

J AISRI A R

Perform the Below Tasks to complete the assignment:- Tasks:-

1. Download the dataset
2. Load the dataset.
3. Perform the Below Visualizations. • Univariate Analysis • Bi - Variate Analysis • Multi-Variate Analysis
4. Perform descriptive statistics on the dataset.
5. Handle the Missing values.

```
from google.colab import files
uploaded = files.upload()
```

Choose Files House Price India.csv

• **House Price India.csv**(text/csv) - 1524561 bytes, last modified: 3/27/2023 - 100% done

Saving House Price India.csv to House Price India.csv

```
import pandas as pd
import seaborn as sns
import numpy as np
import io
df = pd.read_csv('House Price India.csv')
df.head()
```

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house	...	Built Year	Renovation Year	Postal Code	Latt
0	6762810145	42491	5	2.50	3650	9050	2.0	0	4	5	...	1921	0	122003	52
1	6762810635	42491	4	2.50	2920	4000	1.5	0	0	5	...	1909	0	122004	52
2	6762810998	42491	5	2.75	2910	9480	1.5	0	0	3	...	1939	0	122004	52
3	6762812605	42491	4	2.50	3310	42998	2.0	0	0	3	...	2001	0	122005	52
4	6762812919	42491	3	2.00	2710	4500	1.5	0	0	4	...	1929	0	122006	52

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house	...	Built Year	Renovation Year	Postal Code	
14615	6762830250	42734	2	1.5	1556	20000	1.0	0	0	4	...	1957	0	122066	
14616	6762830339	42734	3	2.0	1680	7000	1.5	0	0	4	...	1968	0	122072	
14617	6762830618	42734	2	1.0	1070	6120	1.0	0	0	3	...	1962	0	122056	
14618	6762830709	42734	4	1.0	1030	6621	1.0	0	0	4	...	1955	0	122042	
14619	6762831463	42734	3	1.0	900	4770	1.0	0	0	3	...	1969	2009	122018	

```

    id    Date    number of bedrooms    number of bathrooms    living area    lot area    number of floors    waterfront present    number of views    condition of the house    ...    Built Year    Renovation Year    Postal Code

0    6762810145    42491    5    2.50    3650    9050    2.0    0    4    5    ...    1921    0    122003
1    6762810635    42491    4    2.50    2920    4000    1.5    0    0    5    ...    1909    0    122004
2    6762810998    42491    5    2.75    2910    9480    1.5    0    0    3    ...    1939    0    122004
3    6762812605    42491    4    2.50    3310    42998    2.0    0    0    3    ...    2001    0    122005
45    )    2710

Index(['id', 'Date',
      'number of bedrooms',
      'number of bathrooms',
      'living area',
      'lot area',
      'number of floors',
      'waterfront present',
      'number of views',
      'condition of the house',
      'grade of the house',
      'Area of the house(excluding basement)',
      'Area of the basement',
      'Built Year',
      'Renovation Year',
      'Postal Code',
      'Latitude',
      'Longitude',
      'living_area_renov',
      'lot_area_renov',
      'Number of schools nearby',
      'Distance from the airport',
      'Price'],
      dtype='object',
      name='data')

<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   Latitude                           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
```

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	

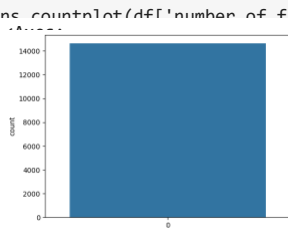
	Renovation Year	Postal Code	Latitude	Longitude	\
count	14620.000000	14620.000000	14620.000000	14620.000000	
mean	90.924008	122033.062244	52.792848	-114.404007	
std	416.216661	19.082418	0.137522	0.141326	
min	0.000000	122003.000000	52.385900	-114.709000	
25%	0.000000	122017.000000	52.707600	-114.519000	
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	

