

# Seasonal adjustment of weekly data by discounted least squares in

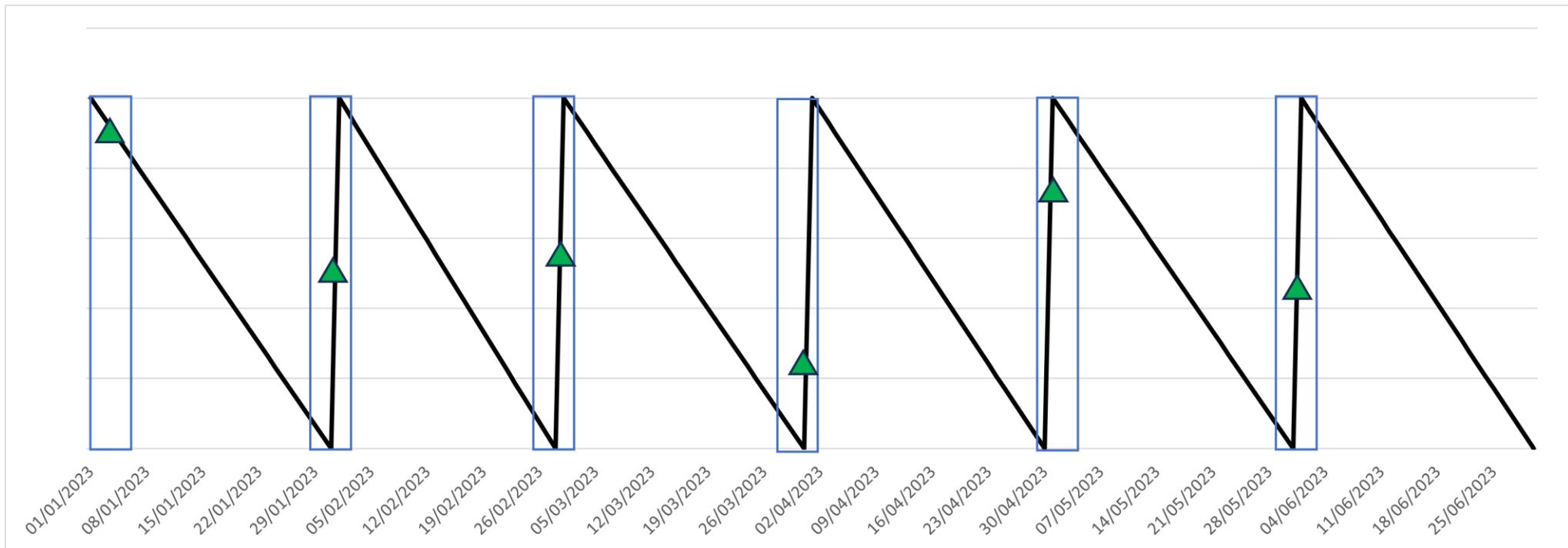
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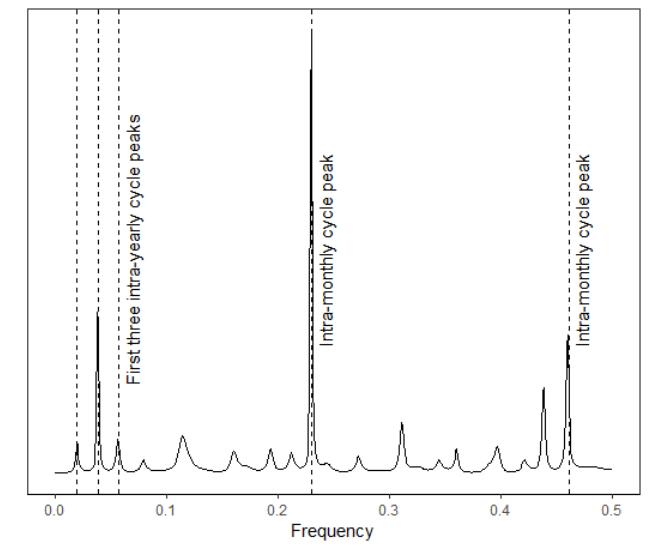
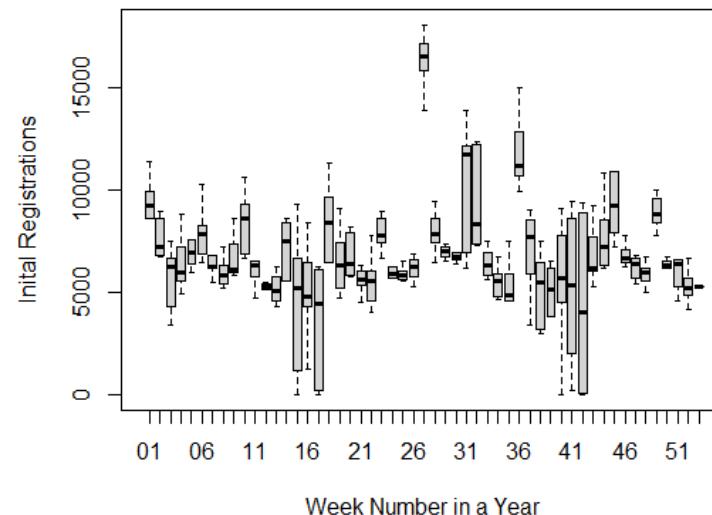
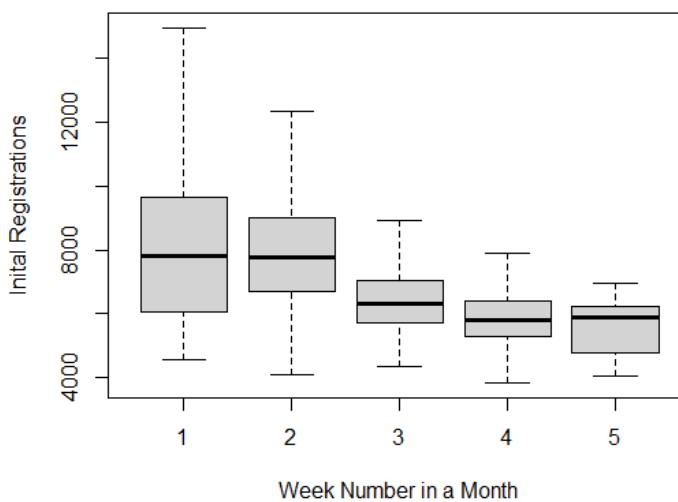
