



HOME PRICE INDEX ANALYSIS

1. OBJECTIVE:

To find **key factors** which will **influence Home Prices** in **US** on a **national level** from time period **2003 till 2022**.

2. KEY FACTORS SELECTED:

1. **GDP Growth** - The **Gross Domestic Product growth** is a **strong** indicator of **economic growth** for a country. When **GDP** is **growing** it will **shoot up economy** with **rising income levels** and **job opportunities**. This will **boost public** to **buy houses** thereby **increasing demand** for houses which leads to **higher home prices** due to **demand**.

2. **Unemployment Rate** - When there is **Unemployment Rate** then **people** will **not** be having any **reliable source of income** to **buy** houses, which will **reduce** the **demand** for **houses** thereby **reducing home prices**.

3. **Inflation Rate** - As the name suggests **Inflation** will lead to **higher home prices** due to **raw materials cost** for **construction** and other **miscellaneous costs**.

4. **Mortgage Interest Rates** - **Mortgage Interest Rates** are crucial factor to **provide loans**, when the **interest rates** are **less** then it will be very much **easier** for the **public** to **borrow more money** to **buy homes** which will **positively influence home prices**.

5. **Housing Starts** - **Supply of new homes** is Housing Starts, an **increase** in **Housing Starts** will **stabilize or lower home prices**, likewise **decrease** in housing starts will **increase prices due to high demand**.

6. **Existing Home Sales** - Existing House Sales is very **good** indicator to get to know the **Real Estate current demand** based on which **prices** are framed.

7. **Consumer Confidence Index** - As the name suggests Consumer Confidence Index **influences** in **capturing** the **confidence levels and sentiments of people** in **deciding** to buy houses.

8. **Housing Credit Availability Index** - With **more availability** of Housing Credit to people, they can **easily lend loans** to purchase properties.

9. **Interest Rate Policy** - **Lower Interest Rates** from Banks will **stimulate** more **purchases** eventually increasing home prices.

These data are collected from US government websites, placing it here for your reference:

I. Gross Domestic Product (GDP) Growth:

Source Website: <https://fred.stlouisfed.org/series/GDP>

II. Unemployment Rate:

Source Website : <https://data.bls.gov/pdq/SurveyOutputServlet>

III. Inflation Rate:

Source Website : <https://data.oecd.org/price/inflation-cpi.htm>

IV. Mortgage Interest Rates:

Source Website : <https://fred.stlouisfed.org/series/MORTGAGE30US>

V. Housing Starts:

Source Website : <https://fred.stlouisfed.org/series/HOUSTNSA>

VI. Existing Home Sales:

Source Website: <https://fred.stlouisfed.org/series/EXHOSLUSM495S>

VII. Consumer Confidence:

Source Website:

<https://data.oecd.org/leadind/consumer-confidence-index-cci.htm>

VIII. Housing Credit Availability Index:

Source Website:

<https://datacatalog.urban.org/dataset/housing-credit-availability-index-hcai>

IX. Interest Rate Policy:

Source Website: <https://fred.stlouisfed.org/series/FEDFUNDS>

3. DATA PREPARATION AND PROCESSING:

Linear Interpolation:

In the **data attributes extracted**, some were in **Monthly format** and some in **Quarterly format**.

As we will be performing **Monthly level analysis** on **Home Price index**, we needed to **perform Linear Interpolation** to **convert** from **Quarterly** to **Monthly**.

```
In [135]: data_monthly['VALUE'].interpolate(method='linear', inplace=True)
```

Noise Detection Test:

Noise Detection Tests are being **performed** on all the features to **check** for the **presence of White Noise**.

Ljung–Box test which is primarily **performed** for **detecting White Noise** is being taken and **results** being arrived at:

Column: GDP GROWTH
Result: Not White Noise

Column: UNEMPLOYMENT RATE
Result: Not White Noise

Column: INFLATION RATE
Result: Not White Noise

Column: MORTGAGE INTEREST RATES
Result: Not White Noise

Column: HOUSING STARTS
Result: Not White Noise

Column: EXISTING HOME SALES
Result: Not White Noise

Column: CONSUMER CONFIDENCE INDEX
Result: Not White Noise

Column: HOUSING CREDIT AVAILABILITY INDEX
Result: Not White Noise

Column: INTEREST RATE POLICY RATE
Result: Not White Noise

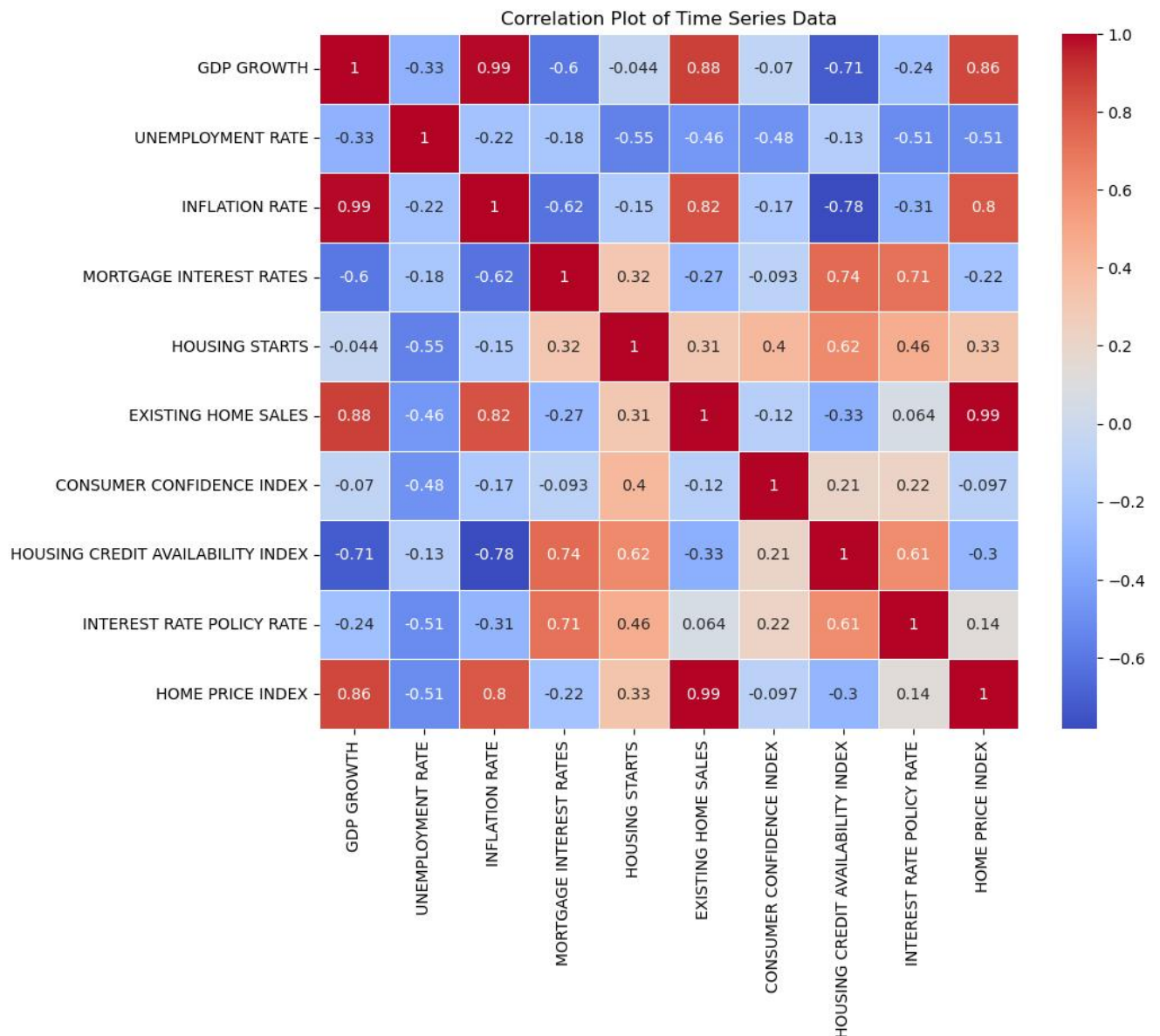
Column: HOME PRICE INDEX
Result: Not White Noise

As per the results we could see **none** of the **columns** contains **White Noise**.

4. DATA VISUALIZATION AND ANALYSIS:

Correlation Plot:

Correlation plot is created to analyze the **correlation of key factors** with **Home Price Index**.



Based on plot we could see **GDP GROWTH, INFLATION RATE, EXISTING HOME SALES** are **positively** impacting **HOME PRICE INDEX**, so basically

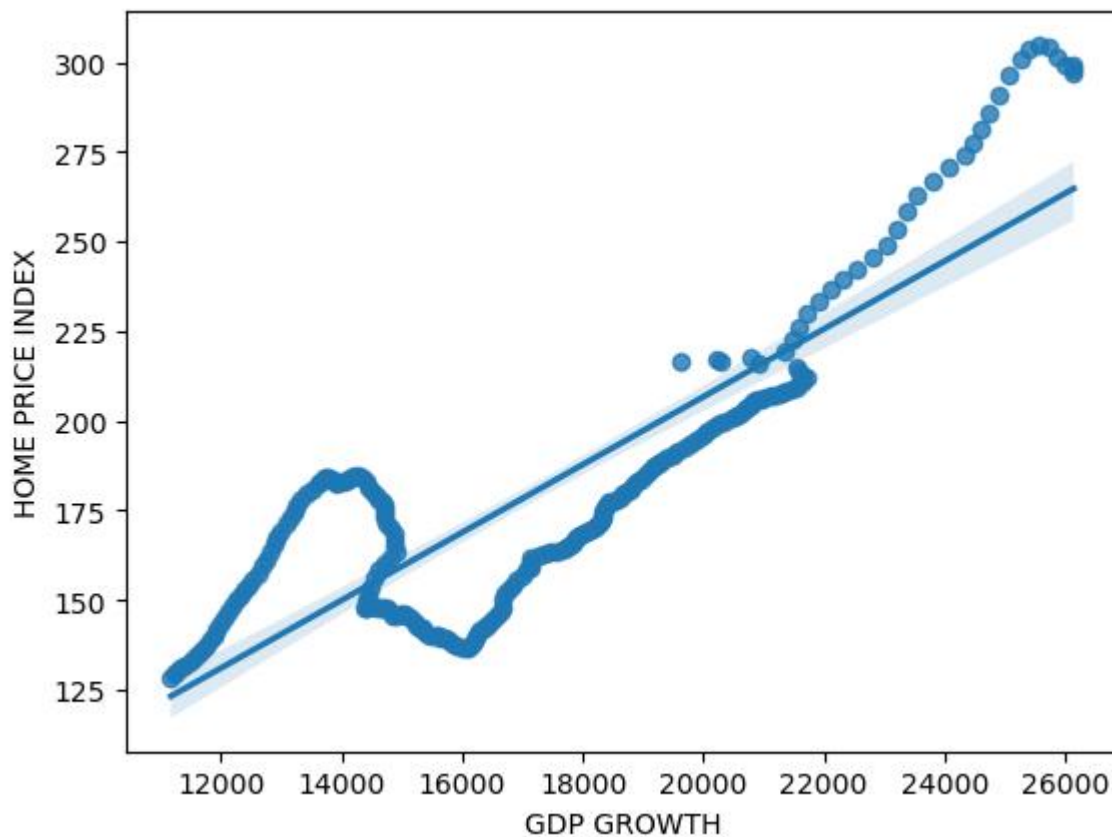
when these attributes **increase** then **HOME PRICE INDEX** also **increases**.

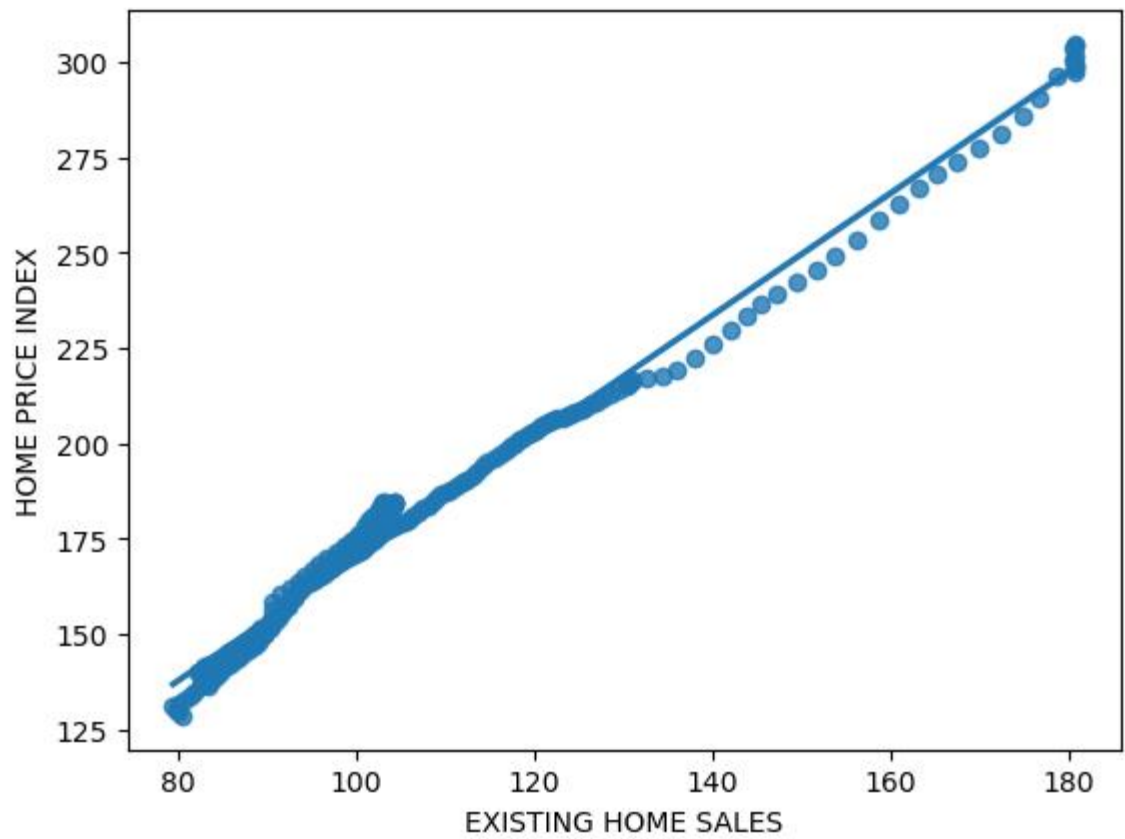
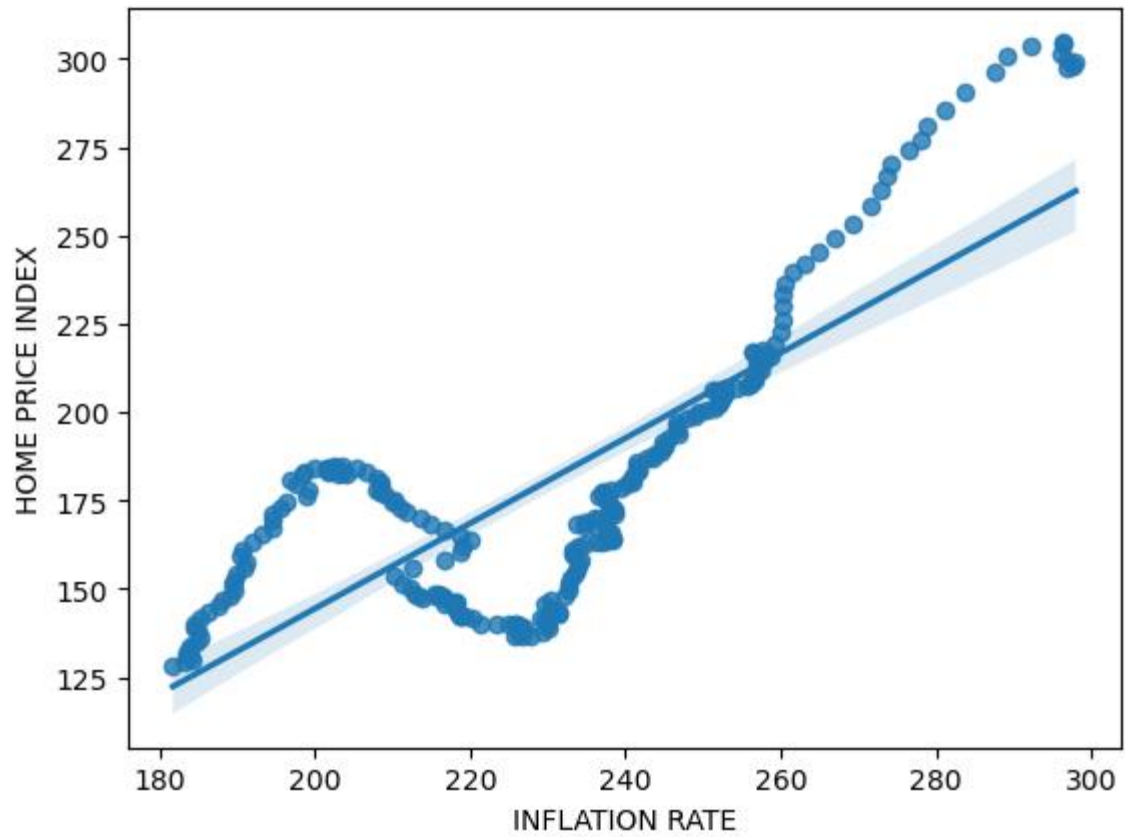
On the other hand **UNEMPLOYMENT RATE, MORTGAGE INTEREST RATES, CONSUMER CONFIDENCE INDEX AND HOUSING CREDIT AVAILABILITY INDEX** are **negatively** impacting **HOME PRICE INDEX**,so when these attributes **increases** then **HOME PRICE INDEX** **decreases**

HOUSING STARTS and **INTEREST RATE POLICY RATE** **doesn't** influence **HOME PRICE INDEX**

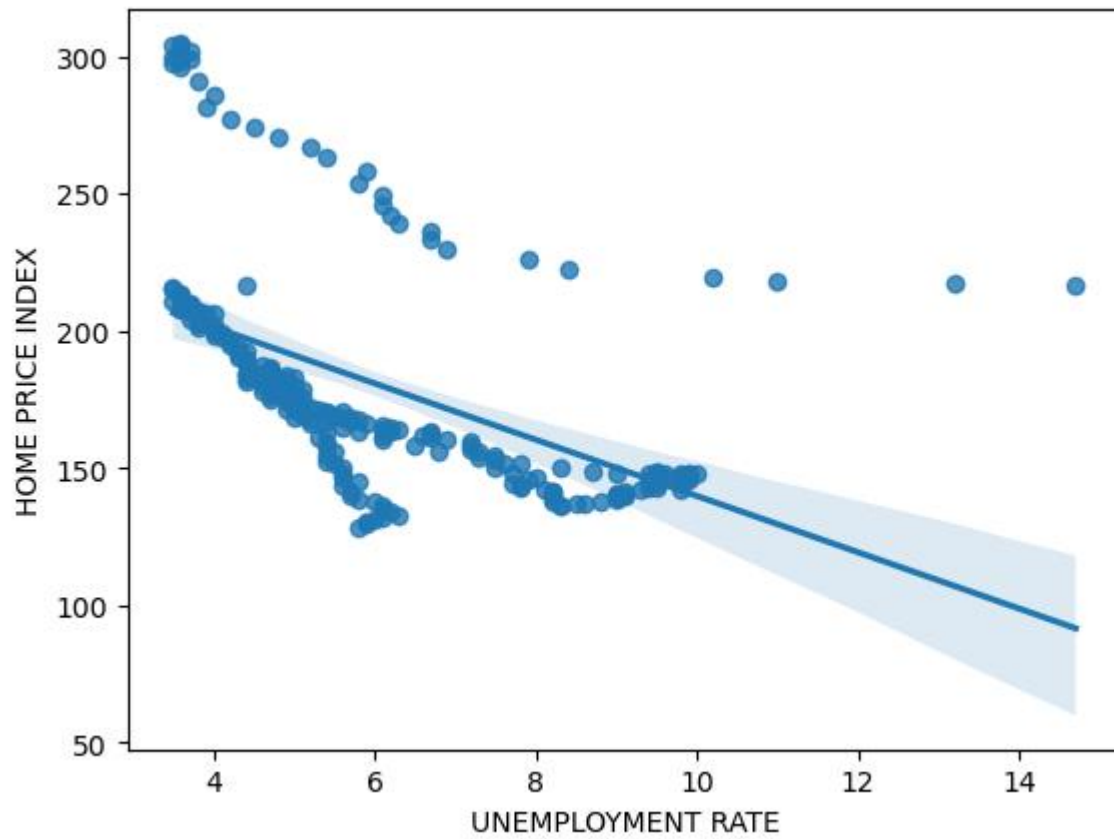
Detailed plots:

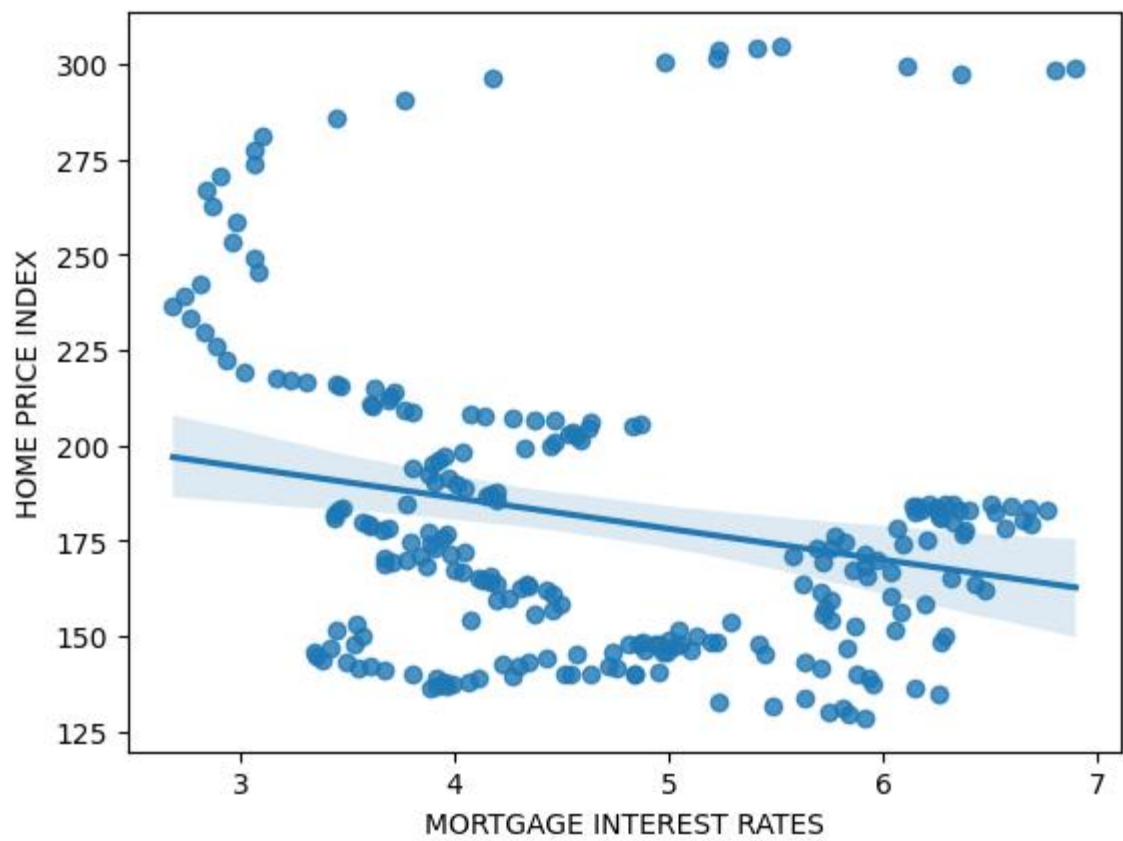
We could also see in detail manner the **Positive** and **Negative** influence.

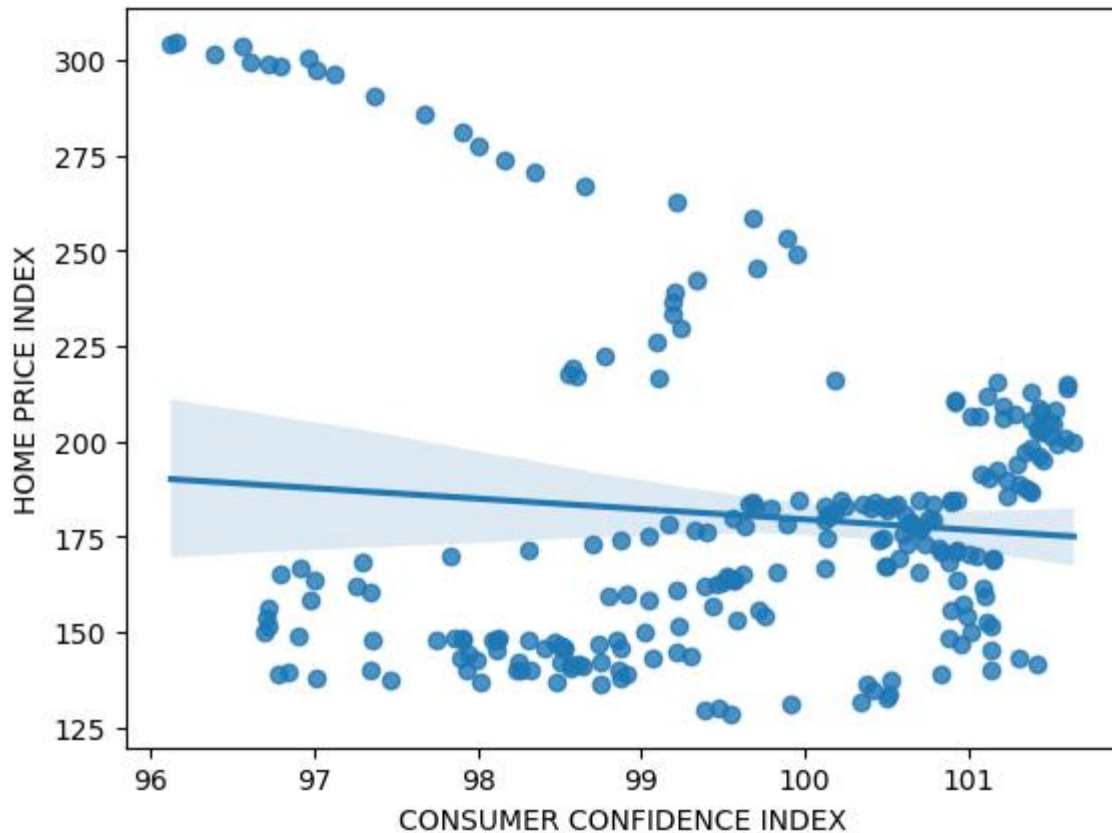




As seen from Correlation plot these attributes are Positively influencing Home Price Index.







We could see these attributes are Negatively impacting Home Price Index as seen in Correlation Plot.

We could see that **Consumer Confidence Index** is **negatively** influencing **Home Price Index** which is **not the case** in **real time** scenario, so lets try to **analyze** this **deeper**.

Past Lags Analyse:

As far as **time series data** is being concerned not all attributes or features in the dataset might always **positively influence** with **present** data alone.

It can also **influence** with **past** data as well.

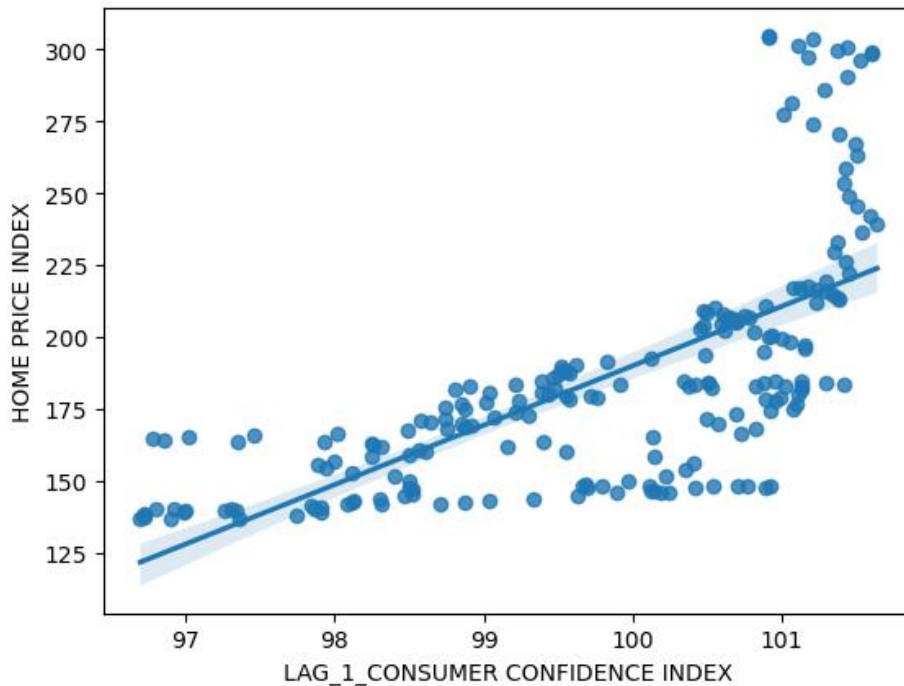
So lets try to **check the correlations for 40 lags** and see in **which lag** we are **observing a positive influence of consumer confidence index** upon home price index.

| | Lag | Correlation |
|----|------|-------------|
| 0 | 1.0 | -0.067422 |
| 1 | 2.0 | -0.035531 |
| 2 | 3.0 | -0.002772 |
| 3 | 4.0 | 0.031346 |
| 4 | 5.0 | 0.068183 |
| 5 | 6.0 | 0.108740 |
| 6 | 7.0 | 0.150632 |
| 7 | 8.0 | 0.189979 |
| 8 | 9.0 | 0.226228 |
| 9 | 10.0 | 0.261245 |
| 10 | 11.0 | 0.293747 |
| 11 | 12.0 | 0.322625 |
| 12 | 13.0 | 0.348632 |
| 13 | 14.0 | 0.373426 |
| 14 | 15.0 | 0.396600 |
| 15 | 16.0 | 0.418010 |
| 16 | 17.0 | 0.436532 |
| 17 | 18.0 | 0.449555 |
| 18 | 19.0 | 0.458220 |
| 19 | 20.0 | 0.465017 |
| 20 | 21.0 | 0.471477 |
| 21 | 22.0 | 0.480605 |
| 22 | 23.0 | 0.493940 |
| 23 | 24.0 | 0.509632 |
| 24 | 25.0 | 0.526341 |
| 25 | 26.0 | 0.543684 |
| 26 | 27.0 | 0.560668 |
| 27 | 28.0 | 0.578696 |
| 28 | 29.0 | 0.599189 |
| 29 | 30.0 | 0.621109 |

| | | |
|-----------|-------------|-----------------|
| 30 | 31.0 | 0.642639 |
| 31 | 32.0 | 0.663168 |
| 32 | 33.0 | 0.677804 |
| 33 | 34.0 | 0.681514 |
| 34 | 35.0 | 0.676793 |
| 35 | 36.0 | 0.668481 |
| 36 | 37.0 | 0.659353 |
| 37 | 38.0 | 0.651027 |
| 38 | 39.0 | 0.644027 |
| 39 | 40.0 | 0.638061 |
| 40 | 41.0 | 0.631717 |
| 41 | 42.0 | 0.622770 |
| 42 | 43.0 | 0.611545 |
| 43 | 44.0 | 0.598808 |
| 44 | 45.0 | 0.585692 |
| 45 | 46.0 | 0.572773 |
| 46 | 47.0 | 0.560671 |
| 47 | 48.0 | 0.548357 |
| 48 | 49.0 | 0.534073 |
| 49 | 50.0 | 0.518300 |

At **34th lag level** we could see **Positive influence** with **Home Price Index**.

Lets try to visualize this



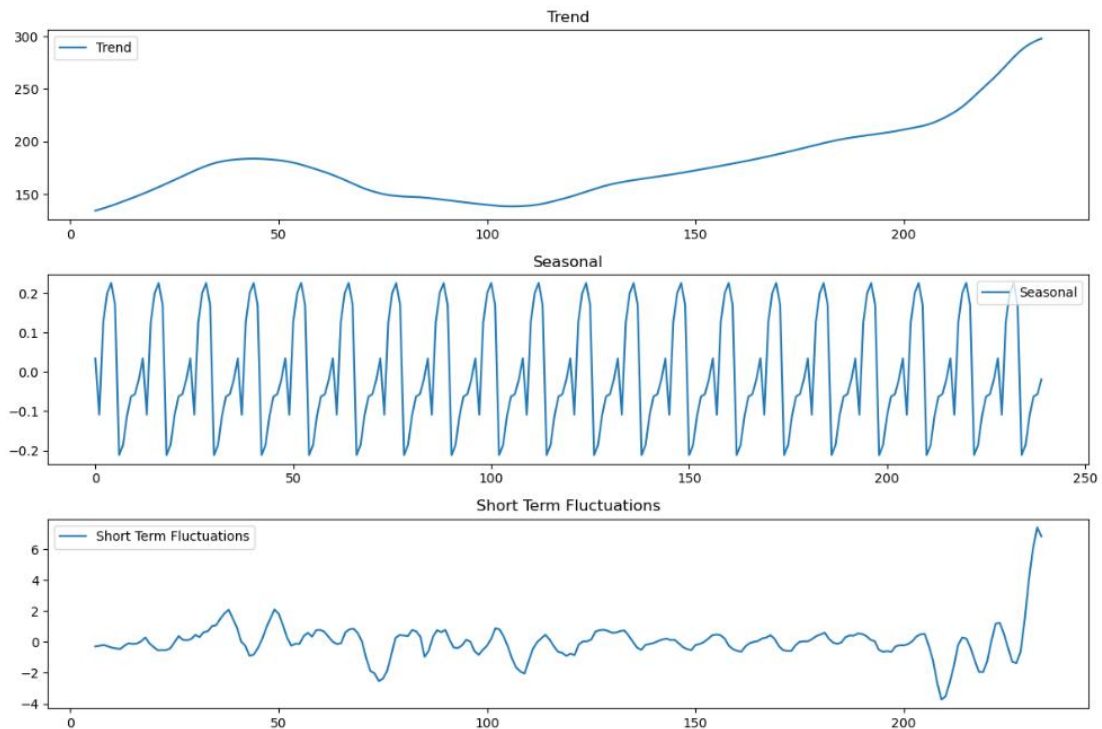
We could **consumer confidence index** at **34 time lag** is **positively** influencing **home price index**.

This can mean that **Consumer Confidence surveys** may have been conducted **34 month interval** prior to **House Sales**, so a **person** would have **appeared** to a **Consumer Confidence survey 34 months before** **purchasing a House or Home**.

As **Consumer Confidence Index** is **helping** only with a **34 time lag**, it will be **difficult** to use this **data** for **forecasting** as we cannot be able to **determine** in which **time level** the **consumer confidence index** will **influence future home price index**.

So its **better** to **not consider Consumer Confidence Index** as a feature for our dataset.

Trend, Seasonality and Decomposition:



TREND:

1. We could see a **slight increase from 2004** and a **flatness from 2005 to 2008**.
2. From **2009 to 2013** there is **flatness in lower level** and **rise from 2014**.
3. **After 2014** we could see **significant rise continuously**.

SEASONALITY:

The data **doesn't show** any **changes in behaviour or trend** on a **seasonal level** and its pretty much **constant**.

SHORT TERM FLUCTUATIONS:

We could see **many small high low dip fluctuations** and **2022** we could see **significant increase level**.

5. MODEL BUILDING:

We will be **building FB Prophet model** where it will be **trained** with data from **January 2003 to December 2021** and **forecasted** for **January 2022 to December 2022 data**.

Here **Home Price Index** will be **forecasted** by the **model**.

Initially **train** and **test data** is being **prepared**.

```
train_data = df_model[df_model['ds'] <= '2021-12-01']  
test_data = df_model[(df_model['ds'] >= '2022-01-01') & (df_model['ds'] <= '2022-12-01')]
```

Model is built

```
model = Prophet(growth='linear',)  
for i in train_data.drop(['y', 'ds'], 1):  
    model.add_regressor(i, mode='multiplicative')  
  
model.fit(train_data)  
  
future = test_data  
  
forecast = model.predict(future)
```


Model is evaluated with Actual Home Price Index.



```
metrics.r2_score(actual,forecast['yhat'])
```

0.6719489563230197

```
metrics.mean_absolute_error(actual,forecast['yhat'])
```

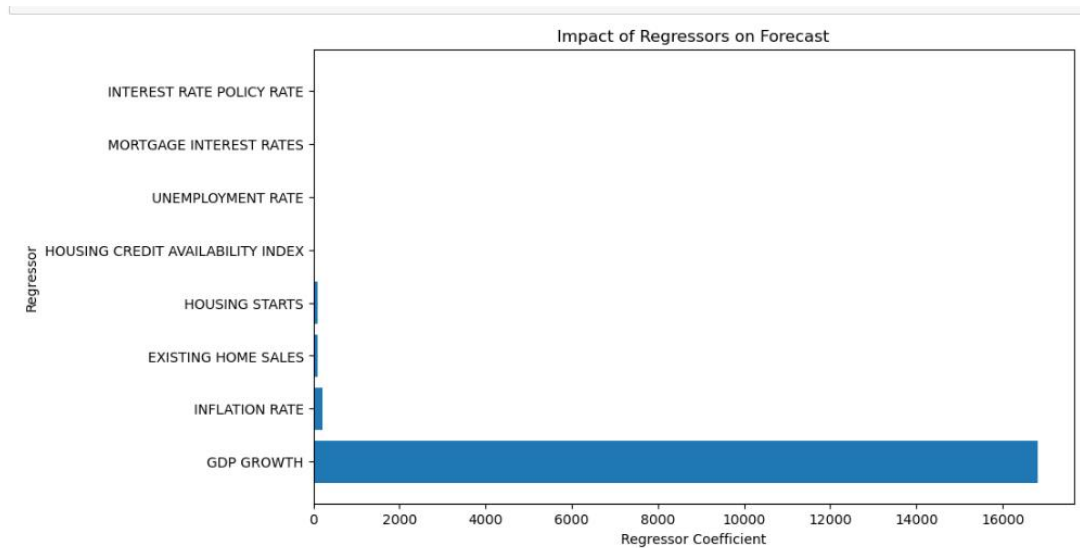
2.658352165569402

We could see only **67% accuracy level** is achieved, this is **due to low records data**, as **totally including train and test data** we have only **240 records**, **splitting** it into **train and test** will be **228** for training and **12** for testing.

Due to low records the model is **Underfitting** if **more data** is collected it will **perform well**.

6. FEATURE IMPORTANCE FOR THE MODEL:

Feature Importance plot is **created** based on **coefficients** from the **model**.



We could see that **GDP GROWTH, INFLATION RATE, EXISTING HOME SALES and HOUSING STARTS** are influencing the model pretty much.

7. FINAL INFERENCE:

Finally we can conclude that **GDP GROWTH, INFLATION RATE and EXISTING HOME SALES** are key factors which influence HOME PRICE INDEX very much.