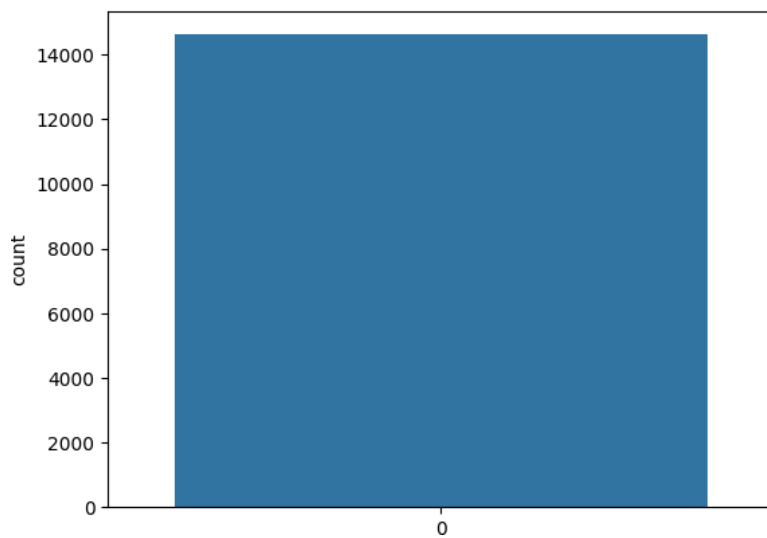
	Distance from the airport	Price
count	14620.000000	1.462000e+04
mean	64.950958	5.389322e+05
std	8.936008	3.675324e+05
min	50.000000	7.800000e+04
25%	57.000000	3.200000e+05
50%	65.000000	4.500000e+05
75%	73.000000	6.450000e+05
max	80.000000	7.700000e+06

[8 rows x 23 columns]

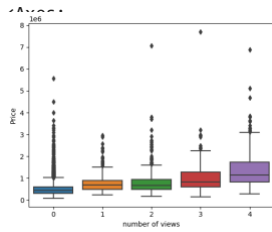
```
(array([1.2916e+04, 1.4260e+03, 1.9100e+02, 6.0000e+01, 1.9000e+01,  
       2.0000e+00, 2.0000e+00, 1.0000e+00, 1.0000e+00, 2.0000e+00]),  
 array([ 78000., 840200., 1602400., 2364600., 3126800., 3889000.,  
       4651200., 5413400., 6175600., 6937800., 7700000.])),  
sns.countplot(df['number of floors'])
```



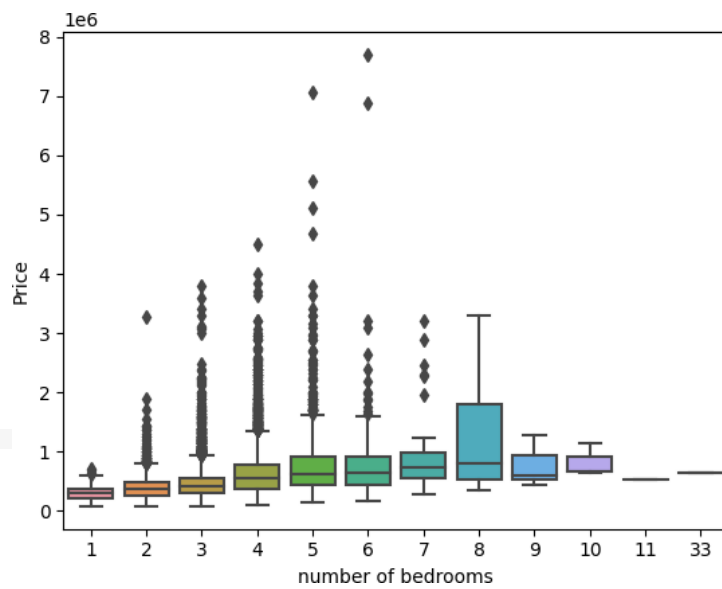
```
sns.countplot(df['number of bathrooms'])  
<Axes: ylabel='count'>
```



```
<Axes: xlabel='Price'>
```



```
<Axes: xlabel='number of bedrooms', ylabel='Price'>
```

