

Web Scraping and Model Development for the Stockholm housing market dataset



Business Problem

To analyze how reduction in interest rates influences property prices and migration patterns in Stockholm's housing market

Objective

To build a Polynomial Regression model to predict prices and investigate the relationship between property types, interest rates and housing transitions.

Data Collection

1. Web Scraping the Housing Market Website: www.Booli.se
2. Downloading Monetary Policy report : <https://www.riksbank.se/>

Data Set Description

Sold Price : Final price at which the house was sold in Swedish krona(SEK)(kr)

Street Address: Specific location of the sold house
Sold_date : Date at which the property sale is legally finalized
Area Name : Descriptive location area name
Object Type : Categorical variable which represents the different type of houses
(e.g., Apartment, Rowhouse, Villa etc)
Date : Interest Rate published date by the Riskbank
Interest Rate : Rate of interest on the specified date

1. Web Scraping

```
In [1]: #Importing the necessary modules
import requests
import json
import pandas as pd
from bs4 import BeautifulSoup
import time
from datetime import datetime
```

Scraping rowhouse, semi-detached house URLs has been done separately to avoid overwhelming the server and time-out error

```
In [2]: # Defining base URLs for each object type
base_urls = {
    'Lägenhet': 'https://www.booli.se/sok/slutpriser?areaIds=1&objectType=Lägenhet&sort=s',
    'Villa': 'https://www.booli.se/sok/slutpriser?areaIds=1&objectType=Villa&sort=s'
}
```

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In [3]: # Initialize a dictionary to store lists for each property type
results = {
    'Lägenhet': [],
    'Villa': []
}

# Set maximum pages to scrape
max_pages = 300

# Loop through each property type
for object_type, base_url in [('Lägenhet', base_urls['Lägenhet']),
                               ('Villa', base_urls['Villa'])]:

    print(f"Scraping {object_type}")

    # Pagination for scraping multiple desired pages
    for page_number in range(1, max_pages + 1):
        url = base_url.format(page_number)
        response = requests.get(url)

        if response.status_code == 200:
            print(f"Scraping page {page_number} for {object_type}")
```

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#Parsing the HTML content using BeautifulSoup
soup = BeautifulSoup(response.content, 'html.parser')

#Inspect the scriptid from HTML structure to retrieve data as there is
script_tag = soup.find('script', id='__NEXT_DATA__')
json_data = json.loads(script_tag.string)
apollo_state = json_data['props']['pageProps'][ '__APOLLO_STATE__']

# Check for sold properties
data_found = False
for key in apollo_state.keys():
    if key.startswith('SoldProperty:'):
        data_found = True
        property_data = apollo_state[key]

    # Only collect data if the object type matches
    if property_data.get('objectType') == object_type:

        sold_price = property_data.get('soldPrice', {}).get('format')

        street_address = property_data.get('streetAddress', 'N/A')

        # Handle living_area extraction safely based on faced challenge
        living_area_data = property_data.get('livingArea', None)

        if living_area_data is not None:
            living_area = living_area_data.get('formatted', 'N/A')
        else:
            living_area = 'N/A'

        object_type = property_data.get('objectType', 'N/A')

        area_name = property_data.get('descriptiveAreaName', 'N/A')

        sold_date = property_data.get('soldDate', 'N/A')

    # Append data to the appropriate list
    results[object_type].append({
        'Sold Price': sold_price,
        'Street Address': street_address,
        'Living Area': living_area,
        'Object Type': object_type,
        'Area Name': area_name,
        'Sold Date': sold_date,
    })

# In case, if no data was found on the page, break the Loop
if not data_found:
    print(f"No more data found for {object_type} on page {page_number}.")
    break
else:
    print(f"Failed to retrieve data from page {page_number} for {object_type}")
    break

```

```
# Delay to avoid rate limiting or overwhelming the server (included based on
time.sleep(2)

# Convert results to DataFrames and save as CSV for data consistency
for object_type, data in results.items():
    df = pd.DataFrame(data)
    df.to_csv(f'{object_type.lower()}.csv', index=False)
```

Scraping Lägenhet

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In [4]: # Read the individual CSV files
df_rowhouse = pd.read_csv('kedjehus_parhus_radhus.csv')
df_apartment = pd.read_csv('lägenhet.csv')
df_villa = pd.read_csv('villa.csv')

2. Monetary Policy rapport data

In [5]: interest_rate_df = pd.read_excel('styrrantan-effektiv.xlsx')

In [6]: interest_rate_df

Out[6]:

	Date	Interest Rate	Year	Month
0	2006-01-26	1.75	2006	1
1	2006-03-02	2.00	2006	3
2	2006-05-03	2.00	2006	5
3	2006-06-22	2.25	2006	6
4	2006-09-07	2.50	2006	9
...
112	2024-07-03	3.75	2024	7
113	2024-08-21	3.50	2024	8
114	2024-10-02	3.25	2024	10
115	2024-11-13	2.75	2024	11
116	2025-01-08	2.50	2025	1

117 rows × 4 columns

Data Cleaning and Data Preprocessing

In [7]:

```
print(df_rowhouse.head())
print(df_apartment.head())
print(df_villa.head())
```

	Sold Price	Street Address	Living Area	Object Type	\
0	6 755 000 kr	Långseleringen 58	112 m ²	Radhus	
1	4 200 000 kr	Kwickrotsgränd 98	119 m ²	Radhus	
2	5 825 000 kr	Skrattmåsvägen 40	115 m ²	Radhus	
3	4 150 000 kr	Valborgsstigen 4	118 m ²	Radhus	
4	5 200 000 kr	Mickelsbergsvägen 144	107 m ²	Kedjehus	

	Area Name	Sold Date
0	Råcksta	2025-01-23
1	Hässelby Norra Villastad	2025-01-23
2	Fagersjö	2025-01-22
3	Nälsta	2025-01-22
4	Hägersten-Liljeholmen	2025-01-21

