

```
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
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

```
df.tail()
```

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	num vi
14615	6762830250	42734	2	1.5	1556	20000	1.0	0	
14616	6762830339	42734	3	2.0	1680	7000	1.5	0	
14617	6762830618	42734	2	1.0	1070	6120	1.0	0	
14618	6762830709	42734	4	1.0	1030	6621	1.0	0	
14619	6762831463	42734	3	1.0	900	4770	1.0	0	

5 rows × 23 columns

```
df
```

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	num vi
0	6762810145	42491	5	2.50	3650	9050	2.0	0	
1	6762810635	42491	4	2.50	2920	4000	1.5	0	

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')
```

14617	6762830618	42734	2	1.00	1070	6120	1.0	0
-------	------------	-------	---	------	------	------	-----	---

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()

```
<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
```

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	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]
```

```
plt.hist(df['living_area_renov'])
```

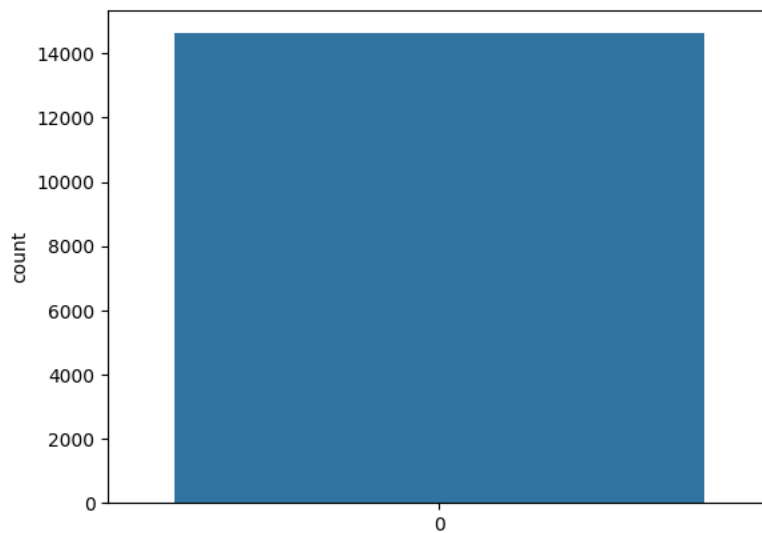


```
(array([ 310., 4449., 4920., 2748., 1406., 525., 187., 60., 9.,  
        6.]),  
 array([ 460., 1025., 1590., 2155., 2720., 3285., 3850., 4415., 4980.,  
        5545., 6110.]),  
 <BarContainer object of 10 artists>)
```

```
5000 |-----|
```

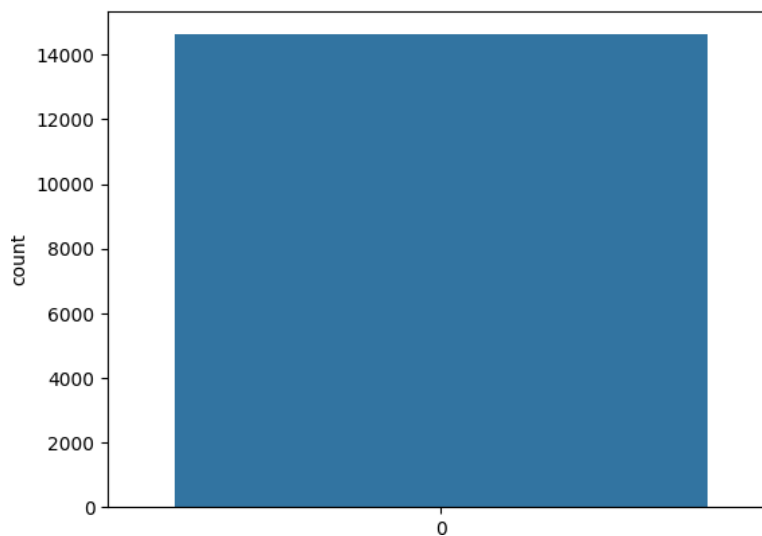
```
sns.countplot(df['number of floors'])
```

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



