

# Data Science Lab 5

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## TIME SERIES ANALYSIS

### Traffic Time Series

#### DATA PROFILING

##### *Data Dimensionality and Granularity*

Figure 1: Traffic Time Series at the most granular detail

Figure 2: Traffic Time Series at the second chosen granularity

Figure 3: Traffic Time Series at the third chosen granularity

##### *Data Distribution*

Figure 4: Boxplot(s) for Traffic Time Series

Figure 5: Histogram(s) for Traffic Time Series

Figure 6: Autocorrelation lag-plots for original Traffic Time Series

Figure 7: Autocorrelation correlogram for original Traffic Time Series

### ***Data Stationarity***

Figure 8: Components study for Traffic Time Series

Figure 9: Stationarity study for Traffic Time Series

## **DATA TRANSFORMATION**

### ***Aggregation***

Figure 10: Forecasting plots after different aggregations on Traffic Time Series

Figure 11: Forecasting results after different aggregations on Traffic Time Series

### ***Differentiation***

Figure 12: Forecasting plots after first and second differentiation of Traffic Time Series

Figure 13: Forecasting plots after first and second differentiation of Traffic Time Series

### ***Smoothing***

Figure 14: Forecasting plots after different smoothing parameterisations on Traffic Time Series

Figure 15: Forecasting results after different smoothing parameterisations on Traffic Time Series

### ***Scaling***

Figure 16: Forecasting plots after different scaling parameterisations on Traffic Time Series

Figure 17: Forecasting results after different scaling parameterisations on Traffic Time Series

# Inflation Rate Time Series

## DATA PROFILING

### *Data Dimensionality and Granularity*



Figure 18: U.S. Inflation Rate dimensionality over time.

