

20.6.2023

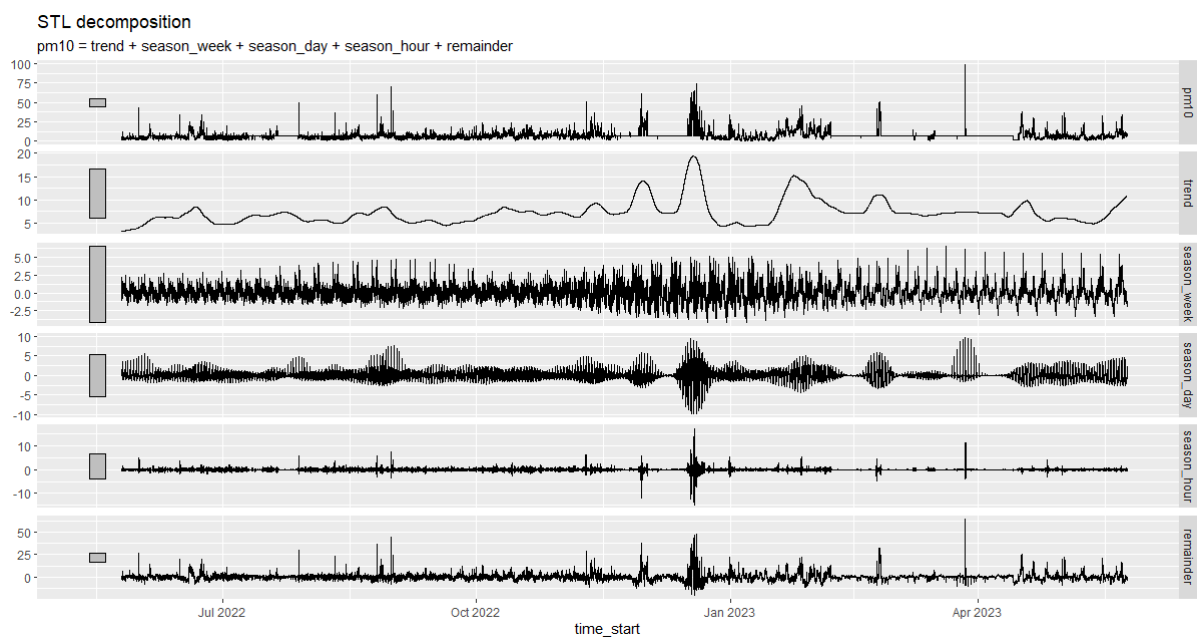
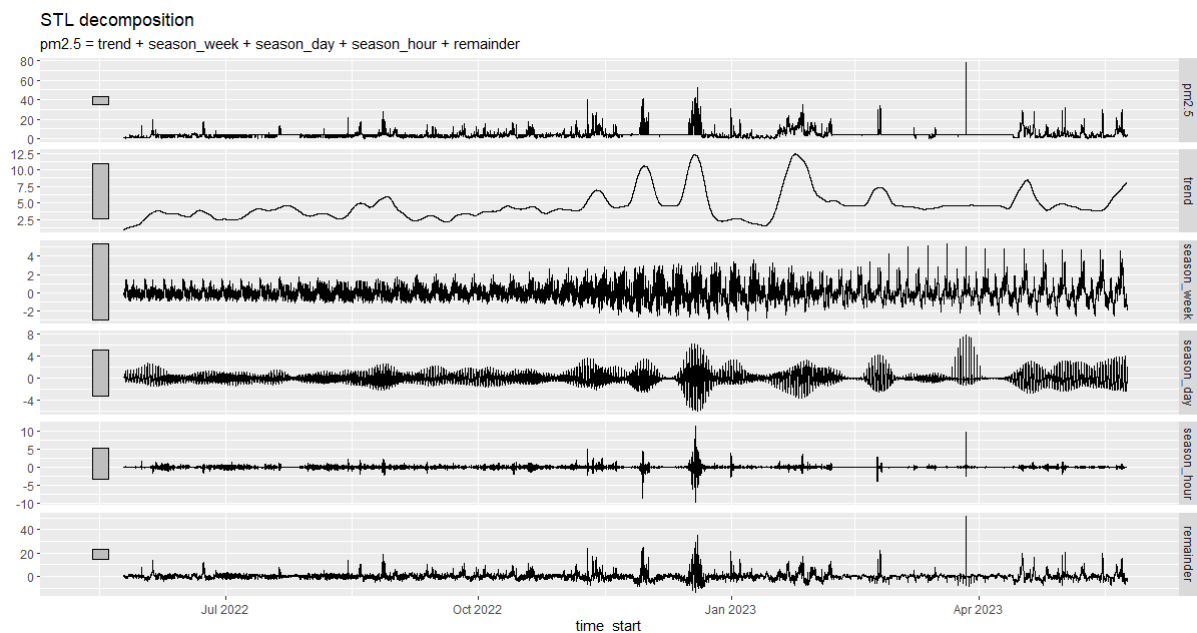
Gil Baram Geosensornetzwerke

