

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'
}
```

```
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
...
110	2024-04-03	4.00	2024	4
111	2024-05-15	3.75	2024	5
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 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	4 420 000 kr	Gräsmarksgränd 33	92 m ²	Radhus	
1	3 799 000 kr	Kattfotsbacken 30	104 m ²	Radhus	
2	1 350 000 kr	Imatragatan 324	39½ m ²	Radhus	
3	4 800 000 kr	Torplyckegränd 206	113 m ²	Parhus	
4	9 450 000 kr	Letstigen 12	102 m ²	Radhus	

	Area Name	Sold Date
0	Farsta	2024-10-24
1	Hässelby Norra Villastad	2024-10-23
2	Akalla Trädgårdsstad	2024-10-22
3	Vinsta	2024-10-21
4	Bagarmossen	2024-10-21

	Sold Price	Street Address	Living Area	Object Type	\
0	6 400 000 kr	Vinodlargatan 11	84½ m ²	Lägenhet	
1	3 300 000 kr	Dalbobranten 31	73 m ²	Lägenhet	
2	8 100 000 kr	Ludvigsbergsgatan 5	81 m ²	Lägenhet	
3	3 750 000 kr	Tellusborgsvägen 66	45 m ²	Lägenhet	
4	13 350 000 kr	Bondegatan 74A	97 m ²	Lägenhet	

	Area Name	Sold Date
0	Liljeholmskajen	2024-11-11
1	Sköndal	2024-11-11
2	Stockholm Södermalm	2024-11-11
3	Midsommarkransen	2024-11-11
4	Södermalm	2024-11-11

	Sold Price	Street Address	Living Area	Object Type	Area Name	\
0	6 950 000 kr	Vagnvägen 1	68 m ²	Villa	Älvsjö	
1	9 400 000 kr	Korpmossevägen 59	69 m ²	Villa	Telefonplan	
2	10 550 000 kr	Långskepsgatan 45	95 m ²	Villa	Norra Ängby	
3	7 850 000 kr	Dalängsvägen 11	131 m ²	Villa	Spånga Sundby	
4	5 350 000 kr	Eksätravägen 22	145 m ²	Villa	Bredäng	

	Sold Date
0	2024-11-08
1	2024-11-08
2	2024-11-07
3	2024-11-07
4	2024-11-07

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
dtype: bool	

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('2024-10-25')
```

```
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()
```

	Sold Price	Street Address	Living Area	Object Type	Area Name	Sold Date	Year	Month
0	4 420 000 kr	Gräsmarksgränd 33	92 m ²	Radhus	Farsta	2024-10-24	2024	10
1	3 799 000 kr	Kattfotsbacken 30	104 m ²	Radhus	Hässelby Norra Villastad	2024-10-23	2024	10
2	1 350 000 kr	Imatragatan 324	39½ m ²	Radhus	Akalla Trädgårdsstad	2024-10-22	2024	10
3	4 800 000 kr	Torplycke gränd 206	113 m ²	Parhus	Vinsta	2024-10-21	2024	10
4	9 450 000 kr	Letstigen 12	102 m ²	Radhus	Bagarmossen	2024-10-21	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]: (30524, 9)
```

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

```
Out[19]: 15057
```

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

	Sold Price	Street Address	Living Area	Object Type	Area Name	Sold Date	Year	Month	Interest Rate
0	4 420 000 kr	Gräsmarksgränd 33	92 m ²	Radhus	Farsta	2024-10-24	2024	10	3.25
1	3 799 000 kr	Kattfotsbacken 30	104 m ²	Radhus	Hässelby Norra Villastad	2024-10-23	2024	10	3.25
2	1 350 000 kr	Imatragatan 324	39½ m ²	Radhus	Akalla Trädgårdsstad	2024-10-22	2024	10	3.25
3	4 800 000 kr	Torplycke gränd 206	113 m ²	Parhus	Vinsta	2024-10-21	2024	10	3.25
4	9 450 000 kr	Letstigen 12	102 m ²	Radhus	Bagarmossen	2024-10-21	2024	10	3.25