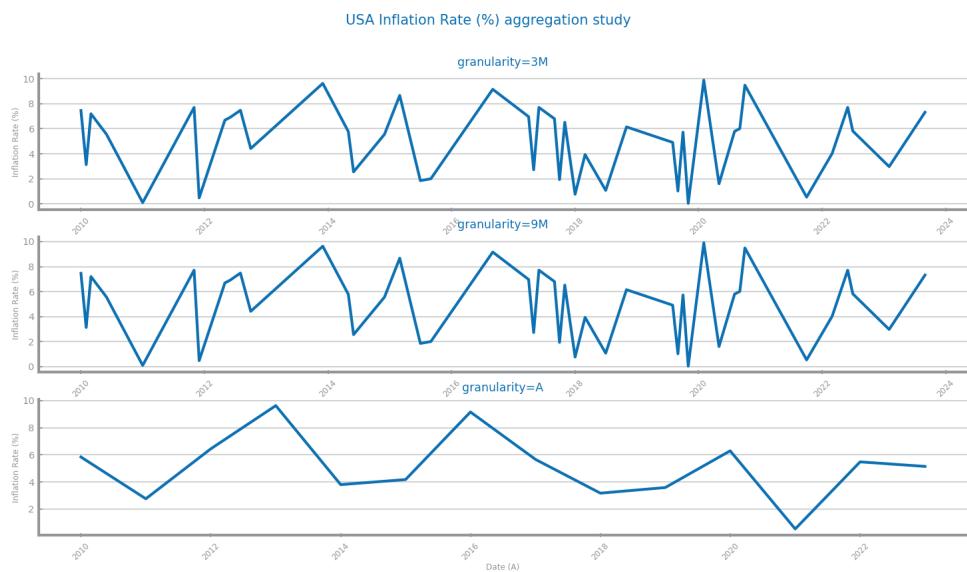


Figure 19: Inflation Rate Time Series at three different granularities

## Data Distribution

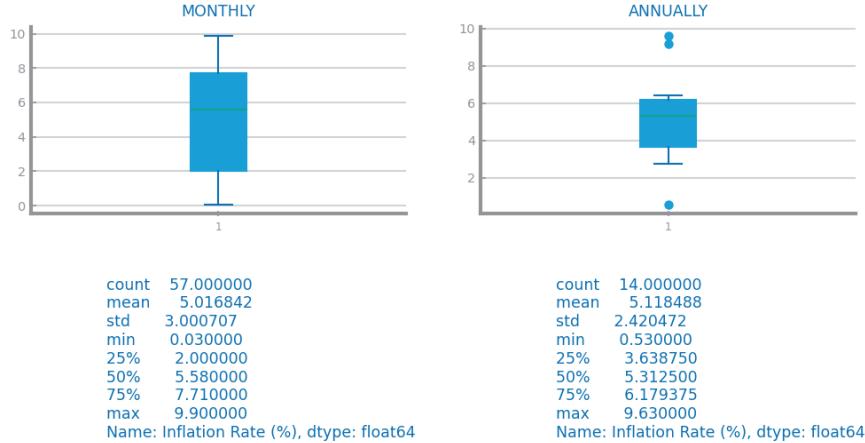


Figure 20: Boxplot(s) for Inflation Rate Time Series

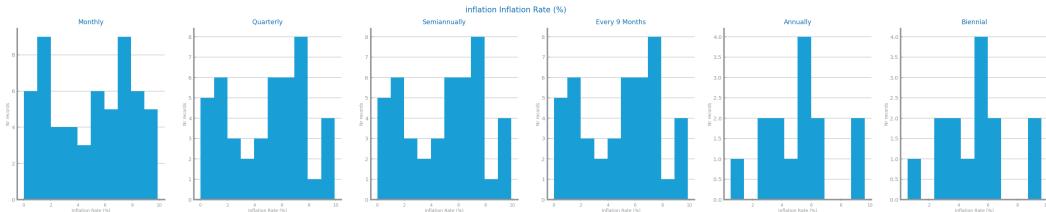


Figure 21: Histogram(s) for Inflation Rate Time Series

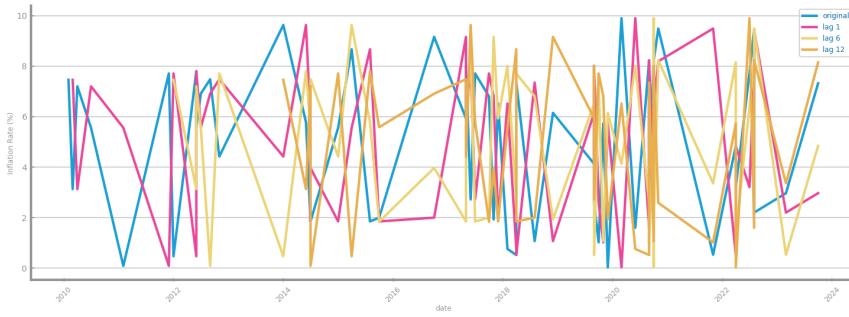


