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

Choose Files House Price India.csv

- **House Price India.csv**(text/csv) - 1524561 bytes, last modified: 10/2/2023 - 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	...	Buil Yea
	6762810145	42491	5	2.50	3650	9050	2.0	0	4	5	...	192
	6762810635	42491	4	2.50	2920	4000	1.5	0	0	5	...	190
	6762810998	42491	5	2.75	2910	9480	1.5	0	0	3	...	193
	6762812605	42491	4	2.50	3310	42998	2.0	0	0	3	...	200
	6762812919	42491	3	2.00	2710	4500	1.5	0	0	4	...	192

ɔws × 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	...
14615	6762830250	42734	2	1.5	1556	20000	1.0	0	0	4	...
14616	6762830339	42734	3	2.0	1680	7000	1.5	0	0	4	...
14617	6762830618	42734	2	1.0	1070	6120	1.0	0	0	3	...
14618	6762830709	42734	4	1.0	1030	6621	1.0	0	0	4	...
14619	6762831463	42734	3	1.0	900	4770	1.0	0	0	3	...

5 rows × 23 columns

```
df
```

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house	...
0	6762810145	42491	5	2.50	3650	9050	2.0	0	4	5	...
1	6762810635	42491	4	2.50	2920	4000	1.5	0	0	5	...
2	6762810998	42491	5	2.75	2910	9480	1.5	0	0	3	...
3	6762812605	42491	4	2.50	3310	42998	2.0	0	0	3	...
4	6762812919	42491	3	2.00	2710	4500	1.5	0	0	4	...
...	...	...	...	...	...	...	...	...	...	...	...
14615	6762830250	42734	2	1.50	1556	20000	1.0	0	0	4	...
14616	6762830339	42734	3	2.00	1680	7000	1.5	0	0	4	...
14617	6762830618	42734	2	1.00	1070	6120	1.0	0	0	3	...
14618	6762830709	42734	4	1.00	1030	6621	1.0	0	0	4	...
14619	6762831463	42734	3	1.00	900	4770	1.0	0	0	3	...

14620 rows × 23 columns

```
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')
```

```
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

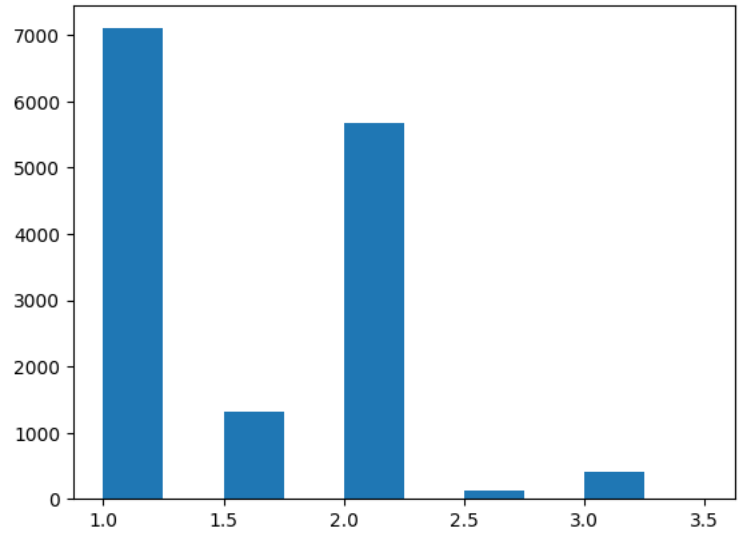
```
print(df.describe())
```

```
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
```

```
        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]
```

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