	Sold Price	Street Address	Living Area	Object Type	\
0	4 130 000 kr	Åkeshovsvägen 36	85 m ²	Lägenhet	
1	4 050 000 kr	Blommensbergsvägen 188	58 m ²	Lägenhet	
2	4 900 000 kr	Banérgatan 50	42 m ²	Lägenhet	
3	3 100 000 kr	Sicklingsvägen 5	56 m ²	Lägenhet	
4	4 600 000 kr	Strandliden 49	73 m ²	Lägenhet	

	Area Name	Sold Date
0	Åkeslund	2025-01-24
1	Aspudden	2025-01-24
2	Östermalm	2025-01-24
3	Farsta Stadsdelsområde	2025-01-24
4	Hässelby Strand	2025-01-24

	Sold Price	Street Address	Living Area	Object Type	Area Name	\
0	11 650 000 kr	Sköldingevägen 40A	172 m ²	Villa	Örby	
1	7 550 000 kr	Vinstavägen 29A	140 m ²	Villa	Spånga	
2	37 500 000 kr	Äppelbovägen 12	235 m ²	Villa	Bromma	
3	10 050 000 kr	Timrågatan 108	169 m ²	Villa	Vällingbyhöjden	
4	7 500 000 kr	Attundavägen 33	157 m ²	Villa	Bromma Kyrka	

	Sold Date
0	2025-01-24
1	2025-01-23
2	2025-01-22
3	2025-01-22
4	2025-01-18

In [8]: `df_rowhouse.isnull().any()`

Out[8]:

Sold Price	False
Street Address	False
Living Area	True
Object Type	False
Area Name	False
Sold Date	False
<code>dtype: bool</code>	

In [9]: `df_apartment.isnull().any()`

```
Out[9]: Sold Price    False
         Street Address  False
         Living Area     True
         Object Type    False
         Area Name      False
         Sold Date      False
         dtype: bool
```

```
In [10]: df_villa.isnull().any()
```

```
Out[10]: Sold Price    False
          Street Address  False
          Living Area     True
          Object Type    False
          Area Name      False
          Sold Date      False
          dtype: bool
```

```
In [11]: #Merging it into a single dataframe
df_merged = pd.concat([df_rowhouse, df_apartment, df_villa], ignore_index = True)
```

```
In [12]: # Converting sold_date into datetime
df_merged['Sold Date'] = pd.to_datetime(df_merged['Sold Date'])

# Initializing start_date and end_date
start_date = pd.to_datetime('2010-01-01')
end_date = pd.to_datetime('2025-1-8')
```

```
In [13]: # Extracting 'year' and 'month' as a separate column from the merged dataset
df_merged['Year'] = pd.DatetimeIndex(df_merged['Sold Date']).year
df_merged['Month'] = pd.DatetimeIndex(df_merged['Sold Date']).month
```

```
In [14]: # Filtering the dataset based on specific dates
real_estate_df = df_merged[(df_merged['Sold Date'] >= start_date) &
                           (df_merged['Sold Date'] <= end_date)]
```

```
In [15]: real_estate_df.head()
```

Out[15]:

	Sold Price	Street Address	Living Area	Object Type	Area Name	Sold Date	Year	Month
241	12 000 000 kr	Martebogatan 32	174 m ²	Parhus	Beckomberga	2024-10-25	2024	10
242	7 750 000 kr	Sten Bergmans väg 25	113 m ²	Radhus	Hammarbyhöjden	2024-10-25	2024	10
243	8 395 000 kr	Nävekvarnsvägen 5A	NaN	Parhus	Örby	2024-10-25	2024	10
244	8 170 000 kr	Sköntorpsvägen 68	123 m ²	Radhus	Årsta	2024-10-25	2024	10
245	5 490 000 kr	Lyckselevägen 85	90 m ²	Radhus	Vällingby Centrum	2024-10-25	2024	10

In [16]: `# Merging interest rate column from the df to the existing df based on year and month`
`real_estate_df = real_estate_df.merge(interest_rate_df[['Year', 'Month', 'Interest Rate']])`

In [17]: `real_estate_df.columns`

Out[17]: `Index(['Sold Price', 'Street Address', 'Living Area', 'Object Type',
 'Area Name', 'Sold Date', 'Year', 'Month', 'Interest Rate'],
 dtype='object')`

In [18]: `real_estate_df.shape`

Out[18]: `(22801, 9)`

In [19]: `real_estate_df['Interest Rate'].isnull().sum()`

Out[19]: `11237`

In [20]: `real_estate_df.head()`

Out[20]:

	Sold Price	Street Address	Living Area	Object Type	Area Name	Sold Date	Year	Month	Interest Rate
0	12 000 000 kr	Martebogatan 32	174 m ²	Parhus	Beckomberga	2024-10-25	2024	10	
1	7 750 000 kr	Sten Bergmans väg 25	113 m ²	Radhus	Hammarbyhöjden	2024-10-25	2024	10	
2	8 395 000 kr	Nävekvarnsvägen 5A	NaN	Parhus	Örby	2024-10-25	2024	10	
3	8 170 000 kr	Sköntorpsvägen 68	123 m ²	Radhus	Årsta	2024-10-25	2024	10	
4	5 490 000 kr	Lyckselevägen 85	90 m ²	Radhus	Vällingby Centrum	2024-10-25	2024	10	

