# The Merit Order Effect in the German Power Market

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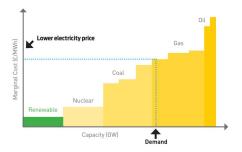
Ladislaus von Bortkiewicz Chair of Statistics Humboldt–Universität zu Berlin Statistical Programming Languages https://github.com/mpff/spl2018-bfm



Motivation — 1-1

#### Motivation

We want to measure the merit order effect for the German power market, based on Clò, Cataldi, and Zoppoli 2015.





Motivation — 1-2

## Regression

#### Regression equation:

price = 
$$\beta_0 + \beta_1 \cdot \text{dem} + \beta_2 \cdot \text{wind} + \beta_3 \cdot \text{solar} + \gamma \cdot \text{ymd} + \epsilon$$
 (1)

price: Mean daily day ahead price

dem: Forecasted daily demand

solar: Forecasted daily electricity production from solar

wind: Forecasted daily electricity production from wind

ymd: Year, month, day dummies

#### **Data Sources**

price: Day Ahead Price, since 2011 (Elspot Prices Data).

dem: Day Ahead Demand, since 2015, DE+AT+LUX (entsoe)

solar: Day Ahead Forecast, since 2012, DE (netztransparenz.de)

Day Ahead Forecast, since 2015, AT (apg.at)

wind: Day Ahead Forecast, since 2012, DE (netztransparenz.de)

Day Ahead Forecast, since 2015, AT (apg.at)

#### **Problems:**

- 1. Different timeframes and -frequencies.
- 2. NA's a problem for daily averages and sums.
- 3. LUX in demand a (big) problem?

# **Data Cleaning**

```
df.pun.0 <- read.csv("source/Elspot_xxx.csv")</pre>
df.pun <- subset(df.pun.0, select = c(HourUTC,</pre>
SpotPriceEUR))
names(df.pun) <- c("TIME", "PUN")</pre>
df.pun$TIME <- ymd_hm(df.pun$TIME)</pre>
df.pun <- time.FRAME(df.pun)</pre>
df.pun <- aggregate(list("PUN" = df.pun$PUN),</pre>
list("TIME" = cut(df.pun$TIME, "1 day")), FUN = mean)
# Bind final Dataframe
df <- cbind(df.pun, df.dm, df.solar, df.wind)</pre>
Github: MOErawdata.R, MOEmergedata.R
```

# **Problems with Missing Values**

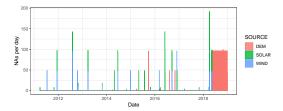


Figure 1: NAs in raw data

How big is the problem after cleaning the data? temp <- DiagMissingValues(df, dlevel=1)...

Number of incomplete cases: 21 (1.916%)

GitHub: MOEplotNAs.R The Merit Order Effect in the German Power Market

## **Problems with Missing Values**

Test: Are NAs handled correctly while calculating daily values?

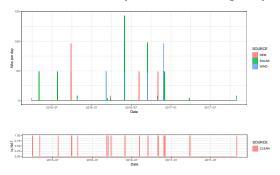


Figure 2: NAs in raw (a) and cleaned data (b)



