

Navigating the Hawkish Fed:

Data Driven Insights on
Trends, Challenges, &
Housing Valuations

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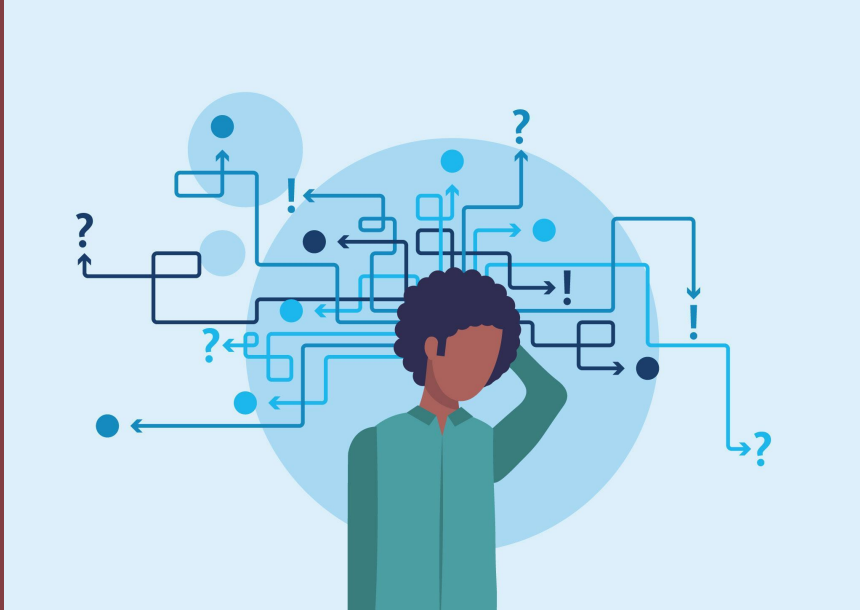
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Problem Statement



By analyzing housing trends across US regions - specifically, Charlotte, New York, and Miami through **descriptive, comparative, and predictive analytics**, this project aims to **uncover actionable insights that can guide housing policy, investment strategies, and affordability initiatives**. The integration of data pipelines and advanced analytics will streamline the process of extracting meaningful patterns from datasets.

```
import requests
import json
import pandas as pd
import time
from keys import x_rapidapi_key, x_rapidapi_host
```

API call

```
def fetch_data(query):
    url = "https://zillow56.p.rapidapi.com/search"

    # query header parameters
    headers = {
        "x-rapidapi-key": x_rapidapi_key,
        "x-rapidapi-host": x_rapidapi_host
    }

    response = requests.get(url, headers=headers, params=query).json()
    time.sleep(1)

    return pd.json_normalize(response['results'])
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.neighbors import NearestNeighbors
from matplotlib.colors import Normalize
from matplotlib.cm import ScalarMappable

# Load data
df = pd.read_csv('USRealEstateTrends.csv')
```



```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
df = pd.read_csv('city_market_tracker.tsv000', delimiter='\t')
```

```
import pandas as pd
import numpy as np
```

```
df = pd.read_excel('historicalweeklydata.xlsx', header=6)
```

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
df = pd.read_csv('API.csv', index_col=False)
```


OUR APPROACH TO DESCRIPTIVE Analysis

"What is the average price of a home in a given year? What are the highs and lows?"

- Identify trends over time
- Single Family Homes
- Calculate summary statistics (mean, median, mode) to understand the central tendency of the data.



Focus on single family homes

```
family_df = cleaned_df[cleaned_df['property_type']== 'Single Family Residential']
```

```
family_df.isna().sum()
```

```
period_begin      0  
period_end        0  
city              0  
state            0  
property_type     0  
median_list_price 303490  
dtype: int64
```