```
sns.countplot(df['number of bathrooms'])
```

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



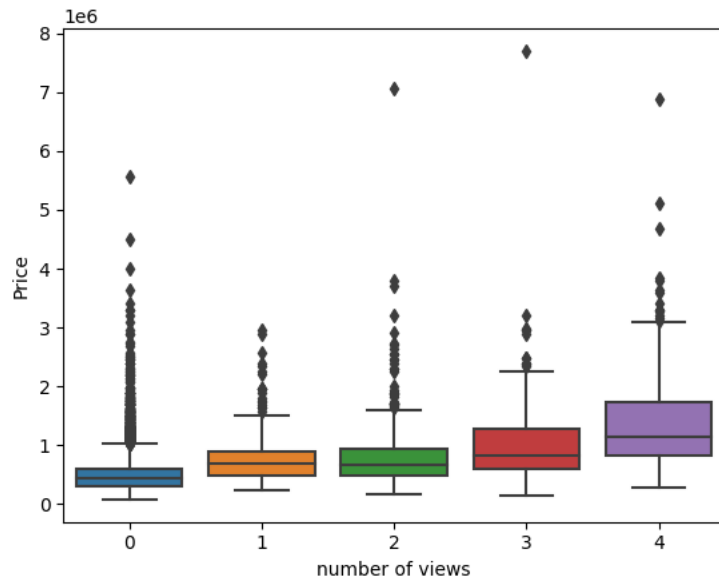
```
sns.boxplot(x=df['Price'])
```

<Axes: xlabel='Price'>

Bivariate Analysis

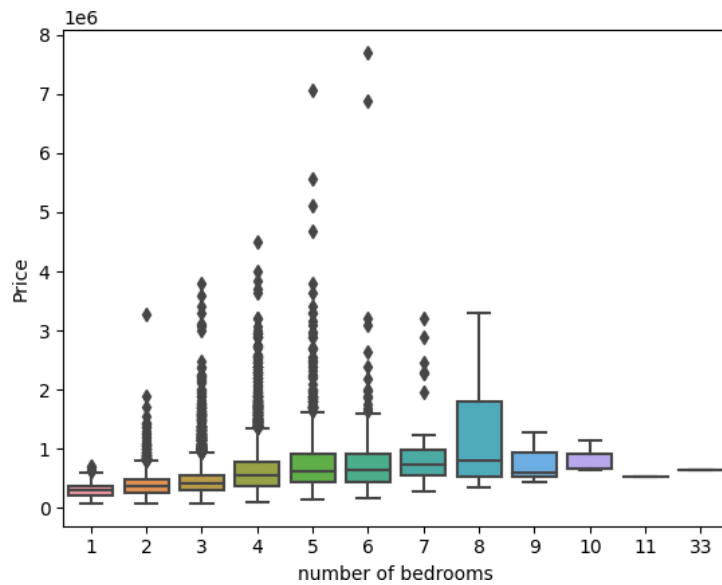
```
sns.boxplot(x=df['number of views'],y=df['Price'])
```

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



