

untitled1

September 19, 2023

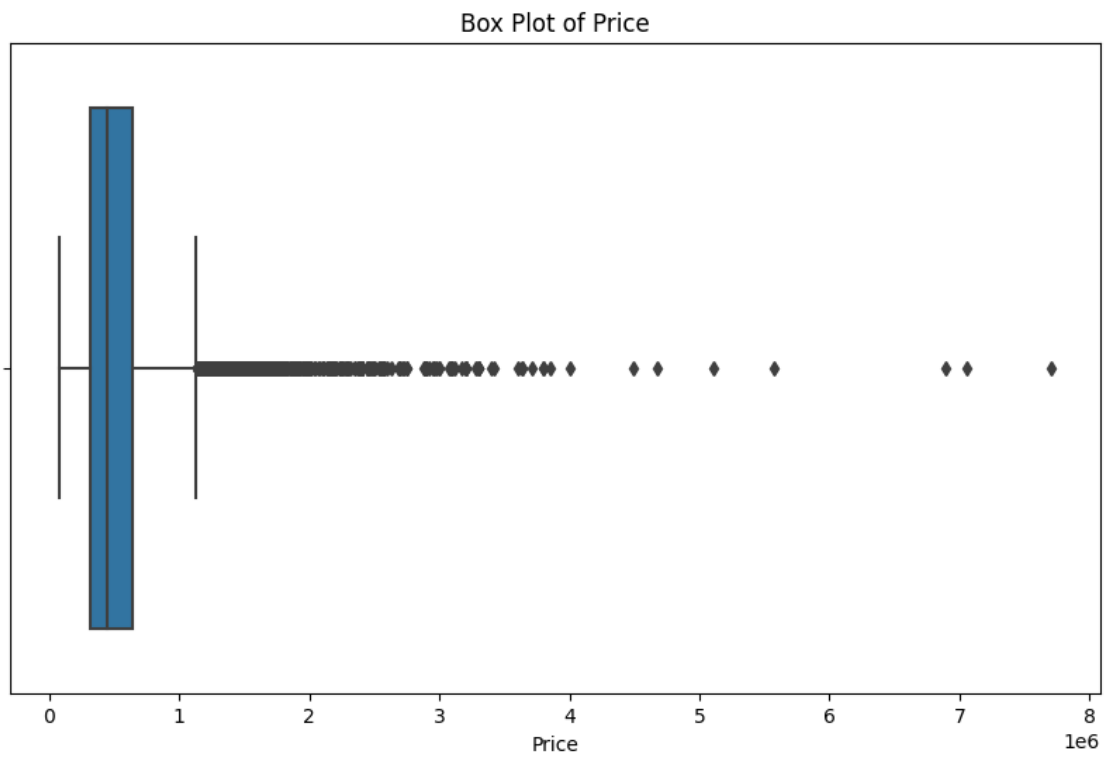
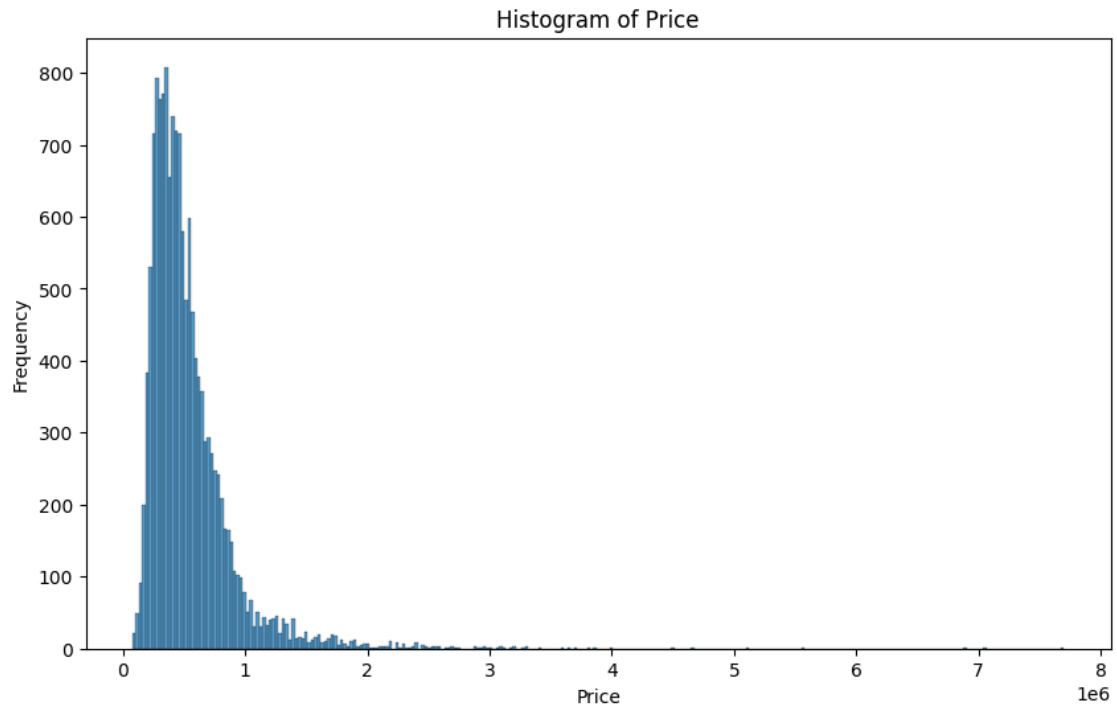
```
[2]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px

[5]: file_path = "/content/House Price India.csv"
df = pd.read_csv(file_path)

[6]: # Univariate Analysis
# Histogram
plt.figure(figsize=(10, 6))
sns.histplot(df['Price'])
plt.title('Histogram of Price')
plt.xlabel('Price')