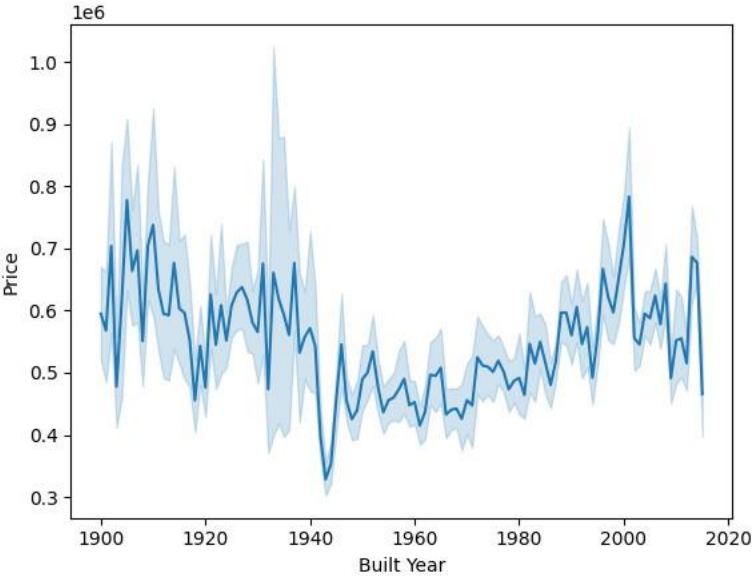


```
sns.lineplot(x=df.g
```

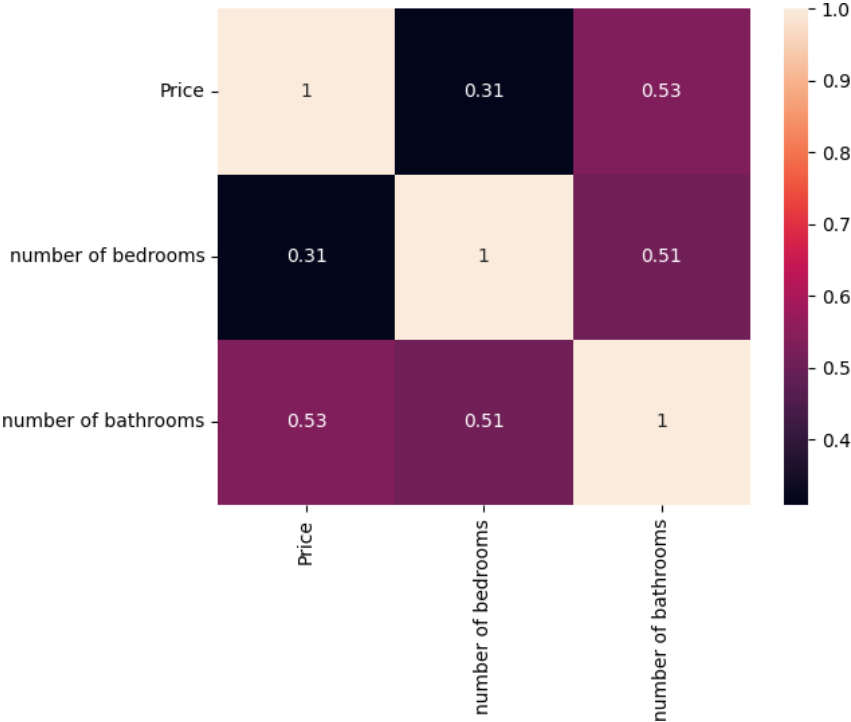


```
sns.lineplot(x=df['Built Year'],y=df['Price'])
```