```
(array([7.103e+03, 0.000e+00, 1.311e+03, 0.000e+00, 5.666e+03, 0.000e+00,
        1.180e+02, 0.000e+00, 4.180e+02, 4.000e+00]),
 array([1.  , 1.25, 1.5 , 1.75, 2.  , 2.25, 2.5 , 2.75, 3.  , 3.25, 3.5 ]),
 <BarContainer object of 10 artists>)
```

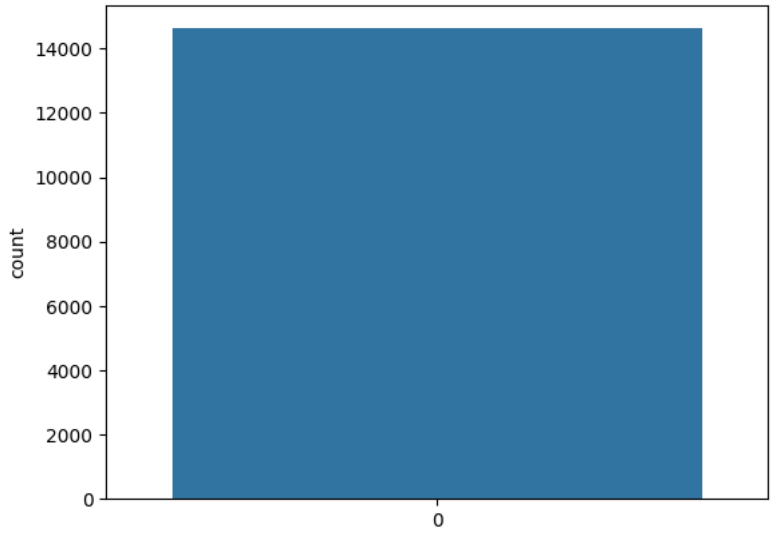



```
sns.countplot(df['number of bedrooms'])
```

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

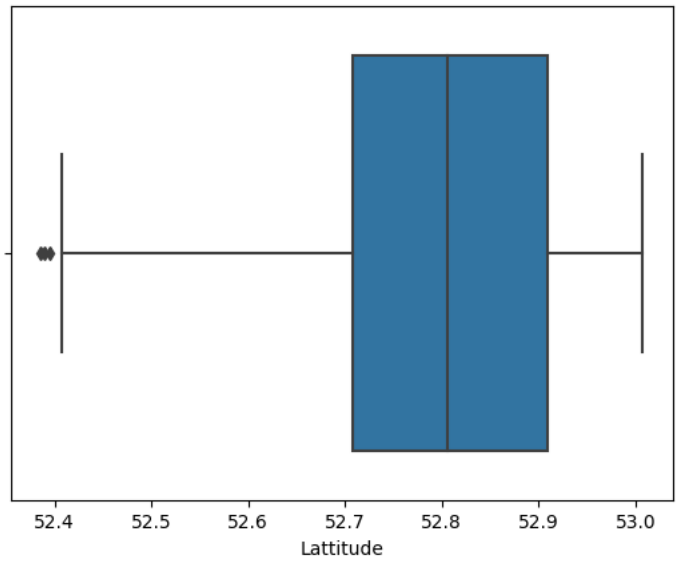