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

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 30524 entries, 0 to 30523
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Sold Price       30524 non-null   object  
 1   Street Address   30524 non-null   object  
 2   Living Area      30439 non-null   object  
 3   Object Type      30524 non-null   object  
 4   Area Name        30524 non-null   object  
 5   Sold Date        30524 non-null   datetime64[ns]
 6   Year             30524 non-null   int64  
 7   Month            30524 non-null   int64  
 8   Interest Rate    15467 non-null   float64 
dtypes: datetime64[ns](1), float64(1), int64(2), object(5)
memory usage: 2.3+ MB
```

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

	Year	Month	Interest Rate
count	30524.000000	30524.000000	15467.000000
mean	2020.728509	6.959802	1.631412
std	3.327671	2.951251	1.851864
min	2013.000000	1.000000	-0.500000
25%	2018.000000	5.000000	-0.250000
50%	2021.000000	7.000000	1.000000
75%	2024.000000	9.000000	3.500000
max	2024.000000	12.000000	4.000000

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

Villa	10810
Lägenhet	9582
Radhus	7735
Kedjehus	1637
Parhus	759
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
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

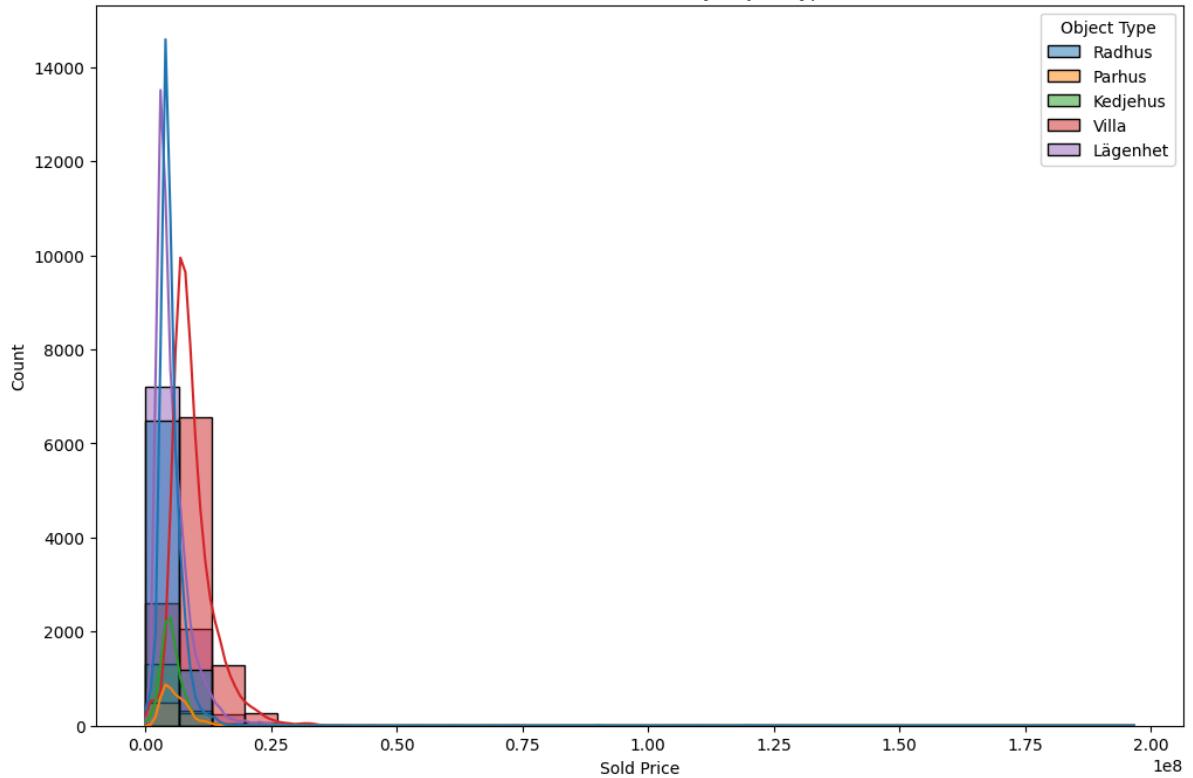
```
In [30]: # Remove non-numeric characters like comma, kr from 'Sold Price' and convert it to
real_estate_df['Sold Price'] = real_estate_df['Sold Price'].str.replace(r'[^\d]', '',
# Verify the result
print(real_estate_df['Sold Price'].head())
0    4420000
1    3799000
2    1350000
3    4800000
4    9450000
Name: Sold Price, dtype: int32
```

Exploratory Data Analysis

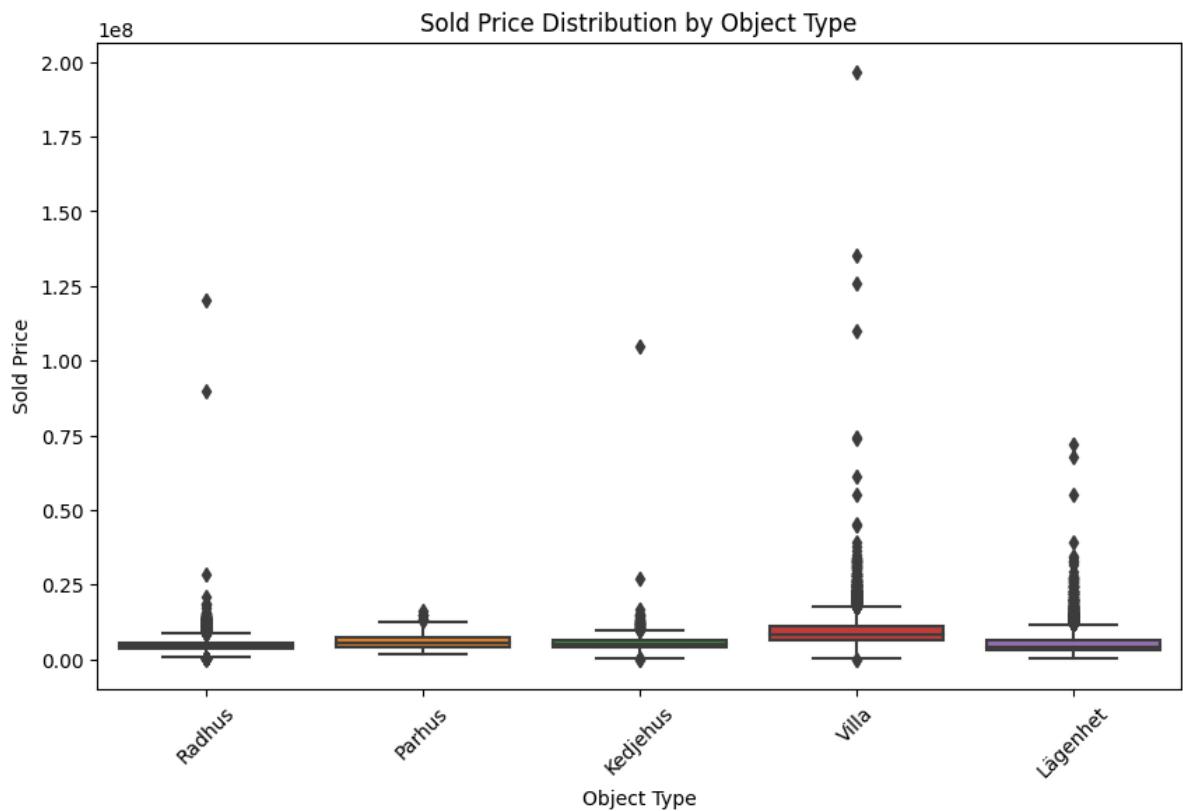
```
In [31]: #Distribution of Sold Price by Object Type
import seaborn as sns
import matplotlib.pyplot as plt