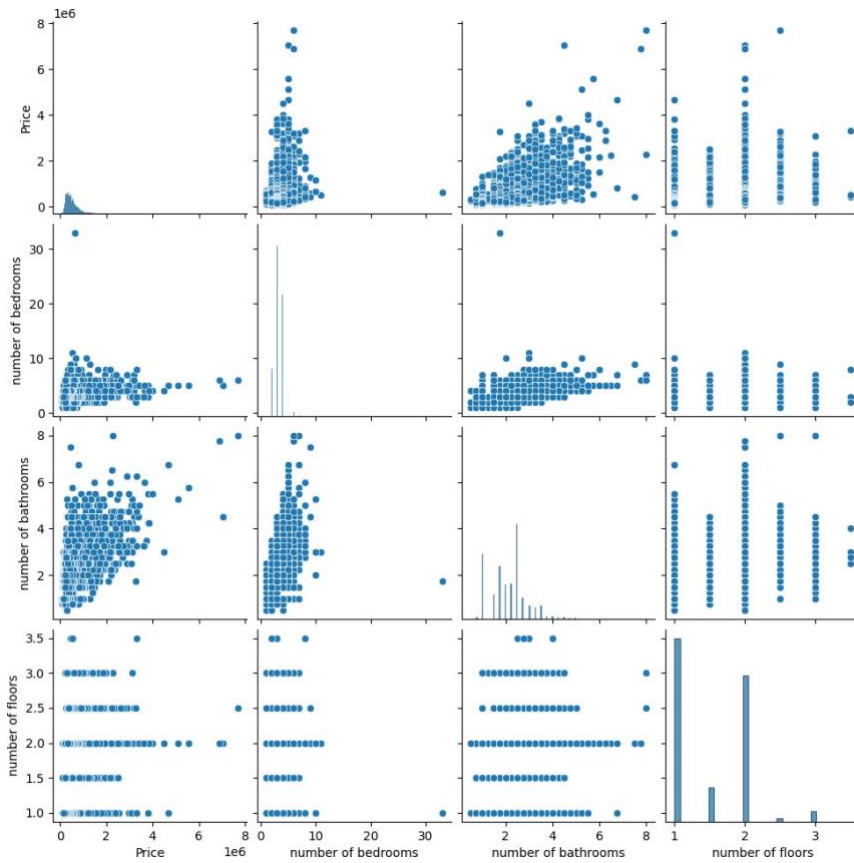
<Axes: xlabel='Built Year', ylabel='Price'>



<Axes: >



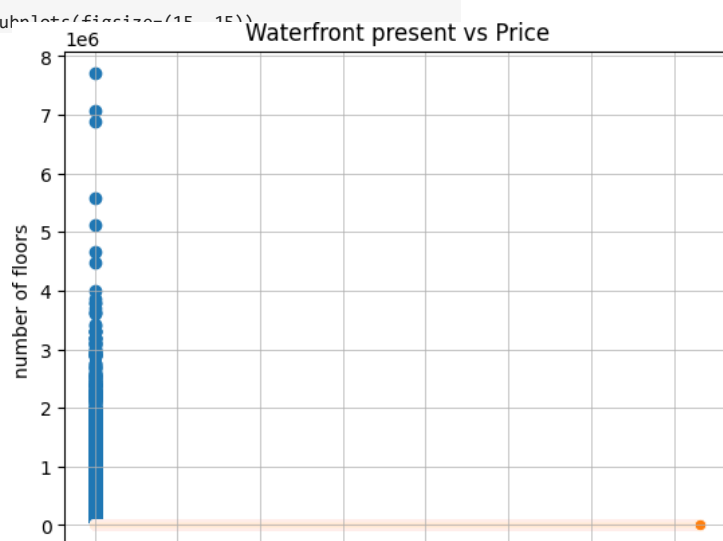
```
<seaborn.axisgrid.PairGrid at 0x7fcd63f242b0>
```