In [21]: `real_estate_df.info()`

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 22801 entries, 0 to 22800
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Sold Price       22801 non-null   object  
 1   Street Address   22801 non-null   object  
 2   Living Area      22743 non-null   object  
 3   Object Type      22801 non-null   object  
 4   Area Name        22801 non-null   object  
 5   Sold Date        22801 non-null   datetime64[ns]
 6   Year              22801 non-null   int64   
 7   Month             22801 non-null   int64   
 8   Interest Rate    11564 non-null   float64 
dtypes: datetime64[ns](1), float64(1), int64(2), object(5)
memory usage: 1.7+ MB
```

In [22]: `real_estate_df.describe()`

	Year	Month	Interest Rate
count	22801.000000	22801.000000	11564.000000
mean	2021.212666	7.233060	1.576358
std	2.617668	3.051542	1.814287
min	2016.000000	1.000000	-0.500000
25%	2019.000000	5.000000	-0.250000
50%	2022.000000	8.000000	0.750000
75%	2024.000000	10.000000	3.500000
max	2024.000000	12.000000	4.000000

In [23]: `real_estate_df['Object Type'].value_counts()`

Villa	10503
Lägenhet	6097
Radhus	4749
Kedjehus	888
Parhus	563
Gård	1
Name:	Object Type, dtype: int64

Handling missing values

In [24]: `real_estate_df['Interest Rate'] = real_estate_df['Interest Rate'].fillna(method = 'ffill')`

In [25]: `real_estate_df['Interest Rate'].isnull().any()`

Out[25]: `False`

```
In [26]: # Remove non-numeric characters like comma,sq.m from the column 'Living Area'
real_estate_df['Living Area'] = real_estate_df['Living Area'].str.replace(r'[^\d]', '',
In [27]: # convert it into numeric thereby handling NaN values
real_estate_df['Living Area'] = pd.to_numeric(real_estate_df['Living Area'], errors='raise')
In [28]: # grouping based on object type before filling missing values for the column Living Area
real_estate_df['Living Area'] = real_estate_df.groupby('Object Type')['Living Area']
```

Removing Outlier

```
In [29]: real_estate_df = real_estate_df[real_estate_df['Object Type'] != 'Gård']
```

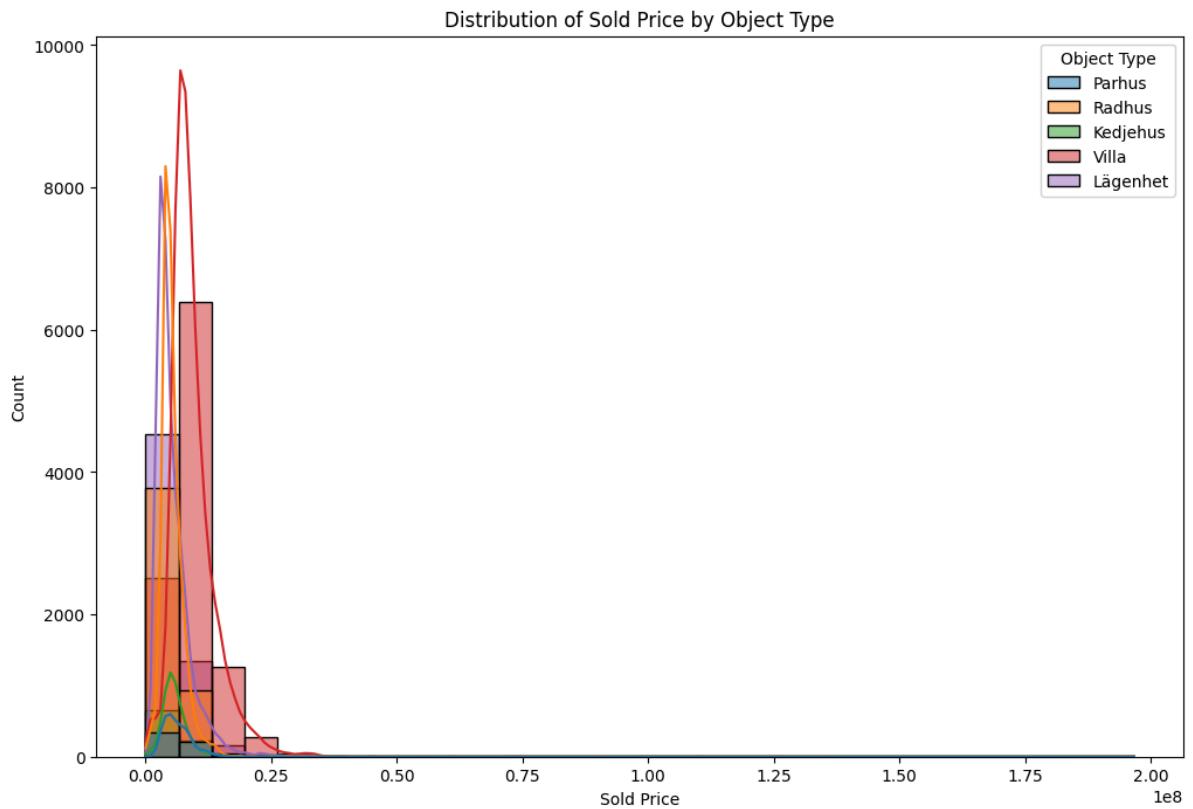
Removing inconsistent dataformat