```
sns.countplot(df['Area of the basement'])
```

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




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

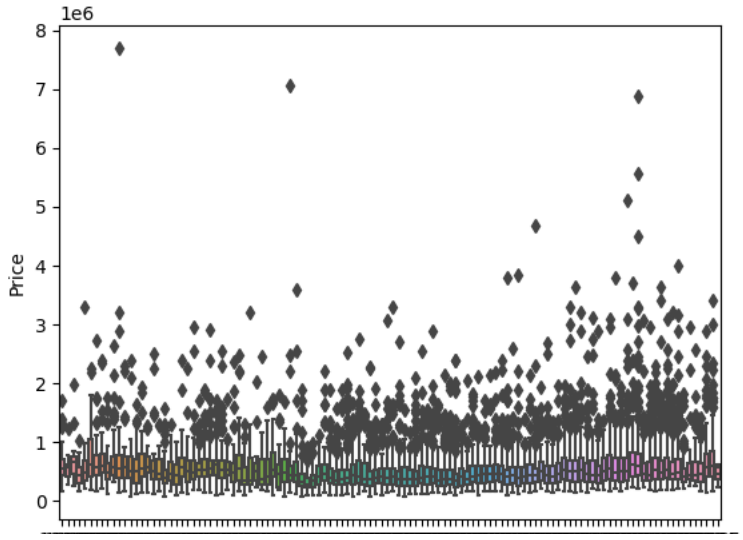
```
<Axes: xlabel='Lattitude'>
```



Bivariate Analysis

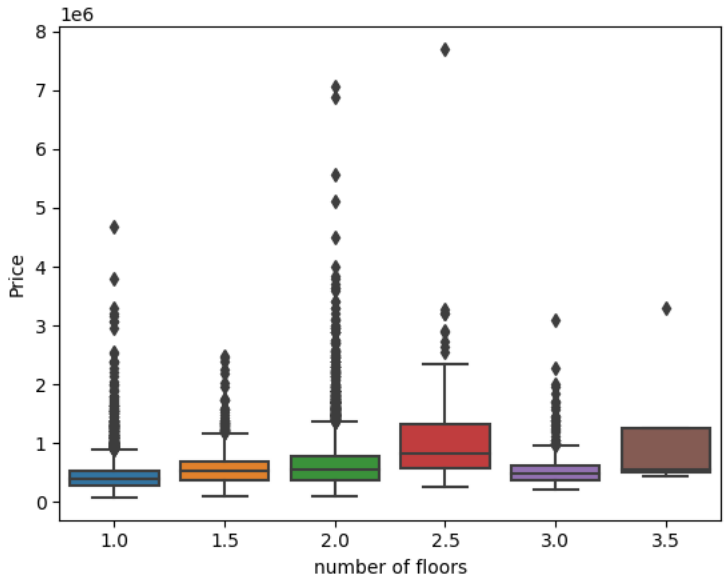
```
sns.boxplot(x=df['Built Year'],y=df['Price'])
```

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



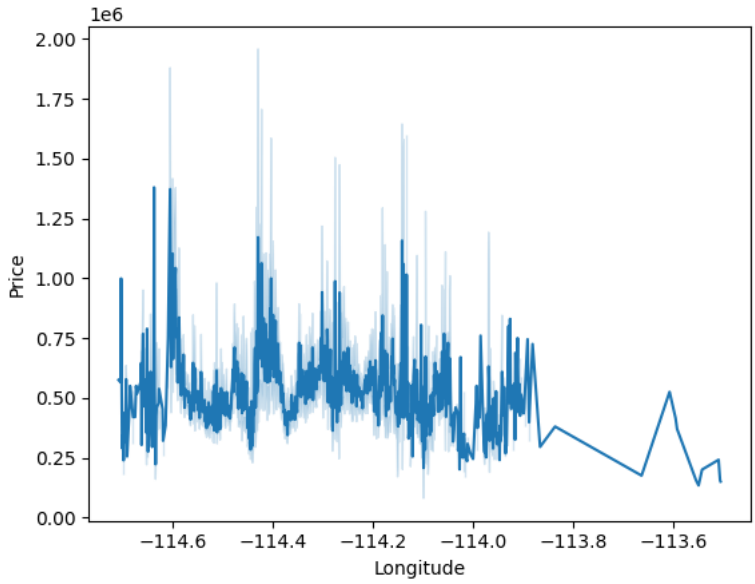

```
sns.boxplot(x=df['number of floors'],y=df['Price'])
```

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

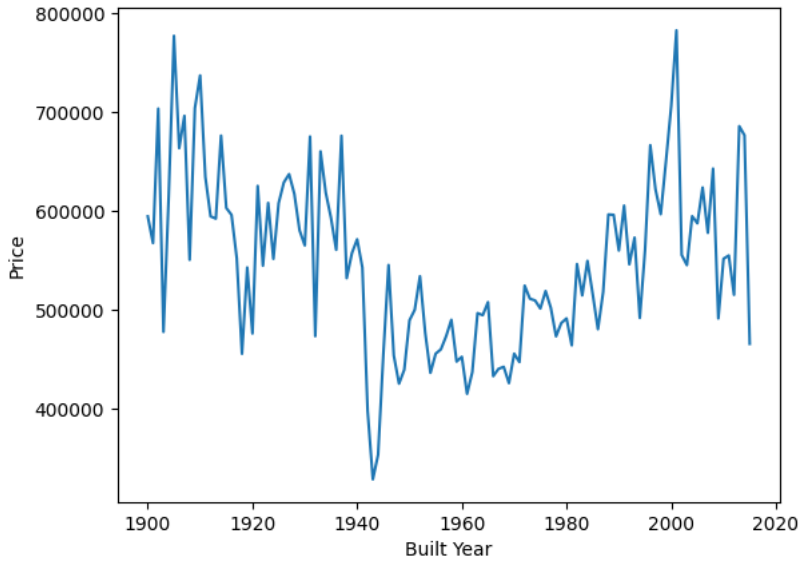