plt.ylabel('Frequency')
plt.show()

# Box Plot (Box-and-Whisker Plot)
plt.figure(figsize=(10, 6))
sns.boxplot(x=df['Price'])
plt.title('Box Plot of Price')
plt.xlabel('Price')
plt.show()

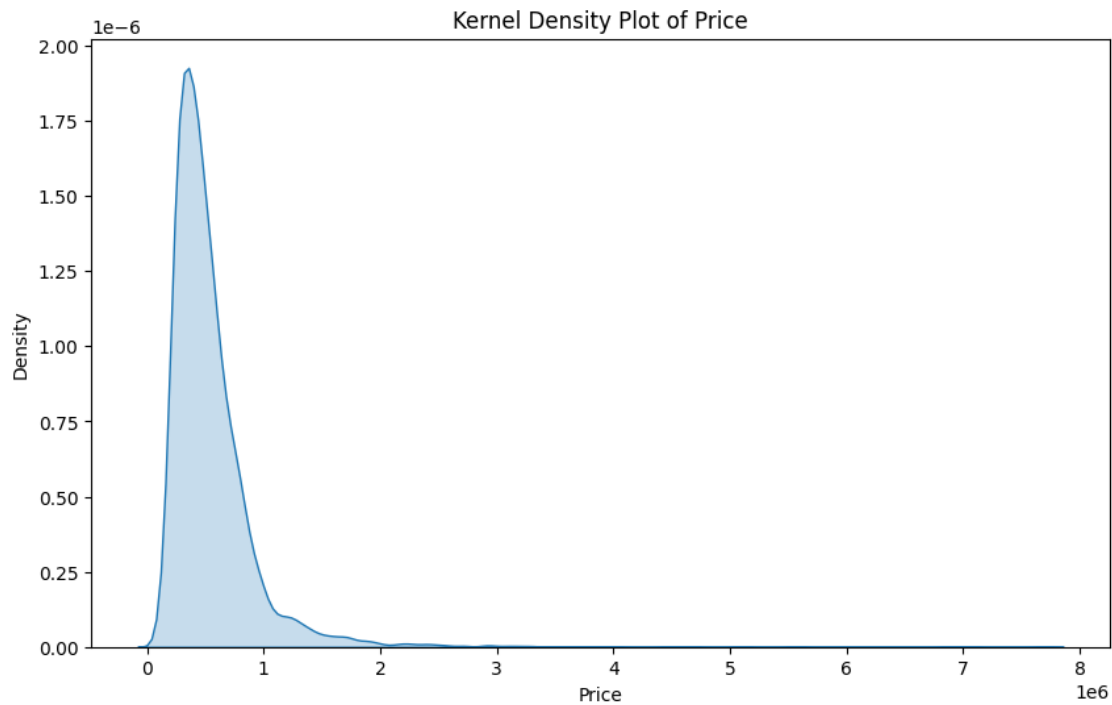
# Kernel Density Plot
plt.figure(figsize=(10, 6))
sns.kdeplot(df['Price'], shade=True)
plt.title('Kernel Density Plot of Price')
plt.xlabel('Price')
plt.show()
```