Figure 22: Autocorrelation lag-plots for original Inflation Rate Time Series

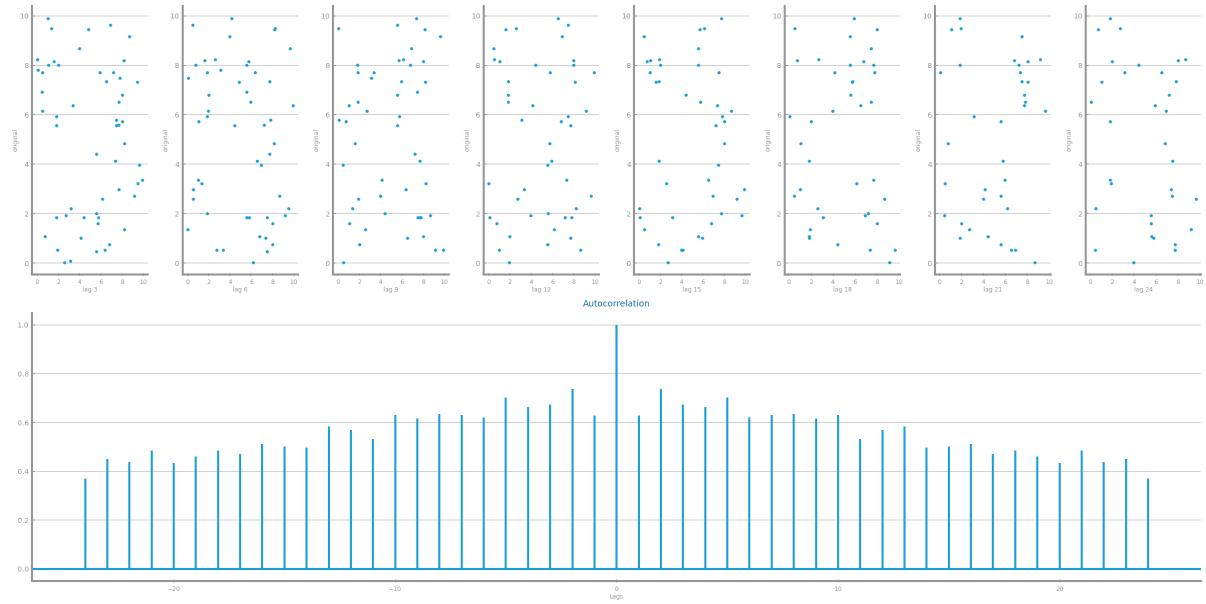


Figure 23: Autocorrelation correlogram for original Inflation Rate Time Series

## Data Stationarity

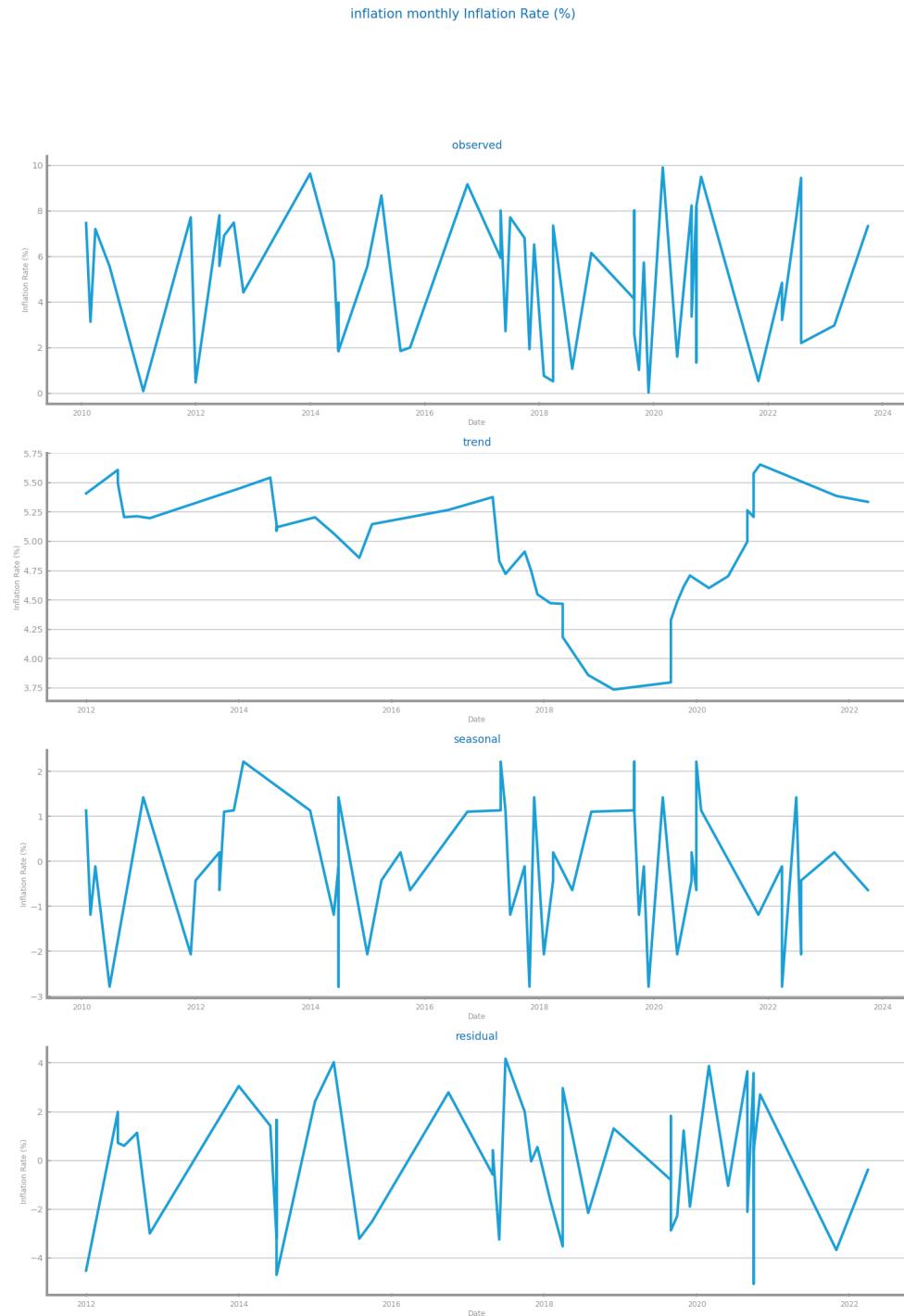


Figure 24: Components study for Inflation Rate Time Series

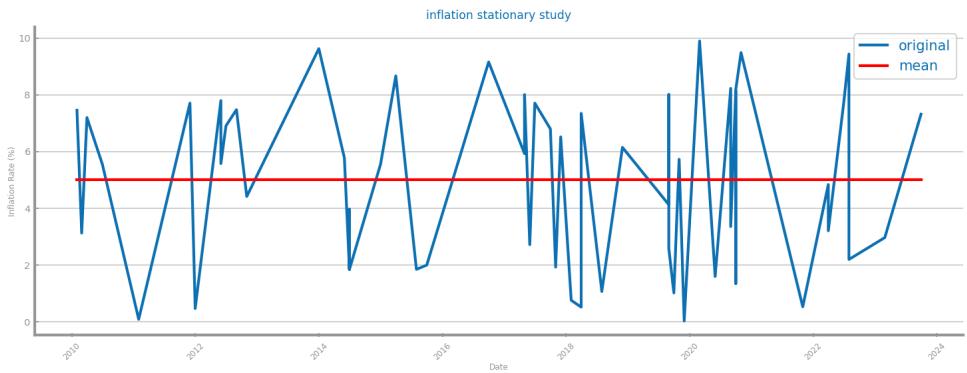