Aufgabe 1: Evaluation der Daten

c)

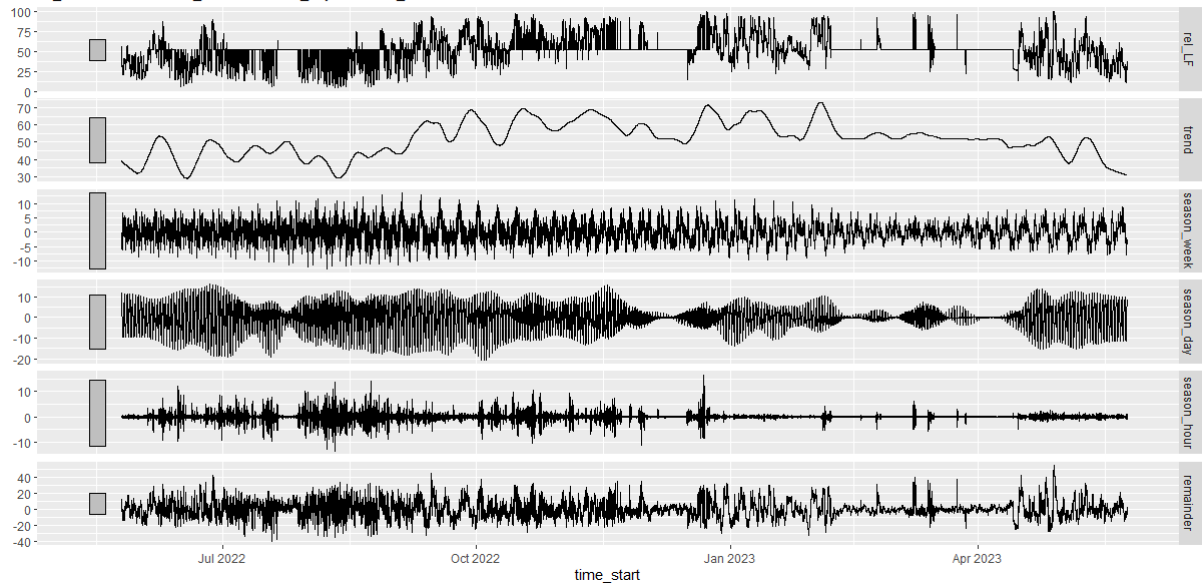
```
> summary(data)
time_start      pm2.5      pm10      rel_LF      temperatur
Length:31542    Min.   : 0.1925    Min.   : 0.225    Min.   : 4.625    Min.   : -6.75
Class :character 1st Qu.: 1.6175    1st Qu.: 3.377    1st Qu.:35.133    1st Qu.:11.57
Mode  :character Median : 2.6913    Median : 5.185    Median :51.075    Median :16.45
                Mean  : 4.5651    Mean  : 7.187    Mean  :51.779    Mean  :16.55
                3rd Qu.: 5.0319    3rd Qu.: 8.467    3rd Qu.:68.146    3rd Qu.:21.37
                Max.   :78.0300    Max.   :98.700    Max.   :99.900    Max.   :38.60
```

The data frame has 31542 rows, with no NA values. After applying gaps filling, it adds 21020 rows.