```
family_df.dropna(inplace=True)
```

▼ Group the 15 & 30 yr FRM by year

```
20]: avg_mortgage = df.groupby(df['Week'].dt.year)[['30 yr FRM', '15yr FRM']].mean()
```

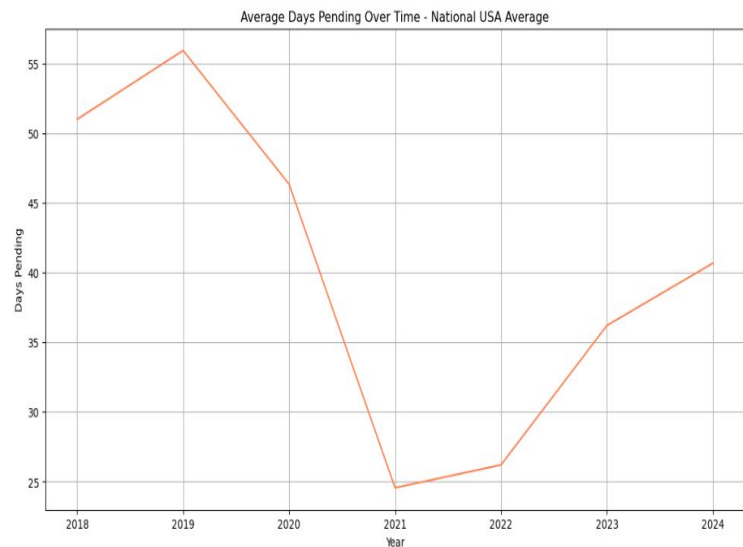
```
21]: # reset the grouped index  
avg_mortgage.reset_index(inplace=True)
```

