

```
In [49]: #Problem Statement: Data Wrangling on Real Estate Market
# Tasks to Perform:
# 1. Import the "RealEstate_Prices.csv" dataset. Clean column names by removing sp
# special characters, or renaming them for clarity.
# 2. Handle missing values in the dataset, deciding on an appropriate strategy (e.g
# imputation or removal).
# 3. Perform data merging if additional datasets with relevant information are avai
# (e.g., neighborhood demographics or nearby amenities).
# 4. Filter and subset the data based on specific criteria, such as a particular ti
# property type, or location.
# 5. Handle categorical variables by encoding them appropriately (e.g., one-hot enc
# label encoding) for further analysis.
# 6. Aggregate the data to calculate summary statistics or derived metrics such as
# sale prices by neighborhood or property type.
# 7. Identify and handle outliers or extreme values in the data that may affect the
# or modeling process.

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

```
In [50]: real_estate = pd.read_csv('Bengaluru_House_Data.csv')
real_estate.head()
```

Out[50]:

	area_type	availability	location	size	society	total_sqft	bath	balcony
0	Super built-up Area	19-Dec	Electronic City Phase II	2 BHK	Coomee	1056	2.0	1.0
1	Plot Area	Ready To Move	Chikka Tirupathi	4 Bedroom	Theanmp	2600	5.0	3.0
2	Built-up Area	Ready To Move	Uttarahalli	3 BHK	NaN	1440	2.0	3.0
3	Super built-up Area	Ready To Move	Lingadheeranahalli	3 BHK	Soiewre	1521	3.0	1.0
4	Super built-up Area	Ready To Move	Kothanur	2 BHK	NaN	1200	2.0	1.0

```
In [51]: print("Number of Rows:",real_estate.shape[0])
print("Number of Columns:",real_estate.shape[1])
```

Number of Rows: 13320  
Number of Columns: 9

```
In [52]: # Step 2: Clean Column Names (remove spaces, etc.)
real_estate.columns = real_estate.columns.str.strip().str.lower().str.replace(' ',
real_estate.columns
```

Out[52]: Index(['area\_type', 'availability', 'location', 'size', 'society',  
                 'total\_sqft', 'bath', 'balcony', 'price'],  
             dtype='object')

```
In [53]: real_estate.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 13320 entries, 0 to 13319
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   area_type       13320 non-null  object
1   availability     13320 non-null  object
2   location        13319 non-null  object
3   size            13304 non-null  object
4   society         7818 non-null   object
5   total_sqft      13320 non-null  object
6   bath            13247 non-null  float64
7   balcony         12711 non-null  float64
8   price           13320 non-null  float64
dtypes: float64(3), object(6)
memory usage: 936.7+ KB
```

```
In [54]: #Step 3: Handle Missing Values
print(real_estate.isnull().sum())
```

```
area_type      0
availability    0
location       1
size           16
society       5502
total_sqft     0
bath           73
balcony       609
price          0
dtype: int64
```

```
In [55]: real_estate['size'].fillna(real_estate['size'].mode()[0], inplace=True)
real_estate['bath'].fillna(real_estate['bath'].median(), inplace=True)
real_estate['balcony'].fillna(real_estate['balcony'].mode()[0], inplace=True)
real_estate['location'].fillna(real_estate['location'].mode()[0], inplace=True)

real_estate.drop(columns=['society'], inplace=True)

real_estate.isnull().sum()
```

```
Out[55]: area_type      0
availability    0
location       0
size           0
total_sqft     0
bath           0
balcony        0
price          0
dtype: int64
```

```
In [56]: real_estate.dtypes
```

```
Out[56]: area_type      object
availability  object
location     object
size         object
total_sqft   object
bath         float64
balcony      float64
price        float64
dtype: object
```