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()
```

Distribution of Sold Price by Object Type



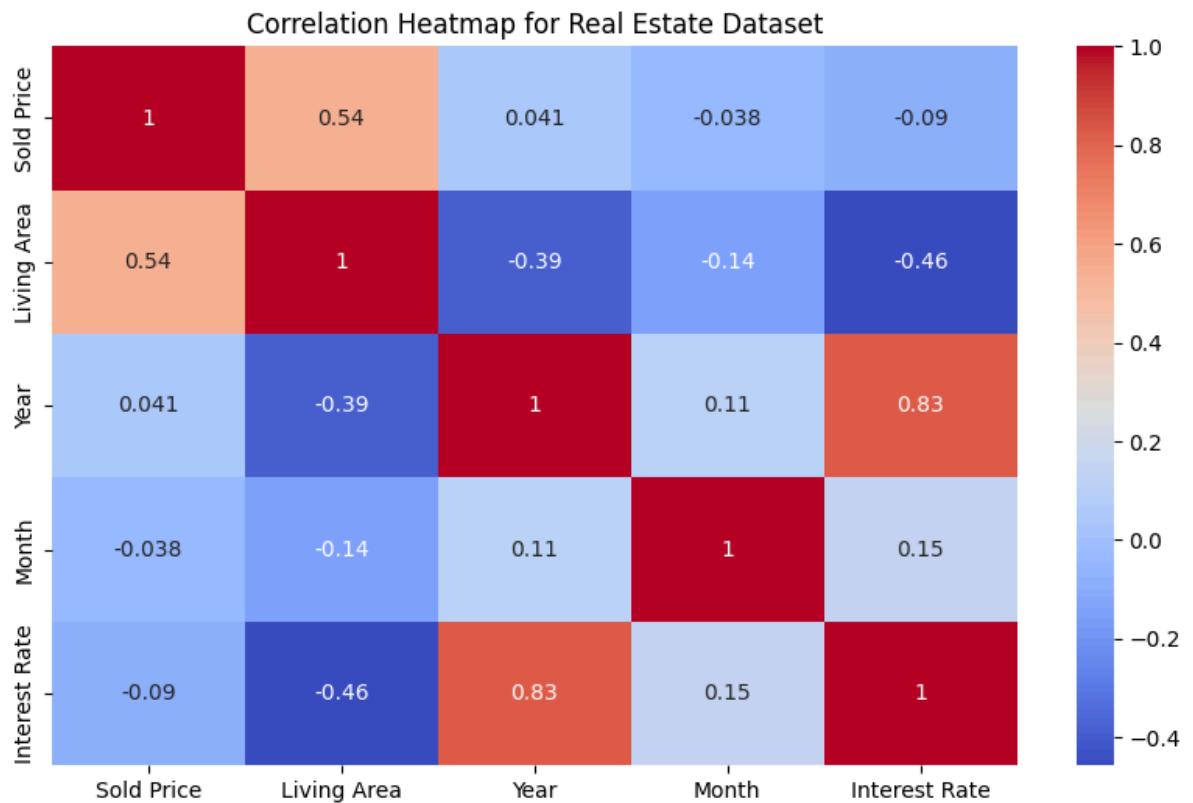
```
In [32]: # to understand IQR, min and max value of sold price for each object type
plt.figure(figsize=(10,6))
sns.boxplot(data=real_estate_df, x='Object Type', y='Sold Price')
plt.title('Sold Price Distribution by Object Type')
plt.xticks(rotation=45)
plt.show()
```