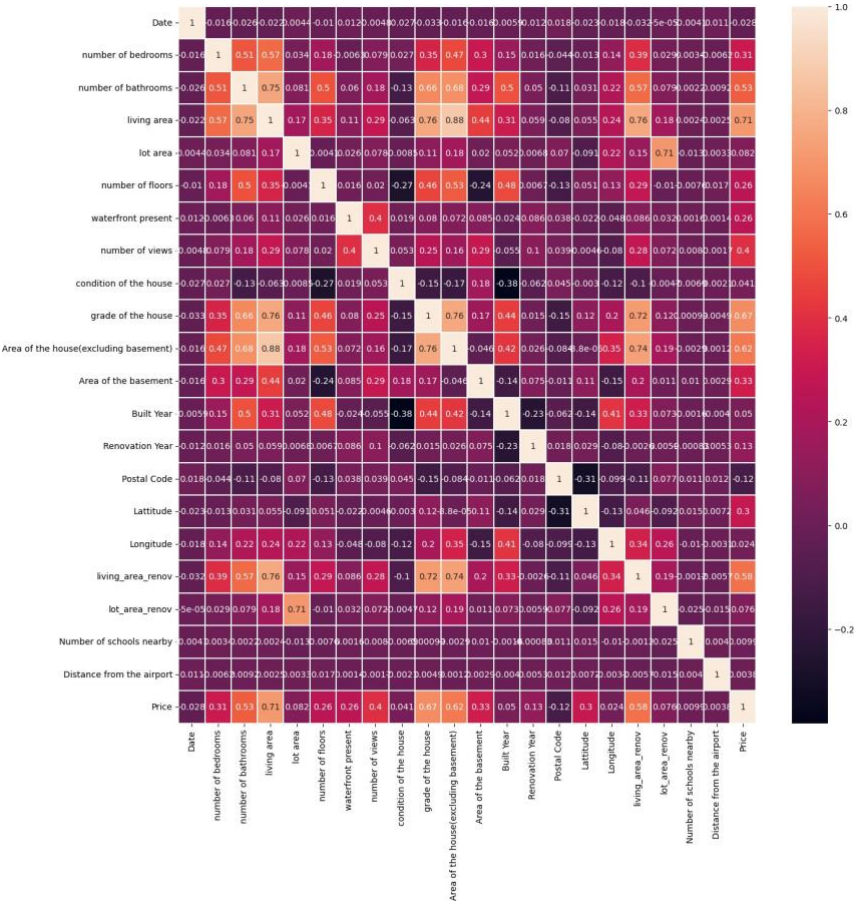


0

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1

```
plt.subplots(figsize=(15, 15))
```





```
id 14620
Date 14620
number of bedrooms 14620
number of bathrooms 14620
living area 14620
lot area 14620
```



```

number of floors          14620
waterfront present        14620
number of views           14620
condition of the house    14620
grade of the house        14620
Area of the house(excluding basement) 14620
Area of the basement      14620
Built Year                14620
Renovation Year           14620
Postal Code               14620
Latitude                  14620
Longitude                 14620
living_area_renov         14620
lot_area_renov            14620
Number of schools nearby   14620
Distance from the airport  14620
Price                     14620
dtype: int64

```

```

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
Latitude              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
Latitude                  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

```

```

std      6.237575e+03      67.347991      0.9387
min      6.762810e+09  42491.000000      1.0000

```

	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	Lattitude	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	\
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std	691.093366	26058.414467	0.817284	
min	460.000000	651.000000	1.000000	
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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

3 4973
2 4853
1 4794
Name: Number of schools nearby, dtype: int64

p
Mean: 64.95095759233926
Median: 0.0
Mode: 0 7
Name: grade of the house, dtype: int64

df.dropna(inplace=True)
d
f

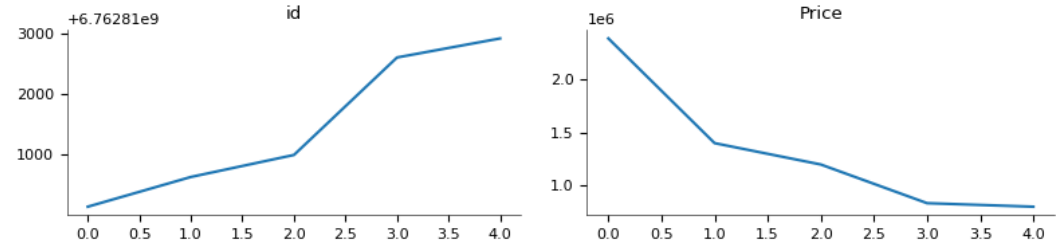
	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house
id								
6762810145	5	2.50	3650	9050	2.0	0	4	5
print(df.isnull().sum())								
6762810998	5	2.75	2910	9480	1.5	0	0	3
id								
6762812605	4	2.50	3310	42998				
number of bedrooms								
number of bathrooms								
living area								
lot area								
number of floors								
waterfront present								
number of views								
condition of the house								
grade of the house								
Area of the house(excluding basement)								
Area of the basement								
Built Year								
Renovation Year								
Postal Code								
Latitude								
Longitude								
living_area_renov								
lot_area_renov								
Number of schools nearby								
Distance from the airport								
Price								
dtype: int64								

