## ANOMALY DETECTION

Anomalies of a time series data can be defined as outliers of the **remainders** once the linear trend and seasonal periodicity are taken into account.

We use library "anomalize" in R to identify anomalies. We call the function time\_decompose() with the option method="stl" (factoring the linear and seasonal components), and the function anomalize() with the option method="iqr". By default, this method defines outliers as observations lying below  $Q1-3\cdot IQR$  or above  $Q3+3\cdot IQR$ , where Q1 is the first quartile (25th percentile), Q3 is the third quartile (75th percentile), and the interquartile range is IQR=Q3-Q1.

The default setting can be changed by specifying a value for alpha other than 0.05 (option "alpha="). Outliers are defined as values that lie  $0.15/\alpha \cdot IQR$  distance away from the quartiles. The default value of  $\alpha = 0.05$ , thus resulting in the multiplicative constant of 3. If alpha is increased, more observations become outliers. If alpha is decreased, fewer observations are labeled as outliers.

**Example.** We use the crude oil data from the previous example to detect and plot anomalies in the daily closing prices.

```
library(tibbletime) #creates indices for date in time series data crudeoil.data tbl <- as tbl time(crudeoil.data, Date)
```

```
library(anomalize) library(tidyverse)
```

 $crudeoil.data\_tbl~\%>\%~time\_decompose(Close, method="stl")~\%>\%~anomalize(remainder, method="iqr", alpha=0.07)~\%>\%~time\_recompose()~\%>\%~plot\_anomalies(time\_recomposed=TRUE, color\_no='navy', color\_yes='red', fill\_ribbon='gray', size\_circles=4) + labs(title="Anomalies in Daily Closing Prices of Crude Oil", subtitle="3/8/2021-3/6/2025")$ 