```
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_18036\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')
```

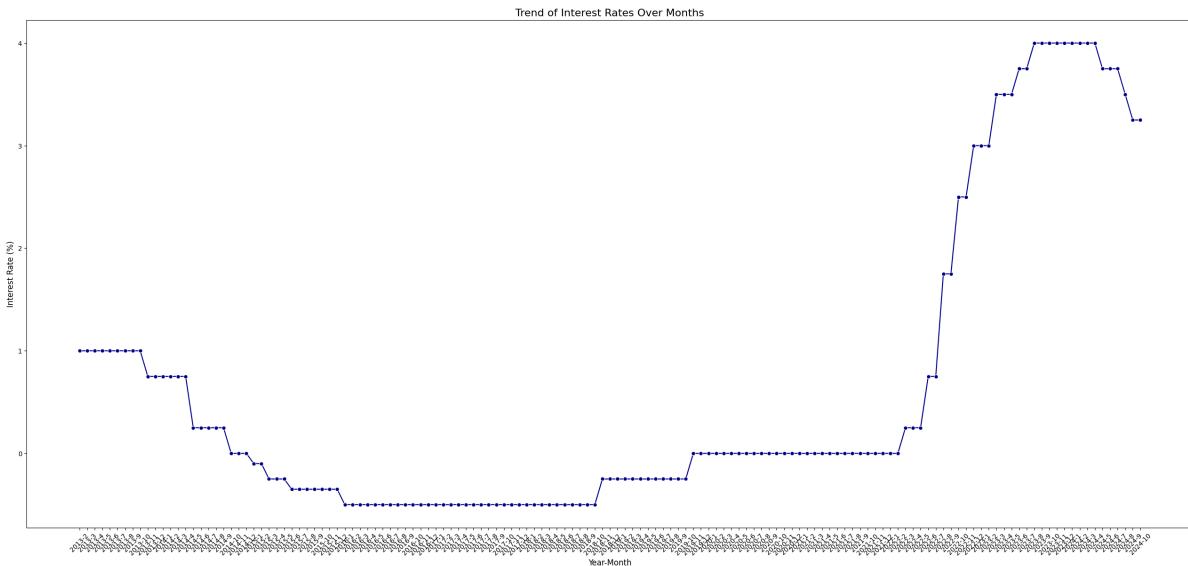