```
22]: # rename headers  
avg_mortgage.rename(str.lower, axis='columns', inplace=True)  
avg_mortgage.rename(columns={'week': 'year', '15yr frm': '15 yr frm'}, inplace=True)
```

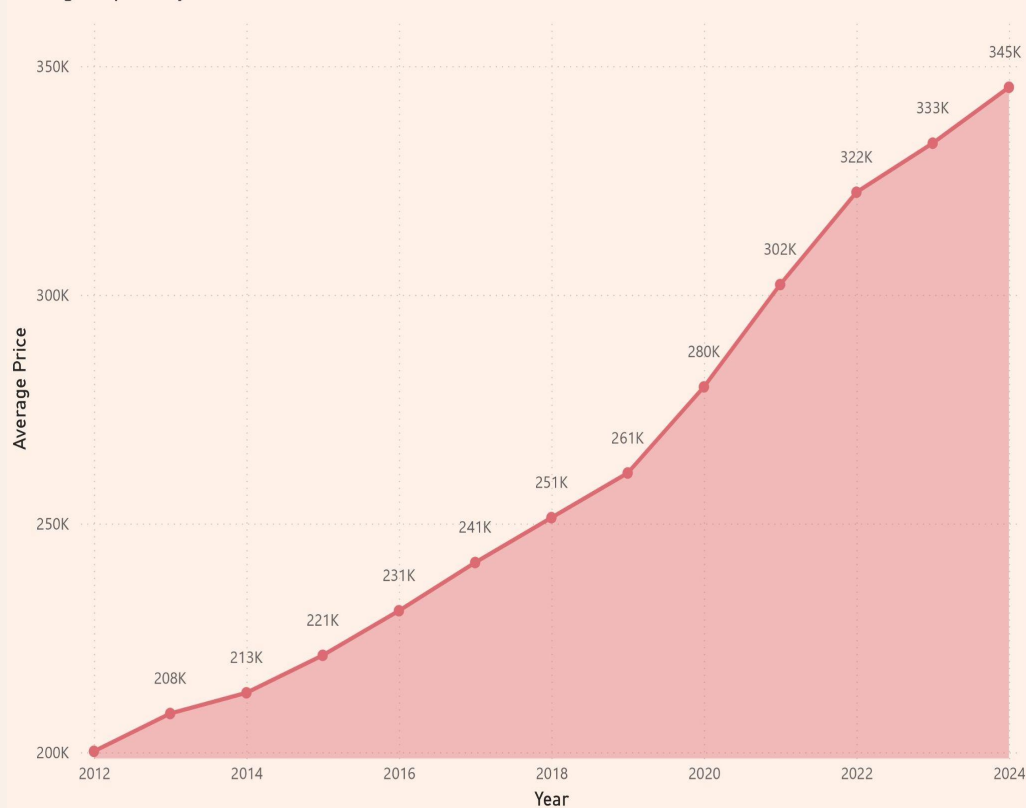
```
23]: # filter out anything before 2012  
avg_mortgage = avg_mortgage.loc[avg_mortgage['year']>=2012].reset_index(drop=True)
```



year	Average Housing Price	30 yr frm	15 yr frm
2012	200,189.15	3.66	2.93
2013	208,456.49	3.98	3.10
2014	212,985.50	4.17	3.29
2015	221,170.90	3.85	3.09
2016	230,922.72	3.65	2.93
2017	241,462.48	3.99	3.27
2018	251,267.83	4.54	4.00
2019	261,027.66	3.94	3.39
2020	279,855.12	3.11	2.60
2021	302,227.49	2.96	2.27
2022	322,399.63	5.34	4.58
2023	333,121.56	6.81	6.11
2024	345,347.68	6.72	5.96



Average of price by Year in USA



OUR APPROACH TO COMPARATIVE ANALYSIS

- Perform year-over-year comparisons to detect growth, decline, or stagnation in housing prices.

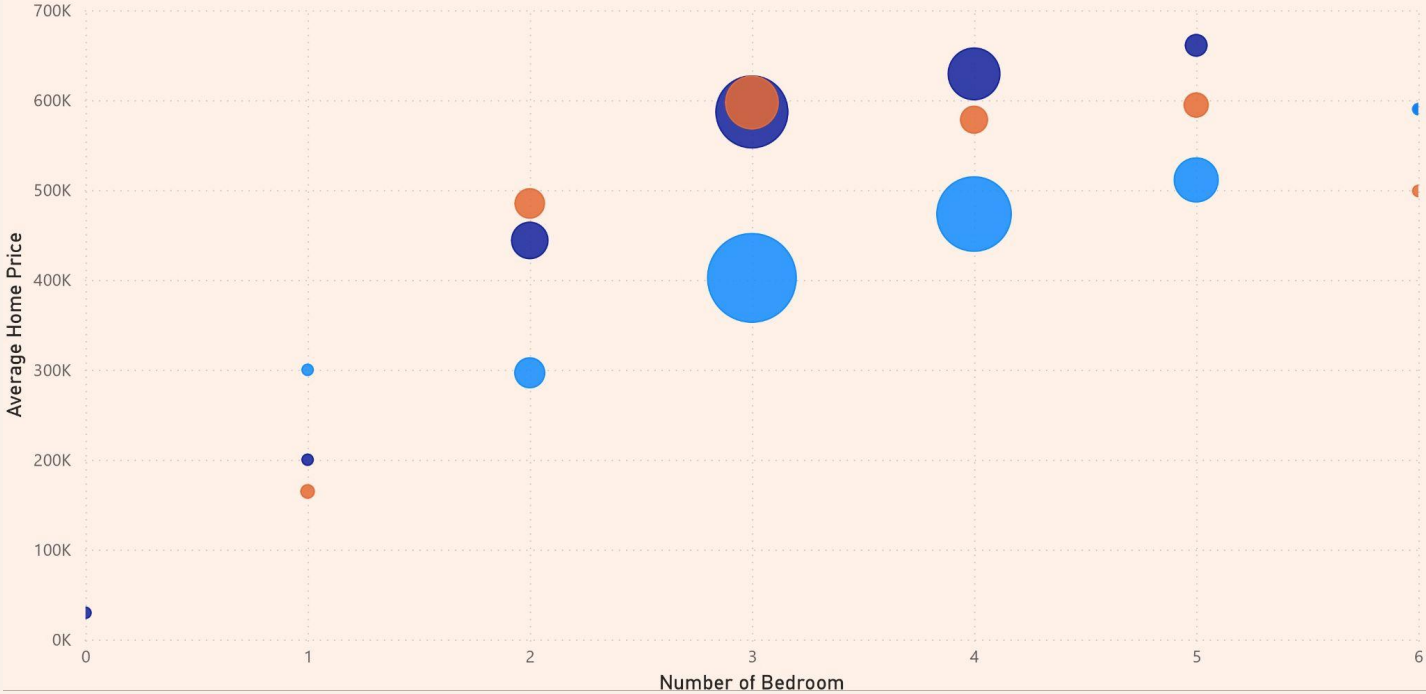
- "Which city saw the largest growth in prices last year?"

- Moderate and steady price growth, supported by relatively stable mortgage rates (~4.5% to 3.7%).



Average Price by Bedrooms

City ● Charlotte ● Miami ● New York



Average Price by State



CONCLUSION

- Many people and investors have to consider the economic downturn caused by the pandemic, thus paving the way for low interest rates. Ultimately, this would be directly correlated to the surge in home buying and rise in property values.
- Inflationary costs on goods which could be contributing to higher home values, especially in single family homes.
- Extend market research to include rental properties.
- Lower mortgage rates encouraged more buyers to enter the market, increasing demand and pushing prices upward. This trend was particularly visible in major cities where affordability was already strained.

