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

Choose Files House Price India.csv

- **House Price India.csv**(text/csv) - 1524561 bytes, last modified: 9/30/2023 - 100% done

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

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

```
import io
df = pd.read_csv(io.BytesIO(uploaded['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

5 rows × 23 columns

```
df.tail()
```

	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

5 rows × 23 columns

```
df
```

```
                              number                          number             number  condition
```

df.columns

```
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', 'Lattitude',
       'Longitude', 'living_area_renov', 'lot_area_renov',
       'Number of schools nearby', 'Distance from the airport', 'Price'],
      dtype='object')
```

```
   4    6762812919  42491         3        2.00    2710    4500      1.5           0       0          4    ...   1929           0  122006
```

df.dtypes

```
id                                      int64
Date                                    int64
number of bedrooms                      int64
number of bathrooms                   float64
living area                             int64
lot area                                int64
number of floors                      float64
waterfront present                      int64
number of views                         int64
condition of the house                  int64
grade of the house                      int64
Area of the house(excluding basement)   int64
Area of the basement                    int64
Built Year                              int64
Renovation Year                         int64
Postal Code                             int64
Lattitude                             float64
Longitude                             float64
living_area_renov                       int64
lot_area_renov                          int64
Number of schools nearby                int64
Distance from the airport               int64
Price                                   int64
dtype: object
```

df.info

```
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

       Renovation Year  Postal Code  Lattitude  Longitude  living_area_renov  \
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
```

```
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

```
print (df.describe())
```

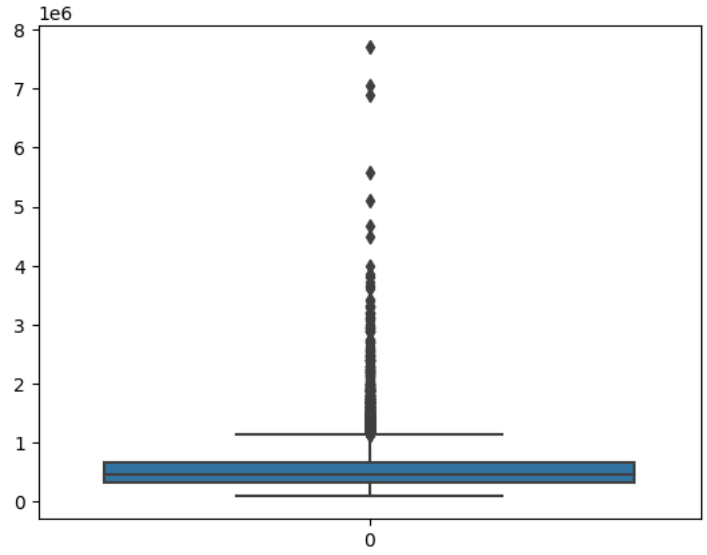
```
std    6.237575e+03     67.347991            0.938719             0.769934
min    6.762810e+09  42491.000000            1.000000             0.500000
25%    6.762815e+09  42546.000000            3.000000             1.750000
50%    6.762821e+09  42600.000000            3.000000             2.250000
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     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  \
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]
```

```
sns.violinplot(x=df['lot_area_renov'])
```

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

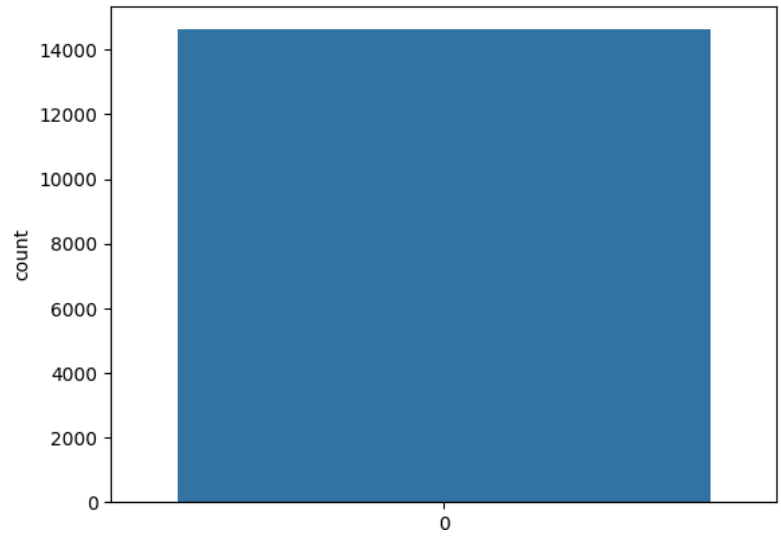
```
sns.boxplot(df['Price'])
```

```
<Axes: >
```




```
sns.countplot(df['grade of the house'])
```

```
<Axes: ylabel='count'>
```