Figure 25: Stationarity study for Inflation Rate Time Series

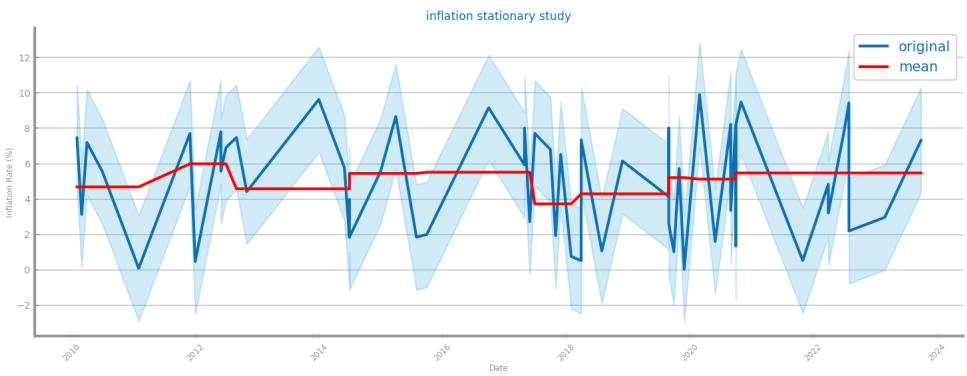


Figure 26: Stationarity study with bins for Inflation Rate Time Series

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ADF Statistic: -10.410
p-value: 0.000
Critical Values:
    1%: -3.553
    5%: -2.915
    10%: -2.595
The series is stationary