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

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



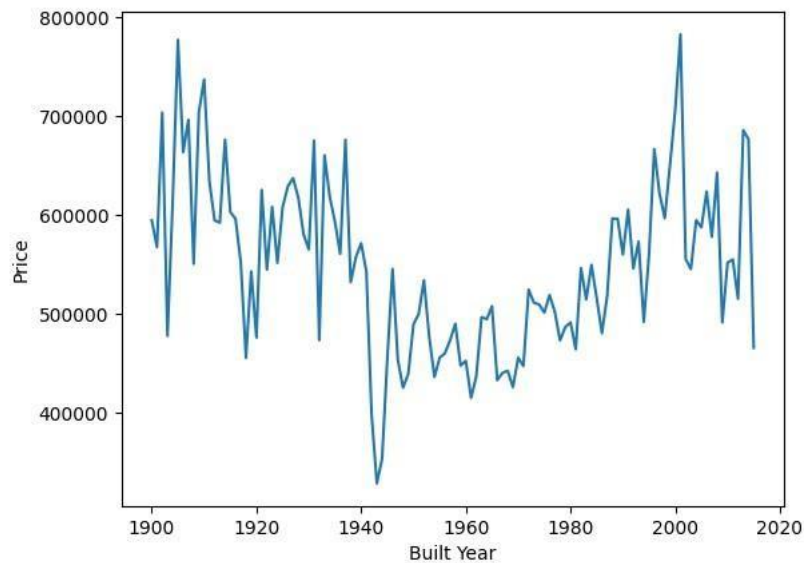
```
sns.lineplot(x=df['Built Year'],y=df['Price'])
```

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



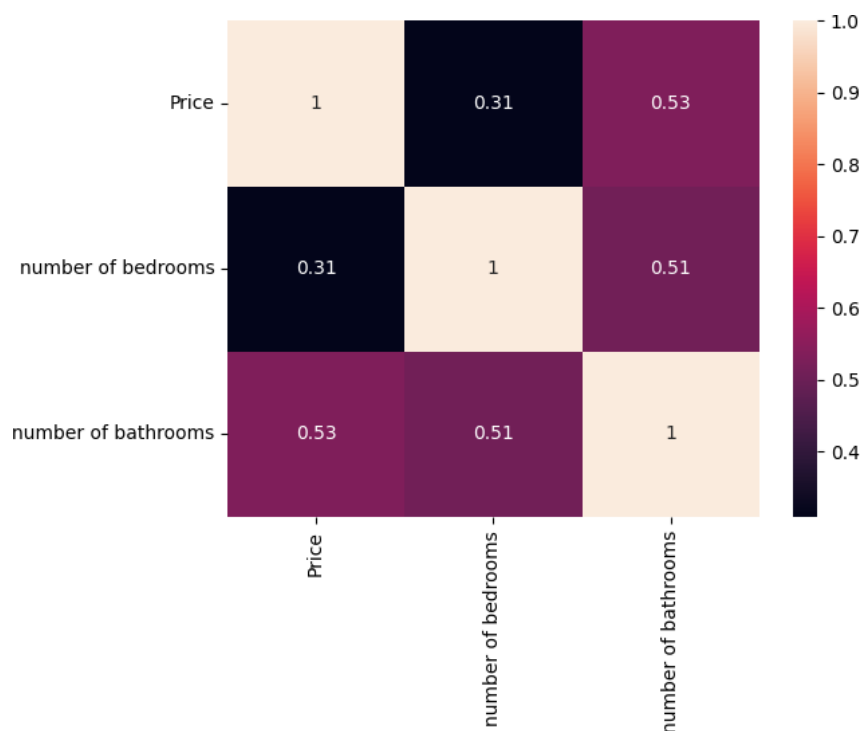
```
sns.lineplot(x=df.groupby('Built Year').mean().index,y=df.groupby('Built Year').mean()['Price'])
plt.show()
```

()



