In []: #Q1. Perform basic EDA

```
In [3]: # Summary statistics
print(data.describe())

# Check for missing values
print(data.isnull().sum())

# Visualize the distribution of price per sqft
sns.histplot(data['price_per_sqft'], bins=30, kde=True)
plt.title('Price per Square Foot Distribution')
plt.xlabel('Price per Square Foot')
plt.ylabel('Frequency')
plt.show()
```

```
total_sqft
                             bath
                                           price
                                                           bhk price_per_sqft
       13200.000000 13200.000000 13200.000000 13200.000000
                                                                  1.320000e+04
count
        1555.302783
                         2.691136
                                     112.276178
                                                      2.800833
                                                                  7.920337e+03
mean
std
        1237.323445
                         1.338915
                                      149.175995
                                                      1.292843
                                                                  1.067272e+05
           1.000000
                         1.000000
                                        8.000000
                                                      1.000000
                                                                  2.670000e+02
min
                         2.000000
                                       50.000000
                                                      2.000000
25%
        1100.000000
                                                                  4.267000e+03
50%
        1275.000000
                         2,000000
                                      71.850000
                                                      3.000000
                                                                  5.438000e+03
        1672.000000
75%
                         3.000000
                                     120,000000
                                                      3,000000
                                                                  7.317000e+03
max
       52272.000000
                        40.000000
                                     3600.000000
                                                     43.000000
                                                                  1.200000e+07
location
                  0
size
{\tt total\_sqft}
                  0
bath
                  0
price
                  0
bhk
                  0
                  0
price_per_sqft
dtype: int64
```


Price per Square Foot

In []: #b) Percentile Method

```
In []: #Q2. Detect the outliers using following methods and remove it using methods like trimming / capping/ imputation using mean or me
In []: #a) Mean and Standard deviation

In [4]: mean = data['price_per_sqft'].mean()
std_dev = data['price_per_sqft'].std()

# Define outlier threshold
lower_limit = mean - 3 * std_dev
upper_limit = mean + 3 * std_dev
# Remove outliers
data_mean_std = data[(data['price_per_sqft'] >= lower_limit) & (data['price_per_sqft'] <= upper_limit)]</pre>
```

1e7

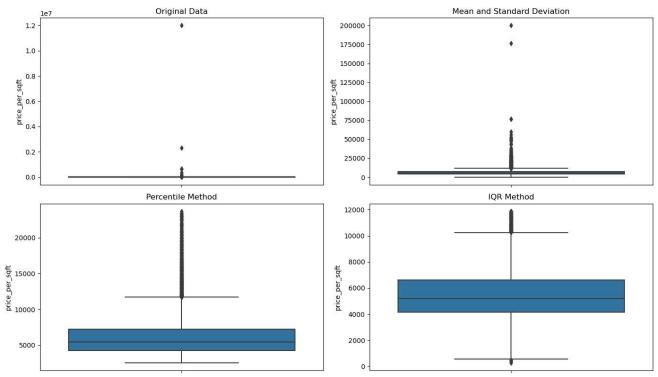
```
In [5]: lower_percentile = data['price_per_sqft'].quantile(0.01)
         upper_percentile = data['price_per_sqft'].quantile(0.99)
         # Remove outliers
         data_percentile = data[(data['price_per_sqft'] >= lower_percentile) & (data['price_per_sqft'] <= upper_percentile)]</pre>
In [ ]: #c) IQR Method
In [6]: Q1 = data['price_per_sqft'].quantile(0.25)
Q3 = data['price_per_sqft'].quantile(0.75)
         IQR = Q3 - Q1
         # Define limits
         lower_iqr = Q1 - 1.5 * IQR
         upper_iqr = Q3 + 1.5 * IQR
         # Remove outliers
         data_iqr = data[(data['price_per_sqft'] >= lower_iqr) & (data['price_per_sqft'] <= upper_iqr)]</pre>
In [ ]: #d) Z Score Method
In [7]: from scipy import stats
         z_scores = np.abs(stats.zscore(data['price_per_sqft']))
         data_z = data[(z_scores < 3)]</pre>
In []: #Q3. Create a box plot and use this to determine which method seems to work best to remove outliers for this data?
```

```
In [8]: plt.figure(figsize=(14, 8))
    plt.subplot(2, 2, 1)
    sns.boxplot(y=data['price_per_sqft'])
    plt.subplot(2, 2, 2)
    sns.boxplot(y=data_mean_std['price_per_sqft'])
    plt.title('Mean and Standard Deviation')

plt.subplot(2, 2, 3)
    sns.boxplot(y=data_percentile['price_per_sqft'])
    plt.title('Percentile Method')

plt.subplot(2, 2, 4)
    sns.boxplot(y=data_iqr['price_per_sqft'])
    plt.title('IQR Method')

plt.tight_layout()
    plt.show()
```