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Figure 27: Augmented Dickey-Fuller test results for Inflation Rate Time Series

# DATA TRANSFORMATION

## *Aggregation*

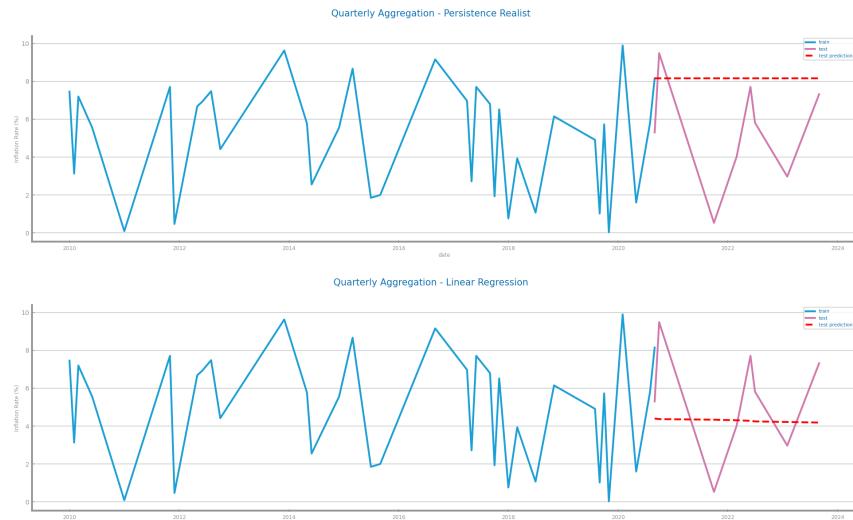


Figure 28: Forecasting predictions from Persistence Realist and Linear Regression on **Quarterly Inflation Aggregation**

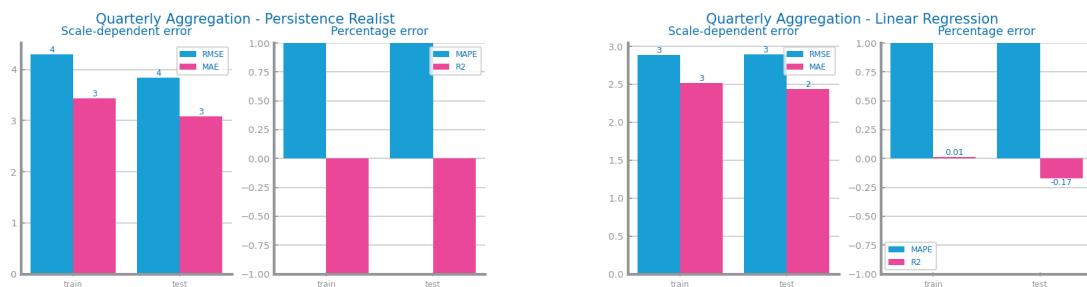


Figure 29: Performance of Persistence Realist and Linear Regression on **Quarterly Inflation Aggregation**