```
sns.heatmap(df[['Price','number of bedrooms','number of bathrooms']].corr(),annot=True)
```

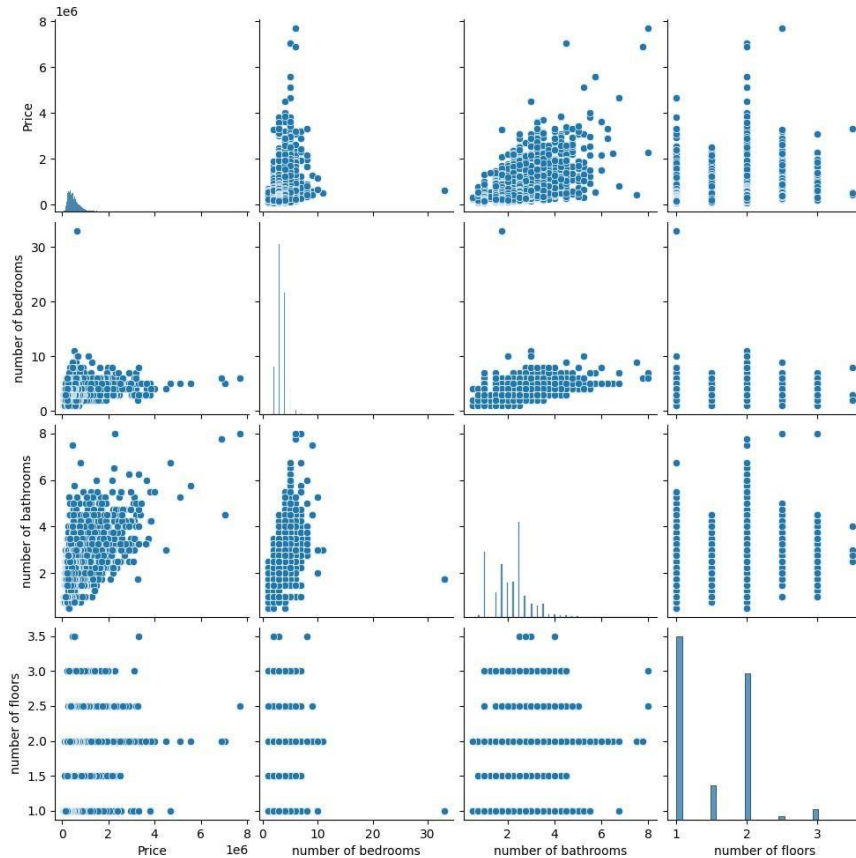
<Axes: >



Multivariate Analysis

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

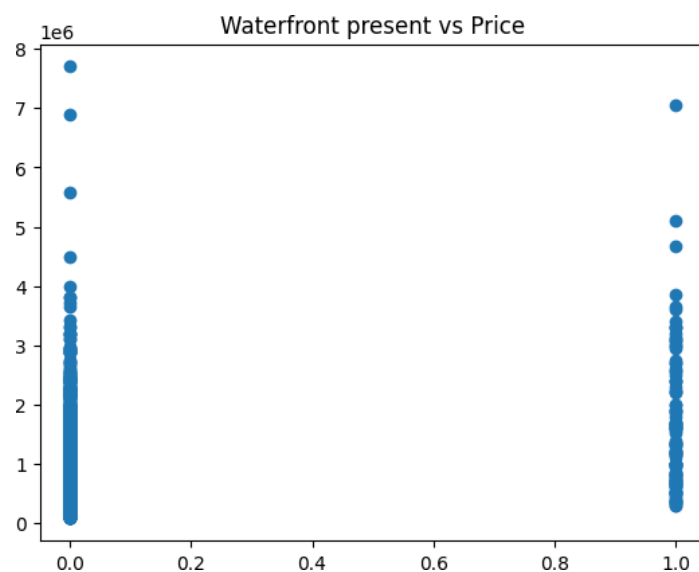
<seaborn.axisgrid.PairGrid at 0x78b131d80130>