<ipython-input-6-4c6ca09cffb4>:19: FutureWarning:

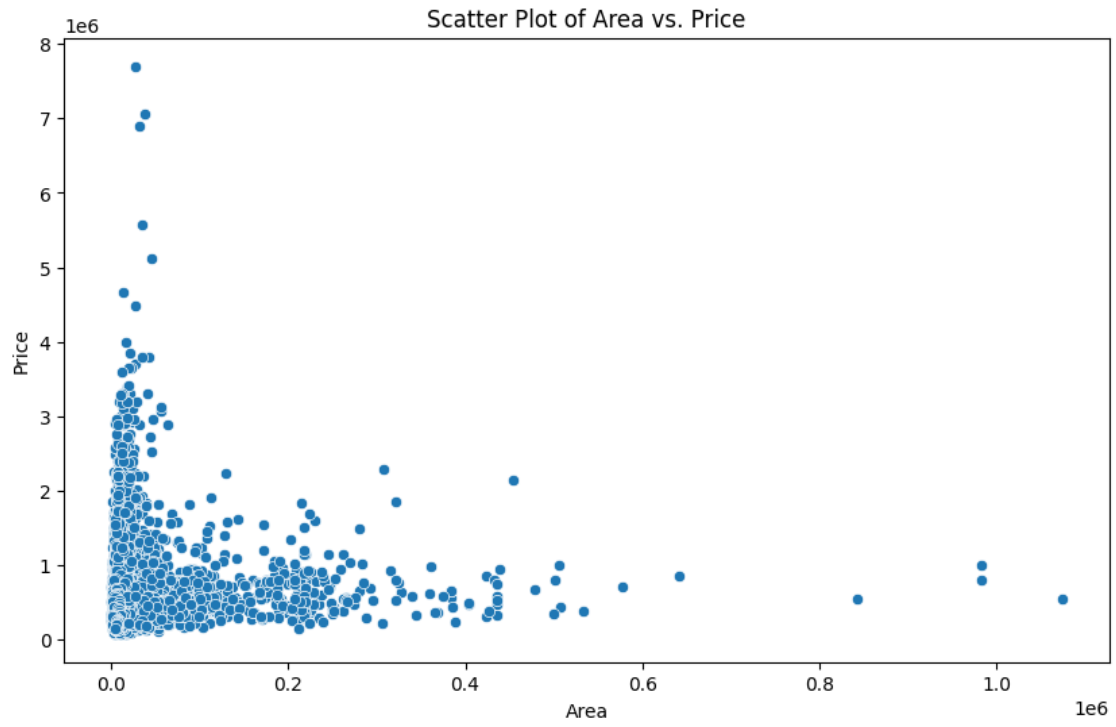
`shade` is now deprecated in favor of `fill`; setting `fill=True`.
This will become an error in seaborn v0.14.0; please update your code.