```
In [34]: # Create a 'Year-Month' column to represent it in the plot
real_estate_df['Year-Month'] = real_estate_df['Year'].astype(str) + '-' + real_esta

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

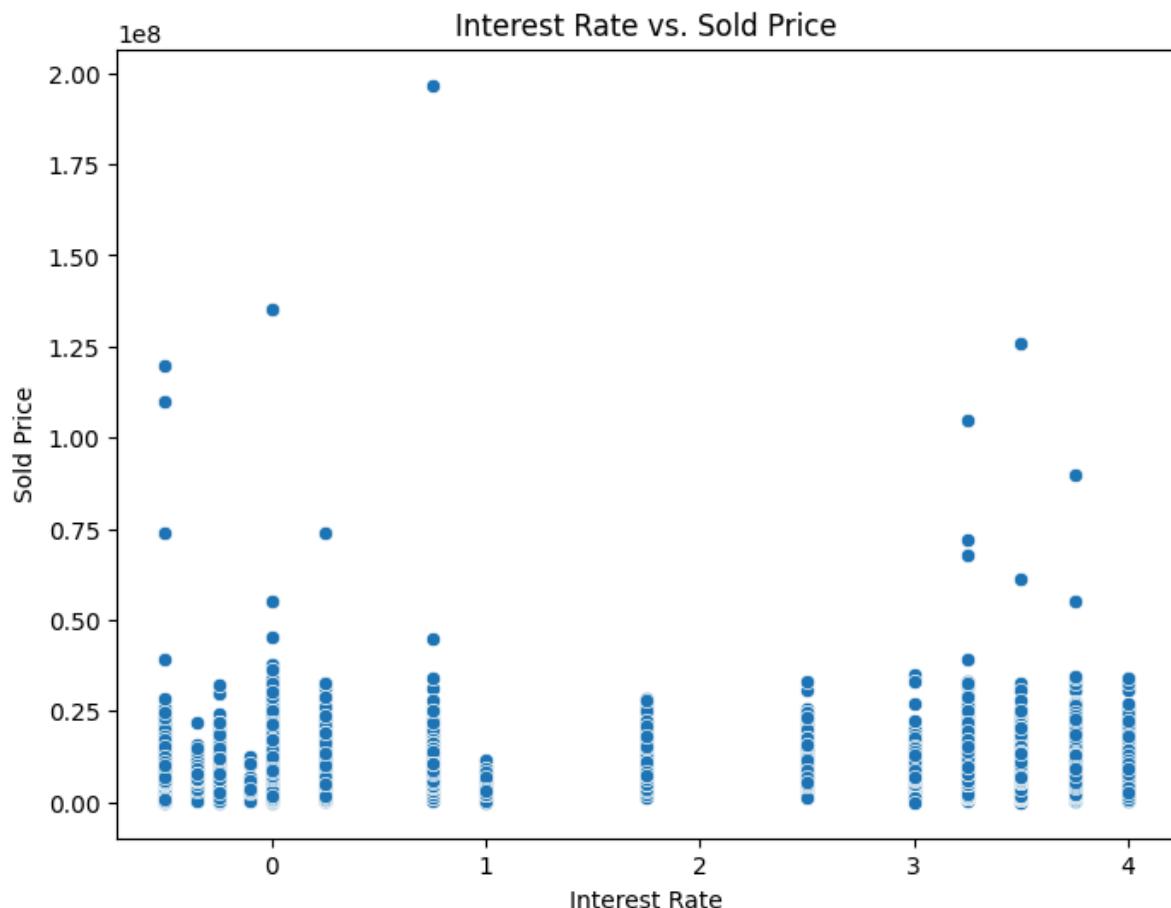
# Plotting the interest rate trend over months
plt.figure(figsize=(32,14))
sns.lineplot(data=real_estate_df, x='Year-Month', y='Interest Rate', color = 'darkblue')