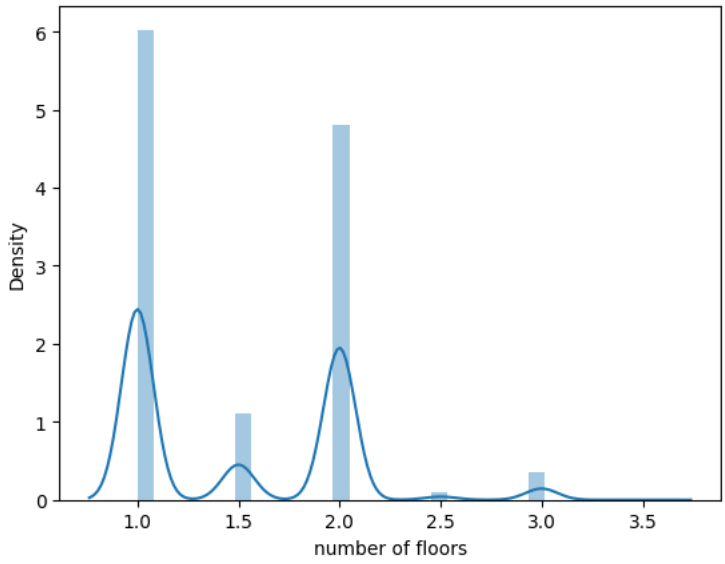



```
plt.hist(df['number of bedrooms'])
```

```
(array([1.3316e+04, 1.2850e+03, 1.7000e+01, 1.0000e+00, 0.0000e+00,
        0.0000e+00, 0.0000e+00, 0.0000e+00, 0.0000e+00, 1.0000e+00]),
 array([ 1. ,  4.2,  7.4, 10.6, 13.8, 17. , 20.2, 23.4, 26.6, 29.8, 33. ]),
 <BarContainer object of 10 artists>)
```


```
sns.distplot(df['number of floors'])
```

```
<ipython-input-19-4bce24ad8e73>: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['number of floors'])
<Axes: xlabel='number of floors', ylabel='Density'>
```



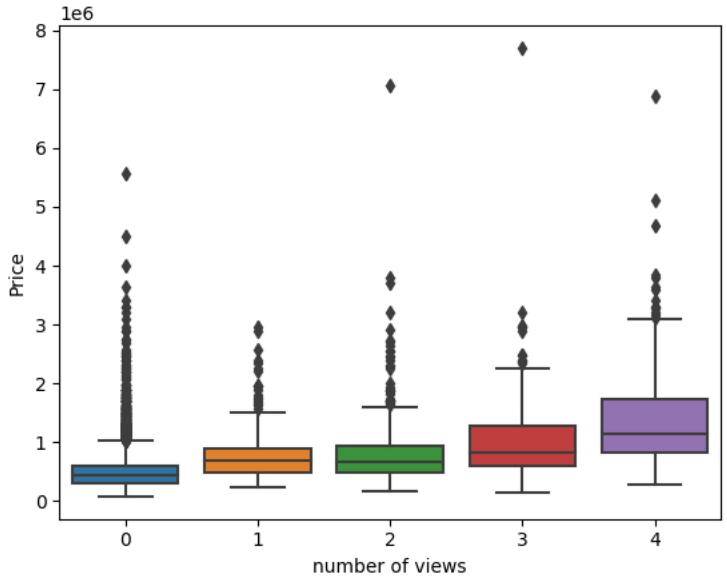
Bivariate Analysis

```
sns.scatterplot(x=df['number of bedrooms'],y=df['living area'])
```