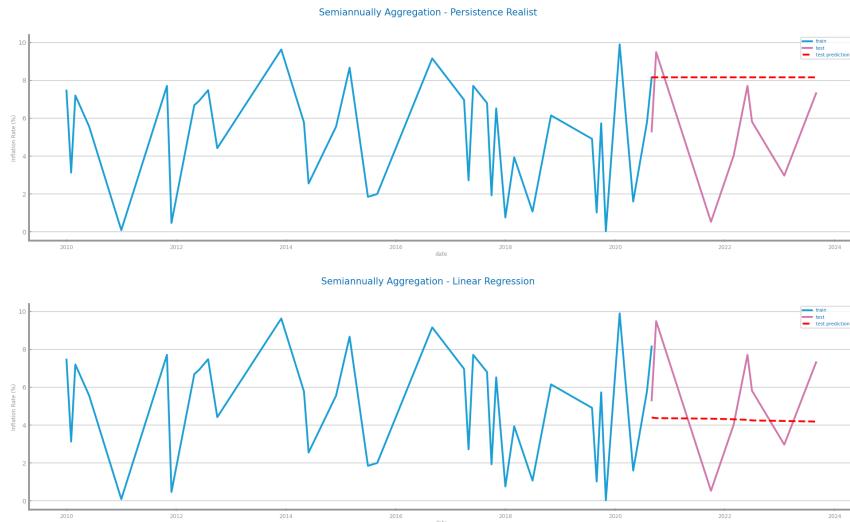


Figure 30: Forecasting predictions from Persistence Realist and Linear Regression on **Semi-Annual Inflation Aggregation**

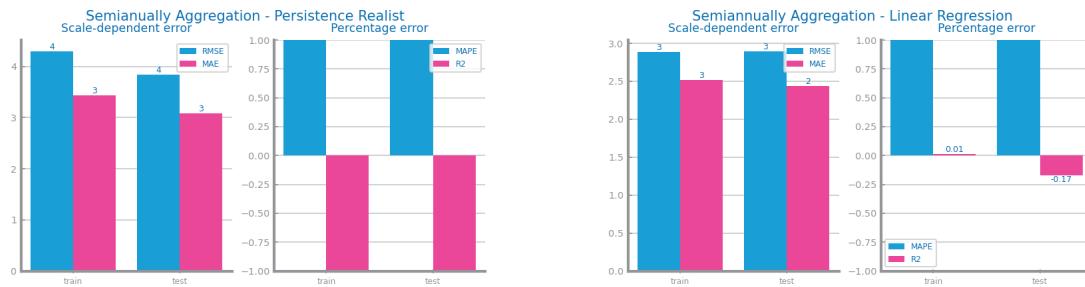


Figure 31: Performance of Persistence Realist and Linear Regression on **Semi-Annual Inflation Aggregation**

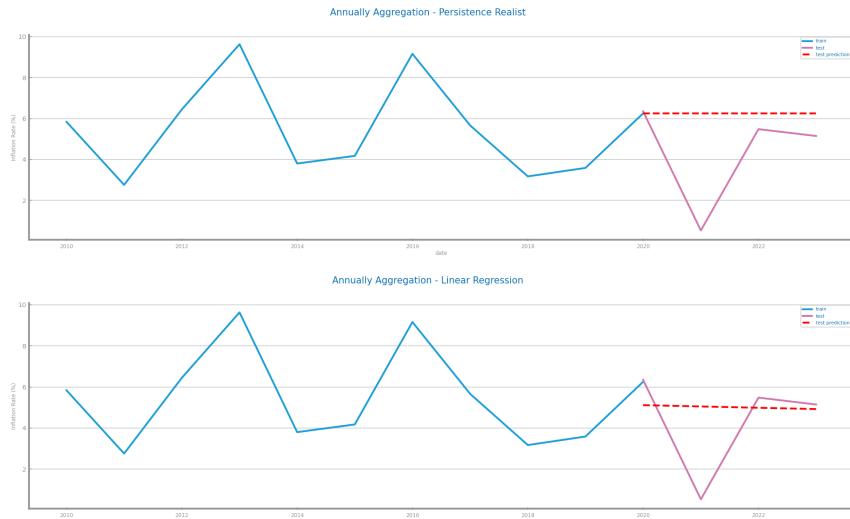


Figure 32: Forecasting predictions from Persistence Realist and Linear Regression on **Annual Inflation Aggregation**

