

Seasonal adjustment of weekly data by discounted least squares in

Tim Ginker

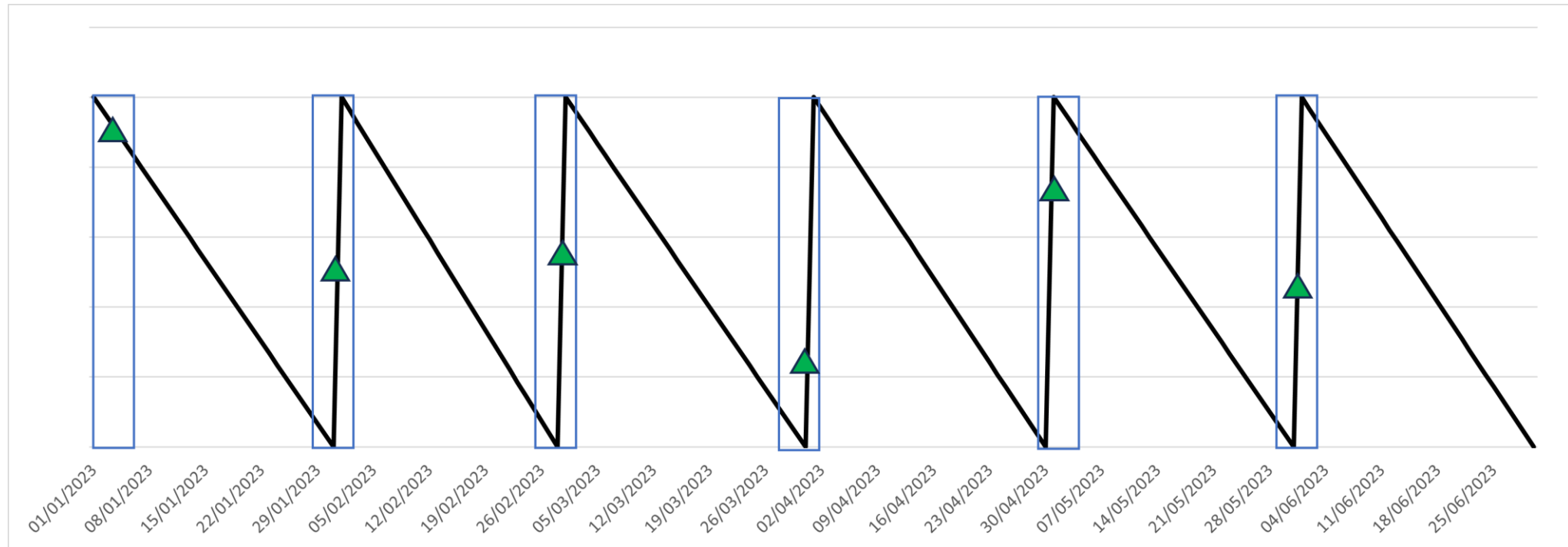
Bank of Israel

Statistical Methods and Data Science Unit

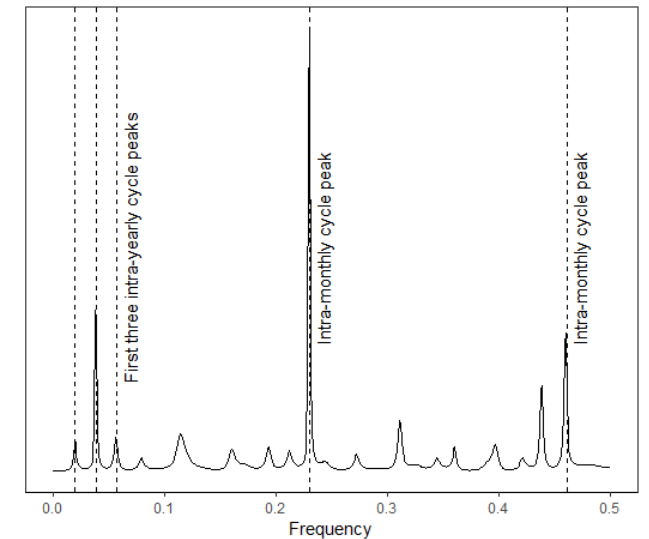