```
sns.lineplot(x=df['Longitude'],y=df['Price'])
```

```
<Axes: xlabel='Longitude', ylabel='Price'>
```




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



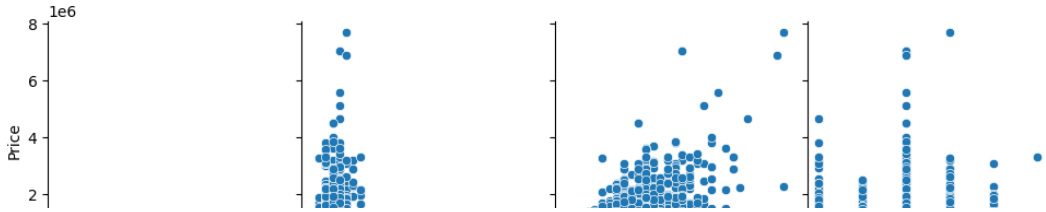

```
sns.heatmap(df[['Price','living area','lot area']].corr(),annot=True)
```

```
<Axes: >
```

Multivariate Analysis

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

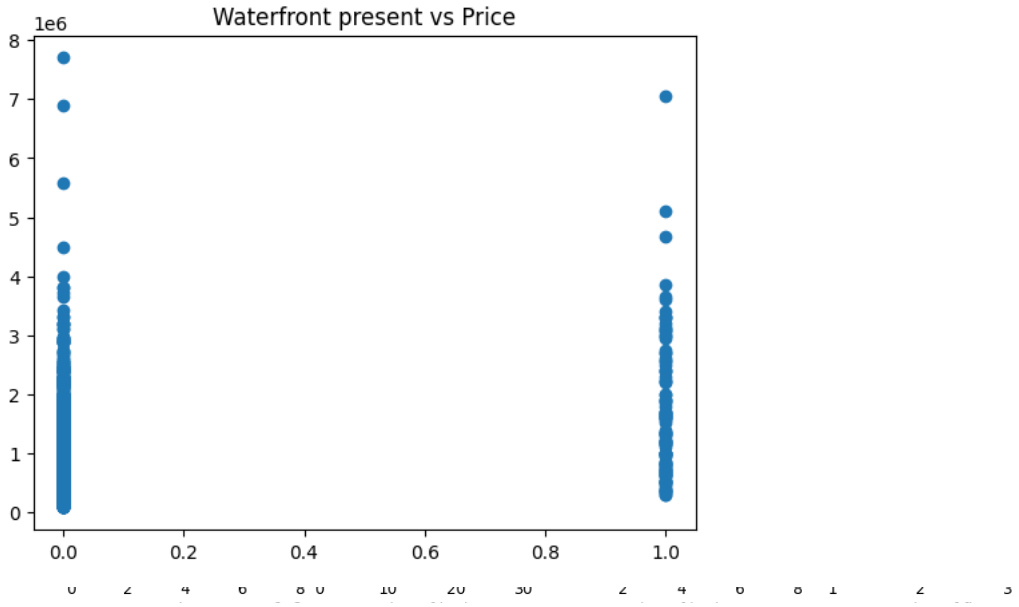
```
<seaborn.axisgrid.PairGrid at 0x781a0c7c2590>
```



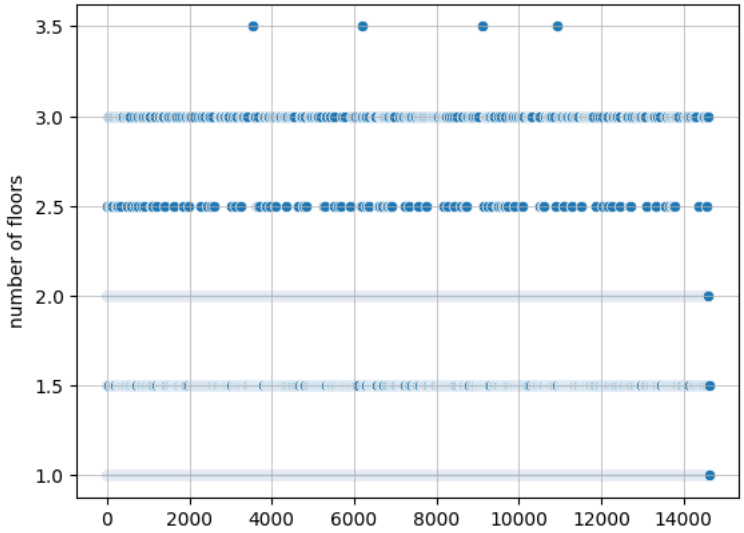