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

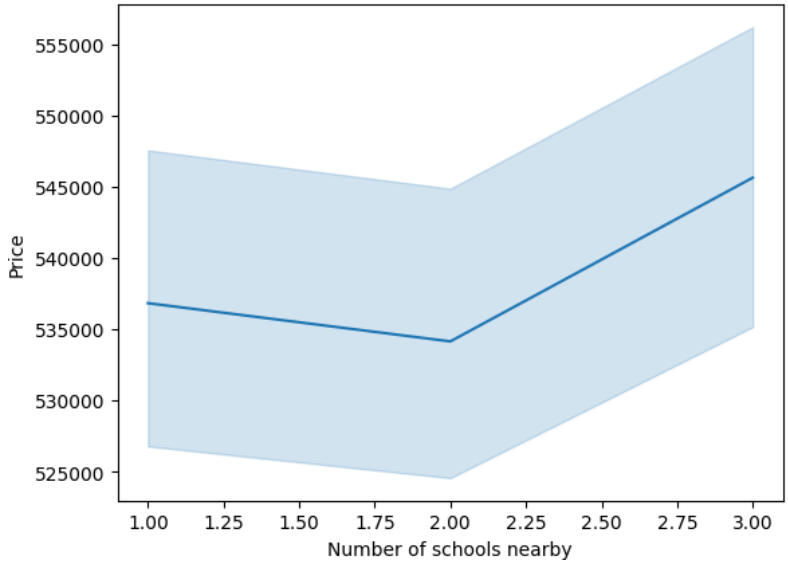

```
sns.boxplot(x=df['number of views'],y=df['Price'])
```

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



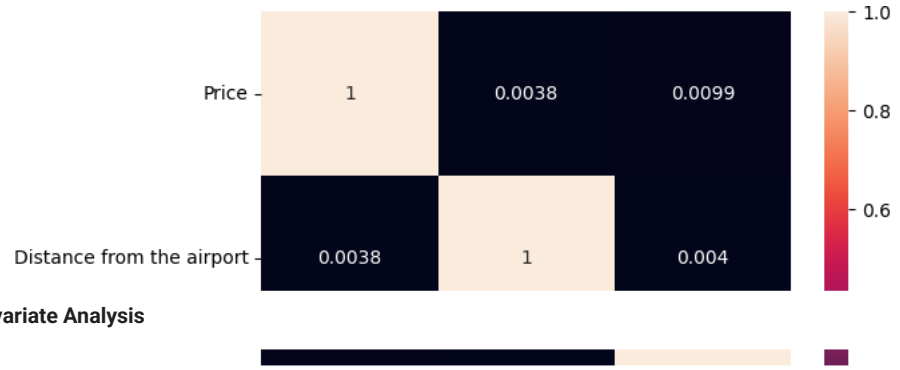

```
sns.lineplot(x=df['Number of schools nearby'],y=df['Price'])
```

```
<Axes: xlabel='Number of schools nearby', ylabel='Price'>
```




```
sns.heatmap(df[['Price','Distance from the airport','Number of schools nearby']].corr(), annot=True)
```