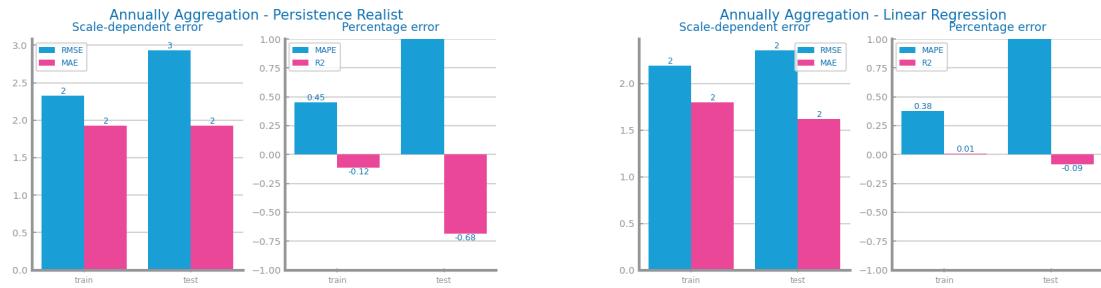


Figure 33: Performance of Persistence Realist and Linear Regression on **Annual Inflation Aggregation**

For all aggregation techniques, the Linear Regression model yields a lower MAE and RMSE than the Persistent Realist Model for the test set. Among the Linear Regression models, the Annual aggregation yields the lowest MAE for the test set ( $\tilde{1.6}$ ).

### Differentiation

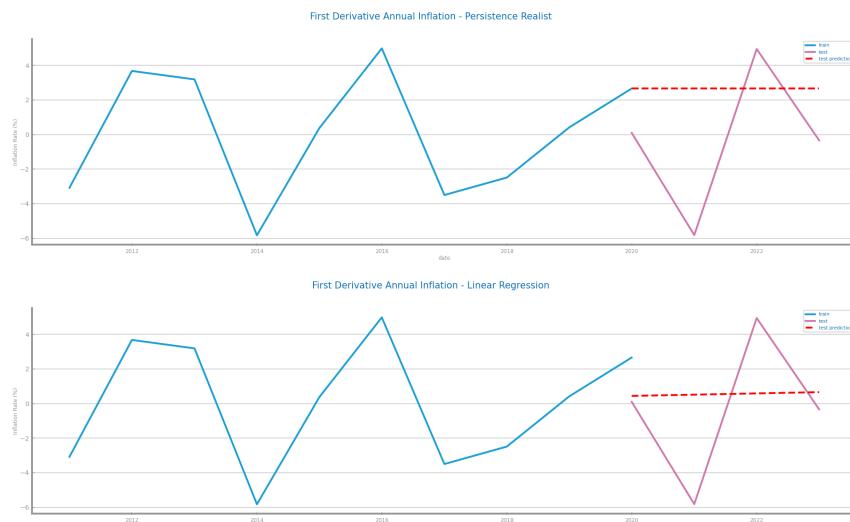


Figure 34: Forecasting predictions from Persistence Realist and Linear Regression on **Annual Inflation Differentiation (1st Order)**



Figure 35: Performance of Persistence Realist and Linear Regression on **Annual Inflation Differentiation (1st Order)**

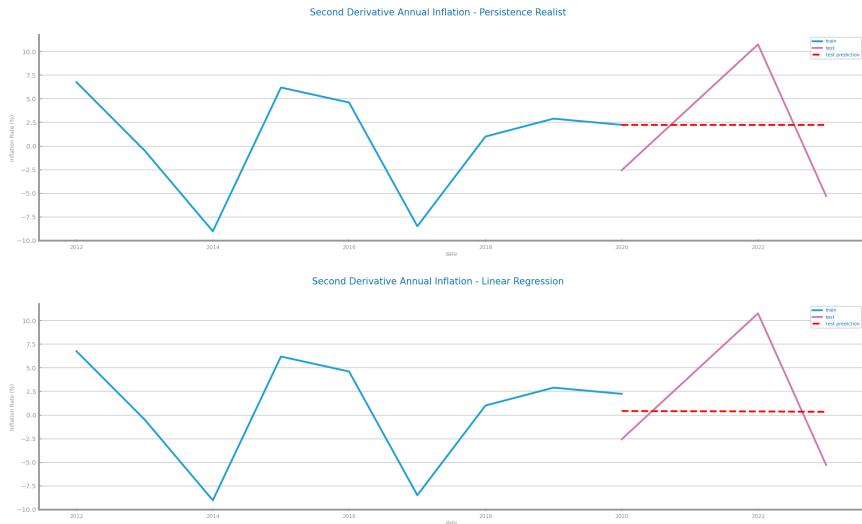


Figure 36: Forecasting predictions from Persistence Realist and Linear Regression on **Annual Inflation Differentiation (2nd Order)**

