal Code	0.011528 -
0.115908	
Lattitude	0.007193
0.297490	
Longitude	-0.003100
0.024414	
living_area_renov	-0.005673
0.584924	
lot_area_renov	-0.014587
0.075535	
Number of schools nearby	0.004035
0.009890	
Distance from the airport	1.000000
0.003804	
Price	0.003804
1.000000	

[23 rows x 23 columns]

```
plt.figure(figsize=(15, 15))
sns.heatmap(df.corr(),annot=True)
```

<Axes: >



Description

```
df.describe()
```

	id	Date	number of bedrooms	number of
bathrooms \				
count	1.462000e+04	14620.000000	14620.000000	
14620.000000				
mean	6.762821e+09	42604.538646	3.379343	
2.129583				
std	6.237575e+03	67.347991	0.938719	

0.769934			
min	6.762810e+09	42491.000000	1.000000
0.500000			
25%	6.762815e+09	42546.000000	3.000000
1.750000			
50%	6.762821e+09	42600.000000	3.000000
2.250000			
75%	6.762826e+09	42662.000000	4.000000
2.500000			
max	6.762832e+09	42734.000000	33.000000
8.000000			

	living area	lot area	number of floors	waterfront
present \				
count	14620.000000	1.462000e+04	14620.000000	
14620.000000				
mean	2098.262996	1.509328e+04	1.502360	
0.007661				
std	928.275721	3.791962e+04	0.540239	
0.087193				
min	370.000000	5.200000e+02	1.000000	
0.000000				
25%	1440.000000	5.010750e+03	1.000000	
0.000000				
50%	1930.000000	7.620000e+03	1.500000	
0.000000				
75%	2570.000000	1.080000e+04	2.000000	
0.000000				
max	13540.000000	1.074218e+06	3.500000	
1.000000				

	number of views	condition of the house	...	Built Year	\
count	14620.000000	14620.000000	...	14620.000000	
mean	0.233105	3.430506	...	1970.926402	
std	0.766259	0.664151	...	29.493625	
min	0.000000	1.000000	...	1900.000000	
25%	0.000000	3.000000	...	1951.000000	
50%	0.000000	3.000000	...	1975.000000	
75%	0.000000	4.000000	...	1997.000000	
max	4.000000	5.000000	...	2015.000000	

	Renovation Year	Postal Code	Latitude	Longitude	\
count	14620.000000	14620.000000	14620.000000	14620.000000	
mean	90.924008	122033.062244	52.792848	-114.404007	
std	416.216661	19.082418	0.137522	0.141326	
min	0.000000	122003.000000	52.385900	-114.709000	
25%	0.000000	122017.000000	52.707600	-114.519000	
50%	0.000000	122032.000000	52.806400	-114.421000	
75%	0.000000	122048.000000	52.908900	-114.315000	
max	2015.000000	122072.000000	53.007600	-113.505000	



	living_area_renov	lot_area_renov	Number of schools nearby \
count	14620.000000	14620.000000	14620.000000
mean	1996.702257	12753.500068	2.012244
std	691.093366	26058.414467	0.817284
min	460.000000	651.000000	1.000000
25%	1490.000000	5097.750000	1.000000
50%	1850.000000	7620.000000	2.000000
75%	2380.000000	10125.000000	3.000000
max	6110.000000	560617.000000	3.000000

	Distance from the airport	Price
count	14620.000000	1.462000e+04
mean	64.950958	5.389322e+05
std	8.936008	3.675324e+05
min	50.000000	7.800000e+04
25%	57.000000	3.200000e+05
50%	65.000000	4.500000e+05
75%	73.000000	6.450000e+05
max	80.000000	7.700000e+06

[8 rows x 23 columns]

Handling the missing values

```

null_values = df.isnull().sum()
null_values

id                                0
Date                              0
number of bedrooms                0
number of bathrooms               0
living area                       0
lot area                          0
number of floors                  0
waterfront present                0
number of views                   0
condition of the house            0
grade of the house                0
Area of the house(excluding basement) 0
Area of the basement              0
Built Year                        0
Renovation Year                   0
Postal Code                       0
Lattitude                         0
Longitude                         0
living_area_renov                 0
lot_area_renov                    0
Number of schools nearby           0

```

```
Distance from the airport      0
Price                          0
dtype: int64
```

```
# Identify columns with missing values
```

```
missing_values = df.columns[df.isnull().any()]
```

```
missing_values
```

```
Index([], dtype='object')
```

```
# Calculate the median for each column with missing values and store it in a dictionary
```

```
median_values = {}
```

```
for column in missing_values:
```

```
    median = df[column].median()
```

```
    median_values[column] = median
```

```
median_values
```

```
df.fillna(median_values, inplace=True)
```