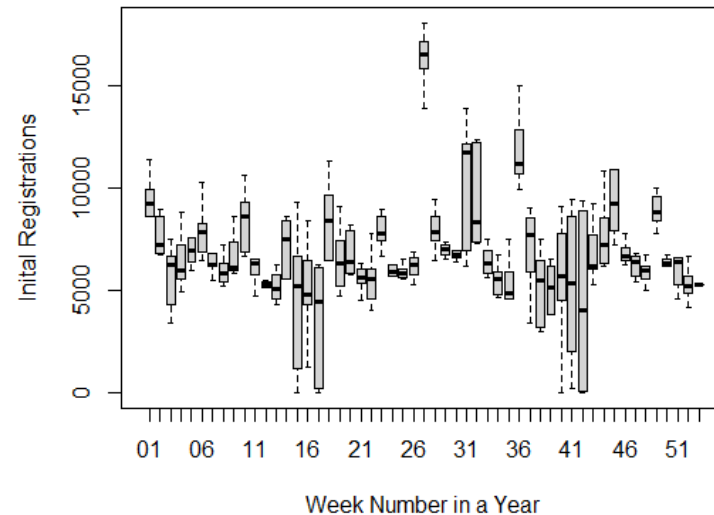
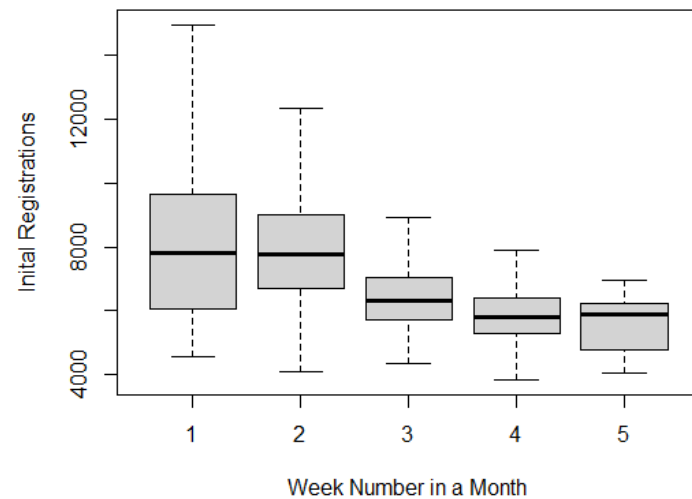
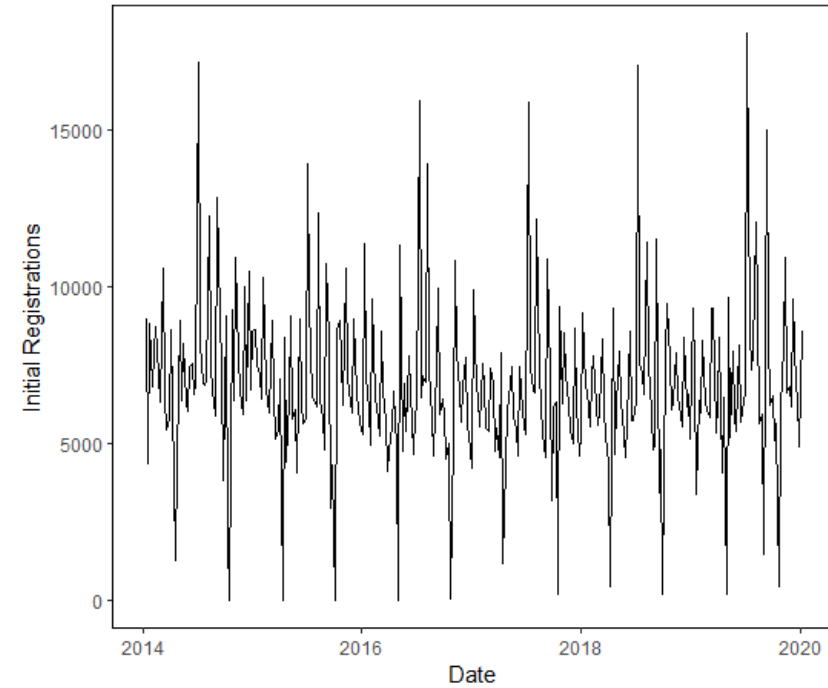
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Why can't we implement conventional statistical methods on weekly data?