In []: . Draw histplot to check the normality of the column(price per sqft column) and perform transformations if needed. Check the skew

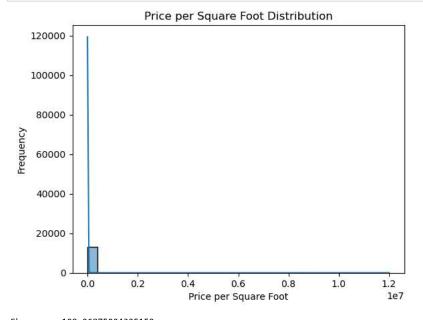
```
In [9]: # Histplot for normality check
sns.histplot(data['price_per_sqft'], bins=30, kde=True)
plt.title('Price per Square Foot Distribution')
plt.xlabel('Price per Square Foot')
plt.ylabel('Frequency')
plt.show()

# Skewness and Kurtosis
from scipy.stats import skew, kurtosis

print("Skewness:", skew(data['price_per_sqft']))
print("Kurtosis:", kurtosis(data['price_per_sqft']))

# Log transformation if needed
data['log_price_per_sqft'] = np.log(data['price_per_sqft'])

# Check skewness and kurtosis after transformation
print("Skewness after log transformation:", skew(data['log_price_per_sqft']))
print("Kurtosis after log transformation:", kurtosis(data['log_price_per_sqft']))
```



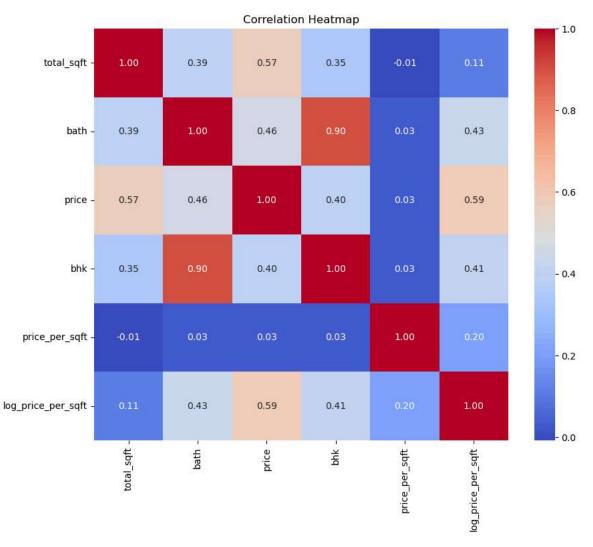
Skewness: 108.26875024325159
Kurtosis: 12090.633538860382
Skewness after log transformation: 1.3997035748119977
Kurtosis after log transformation: 9.199636085376468

In []: #Q5. Check the correlation between all the numerical columns and plot heatmap.

In [10]: # Correlation matrix correlation_matrix = data.corr() # PLot heatmap plt.figure(figsize=(10, 8)) sns.heatmap(correlation_matrix, annot=True, fmt='.2f', cmap='coolwarm') plt.title('Correlation Heatmap') plt.show()

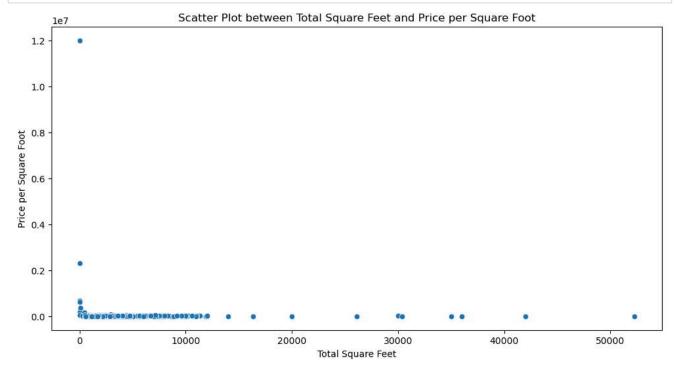
C:\Users\SHONIMA S\AppData\Local\Temp\ipykernel_18008\996203314.py:2: FutureWarning: The default value of numeric_only in DataF rame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numer ic_only to silence this warning.

correlation_matrix = data.corr()



In []: #Q6. Draw Scatter plot between the variables to check the correlation between them.

```
In [13]: plt.figure(figsize=(12, 6))
    sns.scatterplot(data=data, x='total_sqft', y='price_per_sqft') # Replace 'total_sqft' with any relevant variable
    plt.title('Scatter Plot between Total Square Feet and Price per Square Foot')
    plt.xlabel('Total Square Feet')
    plt.ylabel('Price per Square Foot')
    plt.show()
```



In []: