Econometrics Project

EU ETS Spot Volume Modeling and Forecasting

Table Of Contents

1. Introduction

5. Model selection

2. The dataset

6. Exogenous variables

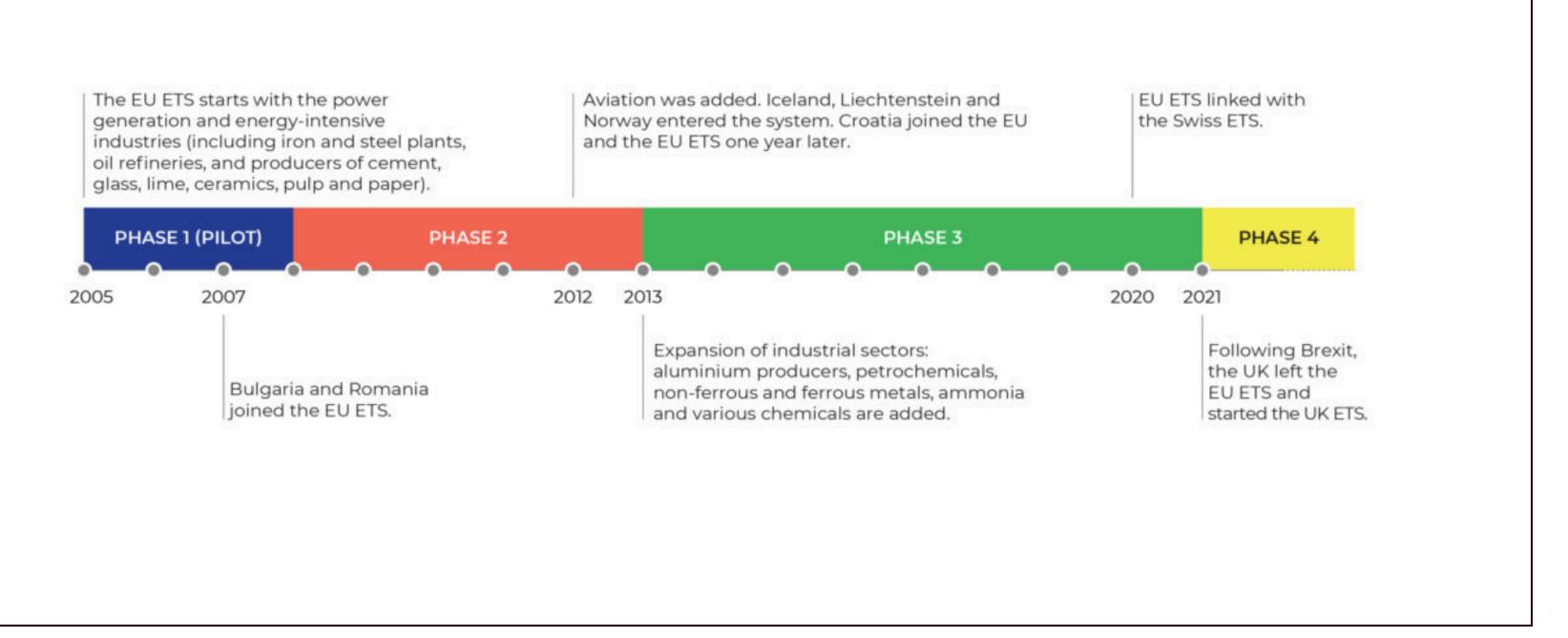
3. Data cleaning

7. Forecasting

4. ARCH effect

8. Conclusions

Introduction



The dataset

Main dataset: EU ETS allowances transactions volumes

- Daily data from third phase of EU ETS
- In-sample data: from 01-01-2013 to 12-09-2019
- Out-of-sample data: from 13-03-2019 to 05-05-2020

The dataset

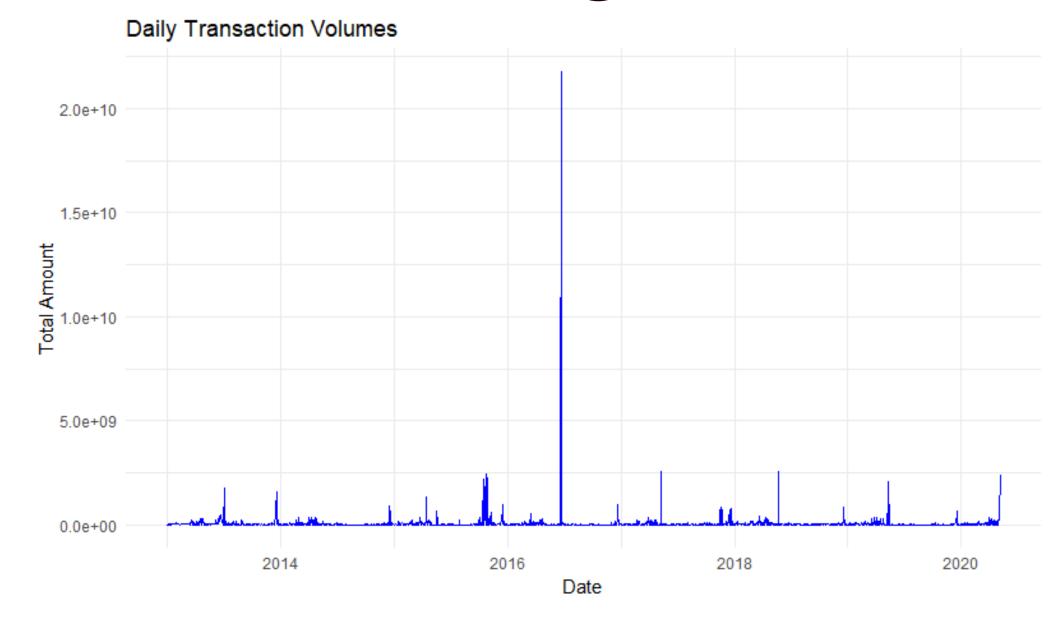
Auxiliary datasets:

Brent oil daily prices

Coal daily prices

• ICE future daily prices

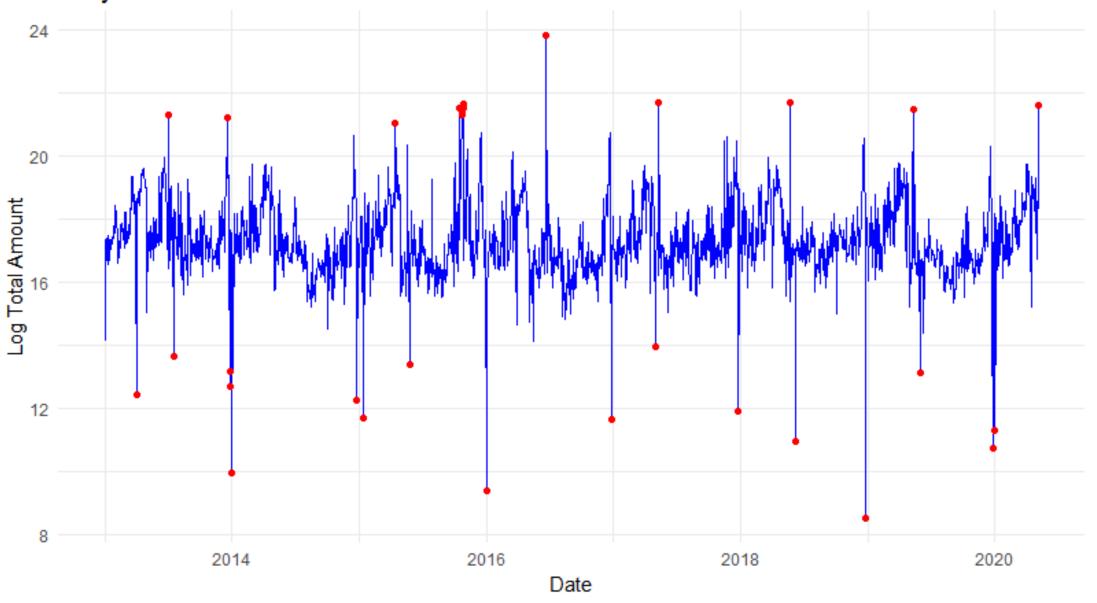
VIX index

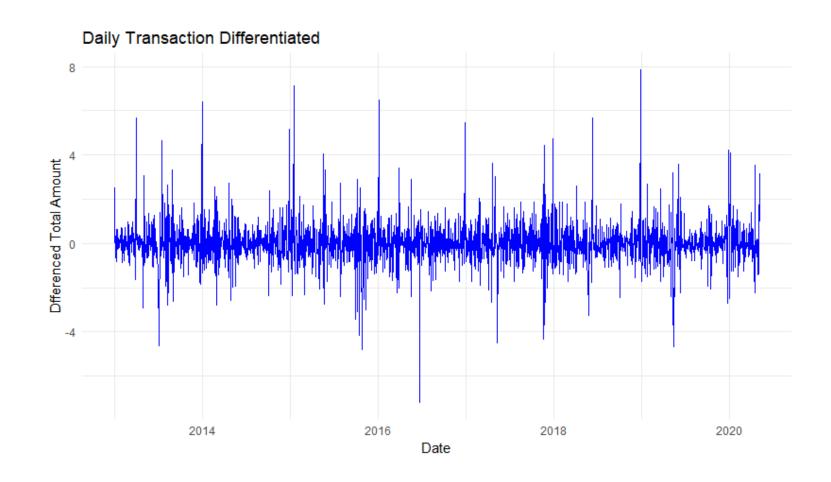


The dataset seems pretty difficult to handle: we do the followings to obtain better results

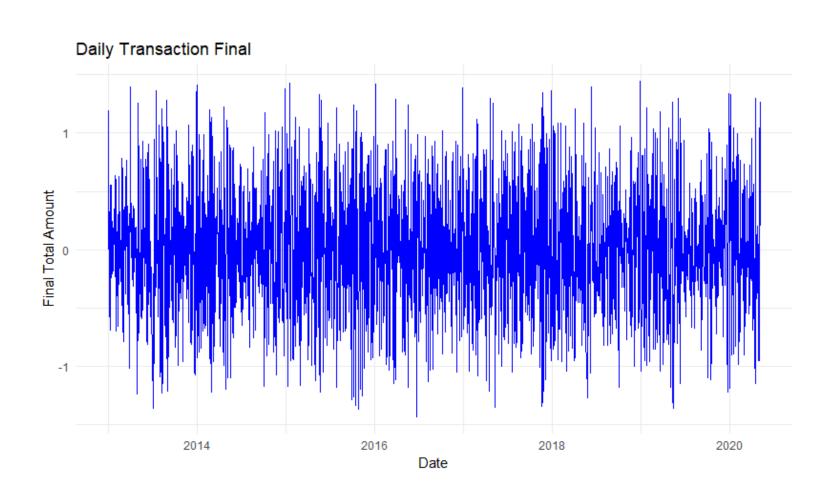
- Delete observations relative to the weekends
- Take the logarithm of the data
- Delete observations that are very distant from the mean
- Differentiate once the logarithmic series
- Use an arctan transformation







Differentiated time series



Final time series

| ADF/KPSS TESTS | Logarithmic time series | Differentiated time series | Final time series |
|----------------|----------------------------|-------------------------------|----------------------|
| KPSS | 0.22 | 0.03 | 0.02 |
| ADF | -0.93 | -57.80 | -60.56 |

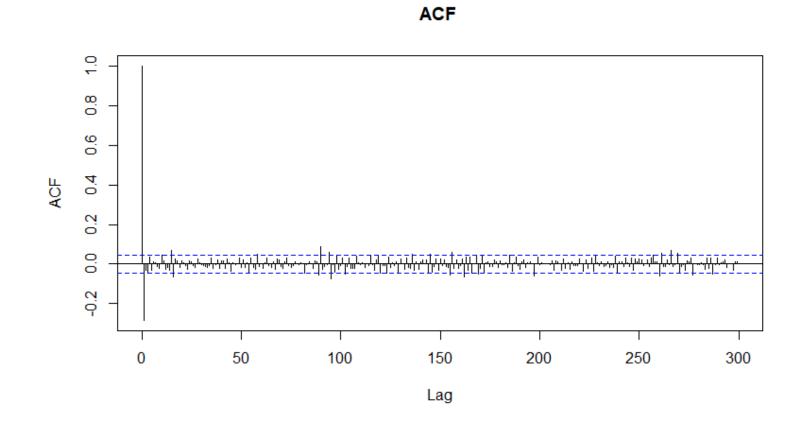
ARCH effect

ARCH LM-test; Null hypothesis: no ARCH effects

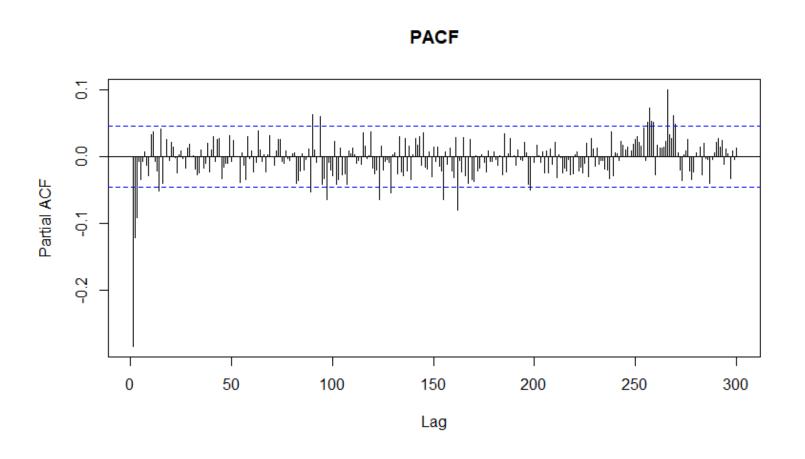
Chi-squared = 39.146, df = 12, p-value = 9.957e-05

We have evidence of ARCH effect

Model selection



Maximum MA order: 1



Maximum AR order: 3

Optimal model selection

| Model | k | \hat{L} | BIC |
|-------------------------|---|-----------|--------|
| AR(0) | 3 | -1653.0 | 1.7743 |
| AR(1) | 4 | -1547.3 | 1.6656 |
| AR(2) | 5 | -1509.8 | 1.6296 |
| AR(3) | 6 | -1498.9 | 1.6221 |
| MA(1) | 4 | -1500.7 | 1.6159 |
| ARMA(1,1) | 5 | -1494.4 | 1.6132 |
| ARMA(2,1) | 6 | -1494.3 | 1.6172 |
| ARMA(3,1) | 7 | -1492.1 | 1.6189 |
| GARCH(1,1) | 5 | -1607.0 | 1.7332 |
| AR(1)- $GARCH(1,1)$ | 6 | -1511.8 | 1.6359 |
| AR(2)- $GARCH(1,1)$ | 7 | -1470.0 | 1.5953 |
| MA(1)- $GARCH(1,1)$ | 7 | -1461.30 | 1.5820 |
| ARMA(1,1)- $GARCH(1,1)$ | 7 | -1454.1 | 1.5783 |
| ARMA(2,1)- $GARCH(1,1)$ | 8 | -1453.5 | 1.5817 |
| ARMA(3,1)- $GARCH(1,1)$ | 9 | -1450.6 | 1.5826 |
| | | | |

 We compare different model orders and choose the one minimizing the BIC

 The selected model is the ARMA(1,1) -GARCH(1,1) model

Exogenous variables

| | \mathbf{A} | В | \mathbf{C} |
|--------------|--------------|-------------|--------------|
| μ | 0.00797 | 0.00715 | 0.00748 |
| | (1.5115) | (1.5273) | (1.5918) |
| ϕ_1 | 0.1878*** | 0.18784*** | 0.18772*** |
| | (3.8754) | (4.1207) | (4.1086) |
| $	heta_1$ | -0.62732*** | -0.63334*** | -0.63277*** |
| | (-16.7109) | (-19.7662) | (-19.5837) |
| α_{0} | 0.03508*** | 0.0341** | 0.03392** |
| | (2.8485) | (2.4525) | (2.4725) |
| α_1 | 0.12911*** | 0.12696*** | 0.12854*** |
| | (4.7835) | (4.5189) | (4.5959) |
| eta_1 | 0.75122*** | 0.7561*** | 0.75568*** |
| | (12.2784) | (10.9682) | (11.0859) |
| γ_1 | | -0.00515 | |
| | | (-0.2094) | |
| γ_2 | | -0.00556 | |
| | | (-0.5802) | |
| γ_3 | | 0.00388 | |
| | | (0.7524) | |
| γ_4 | | 1.13573* | 1.18295* |
| | | (1.6942) | (1.7538) |

A = ARMA(1,1) - GARCH(1,1)

B = ARMAX(1,1,4) - GARCH(1,1)

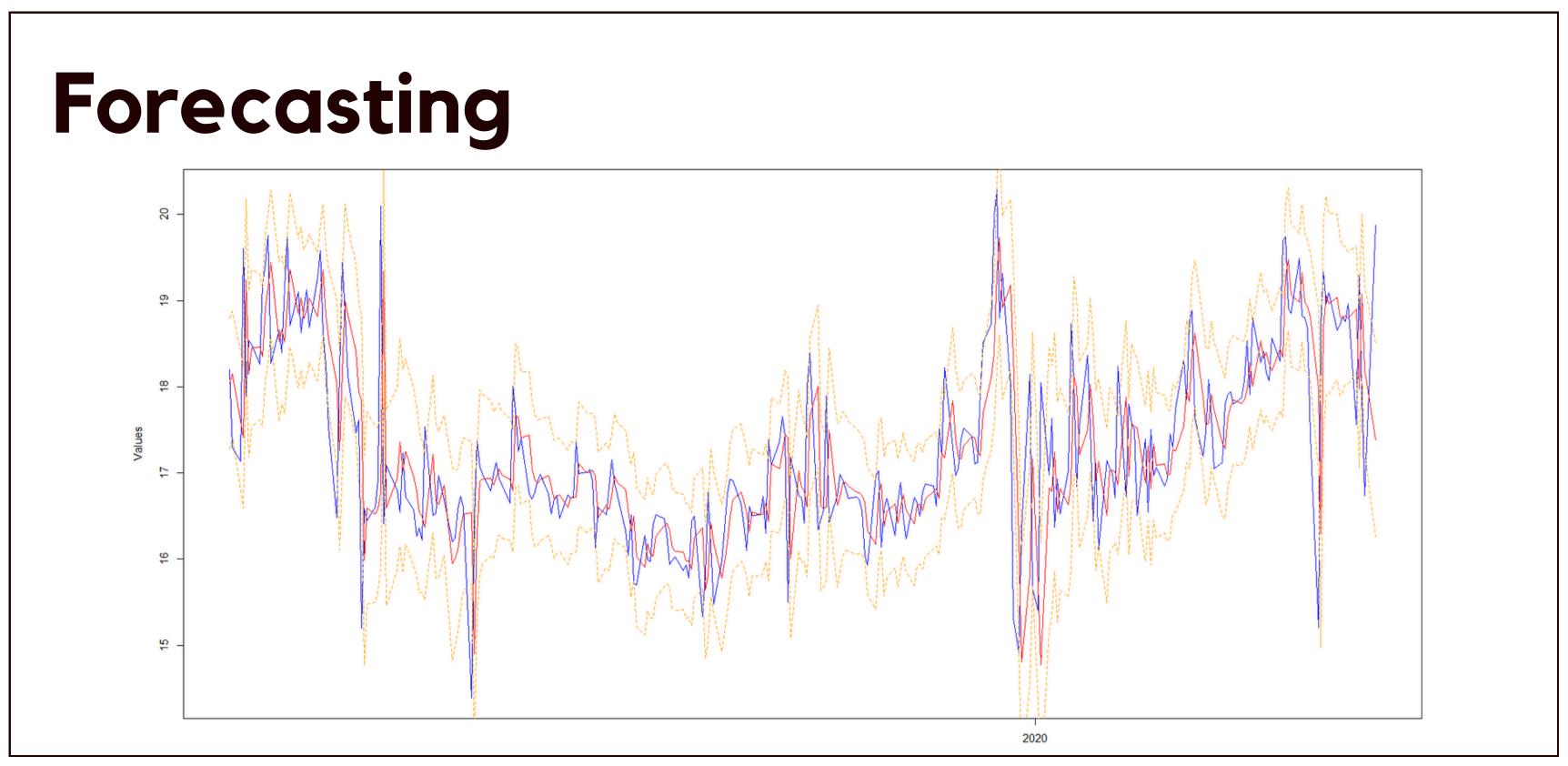
C = ARMAX(1,1,1) - GARCH(1,1)