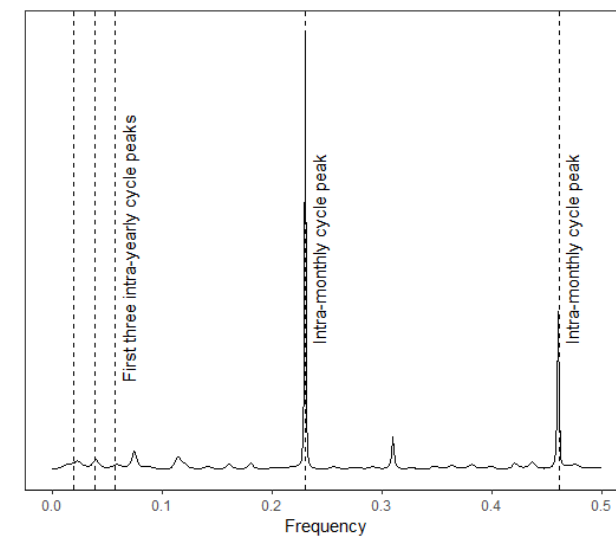
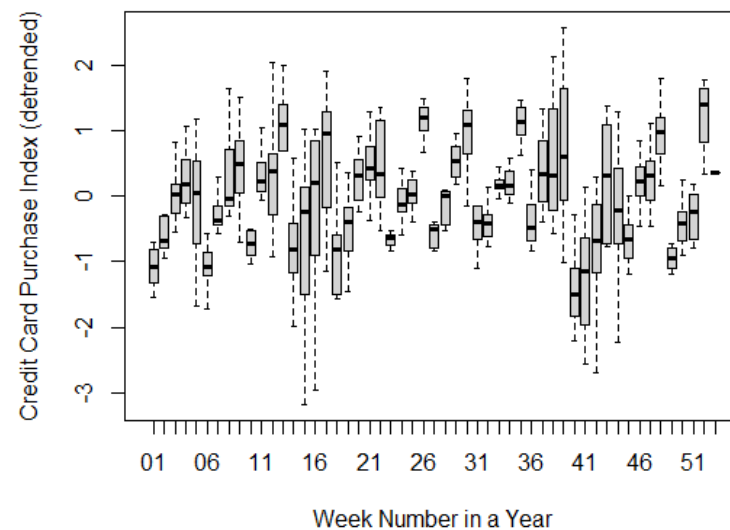
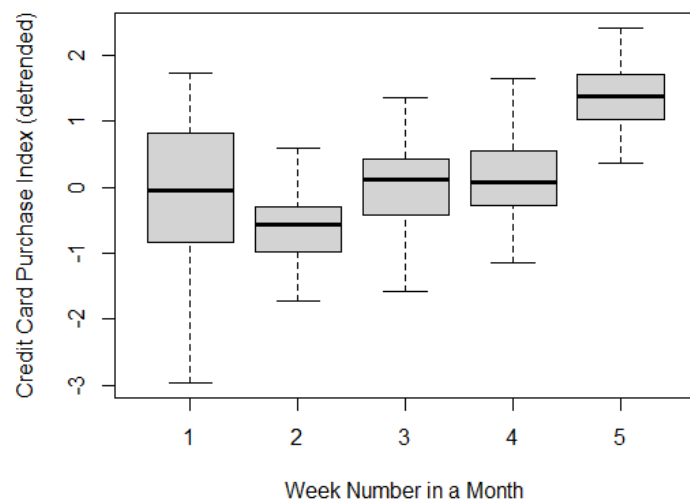
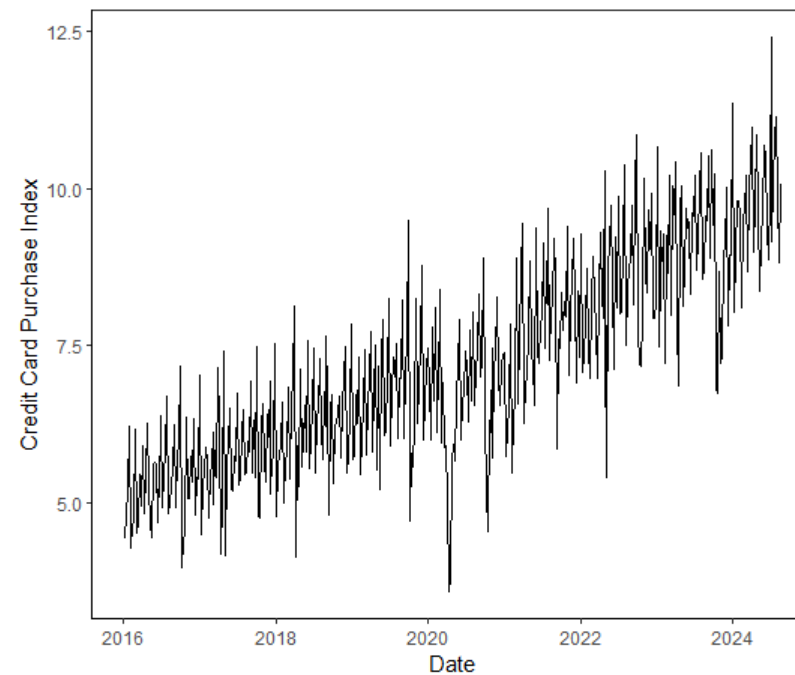
- Varying periodicity (number of weeks in a month/year)
- Multiple seasonal cycles – intra-monthly and intra-yearly cycles.
- Moving window - for weekly data, each week ends at a different position within a month.