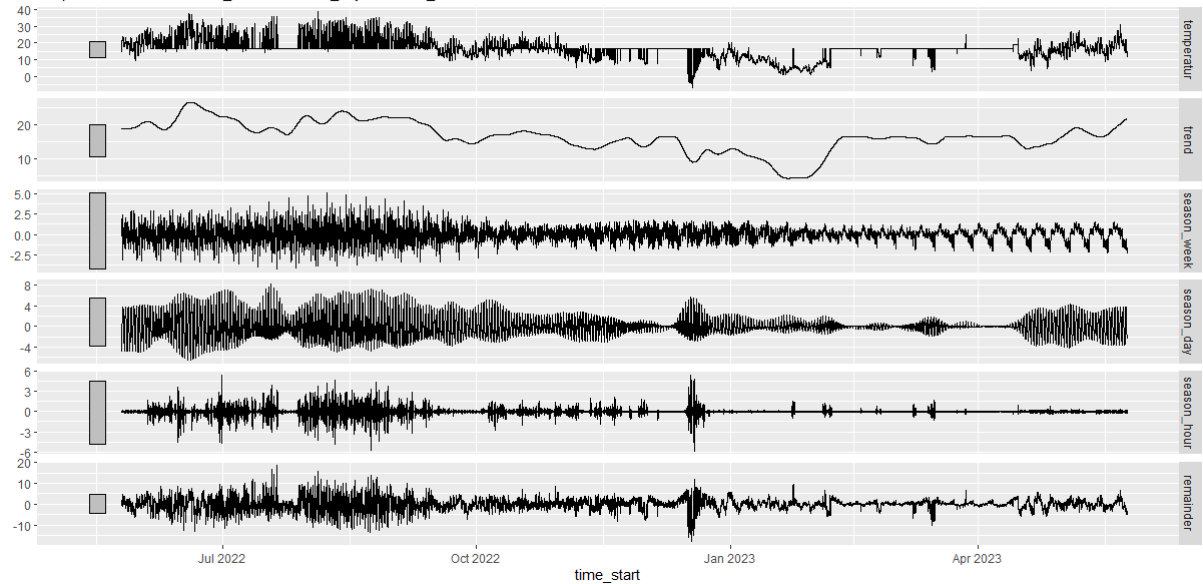
STL decomposition

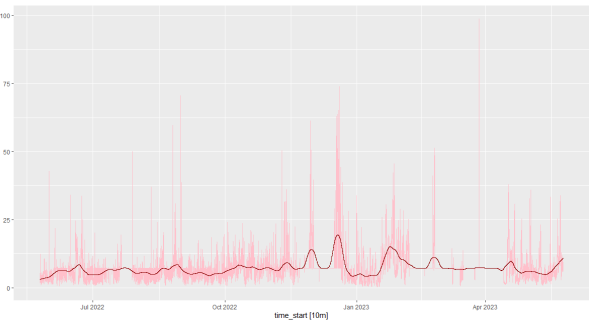
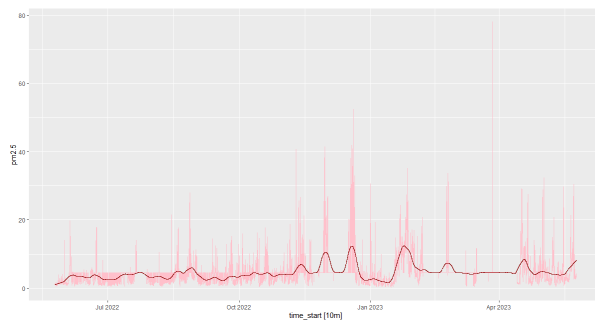
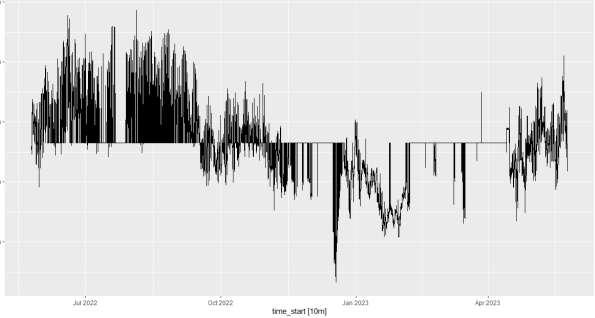
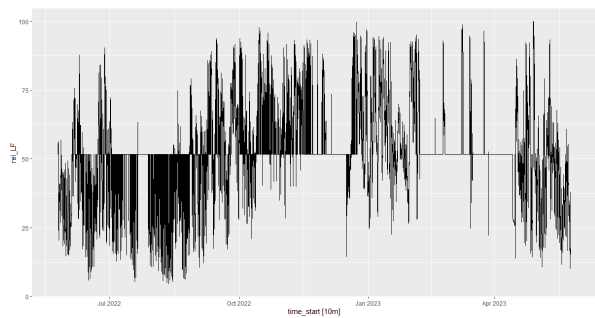
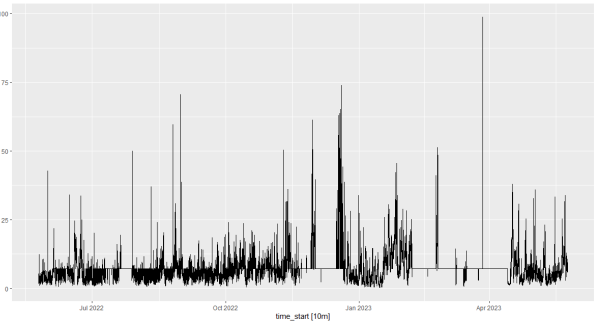
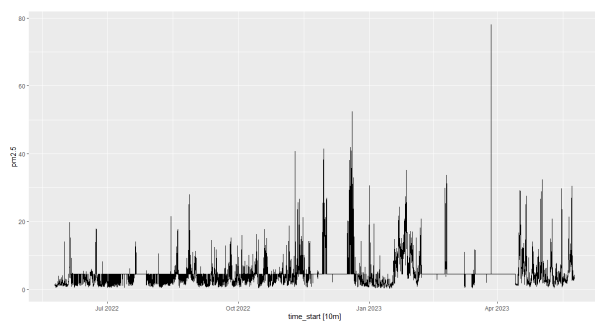
$\text{rel_LF} = \text{trend} + \text{season_week} + \text