```
In [30]: # Remove non-numeric characters like comma, kr from 'Sold Price' and convert it to
real_estate_df['Sold Price'] = real_estate_df['Sold Price'].str.replace(r'[^\d]', '',
# Verify the result
print(real_estate_df['Sold Price'].head())
0    12000000
1    7750000
2    8395000
3    8170000
4    5490000
Name: Sold Price, dtype: int32
```

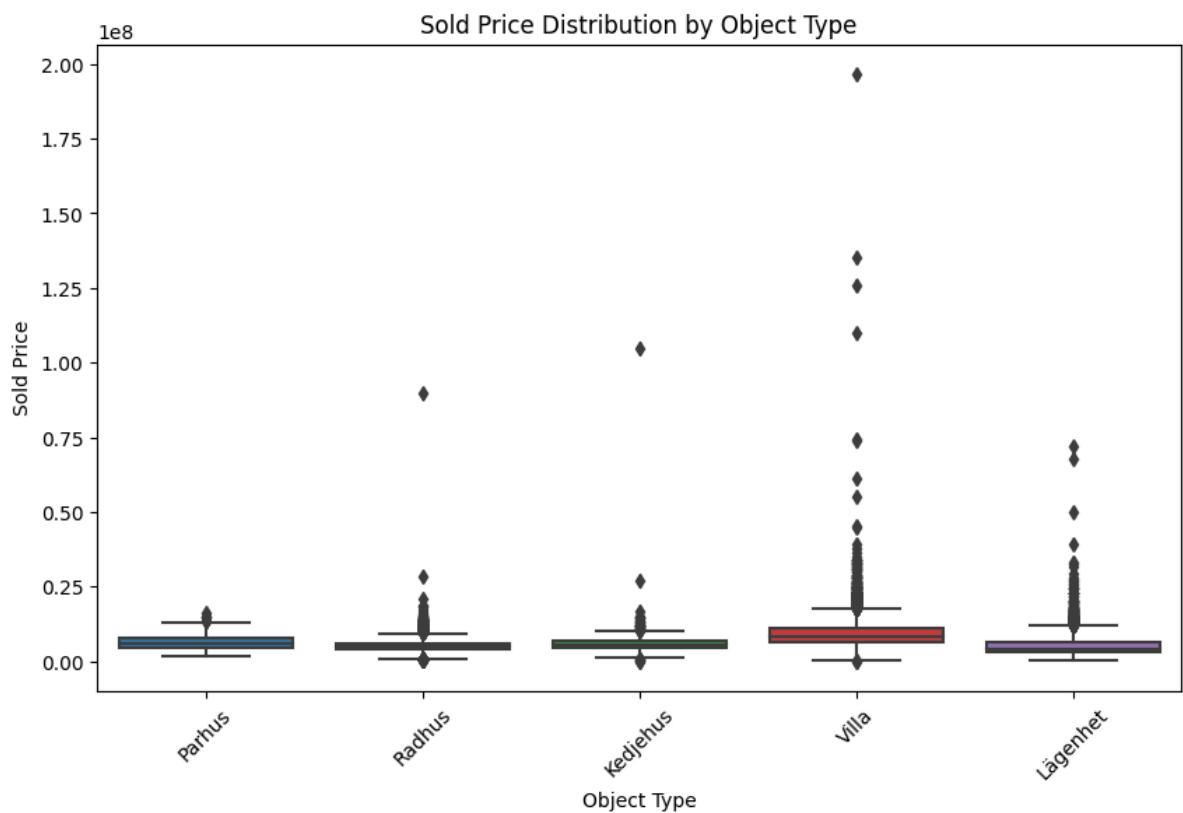
Exploratory Data Analysis

```
In [31]: #Distribution of Sold Price by Object Type
import seaborn as sns
import matplotlib.pyplot as plt

plt.figure(figsize=(12,8))
sns.histplot(data=real_estate_df, x='Sold Price', hue='Object Type', bins=30, kde=True)
plt.title('Distribution of Sold Price by Object Type')
plt.show()
```



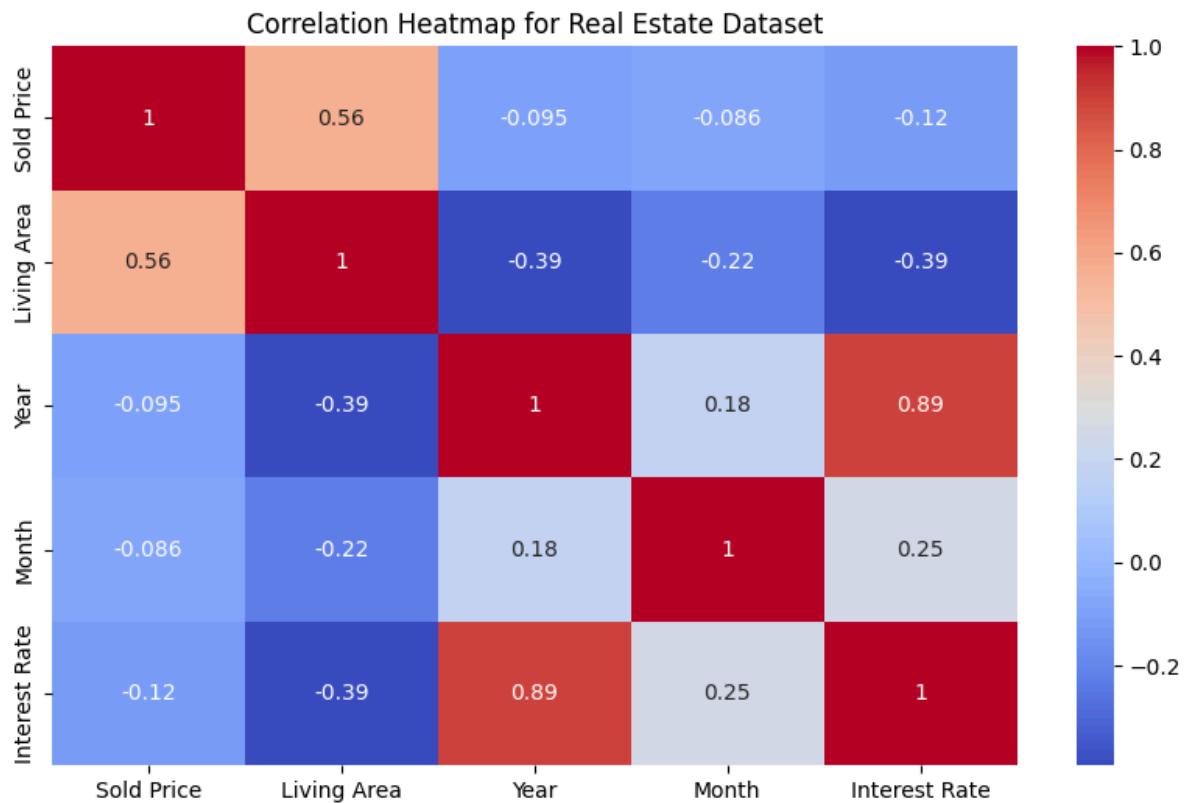
```
In [32]: # to understand IQR, min and max value of sold price for each object type
plt.figure(figsize=(10,6))
sns.boxplot(data=real_estate_df, x='Object Type', y='Sold Price')
plt.title('Sold Price Distribution by Object Type')
plt.xticks(rotation=45)
plt.show()
```



```
In [33]: # to understand the relationship between numerical variables
plt.figure(figsize = (10,6))
sns.heatmap(real_estate_df.corr(), annot = True, cmap = 'coolwarm')
plt.title('Correlation Heatmap for Real Estate Dataset')
plt.show()
```

C:\Users\nklmy\AppData\Local\Temp\ipykernel_5000\1080443470.py:3: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric_only to silence this warning.