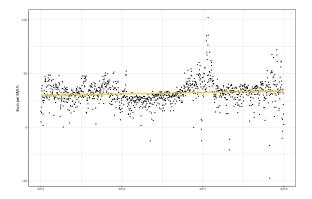


Figure 3: Mean Daily Day-Ahead Price



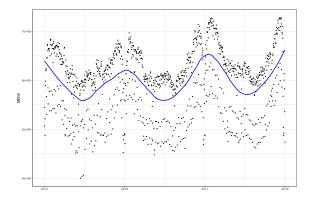


Figure 4: Daily Day-Ahead Demand



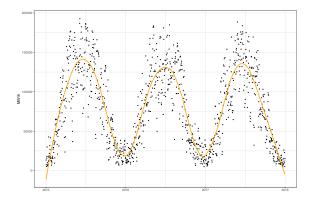


Figure 5: Daily Day-Ahead Solar Production



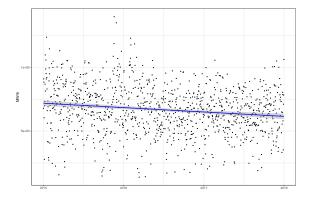


Figure 6: Daily Day-Ahead Wind Production



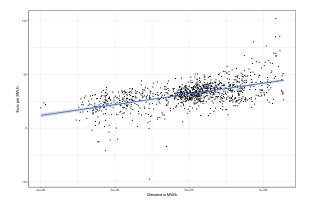


Figure 7: Price on Demand

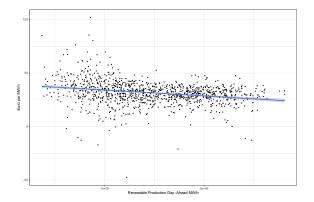


Figure 8: Price on Renewables Production



#### Time-Dummy generating function:

#Step1: Create the dummy variables for years, months and days

- ☐ Creates a dummy matrix for: year, month, day o.t. week.
- □ Function adapts the amount of years directly from the data.
- y/m/d to be omitted, bc. of the dummy variable trap, can be specified.GitHub: MOEtimedummies.R

price = 
$$\beta_0 + \beta_1 \cdot \text{dem} + \beta_2 \cdot \text{wind} + \beta_3 \cdot \text{solar} + \gamma \cdot \text{ymd} + \epsilon$$

- Step 1: Check for stationarity of the variables  $\rightarrow$  (augmented dickey fuller Test + Philipps-Perron)
- Step 2: Perform the OLS regression.
- Step 3: Check for Heteroscedasticity and Autocorrelation.
  - ightarrow Breusch-Pagan Test for heteroscedasticity.
  - $\rightarrow$  Durbin-Watson test for autocorrelation.
  - $\rightarrow$  Plot the autocorrelation structure.
- Step 4: If okay [AR(1)], use the Prais-Winsten generalized estimation method.
- Step 5: Check the new autocorrelation structure.

GitHub: MOEregression.R

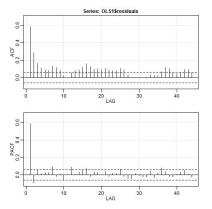


Figure 9: **OSL**, **ACF** and **PACF** Autocorrelation = TRUE (durbin-watson)



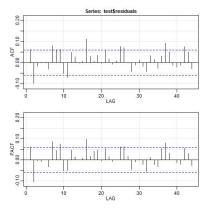


Figure 10: Prais-Winsten, ACF and PACF



```
price = \beta_0 + \beta_1 \cdot \text{dem} + \beta_2 \cdot \text{wind} + \beta_3 \cdot \text{solar} + \gamma \cdot \text{ymd} + \epsilon
```

```
coefficients:
           Estimate Std. Error t value Pr(>|t|)
Intercept -3.541e+00 5.205e+00 -0.680 0.496445
          6.751e-06 8.914e-07
                               7.573 7.82e-14
SOLAR
         -9.209e-06 9.370e-06 -0.983 0.325906
WTND
         -7.679e-05 1.029e-05 -7.465 1.72e-13 ***
X2016
         -2.862e+00 1.352e+00 -2.116 0.034558
x2017
          6.009e-01 1.377e+00
                               0.436 0.662736
         -4.393e+00 2.302e+00 -1.908 0.056604
         -6.457e+00 2.449e+00
                              -2.637 0.008495 **
         -3.994e+00 2.605e+00 -1.533 0.125561
         -5.292e+00 2.656e+00 -1.992 0.046614
X06
         -2.150e+00 2.692e+00 -0.798 0.424759
         -1.322e-01 2.652e+00 -0.050 0.960263
         -1.724e+00 2.648e+00 -0.651 0.515092
         -1.009e+00 2.542e+00 -0.397 0.691371
x10
          1.165e+00 2.442e+00
                               0.477 0.633527
X11
         -4.452e-01 2.420e+00 -0.184 0.854089
         -3.498e+00 2.321e+00 -1.507 0.131985
Monday
          4.745e+00 1.129e+00
                               4.203 2.86e-05 ***
          5.336e+00 1.350e+00
                               3,952 8,25e-05
Tuesday
Wednesday 4.771e+00 1.402e+00
                               3,404 0,000689 ***
          4.210e+00 1.392e+00
                               3.025 0.002543 **
Friday
          3.994e+00 1.303e+00
                               3.064 0.002237 **
Saturday
                               3.652 0.000273 ***
         2.513e+00 6.881e-01
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 6.595 on 1073 degrees of freedom
Multiple R-squared: 0.7858, Adjusted R-squared: 0.7812
F-statistic: 171.1 on 23 and 1073 DF. p-value: < 2.2e-16
```

Figure 11: **Prais-Winsten** 

#### Outlook

- 1: Clean code and adhere to styleguide
- 2: Improve and add graphics (e.g. Fraunhofer Energycharts)
- 3: How much money have consumers saved due to renewables?
- 4: Testing of code