# 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()
```



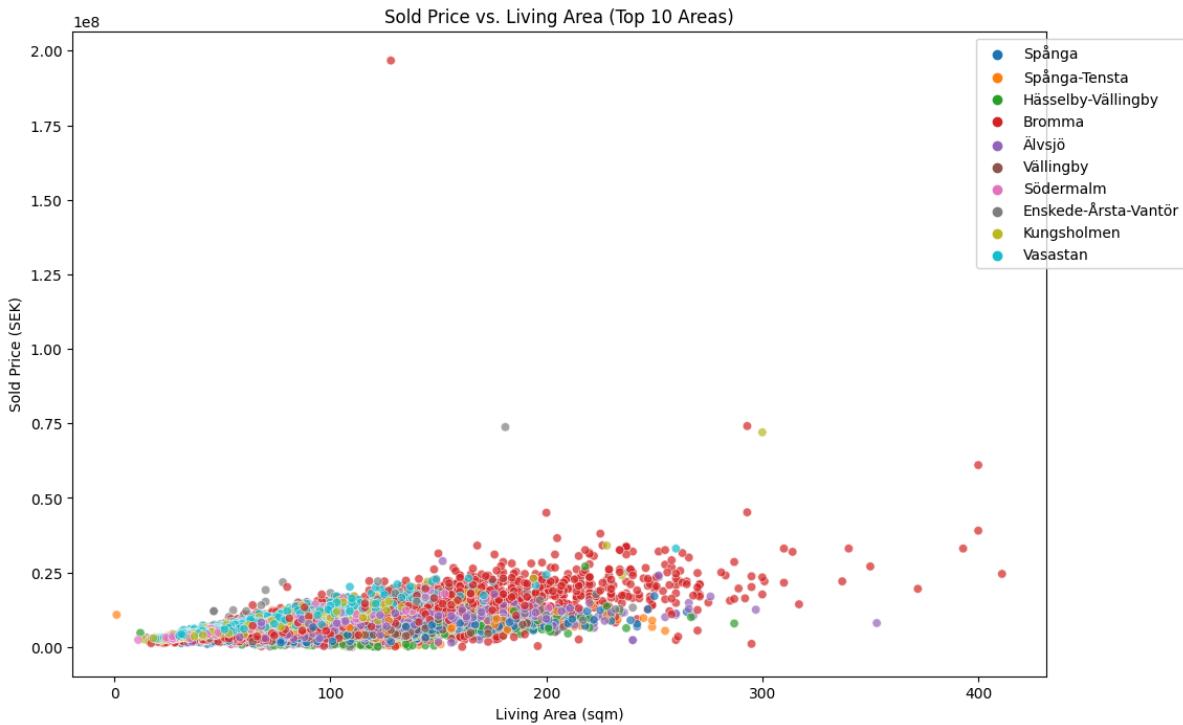
In [35]: # Scatter plot: Interest Rate vs. Sold Price

```
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 matter)
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)