```
sns.heatmap(real_estate_df.corr(), annot = True, cmap = 'coolwarm')
```

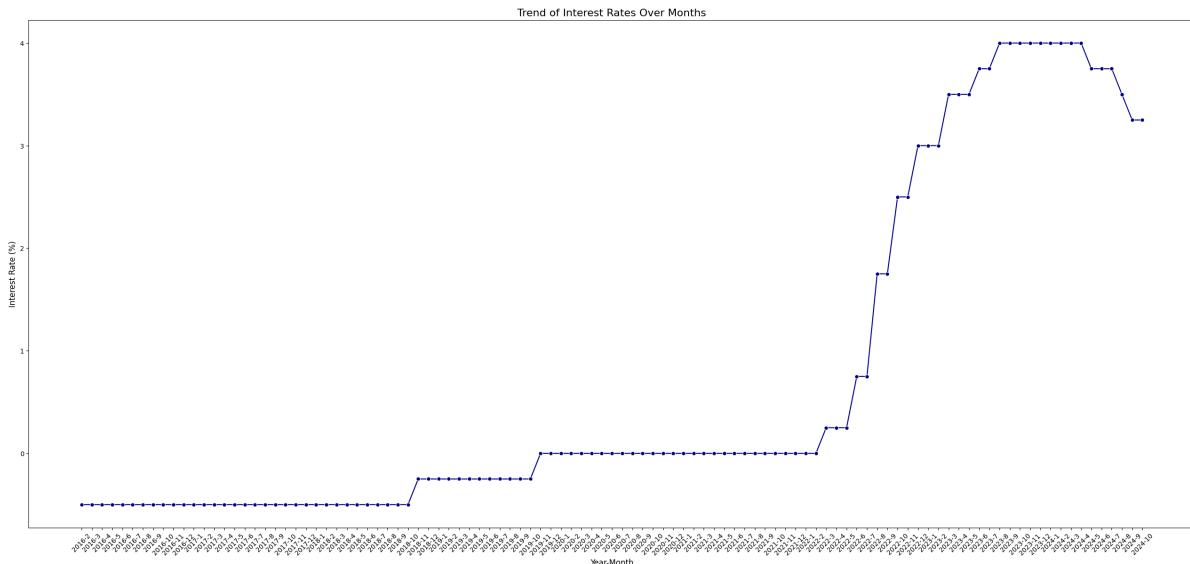


```
In [34]: # Create a 'Year-Month' column to represent it in the plot
real_estate_df['Year-Month'] = real_estate_df['Year'].astype(str) + '-' + real_estate_df['Month'].astype(str)

# Sort by 'Year-Month' to ensure proper order
real_estate_df = real_estate_df.sort_values(by=[ 'Year', 'Month'])

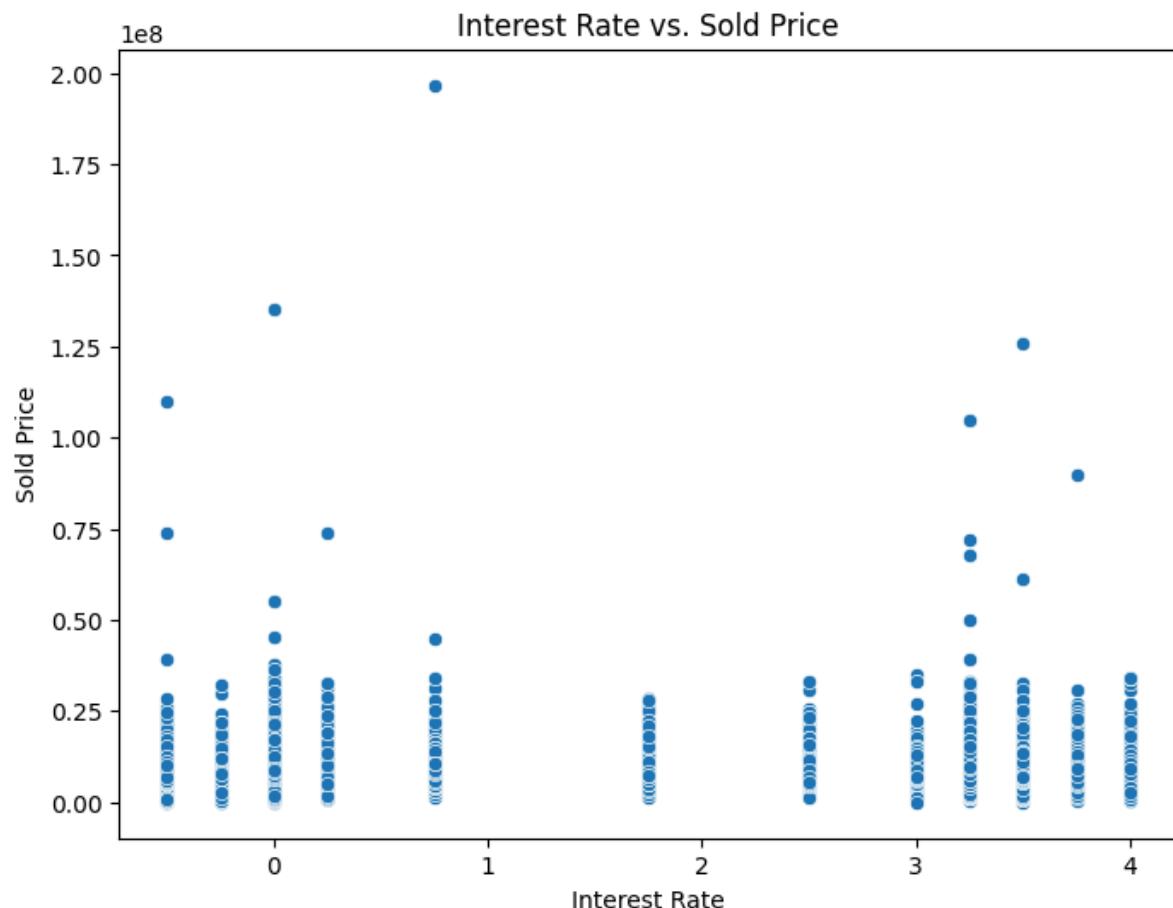
# Plotting the interest rate trend over months
plt.figure(figsize=(32,14))
sns.lineplot(data=real_estate_df, x='Year-Month', y='Interest Rate', color = 'darkblue')