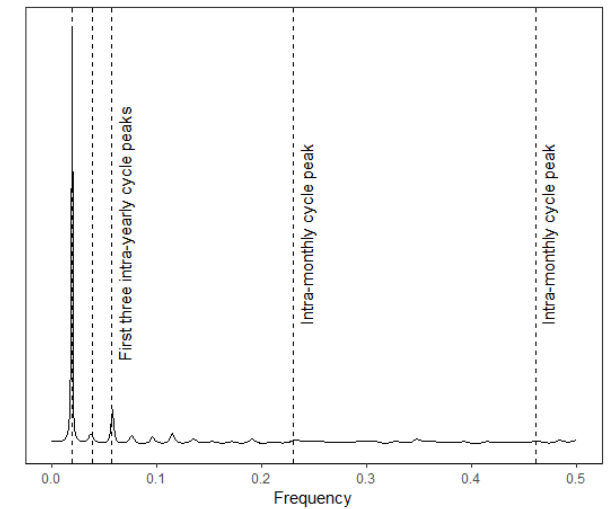
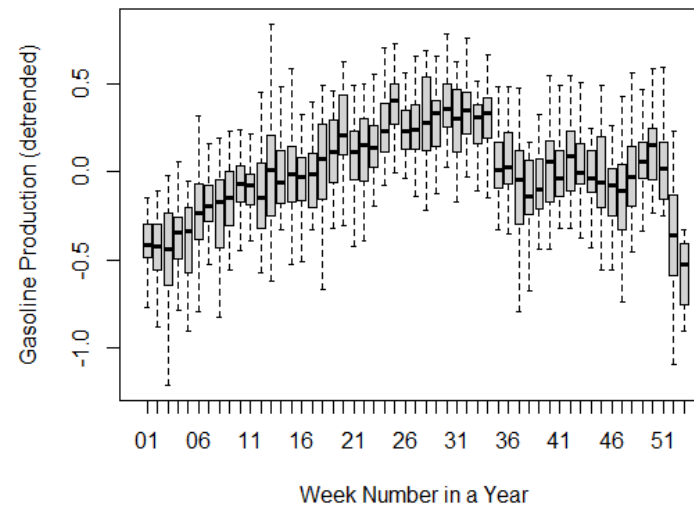
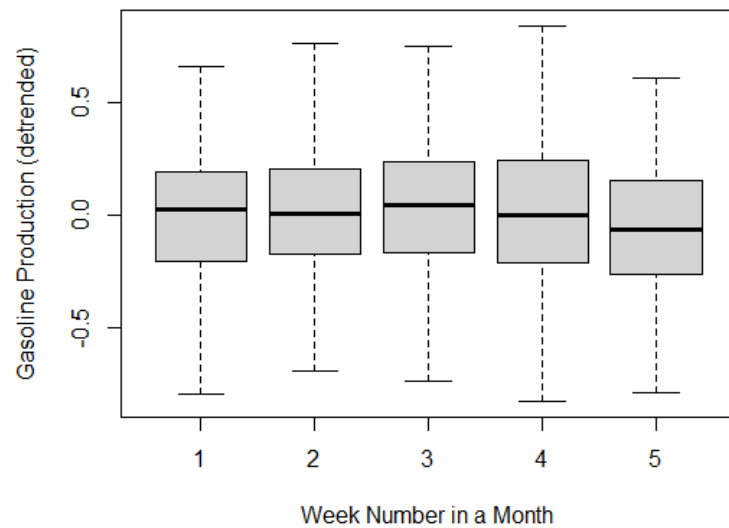
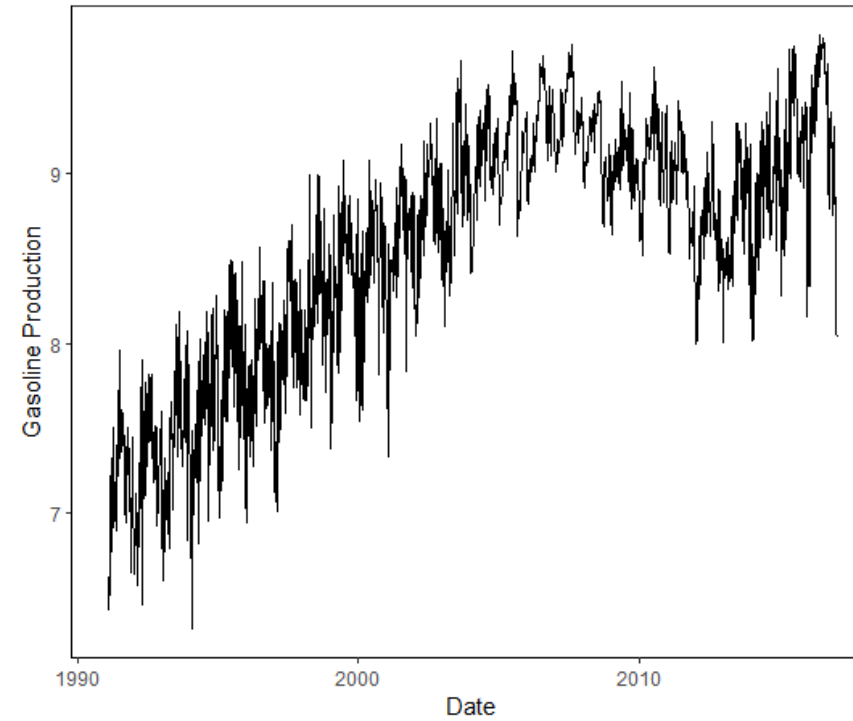
Weekly Number of Initial Registrations in the Israeli Employment Service