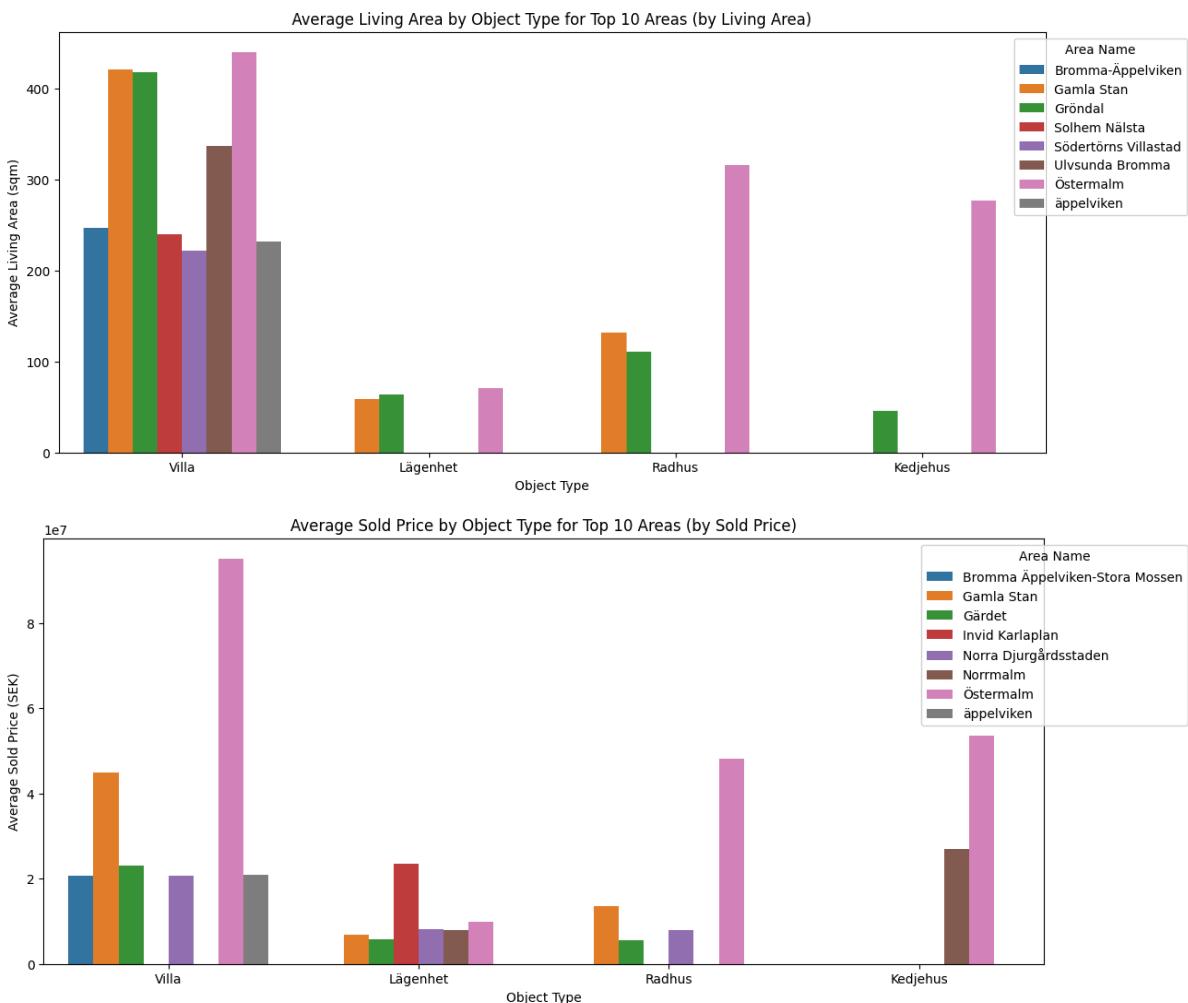
# Plot for Average Living Area
plt.figure(figsize=(14, 6))
sns.barplot(
    data=filtered_living_area,
    x='Object Type',
```