# Customize the plot
plt.title('Trend of Interest Rates Over Months', fontsize=16)
plt.xlabel('Year-Month', fontsize=12)
plt.ylabel('Interest Rate (%)', fontsize=12)
plt.xticks(rotation=45)
plt.show()
```



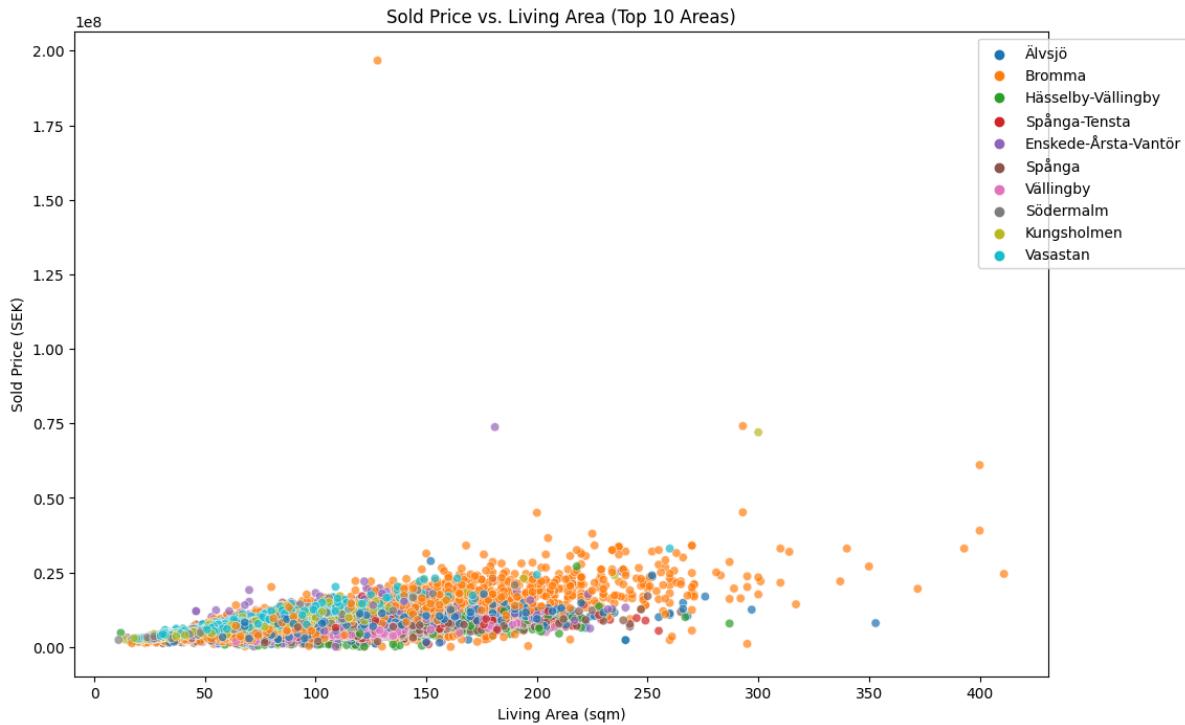
In [35]: # Scatter plot: Interest Rate vs. Sold Price

```
plt.figure(figsize=(8,6))
sns.scatterplot(x=real_estate_df['Interest Rate'], y=real_estate_df['Sold Price'])
plt.title('Interest Rate vs. Sold Price')
plt.show()
```



In [36]: # Extracting top 10 areas by count (regardless of the property type, total count mat
top_areas = real_estate_df['Area Name'].value_counts().nlargest(10).index
filtered_df = real_estate_df[real_estate_df['Area Name'].isin(top_areas)]

```
# Plot
plt.figure(figsize=(12,8))
sns.scatterplot(data=filtered_df, x='Living Area', y='Sold Price', hue='Area Name',
plt.title('Sold Price vs. Living Area (Top 10 Areas)')
plt.ylabel('Sold Price (SEK)')
plt.xlabel('Living Area (sqm)')
plt.legend(loc='upper right', bbox_to_anchor=(1.15, 1))
plt.show()
```



In [37]:

```
# Calculate average Living Area and Sold Price by Area and Object Type
area_stats = real_estate_df.groupby(['Area Name', 'Object Type']).agg(
    avg_living_area=('Living Area', 'mean'),
    avg_sold_price=('Sold Price', 'mean')
).reset_index()

# Get top 10 areas by average Living Area
top_areas_living_area = area_stats.nlargest(10, 'avg_living_area')['Area Name']
# Get top 10 areas by average Sold Price
top_areas_sold_price = area_stats.nlargest(10, 'avg_sold_price')['Area Name']
```

In [38]:

```
# Plot Average Living Area and Sold Price Across Object Types for Top 10 Areas
import seaborn as sns
import matplotlib.pyplot as plt

# Filter data for top 10 areas by Living Area and Sold Price
filtered_living_area = area_stats[area_stats['Area Name'].isin(top_areas_living_are
filtered_sold_price = area_stats[area_stats['Area Name'].isin(top_areas_sold_price)

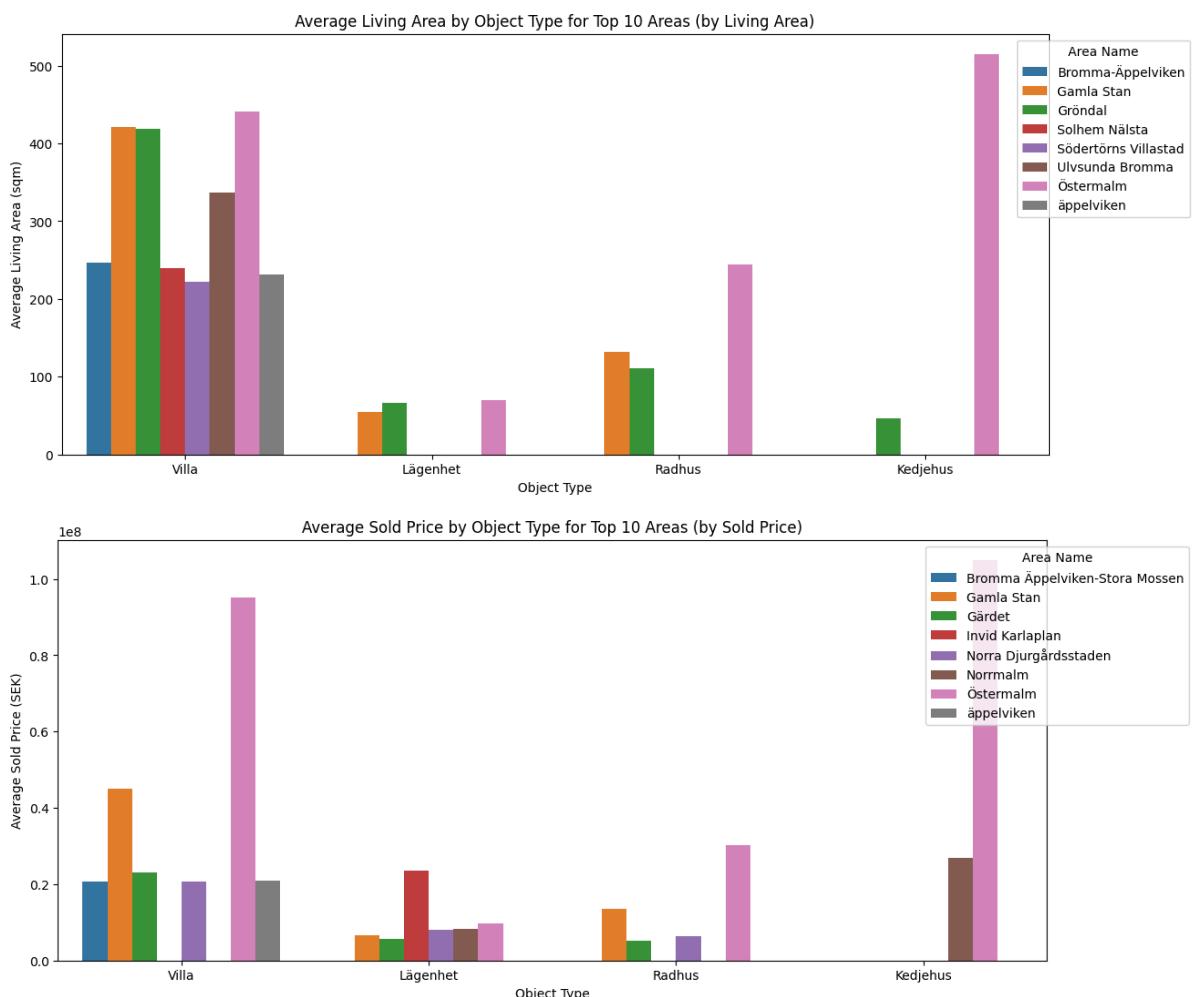
# Plot for Average Living Area
plt.figure(figsize=(14, 6))
sns.barplot(
    data=filtered_living_area,
    x='Object Type',
```

```

y='avg_living_area',
hue='Area Name',
palette='tab10'
)
plt.title('Average Living Area by Object Type for Top 10 Areas (by Living Area)')
plt.ylabel('Average Living Area (sqm)')
plt.xlabel('Object Type')
plt.legend(title='Area Name', loc='upper right', bbox_to_anchor=(1.15, 1))
plt.show()

# Plot for Average Sold Price
plt.figure(figsize=(14, 6))
sns.barplot(
    data=filtered_sold_price,
    x='Object Type',
    y='avg_sold_price',
    hue='Area Name',
    palette='tab10'
)
plt.title('Average Sold Price by Object Type for Top 10 Areas (by Sold Price)')
plt.ylabel('Average Sold Price (SEK)')
plt.xlabel('Object Type')
plt.legend(title='Area Name', loc='upper right', bbox_to_anchor=(1.15, 1))
plt.show()

```



Encoding Categorical Variable

```
In [39]: real_estate_df = pd.get_dummies(real_estate_df, columns = ['Object Type'], drop_fir
```