{season_day} + \text{season_hour} + \text{remainder}$



STL decomposition

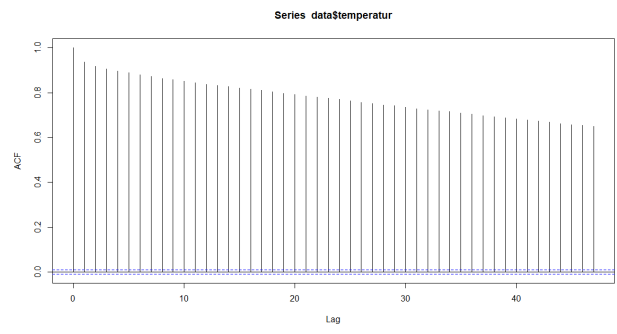
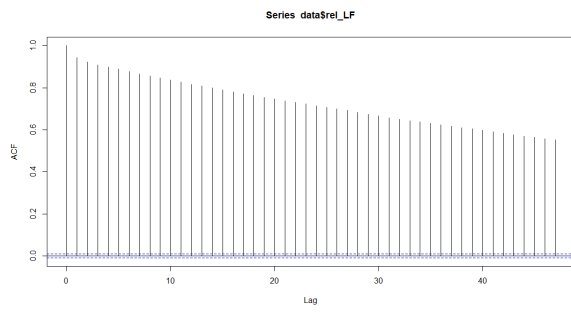
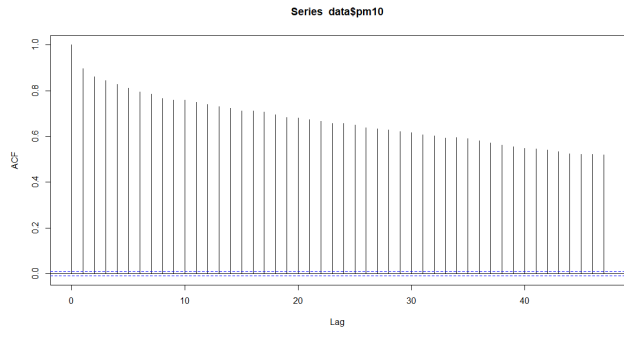
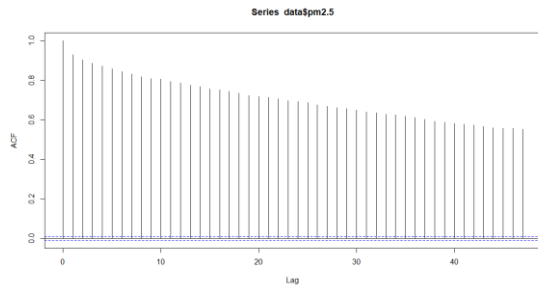
$\text{temperatur} = \text{trend} + \text{season_week} + \text{season_day} + \text{season_hour} + \text{remainder}$



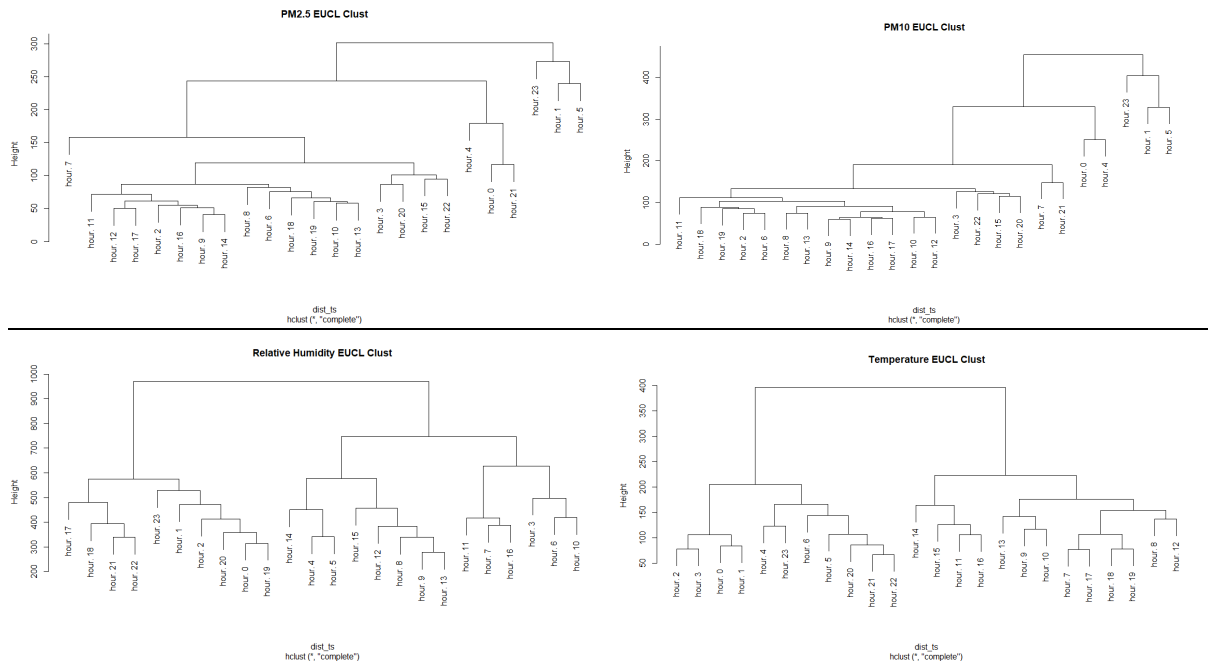


There is an outlier before Apr 2023. It reached a peak 80-100 $\mu\text{g}/\text{m}^3$.