```
df.duplicated().sum()
```

```
0
```

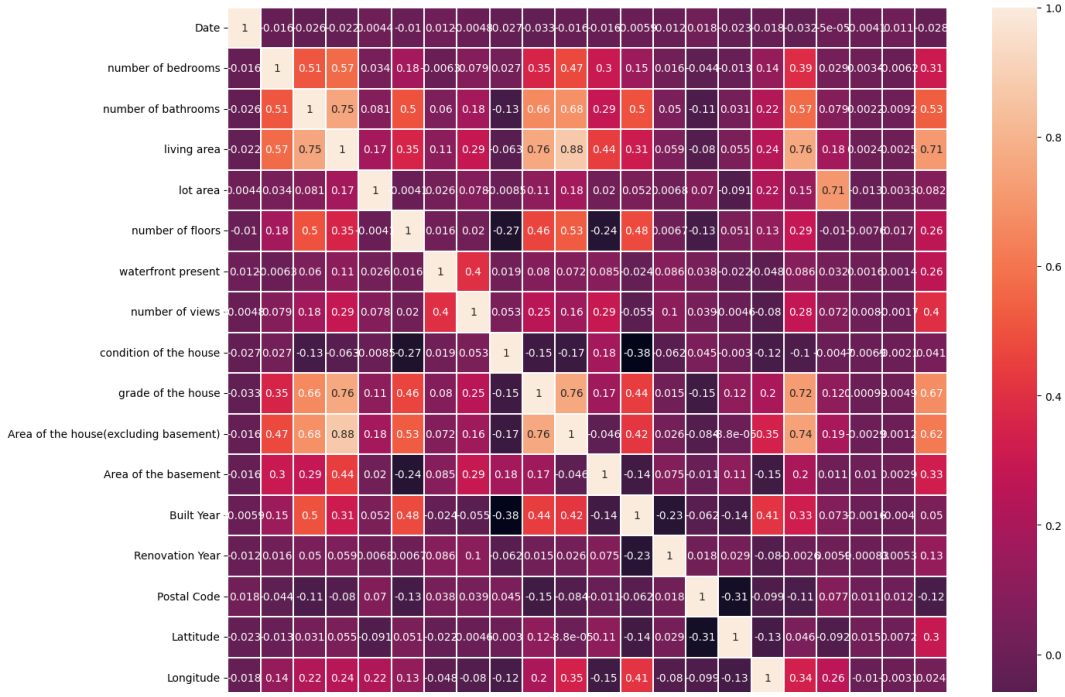
```
plt.scatter(df['waterfront present'],df['Price'])
plt.title("Waterfront present vs Price")
plt.grid(linestyle='-', linewidth=0.)
```




```
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())
```

```
              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
```

```
[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())
```

```
print(df['number of floors'].value_counts())
```

```
1.0    7103
2.0    5666
1.5    1311
3.0     418
2.5     118
3.5       4
Name: number of floors, 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
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 house(exclud baseme
6762810145	5	2.50	3650	9050	2.0	0	4	5	10	3
6762810635	4	2.50	2920	4000	1.5	0	0	5	8	1
6762810998	5	2.75	2910	9480	1.5	0	0	3	8	2
6762812605	4	2.50	3310	42998	2.0	0	0	3	9	3
6762812919	3	2.00	2710	4500	1.5	0	0	4	8	1

```
y.head()
```

	id	Price
0	6762810145	2380000
1	6762810635	1400000
2	6762810998	1200000

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

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

```
0.9130610931235753
```

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

```
0.9025174350550257
```

```
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