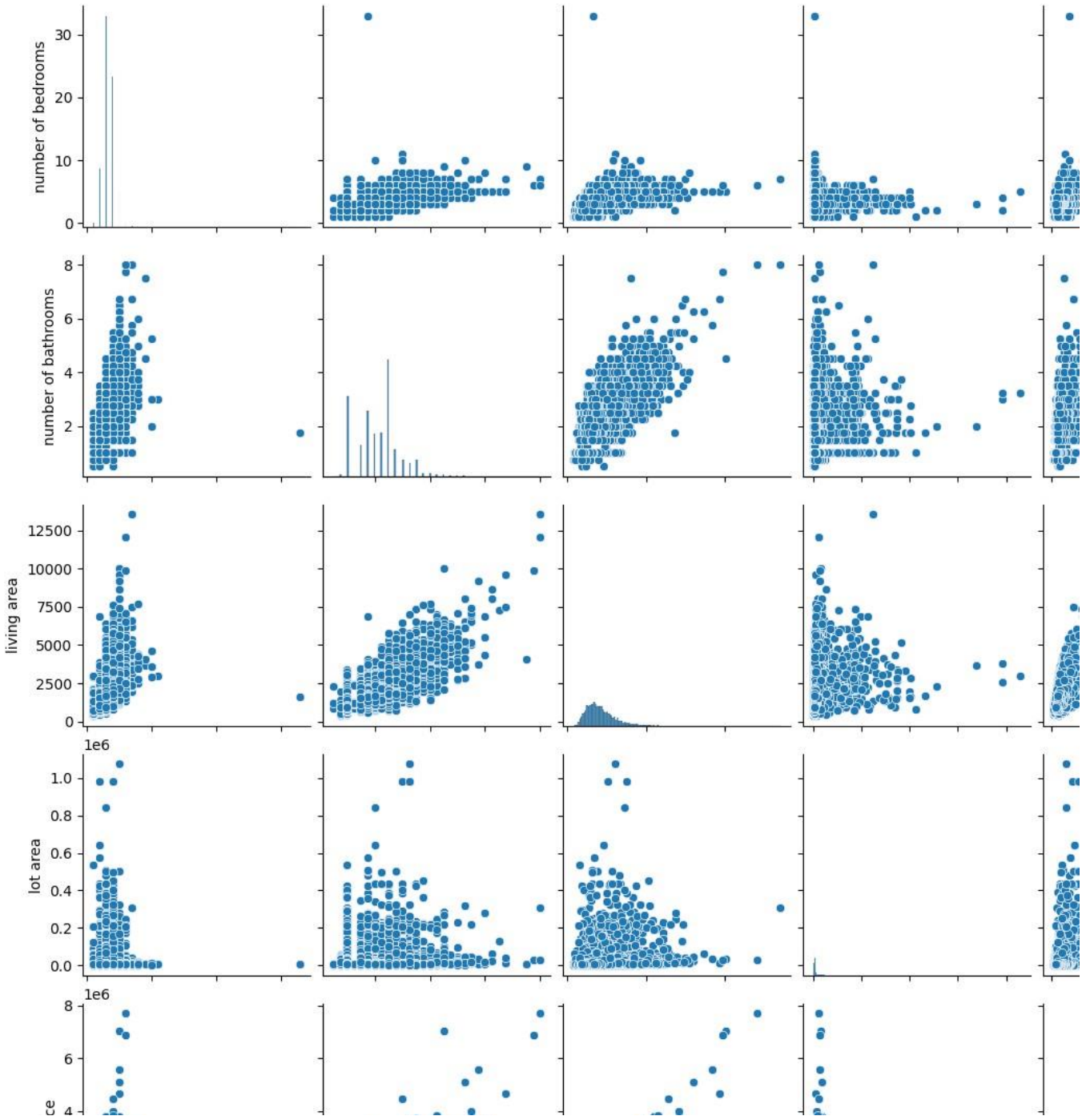
```
<Axes: >
```



Multivariate Analysis

```
sns.pairplot(df[['number of bedrooms','number of bathrooms','living area','lot area','Price']])
```

```
<seaborn.axisgrid.PairGrid at 0x7aff98f5eb60>
```




```
sns.heatmap(df.drop(['id','Lattitude','Longitude'], axis=1).corr(),linewidth=0.3,annot=True)
plt.show()
```

Perform descriptive statistics on the dataset

```
print(df.describe())
```

```
       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     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  \
       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
```

```
       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]
```

```
print(df.count())
```

```
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
Lattitude                             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
```

```
print(df.corr())
```

```
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]
```

```
print(df['Number of schools nearby'].value_counts())
```

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

```
print('Mean:',df['Area of the house(excluding basement)'].mean())
print('Median:',df['Distance from the airport'].median())
print('Mode:',df['Price'].mode())
```

```
Mean: 1801.783926128591
Median: 65.0
Mode: 0    450000
Name: Price, dtype: int64
```

Handle the Missing Values

```
print(df.isnull().sum())
```

```
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
```

```
df.dropna(inplace=True)
```

```
df.fillna(0, inplace=True)
```

```
df.interpolate(inplace=True)
```

```
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
```

```
x=df.drop(['Price','Date'],axis=1)
x.set_index(['id'],inplace=True)
y=df[['id','Price']]
```

```
x.head()
```

id	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	Renov
6762810145	5	2.50	3650	9050	2.0	0	4	5	10	3370	280	1921	
6762810635	4	2.50	2920	4000	1.5	0	0	5	8	1910	1010	1909	

```
y.head()
```

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

```
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.ensemble import GradientBoostingRegressor
from sklearn.metrics import r2_score
```

```
x_train,x_test,y_train,y_test = train_test_split(x , y['Price'],test_size =0.1,random_state=2)
model = GradientBoostingRegressor(n_estimators= 400,max_depth=5,min_samples_split=2,learning_rate=0.1)
model.fit(x_train,y_train)
```

```
GradientBoostingRegressor
GradientBoostingRegressor(max_depth=5, n_estimators=400)
```

```
y_pred = model.predict(x_test)
model.score(x_test,y_test)
```

```
0.912295301288466
```

```
r2_score(y_pred,y_test)
```

```
0.9017337864944384
```

```
y_pred
```

```
array([497766.12740438, 244495.3776842 , 293819.40063242, ...,
       698495.60350629, 297006.00386358, 245881.76921871])
```

```
y_pred_list = y['id'][-len(y_pred):].tolist()
y_pred_df=pd.DataFrame(y_pred_list,columns=['ID'])
y_pred_df['Predicted Price'] = y_pred.round(2)
y_pred_df
```

	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 × 2 columns