

PA6 Cost Prediction

Schneider Electric Data Challenge

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Business Issue



Schneider Electric's reliance on plastic materials in its operations



Difficulty in accurately predicting costs due to fluctuating market factors



Inaccurate cost predictions can decrease profitability



Estimate future costs of Polyamide 6 (PA6) plastic raw material over 3, 6, and 9 months

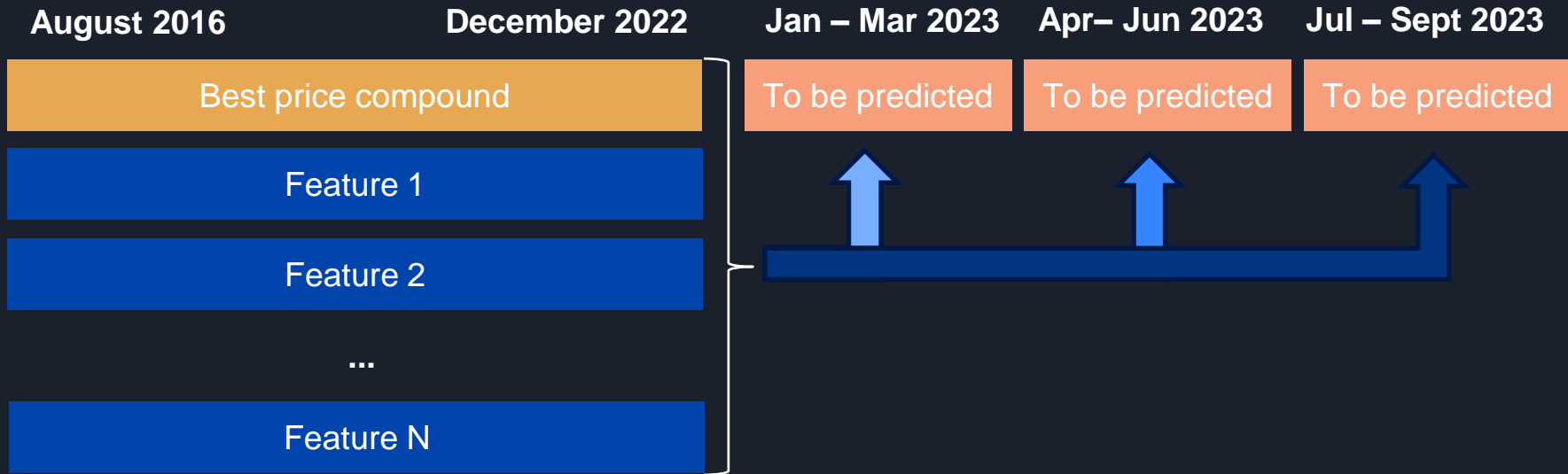


Informed decision-making, optimized budgeting and minimized financial risks



Problem Statement:

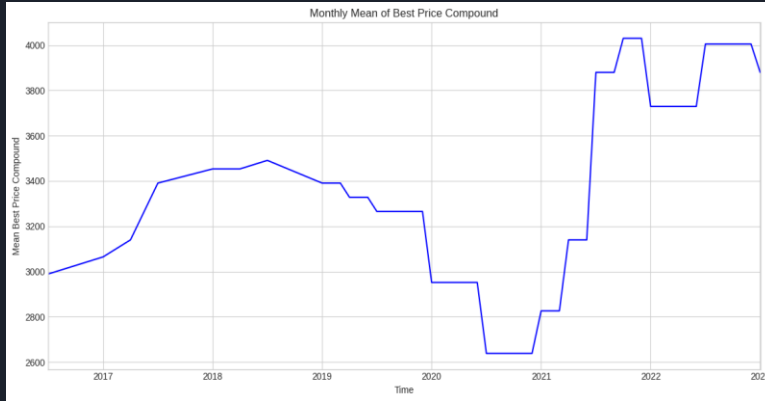
- Predict the best supplier price for 3 successive quarters, starting 01/01/2023
- We assume 3-month contracts with suppliers





EDA: Seasonal Decomposition

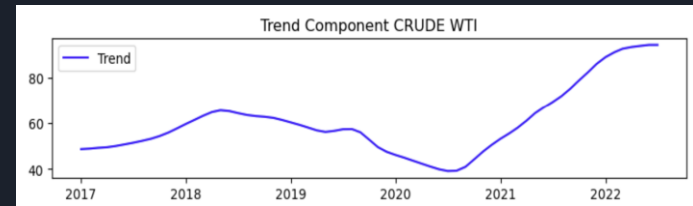
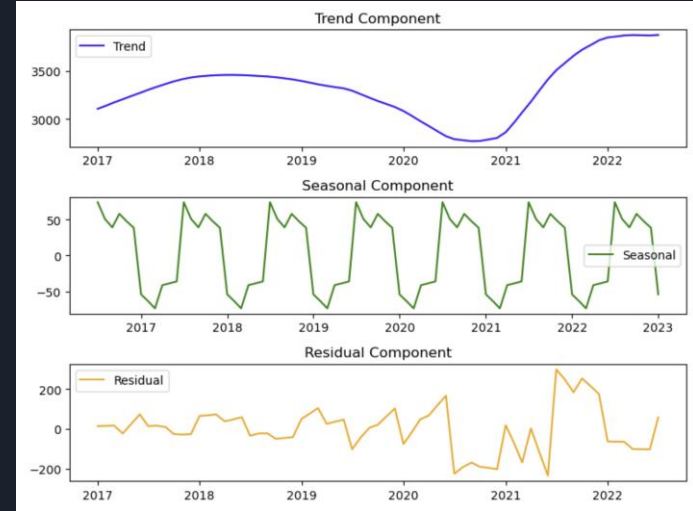
"Best price" varies every month initially, then every 3 or 6 months after 2019, due to supplier contracts.



The trend seems to be macroeconomic.

Shows high growth in the past 2 years and a strong dip during the year 2020.

Seasonality is marginal but highlights a mid-year peak.





EDA: Lag Observations

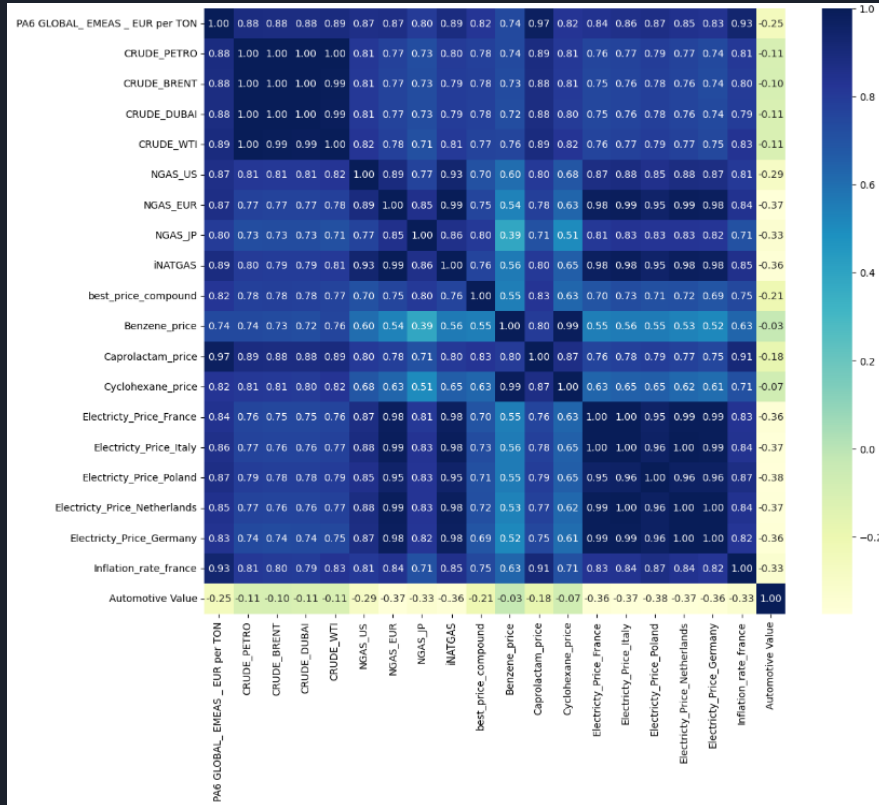


same_month_correlation =0.83
correlation_shift_1 =0.86
correlation_shift_2 =0.91
correlation_shift_3 =0.93

Caprolactam prices from 3 months ago are most highly correlated with best price compound!



EDA: Correlational Analysis



The dataset shows very **high correlations** amongst almost all values.

There is an almost **perfect positive correlation** between the **electricity prices** in Europe and between the **crude oil prices** from different continents.

The prices of hydrocarbons in the market showcase only a moderate correlation with the prices of different types of natural gases.

Automotive value is **weakly negatively correlated** to all other features.



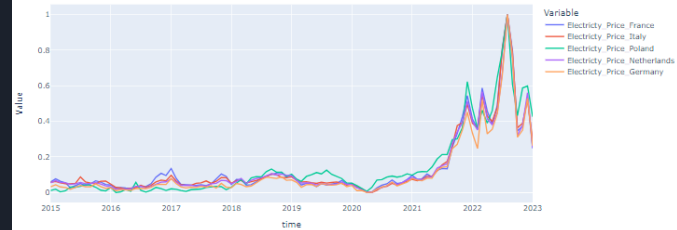
Feature Engineering: Reducing Multicollinearity

Petro variables normalized



→ Keep only one of crude oil variables

Electricity variables normalized



→ Only keep one country at random

NGas variables normalized



→ Take mean of natural gas

Chemicals variables normalized



→ Keep either Benzene or Cyclohexane



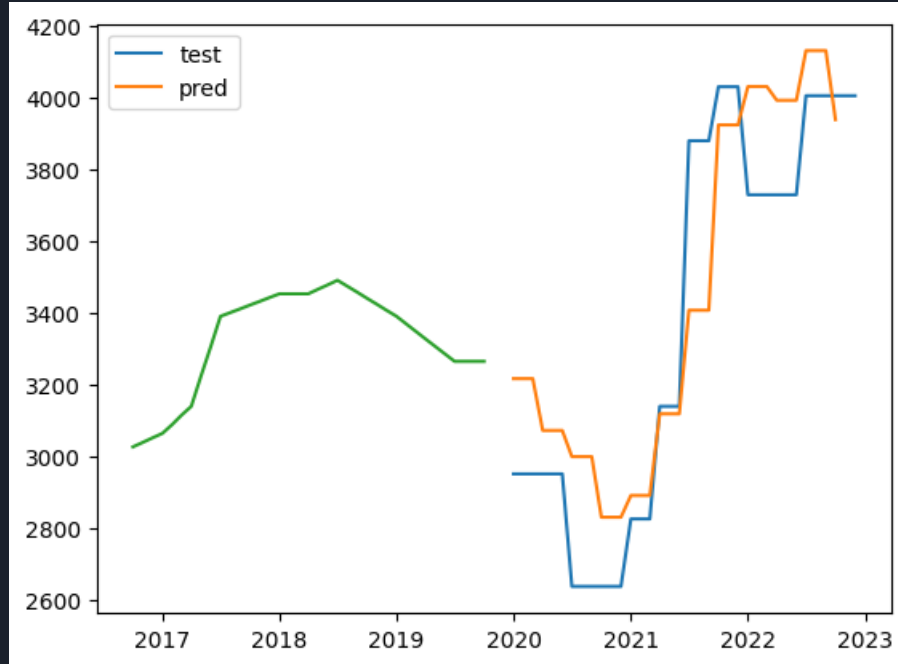
Prediction horizon 1: January-March 2023

- For this prediction we trained a model that predicts **next** months "best price"
- This model is most useful at end of quarter, to predict next quarter prices





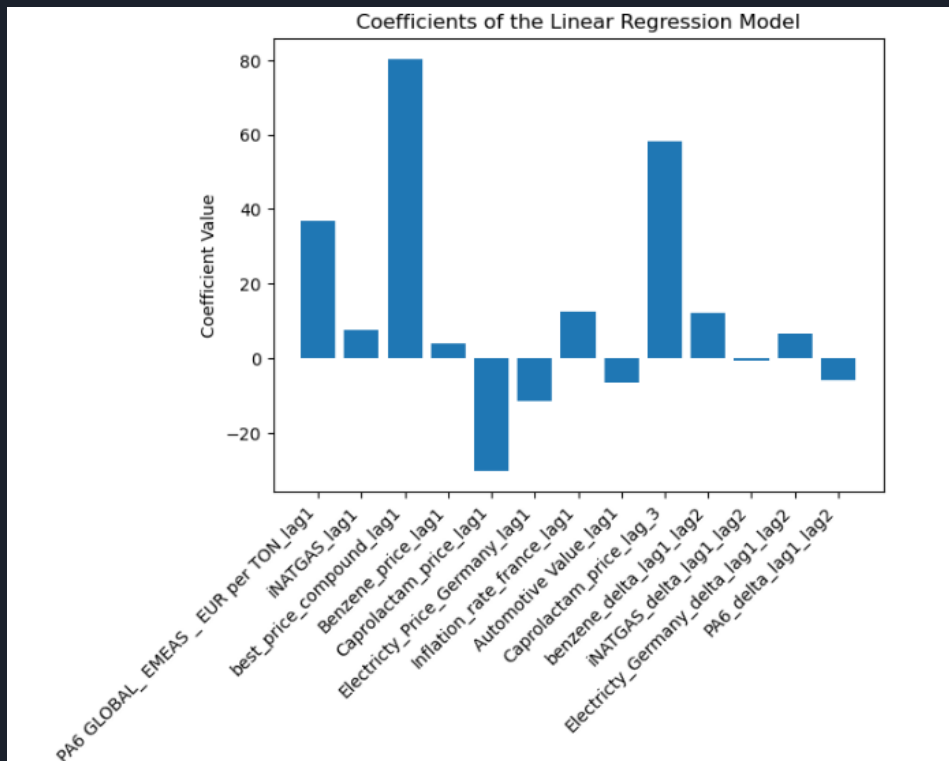
Modelling: Linear Model for 1 month prediction



- Trained in 2016-2020
- Tested in 2020-2022
- All predictions made EOQ for next month
- **MAPE: 5.77%**
- **MAE: 196.8**
- Target prediction for 01/01/2023: 3665
- True value: 3879



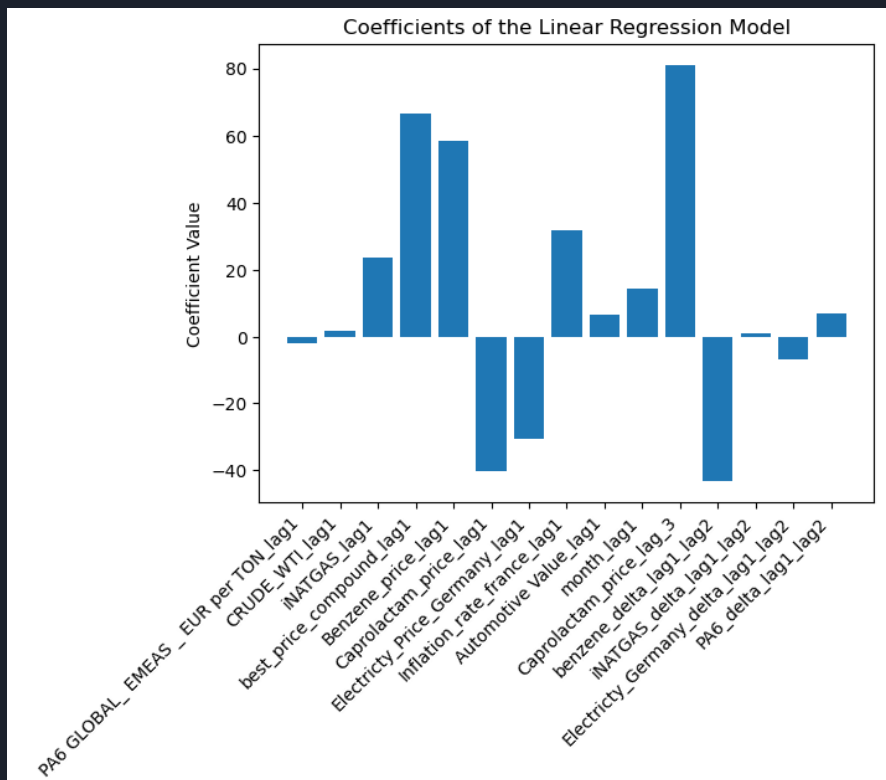
Modelling: Feature Importance up to 2020



- **High autocorrelation**
(dependance on best price lag 1)
- High importance of **Caprolactam lag 3** as anticipated
- High importance of **PA6 Global lag 1**



Modelling: Feature Importance trained up to 2022

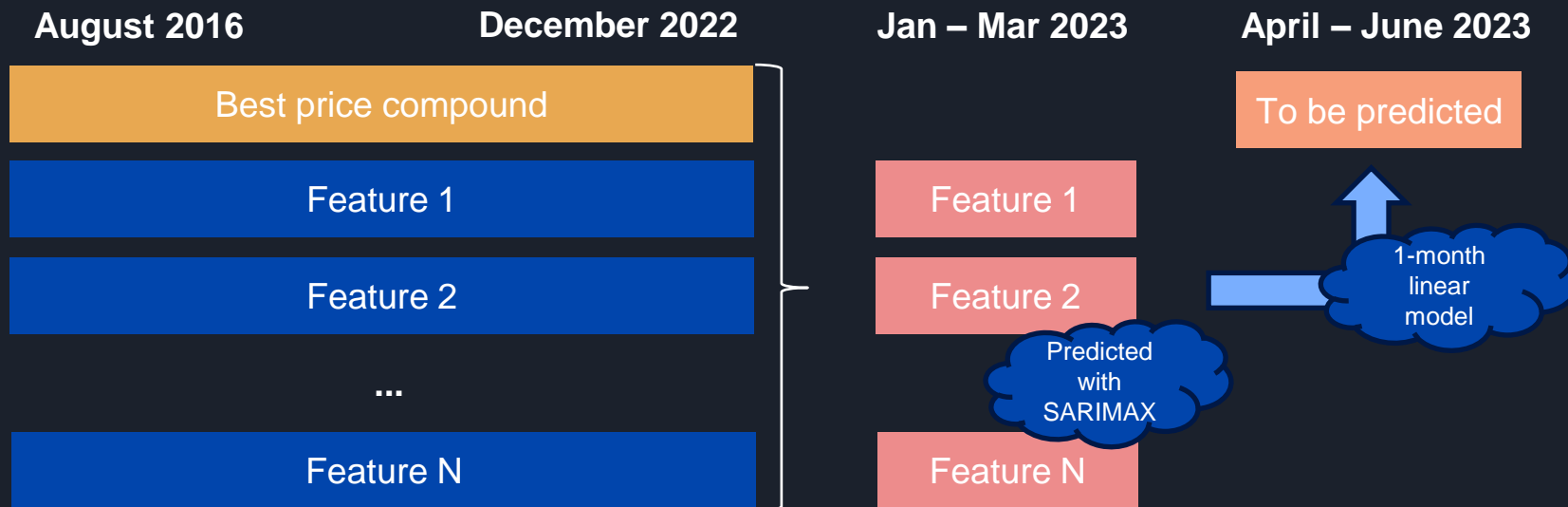


- Still high autocorrelation
- Even higher importance of **Caprolactam lag 3** as anticipated
- Big increase Benzene importance



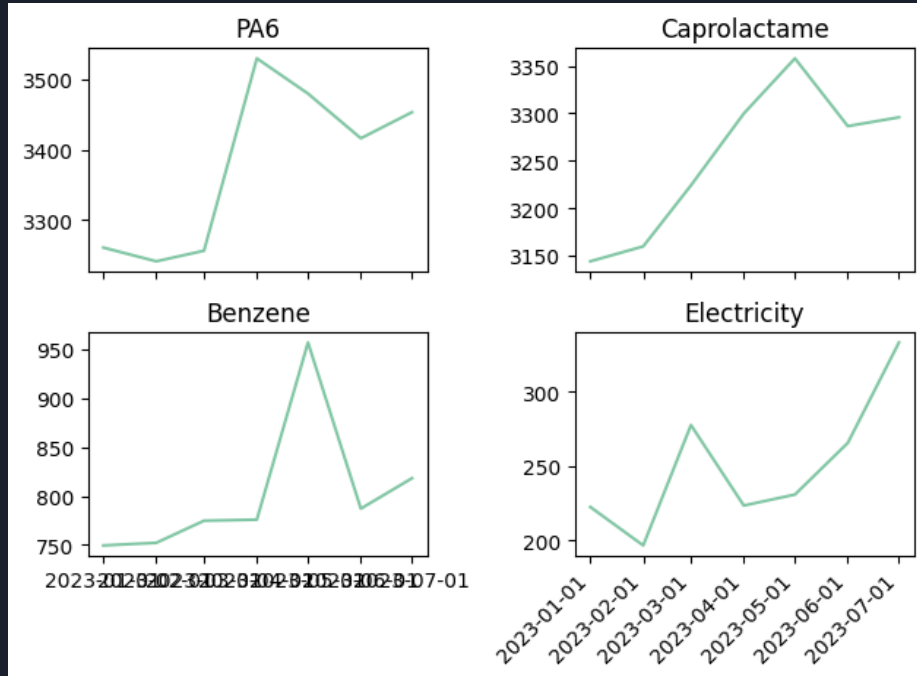
Prediction horizon 2: April-June 2023

- Model predicts best price **4 months ahead**
- First predicts X features 3 months ahead using SARIMAX
- Then uses 1-month linear model on top of SARIMAX predictions