s

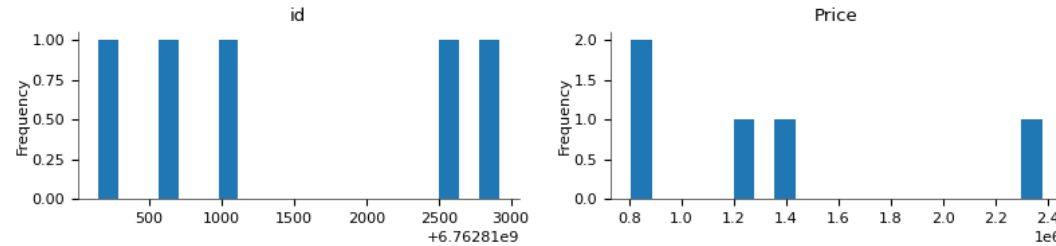
	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house
0	0.12500	0.266667	0.249051	0.007945	0.4			
1	0.09375	0.266667	0.193622	0.003241	0.2			
2	0.12500	0.266667	0.193622	0.003241	0.2			

	id	Price
0	6762810145	2380000
1	6762810635	1400000
2	6762810998	1200000
3	6762812605	838000
4	6762812919	805000

Values



Distributions



2-d distributions

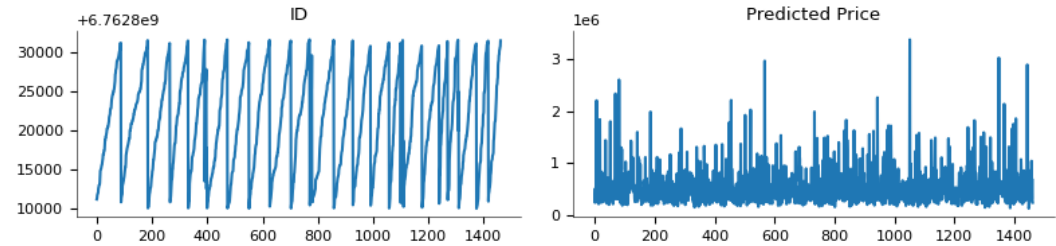


```
array([497766.12740438, 244495.3776842 , 293819.40063242, ...,  
       698495.60350629, 297006.00386358, 245881.76921871])  
  
Time series  
  
y_pred_list = y['id'][-len(y_pred):].tolist()
```

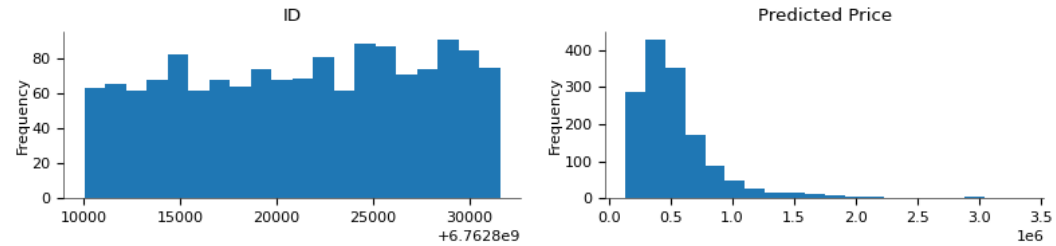
	ID	Predicted Price
0	6762811233	497766.13
1	6762811403	244495.38
2	6762811775	293819.40
3	6762811861	397555.35
4	6762812009	474843.29
...
1457	6762830250	1041014.57
1458	6762830339	317512.59
1459	6762830618	698495.60
1460	6762830709	297006.00
1461	6762831463	245881.77

1462 rows x 2 columns

Values



Distributions



2-d distributions