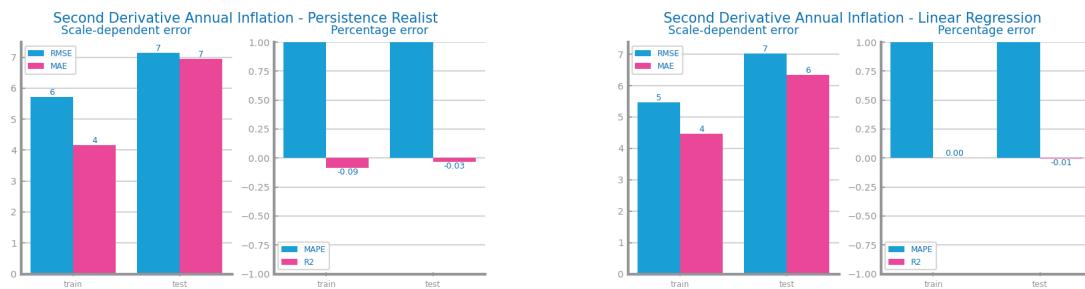


Figure 37: Performance of Persistence Realist and Linear Regression on **Annual Inflation Differentiation (2nd Order)**

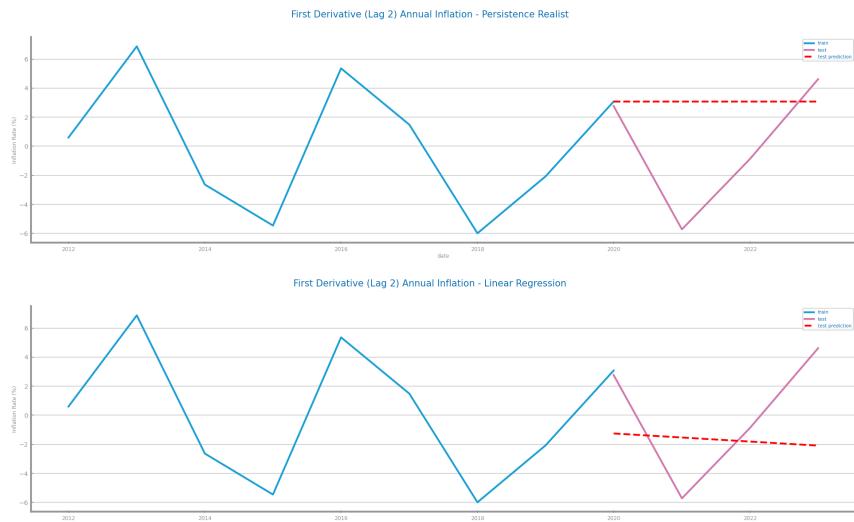


Figure 38: Forecasting predictions from Persistence Realist and Linear Regression on **Annual Inflation Differentiation with Lag 2**



Figure 39: Performance of Persistence Realist and Linear Regression on **Annual Inflation Differentiation with Lag 2**

for diff\_1 and diff\_2, the Linear Regression Model yields a lower MAE and RMSE for the test set than the Persistent Realist models. For the diff\_lag2 treatment, the Persistent Realist has a lower MAE (but higher RMSE) than the Linear Regression Model. But overall, the best model, a Linear Regression Model trained on the diff\_1 treatment, is worse than the best model without differential treatment. So, we do not differentiate.

### *Smoothing*

Figure 40: Forecasting plots after different smoothing parameterisations on Inflation Rate Time Series

Figure 41: Forecasting results after different smoothing parameterisations on Inflation Rate Time Series

### *Scaling*

Figure 42: Forecasting plots after different scaling parameterisations on Inflation Rate Time Series

Figure 43: Forecasting results after different scaling parameterisations on Inflation Rate Time Series