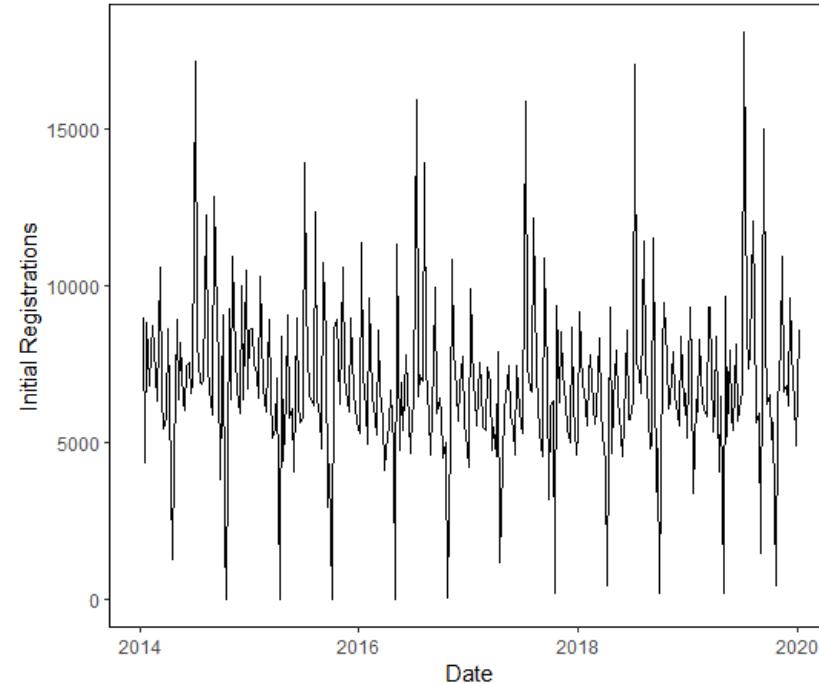
This presentation does not necessarily reflect the views of the Bank of Israel

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

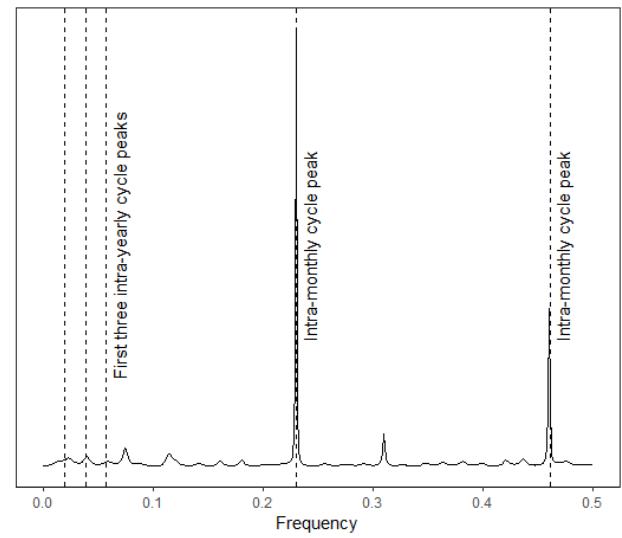
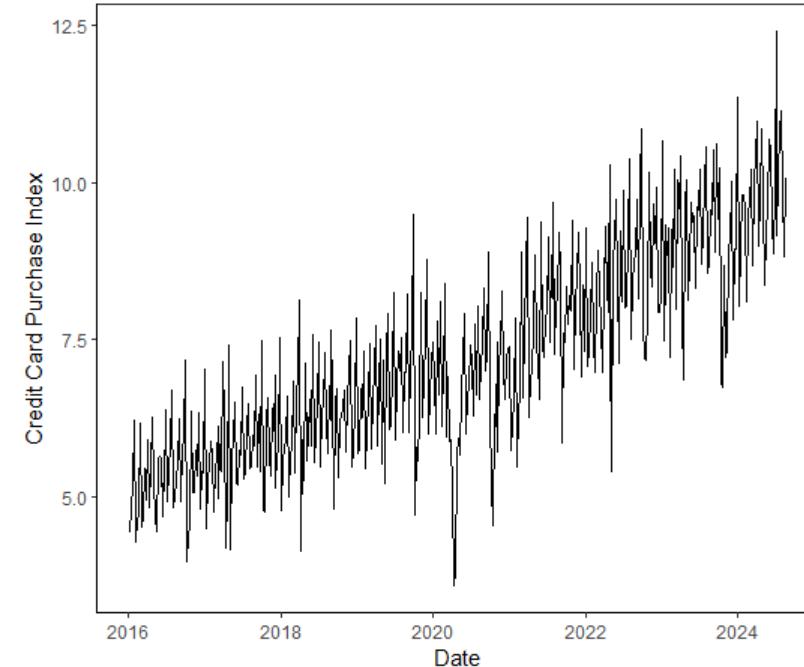