Weekly Credit Card Purchases in Israel





Gasoline Production in the US



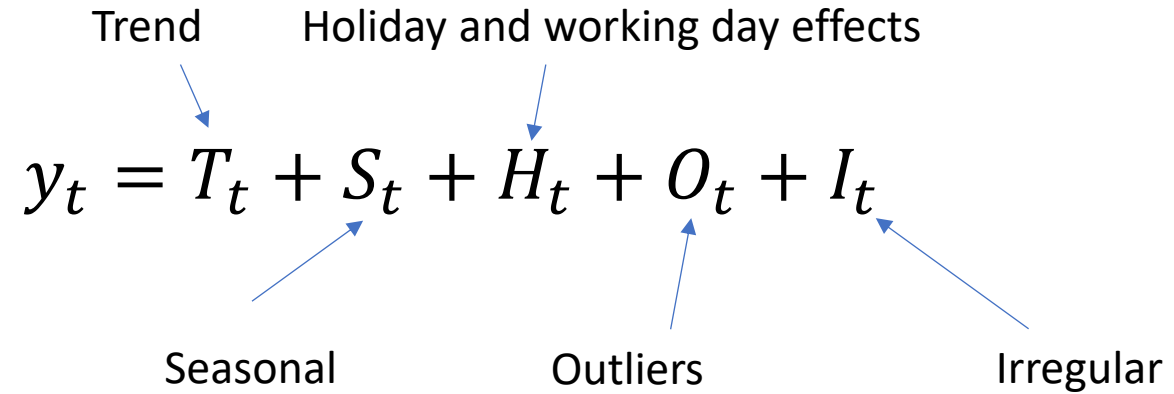
R programs, occasionally employed for seasonal adjustment of weekly data

- **Multiple Seasonal-Trend decomposition using Loess (MSTL)** (Bandara et al., 2021) 
- **Prophet** (Taylor and Letham, 2018) 
- **TBATS** (De Livera et al., 2011) 
- **stR** (Dokumentov and Hyndman, 2022) 

Software tools for the seasonal adjustment of weekly data

- MoveReg (SAS, Eviews) (Cleveland et al., 2014)
- FAM (Ecce Signum)  (McElroy and Livsey, 2022)
- boiwsa 😊 

Methodology



The diagram shows the equation $y_t = T_t + S_t + H_t + O_t + I_t$ with five descriptive labels and arrows pointing to each term: 'Trend' points to T_t , 'Seasonal' points to S_t , 'Holiday and working day effects' points to H_t , 'Outliers' points to O_t , and 'Irregular' points to I_t .

$$y_t = T_t + S_t + H_t + O_t + I_t$$

Trend

Holiday and working day effects

Seasonal

Outliers

Irregular

t denotes the date of the last day within a given week.

Methodology

Similarly to Cleveland et al. (2014), the seasonal component is modeled as

$$S_t = \sum_{k=1}^K \left(\alpha_k^y \sin \left(\frac{2\pi k D_t^y}{n_t^y} \right) + \beta_k^y \cos \left(\frac{2\pi k D_t^y}{n_t^y} \right) \right) + \sum_{l=1}^L \left(\alpha_l^m \sin \left(\frac{2\pi l D_t^m}{n_t^m} \right) + \beta_l^m \cos \left(\frac{2\pi l D_t^m}{n_t^m} \right) \right),$$

where D_t^y and D_t^m are the day of the year and the day of the month, and n_t^y and n_t^m are the number of days in the given year or month.

Thus, the seasonal adjustment procedure takes into account the existence of two cycles, namely intra-yearly and intra-monthly.

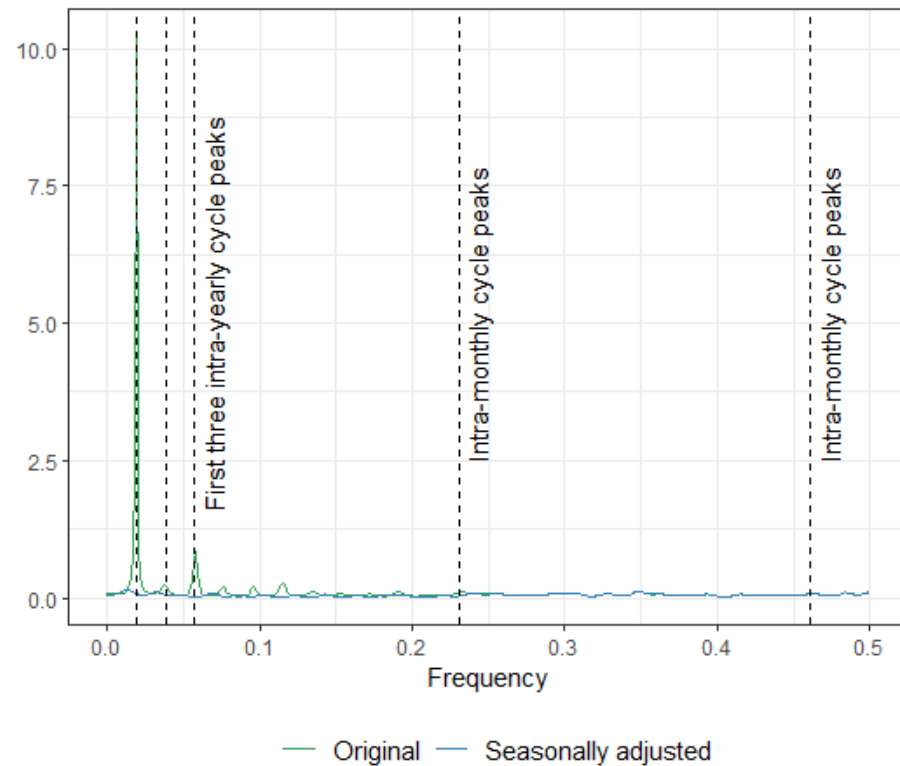
Methodology

Similarly to the X-11 method (Ladiray and Quenneville, 2001), the boiwsa procedure employs an iterative approach to estimate the different components. The seasonal adjustment algorithm comprises eight steps, which are detailed below:

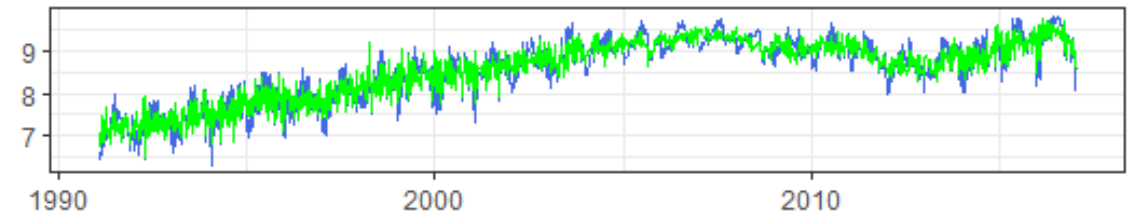
1. Estimation of trend using `stats::supsmu()`.
2. Estimation of the Seasonal-Irregular component; Searching for additive outliers; Identifying the optimal number of trigonometric variables.
3. Calculation of seasonal factors.
4. Estimation of trend from seasonally and outlier adjusted series.
5. Estimation of the Seasonal-Irregular component.
6. Computing the final seasonal factors.
7. Estimation of the final seasonally adjusted series.
8. Computing the final trend.

Gasoline Consumption in the US

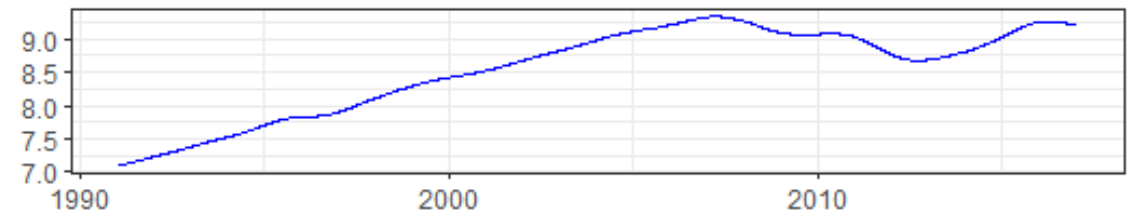
```
## {r}  
res=boiwsa(x=gasoline.data$y,  
          dates=gasoline.data$date)|  
##
```



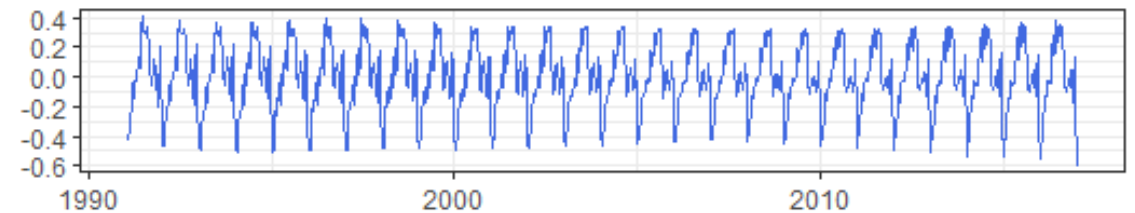
Original (blue) and Seasonally adjusted (green)



Trend



Seasonal



Weekly number of initial registrations in the Israeli Employment Service

```
```{r}

creating an input for simple_td
dates_il%>%
 dplyr::select(DATE_VALUE, ISR_WORKING_DAY_PART)%>%
 `colnames<-`(c("date", "WORKING_DAY_PART"))%>%
 dplyr::mutate(date=as.Date(date))>df.td
creating a matrix with a working day variable
td=simple_td(dates = lbm$date, df.td = df.td)

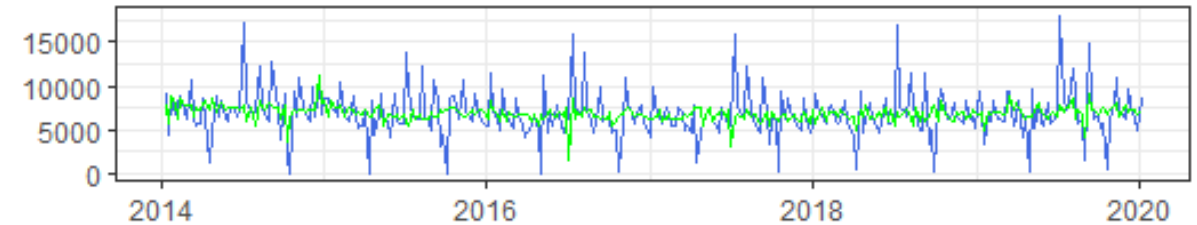
generating Rosh Hashanah and Pesach moving holiday variables
rosh=my_rosh(dates=lbm$date,
 holiday.dates = holiday_dates_il$rosh)
naming (make sure that all the variables in H have distinct names)
colnames(rosh)=paste0("rosh", colnames(rosh))

pesach=my_rosh(dates=lbm$date,
 holiday.dates = holiday_dates_il$pesach,
 start=3, end=-1)
colnames(pesach)=paste0("pesach", colnames(pesach))