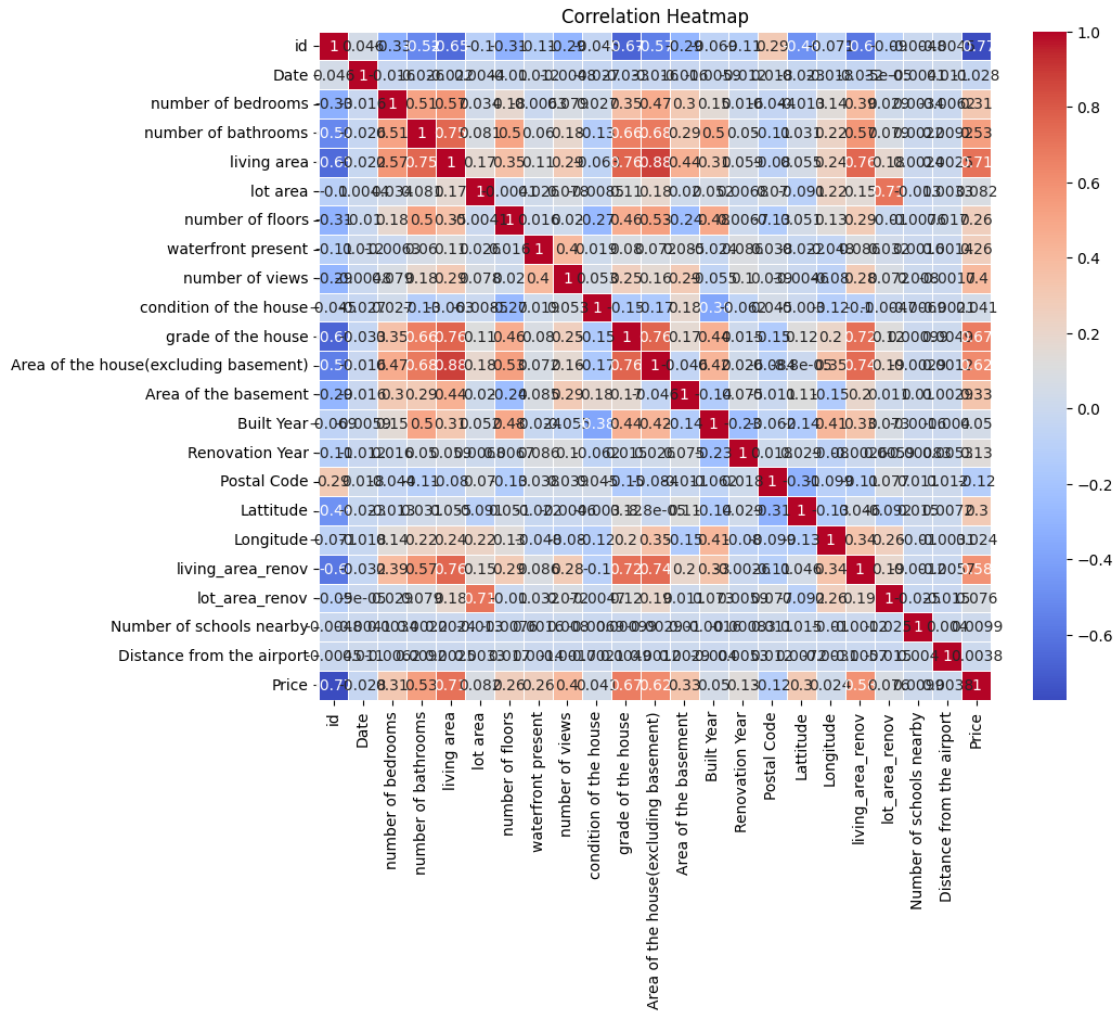
```
sns.kdeplot(df['Price'], shade=True)
```