 γ_1 =future daily price differences

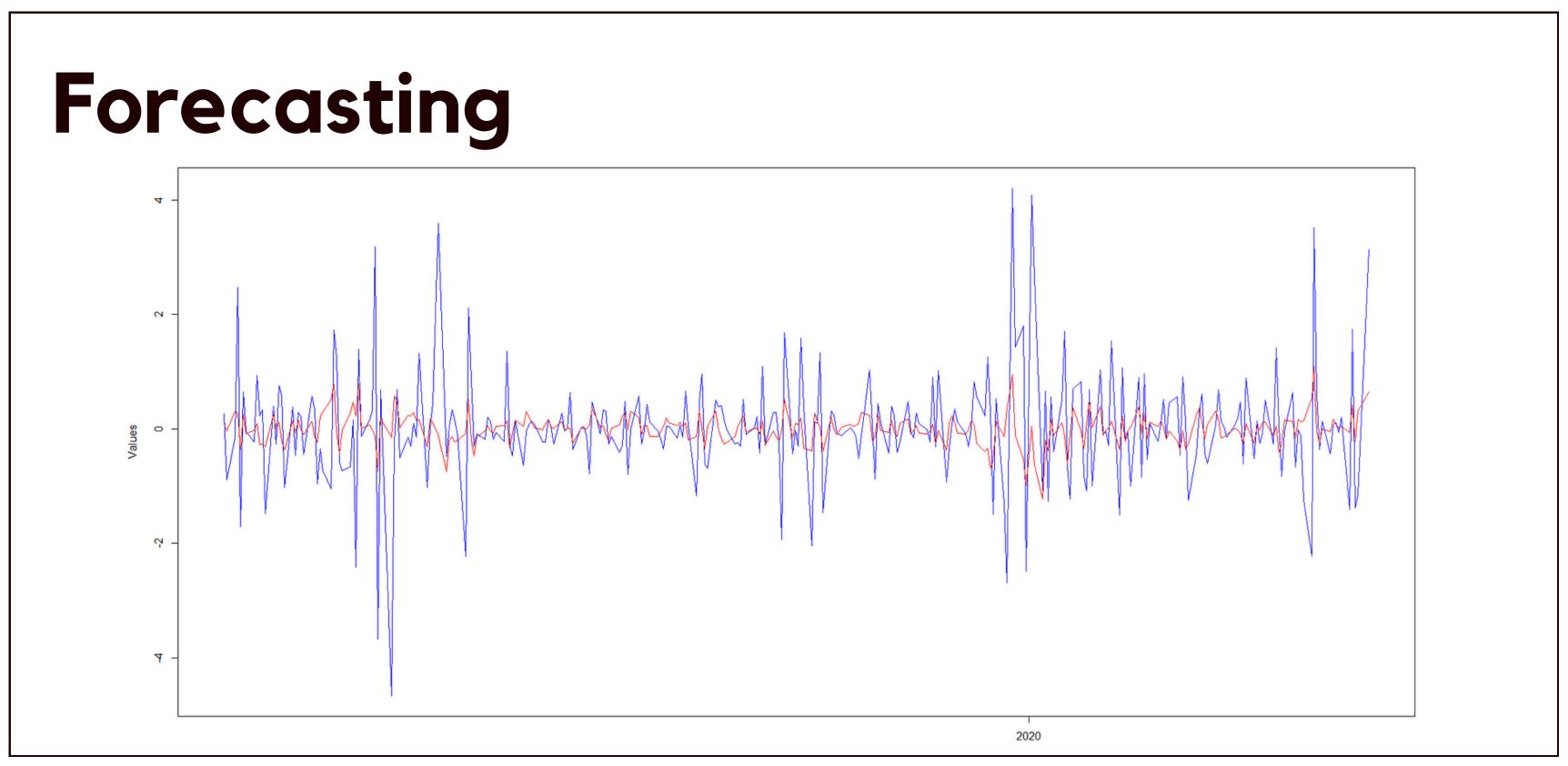
 γ_2 =Brent oil daily price differences

 γ_3 =VIX daily differences

 γ_4 =coal daily price differences



Real data and predicted data with their 90% confidence interval



Real data and predicted daily differences

Forecasting metrics results

| Set | MAPE | Pct | MAE | MSE |
|------------|--------|--------|----------|-------------|
| Volume | 77.72% | 63.17% | 34509452 | 5.539e + 15 |
| Log Volume | 3.08% | 63.17% | 0.534 | 0.635 |

Table 4: Performance metrics for the volume time series and the log-volume time series

Conclusions

- Our model performance seems good but is not able to perform well in different scenarios, especially if there are peaks in log scale differences
- Although the same parametric model class used in previous literature to predict EU ETS prices seems to work for volumes, the same cannot be said for the exogenous variables, which don't seem to be highly significant in our case

Bibliography

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