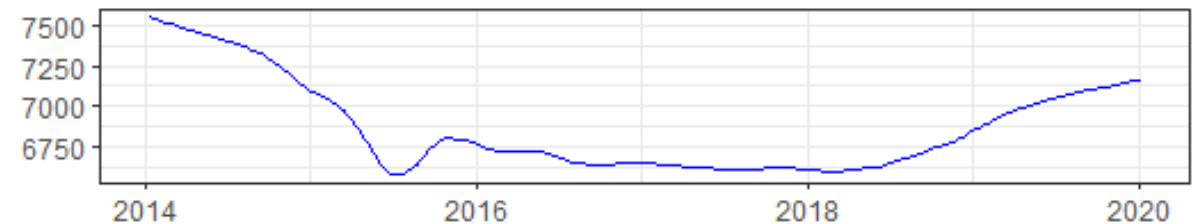
combining our working day and moving holiday variables
H=as.matrix(cbind(rosh[, -1], pesach[, -1], td[, -1]))
running seasonal adjustment routine
res=boiwsa(x=lbm$IES_IN_W_ADJ,
 dates = lbm$date,
 H=H,
 out.threshold = 3.8)

```
```

Original (blue) and Seasonally adjusted (green)



Trend

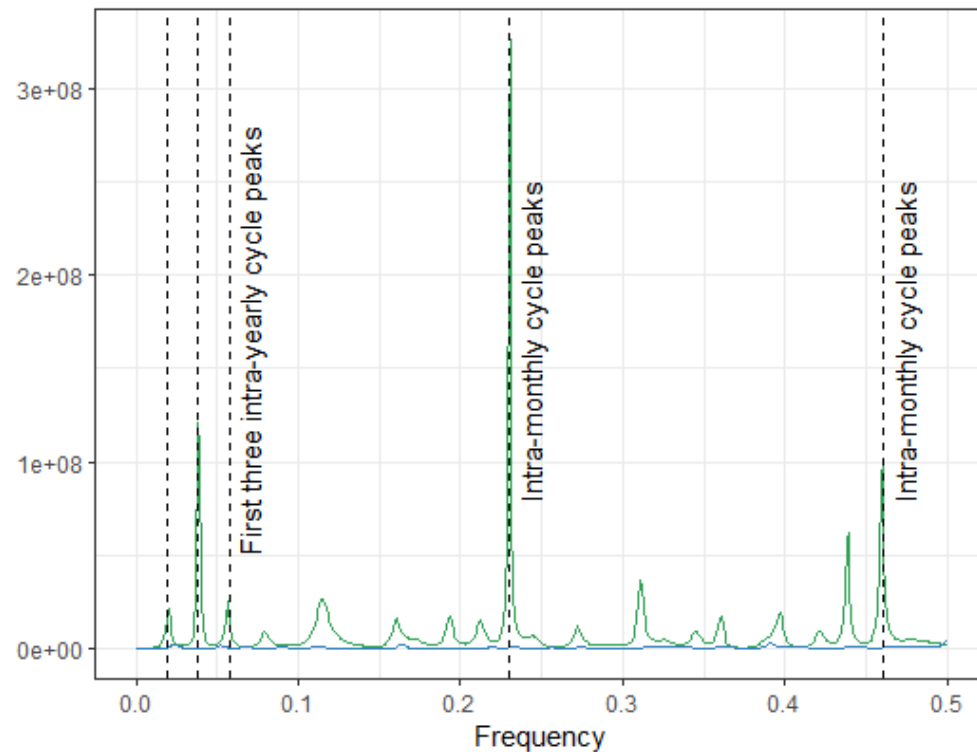


Seasonal



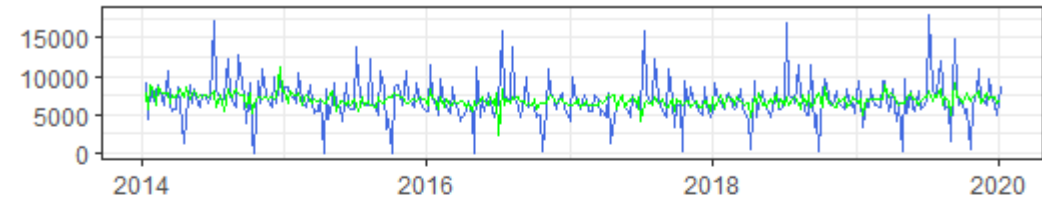
Weekly number of initial registrations in the Israeli Employment Service

```
res=boiwsa(x=lbm$IES_IN_W_ADJ,  
           dates = lbm$date,  
           H=H,  
           out.threshold = 5)
```