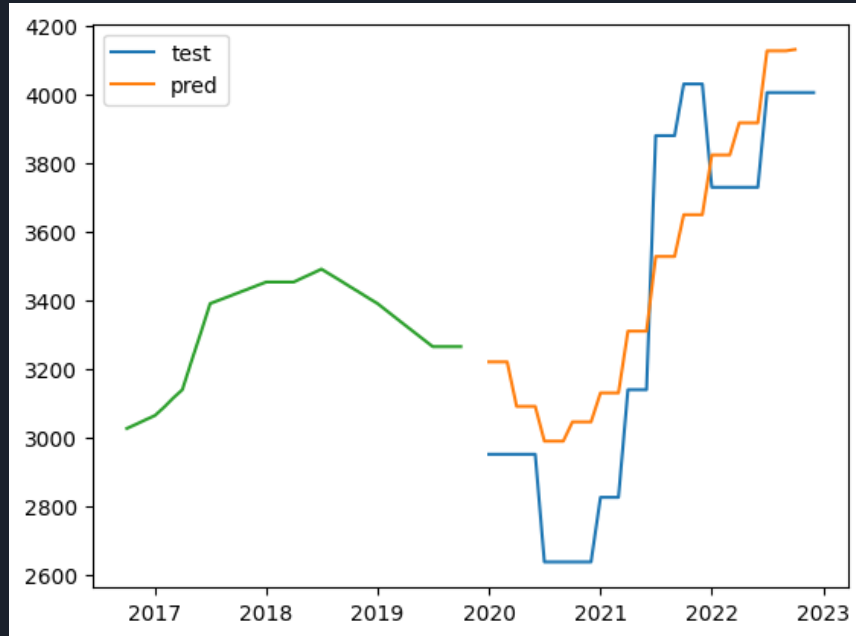
Predicted features with SARIMAX



- Prediction of the 4 most important features using SARIMAX model



Modelling: linear model for long term prediction



- Only trained on the important selected features
- **Trained in 2016-2020**
- **Tested in 2020-2022**
- MAPE: 7.50%
- MAE: 242
- Less precise than 1 month model



Link between Raw Material and Business Trends

Expectation 1 : Prices for products with highest PA6 % are most correlated with PA6 global compound prices

Expectation 2 : Most likely lag between PA6 global price and finished product prices, due to production time

Odance is made of 20% of PA6



PA6 price correlation

Correlation Coefficients for price_odace_PA6:

Mean_price_Schneider	0.148754
Mean_price_Legrand	NaN
Mean_price_Hager	NaN

Best price correlation

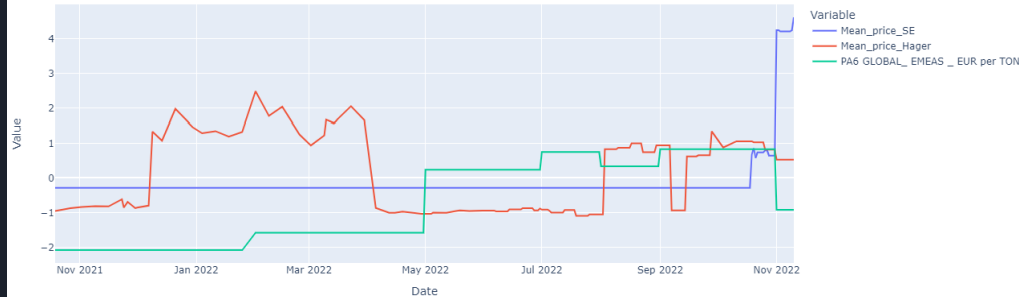
Correlation Coefficients for price_odace_

Mean_price_Schneider	-0.055981
Mean_price_Legrand	NaN
Mean_price_Hager	NaN



Price analysis – other products

All Variables Normalized - price_IC60



IC60 product (33,2% PA6)

Correlation Coefficients for price_IC60_PA6:

Mean_price_SE 0.759527

Mean_price_Hager -0.120425

All Variables Normalized - price_GV2



GV2 product (2,2% of PA6)

Correlation Coefficients for price_GV2_PA6:

Mean_price_SE 0.841660

Mean_price_ABB 0.418522

Mean_price_Siemens 0.060293



Executive Summary

- Objective: using time series data to predict next 3/6/9 months PA6 cost and provide pricing strategies
- Methods:
 1. EDA highlighted importance of feature engineering to remove multicollinearity and add lagged features, such as Caprolactam_lag_3
 2. Utilization of 3 simple models, with decreasing precision, one for each horizon
 3. Using predicted features from SARIMAX model together with a linear model to predict the 6/9-months price

- Prediction results:

3-month	6-month	9-month
3665	3539	3726

- Recommendations:
 1. Whenever the 1-month prediction is significantly lower than 4-month prediction, negotiate a 6-month supplier contract instead of 3-months
 2. Keep an eye on the highlighted compounds from our analysis, and macroeconomic trends