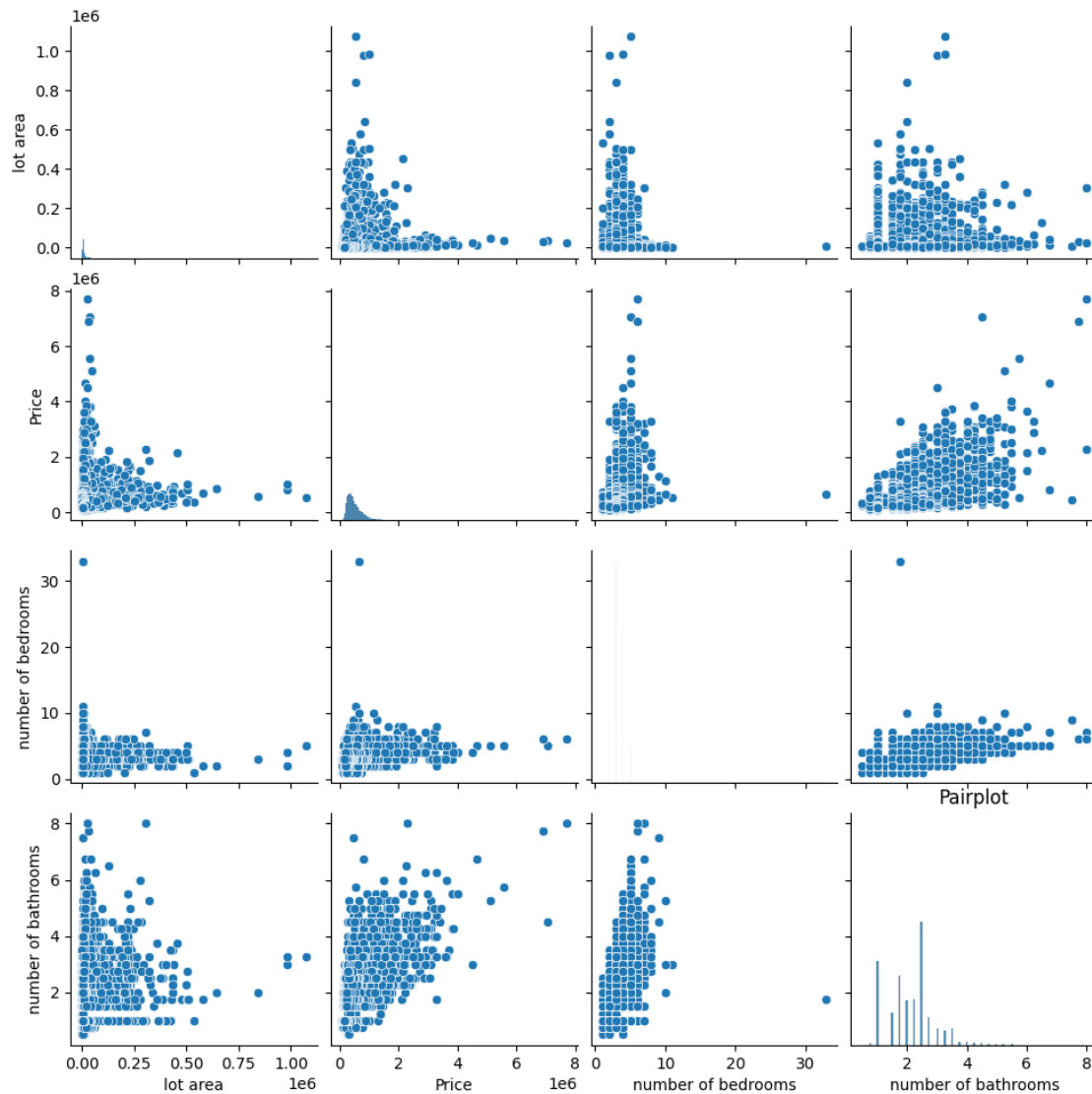
```
[18]: # Bivariate Analysis
# Scatter Plot
plt.figure(figsize=(10, 6))
sns.scatterplot(x=df['lot area'], y=df['Price'])
plt.title('Scatter Plot of Area vs. Price')
plt.xlabel('Area')
plt.ylabel('Price')
plt.show()

# Heatmap
correlation_matrix = df.corr()
plt.figure(figsize=(10, 8))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', linewidths=.5)
plt.title('Correlation Heatmap')
plt.show()
```





```
[16]: # Pairplot
sns.pairplot(df[['lot area', 'Price', 'number of bedrooms', 'number of_bathrooms']])
plt.title('Pairplot')
plt.show()
```



```
[21]: # Multivariate Analysis
# Parallel Coordinates Plot
fig = px.parallel_coordinates(df, dimensions=['lot area', 'Price', 'number of_
↵bedrooms', 'number of bathrooms'], color="Price")
fig.show()

# 3D Scatter Plot
fig = px.scatter_3d(df, x='lot area', y='number of bedrooms', z='Price',_
↵color='Price')
fig.show()

# Radial Plot (Polar plot)
fig = px.line_polar(df, r='Price', theta='lot area', line_close=True)
```

```
fig.update_traces(fill='toself')
fig.show()
```

```
[23]: #Descriptive Statistics
descriptive_stats = df.describe()
print(descriptive_stats)
```

	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	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]

```
[27]: df.isnull().any()
```

```
[27]: id                False
Date                  False
number of bedrooms    False
number of bathrooms   False
living area           False
lot area              False
number of floors      False
waterfront present    False
number of views       False
condition of the house False
grade of the house    False
Area of the house(excluding basement) False
Area of the basement  False
Built Year            False
Renovation Year       False
Postal Code           False
Latitude              False
Longitude             False
living_area_renov     False
lot_area_renov        False
Number of schools nearby False
Distance from the airport False
Price                 False
```


dtype: bool

```
[8]: print(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', 'Latitude',  
      'Longitude', 'living_area_renov', 'lot_area_renov',  
      'Number of schools nearby', 'Distance from the airport', 'Price'],  
      dtype='object')
```