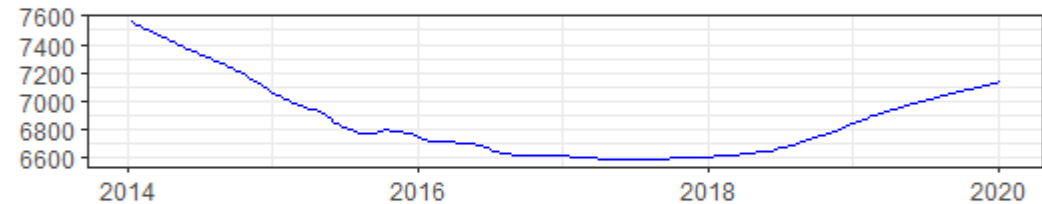


— Original — Seasonally adjusted

Original (blue) and Seasonally adjusted (green)



Trend



Seasonal



Weekly Credit Card Purchases

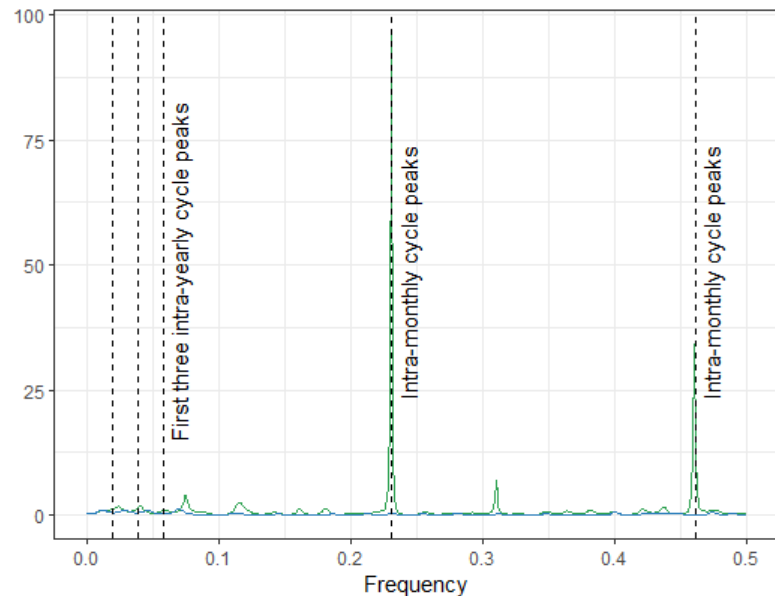
```
##{r}
pesach=genhol(dates = dates,holiday.dates = holiday_dates_il$pesach,start=0, end=15)
rosh=genhol(dates = dates,holiday.dates = holiday_dates_il$rosh,start=4, end=28)

dates_il%>%
  dplyr::select(DATE_VALUE,ISR_WORKING_DAY_PART)%>%
  `colnames`-<-`c("date","WORKING_DAY_PART")`%>%
  dplyr::mutate(date=as.Date(date))>df.td

td=simple_td(dates = dates,df.td = df.td)

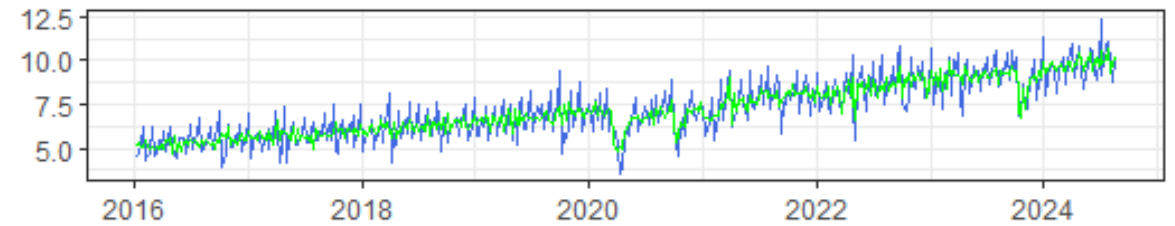
H=cbind(rosh[,-1],pesach[,-1],td[,-1])
colnames(H)=c("rosh","pesach","td")

res=boiwsa(x=y,
  dates=dates,
  H=as.matrix(H),
  ic = "bic",
  my.k_1 = c(6,12))
...
```



— Original — Seasonally adjusted

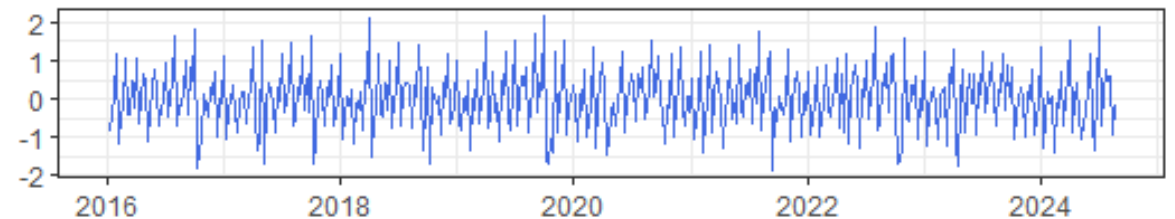
Original (blue) and Seasonally adjusted (green)



Trend



Seasonal



Thank You!