```
In [57]: #Step 4: Filter and Subset the Data
available = real_estate[real_estate['availability'] != 'Ready To Move']
print("Properties available for sale", available)
```

Properties available for sale				area_type availability	
location	size	\			
0	Super built-up	Area	19-Dec	Electronic City Phase II	2 BHK
6	Super built-up	Area	18-May	Old Airport Road	4 BHK
10	Super built-up	Area	18-Feb	Whitefield	3 BHK
21	Super built-up	Area	19-Dec	Binny Pete	3 BHK
24	Super built-up	Area	18-Nov	Thanisandra	1 RK
...		...	...	...	...
13291	Plot	Area	18-Jan	Weavers Colony	1 Bedroom
13292	Super built-up	Area	18-Jul	Udayapur Village	3 BHK
13295	Super built-up	Area	18-Feb	Haralur Road	3 BHK
13299	Super built-up	Area	18-Dec	Whitefield	4 BHK
13318	Super built-up	Area	18-Jun	Padmanabhanagar	4 BHK

  

	total_sqft	bath	balcony	price
0	1056	2.0	1.0	39.07
6	2732	4.0	2.0	204.00
10	1800	2.0	2.0	70.00
21	1755	3.0	1.0	122.00
24	510	1.0	0.0	25.25
...	...	...	...	...
13291	812	1.0	0.0	26.00
13292	1440	2.0	2.0	63.93
13295	1810	3.0	2.0	112.00
13299	2830 - 2882	5.0	0.0	154.50
13318	4689	4.0	1.0	488.00

[2739 rows x 8 columns]

```
In [58]: high_value = real_estate[real_estate['price'] > 100]
print('Properties with high price:', high_value)
```

Properties with high price:				area_type	availability	location
location	size	\				
1	Plot	Area	Ready To Move	Chikka Tirupathi	4 Bedroom	
6	Super built-up	Area	18-May	Old Airport Road	4 BHK	
7	Super built-up	Area	Ready To Move	Rajaji Nagar	4 BHK	
9	Plot	Area	Ready To Move	Gandhi Bazar	6 Bedroom	
11	Plot	Area	Ready To Move	Whitefield	4 Bedroom	
...		...	...	...	...	
13311	Plot	Area	Ready To Move	Ramamurthy Nagar	7 Bedroom	
13314	Super built-up	Area	Ready To Move	Green Glen Layout	3 BHK	
13315	Built-up	Area	Ready To Move	Whitefield	5 Bedroom	
13316	Super built-up	Area	Ready To Move	Richards Town	4 BHK	
13318	Super built-up	Area	18-Jun	Padmanabhanagar	4 BHK	

  

	total_sqft	bath	balcony	price
1	2600	5.0	3.0	120.0
6	2732	4.0	2.0	204.0
7	3300	4.0	2.0	600.0
9	1020	6.0	2.0	370.0
11	2785	5.0	3.0	295.0
...	...	...	...	...
13311	1500	9.0	2.0	250.0
13314	1715	3.0	3.0	112.0
13315	3453	4.0	0.0	231.0
13316	3600	5.0	2.0	400.0
13318	4689	4.0	1.0	488.0

[4103 rows x 8 columns]

```
In [59]: #Step 5: Aggregate Data for Insights
avg_price_by_location = real_estate.groupby('location')['price'].mean().sort_values
print("Average price by location:\n",avg_price_by_location)

avg_price_by_area = real_estate.groupby('area_type')['price'].mean()
print("Average price by area type:\n",avg_price_by_area)
```

```
Average price by location:
  location
Cubbon Road      1900.000000
Ashok Nagar      1486.000000
Defence Colony   1167.714286
Yemlur           1093.388889
Church Street    1068.000000
D Souza Layout   1015.000000
Sadashiva Nagar  1011.100000
Sindhi Colony    988.000000
Srinivas Colony  922.000000
5th Block Jayanagar 905.000000
Name: price, dtype: float64
Average price by area type:
  area_type
Built-up Area    104.285498
Carpet Area       89.502356
Plot Area        208.495486
Super built-up Area 92.971757
Name: price, dtype: float64
```

```
In [60]: # Step 8: Identify and Handle Outliers
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```
Q1 = real_estate['price'].quantile(0.25)
Q3 = real_estate['price'].quantile(0.75)
IQR = Q3 - Q1

lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

real_estate = real_estate[(real_estate['price'] >= lower_bound) & (real_estate['price'] <= upper_bound)]

print("Outliers handled successfully!")
```

Outliers handled successfully!

```
In [61]: # Step 9: Export Cleaned Data for Further Analysis
real_estate.to_csv("Cleaned_Bengaluru_House_Data.csv", index=False)
print("Cleaned dataset saved successfully!")
```

Cleaned dataset saved successfully!