```

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-Month
10447	2000000	Torpstugegränd 32	90.0	Spånga	2013-02-28	2013	2	1.00	2013-2
10448	4100000	Lisebergsvägen 109A	136.0	Älvsjö Liseberg	2013-02-28	2013	2	1.00	2013-2
10449	3250000	Ekshäradsgatan 228	87.0	Farsta	2013-02-28	2013	2	1.00	2013-2
10450	2800000	Gillsätragränd 88	126.0	Sätra	2013-02-28	2013	2	1.00	2013-2
10451	4460000	Älvkvarnsvägen 76	150.0	Spånga-Tensta	2013-02-28	2013	2	1.00	2013-2
...
20211	8000000	Doktor Abrahams väg 46	85.0	Bromma Kyrka	2024-10-01	2024	10	3.25	2024-10
20212	9200000	Bollstavägen 2	101.0	Stureby	2024-10-01	2024	10	3.25	2024-10
20213	7700000	Granskogsvägen 31	184.0	Hässelby Norra Villastad	2024-10-01	2024	10	3.25	2024-10
20214	9925000	Mälarhöjdsvägen 69	95.0	Hägersten-Liljeholmen	2024-10-01	2024	10	3.25	2024-10
20215	15250000	Tussmötevägen 223	140.0	Enskede-Årsta-Vantör	2024-10-01	2024	10	3.25	2024-10

30523 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 [43]: # 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 = 0.2, random_st
```

2. Training the Model

```
In [45]: # 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=3)

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

4. Model Evaluation

```
In [47]: # 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)

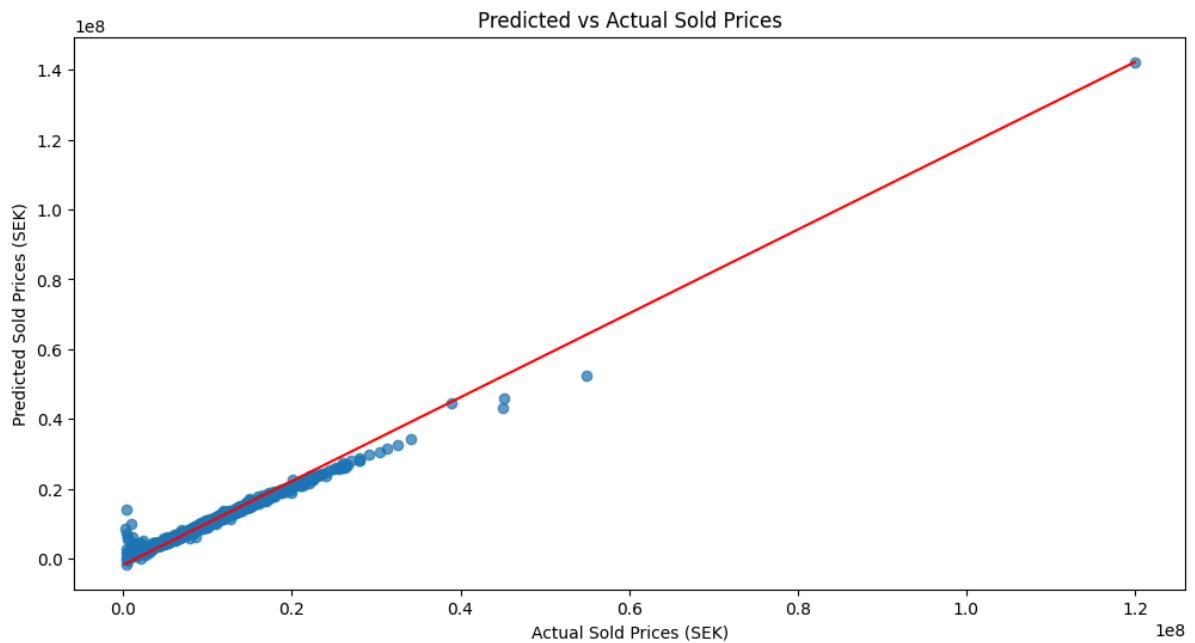
# Calculate Root Mean Squared Error (RMSE)
rmse = np.sqrt(mse)

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

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

Root Mean Squared Error: 501764.27 SEK
R-squared: 0.9861467444091133

```
In [48]: # 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 [49]: # Save the model to a file
import joblib
joblib.dump(lin_reg_of_poly, 'stockholm_property_price_model.pkl')
```

```
Out[49]: ['stockholm_property_price_model.pkl']
```