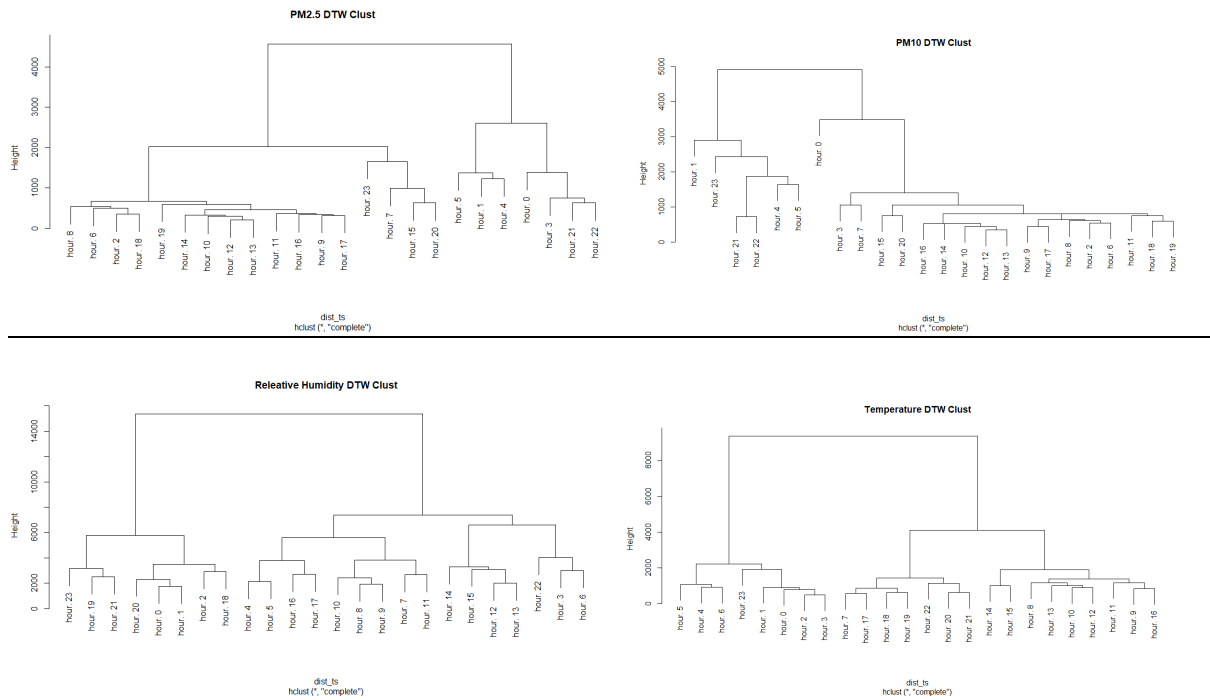


Aufgabe 2: Zeitreihen Clustering



The Euclidean distance process determines the proximity between observations by drawing a straight line between pairs of observations. Therefore, this process measures the distance between observations by looking at the length of this line between observations.

The clusters of PM2.5 and PM10 are identical. The measures during the night (hours = 1,5,23) are at same level, at this is the largest cluster.



In hierarchical clustering, the algorithm builds clusters by measuring the dissimilarities between data.

Aufgabe 3: Zeitreihen Forecasting

c) I calculated MAE for three different Forecasts for all the 4 features.

```
> fit_CV<- data_stretch |> model(MEAN(pm2.5))
> mae_display(fit_cv)
[1] 2.231458
[1] 2.288092
[1] 2.156506
> fit_CV<- data_stretch |> model(MEAN(pm10))
> mae_display(fit_cv)
[1] 2.781453
[1] 2.848204
[1] 2.72481
> fit_CV<- data_stretch |> model(MEAN(rel_LF))
> mae_display(fit_cv)
[1] 12.22813
[1] 12.10388
[1] 11.82352
> fit_CV<- data_stretch |> model(MEAN(temperatur))
> mae_display(fit_cv)
[1] 4.205818
[1] 4.133279
[1] 4.153775
```