```
In [40]: real_estate_df
```

Out[40]:

	Sold Price	Street Address	Living Area	Area Name	Sold Date	Year	Month	Interest Rate	Year Mont
22734	6950000	Humlestigen 5	118.0	Älvsjö	2016-02-29	2016	2	-0.50	2016
22735	7360000	Timrågatan 31	160.0	Vällingbyhöjden	2016-02-29	2016	2	-0.50	2016
22736	12800000	Virvelvindsvägen 58	106.0	Bromma	2016-02-29	2016	2	-0.50	2016
22737	6900000	Kålgårdsvägen 30	64.0	Enskede Gård	2016-02-29	2016	2	-0.50	2016
22738	6900000	Brändövägen 43	202.0	Hässelby-Vällingby	2016-02-29	2016	2	-0.50	2016
...
12750	17000000	Ensittarvägen 17	146.0	Essingeöarna	2024-10-01	2024	10	3.25	2024
12751	8000000	Doktor Abrahams väg 46	85.0	Bromma Kyrka	2024-10-01	2024	10	3.25	2024
12752	9200000	Bollstavägen 2	101.0	Stureby	2024-10-01	2024	10	3.25	2024
12753	9925000	Mälarhöjdsvägen 69	95.0	Hägersten-Liljeholmen	2024-10-01	2024	10	3.25	2024
12754	15250000	Tussmötevägen 223	140.0	Enskede-Årsta-Vantör	2024-10-01	2024	10	3.25	2024

22800 rows × 13 columns



Feature Scaling

```
In [41]: import numpy as np
```

```
In [42]: real_estate_df['Sold Price Log'] = np.log(real_estate_df['Sold Price'])
```

Model Development

1. Train-Test data split

```
In [61]: # importing the necessary library
from sklearn.model_selection import train_test_split

X = real_estate_df.drop(columns = ['Sold Price', 'Street Address', 'Area Name', 'Sold
y = real_estate_df['Sold Price']

# splitting the dataset into training and test data
X_train, X_test, y_train, y_test = train_test_split(X,y, test_size = 0.3, random_st
```

2. Training the Model

```
In [62]: # training the multiple Linear regression model on the training data set'
from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
pol_reg = PolynomialFeatures(degree=4)

# Transforming the features
X_pol = pol_reg.fit_transform(X_train)

# Fitting the Linear Regression model
lin_reg_of_poly = LinearRegression()
lin_reg_of_poly.fit(X_pol, y_train)

# Transforming the test data
X_test_pol = pol_reg.transform(X_test)
```

3. Model Prediction

```
In [63]: # Making predictions on the transformed test data
y_pred = lin_reg_of_poly.predict(X_test_pol)
```

4. Model Evaluation

```
In [64]: # Mean Squared Error and R-squared for model evaluation
from sklearn.metrics import mean_squared_error, r2_score

mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)

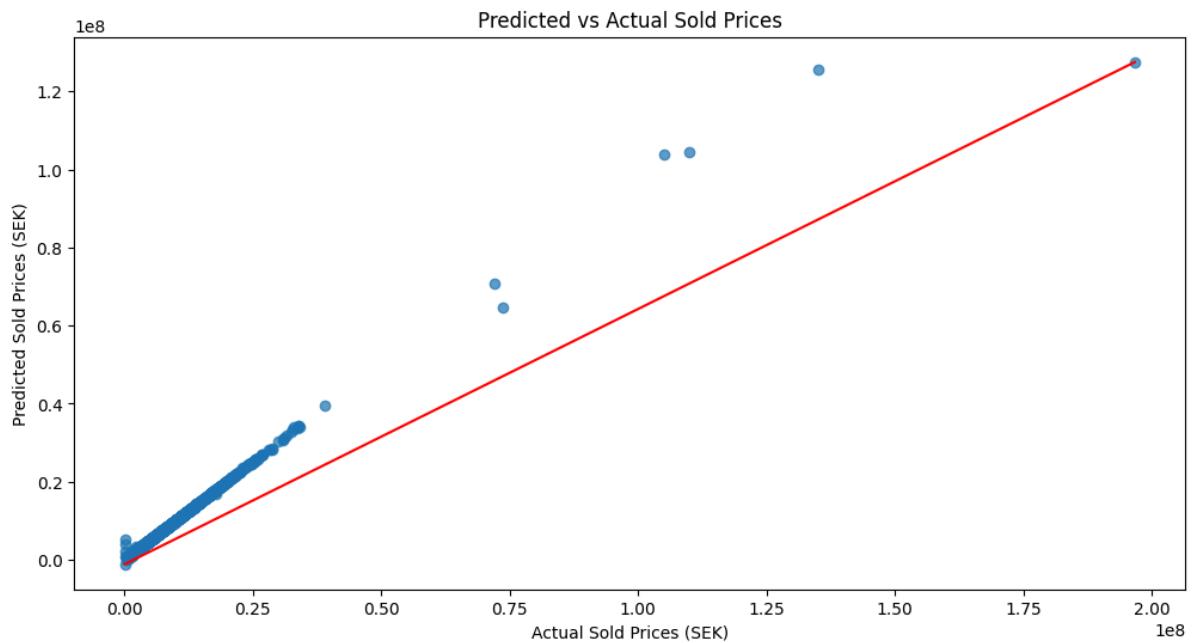
# Calculate Root Mean Squared Error (RMSE)
rmse = np.sqrt(mse)

print(f'Root Mean Squared Error: {rmse:.2f} SEK')

print(f'R-squared: {r2}')
```

Root Mean Squared Error: 860193.63 SEK
R-squared: 0.973746235485654

```
In [65]: # Plotting predicted vs actual values
plt.figure(figsize=(12, 6))
plt.scatter(y_test, y_pred, alpha=0.7)
plt.plot([y_test.min(), y_test.max()], [y_pred.min(), y_pred.max()], color='red')
plt.title('Predicted vs Actual Sold Prices')
plt.xlabel('Actual Sold Prices (SEK)')
plt.ylabel('Predicted Sold Prices (SEK)')
plt.show()
```



5. Saving the model

```
In [66]: # Save the model to a file
import joblib
joblib.dump(lin_reg_of_poly, 'stockholm_property_price_model.pkl')
```

```
Out[66]: ['stockholm_property_price_model.pkl']
```