```
df.duplicated().sum()
```

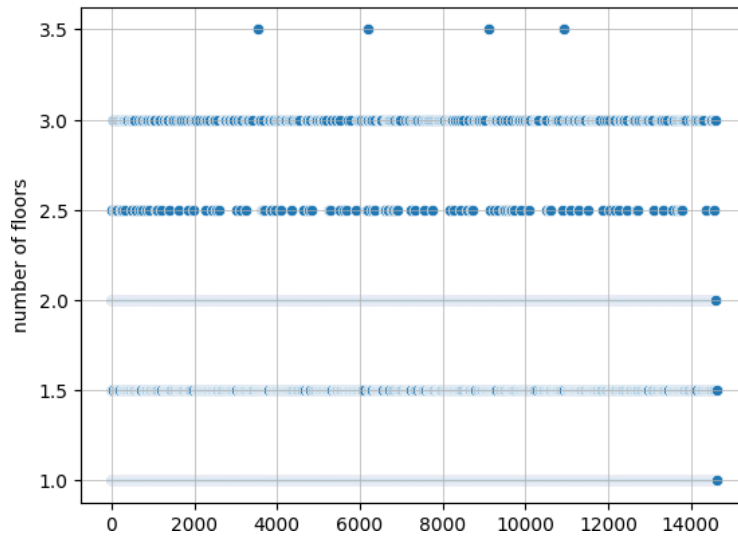
0

```
plt.scatter(df['waterfront present'],df['Price'])
plt.title("Waterfront present vs Price")
plt.grid(linestyle='-', linewidth=0.)
```

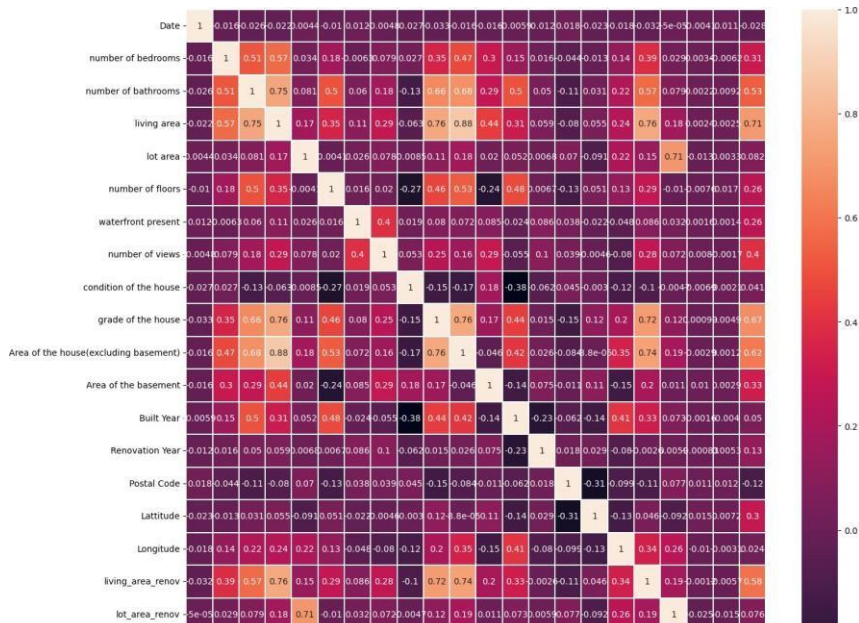


```
sns.scatterplot(df['number of floors'])
```

```
plt.grid(linestyle='-',linewidth=0.5)
```



```
plt.subplots(figsize=(15,15))  
sns.heatmap(df.drop(['id'],axis=1).corr(),linewidth=0.3,annot=True)  
plt.show()
```

```
print(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

```
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
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
```

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

```
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
Latitude          -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
Latitude          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
```

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

```
3    4973
2    4853
1    4794
```

```
Name: Number of schools nearby, dtype: int64
```

```
print('Mean:',df['Distance from the airport'].mean())
```

```
print('Median:',df['Area of the basement'].median())
```

```
print('Mode:',df['grade of the house'].mode())
```

```
Mean: 64.95095759233926
```

```
Median: 0.0
```

```
Mode: 0    7
```

```
Name: grade of the house, 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
Latitude          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()
```

	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
6762810635	4	2.50	2920	4000	1.5	0	0	5
6762810998	5	2.75	2910	9480	1.5	0	0	3
6762812605	4	2.50	3310	42998	2.0	0	0	3
6762812919	3	2.00	2710	4500	1.5	0	0	4

```
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)

0.91198551249072
```

```
r2_score(y_pred,y_test)

0.9015923458012287
```



```
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			

Double-click (or enter) to edit

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