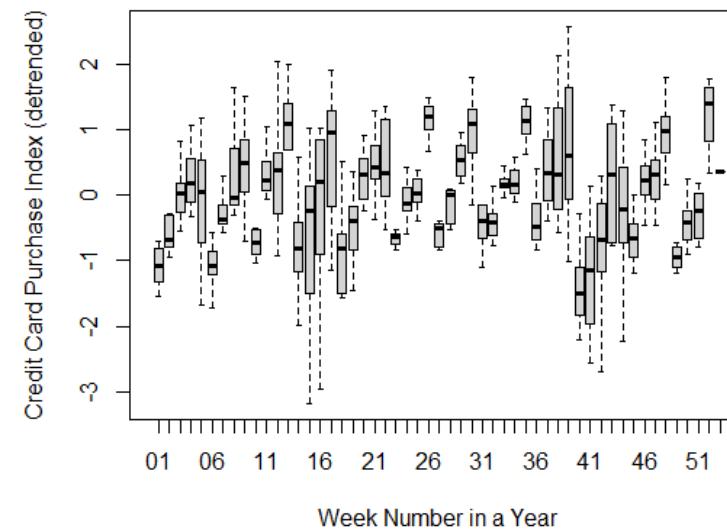
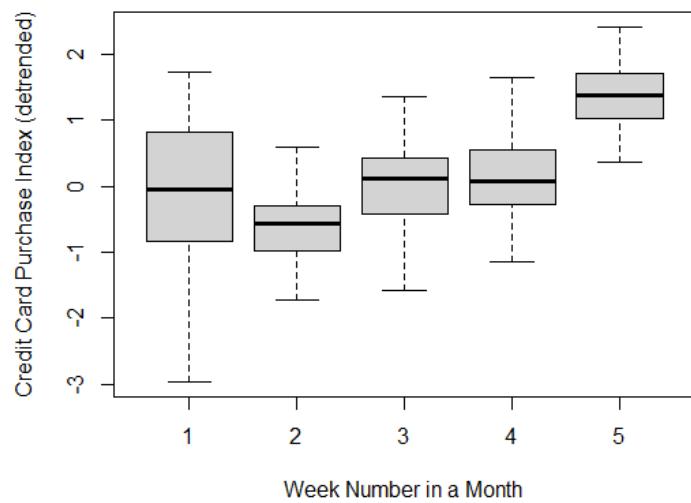
- Varying periodicity (number of weeks in a month/year)
- Multiple seasonal cycles – intra-monthly and intra-yearly cycles.
- Moving window - for weekly data, each week ends at a different position within a month.



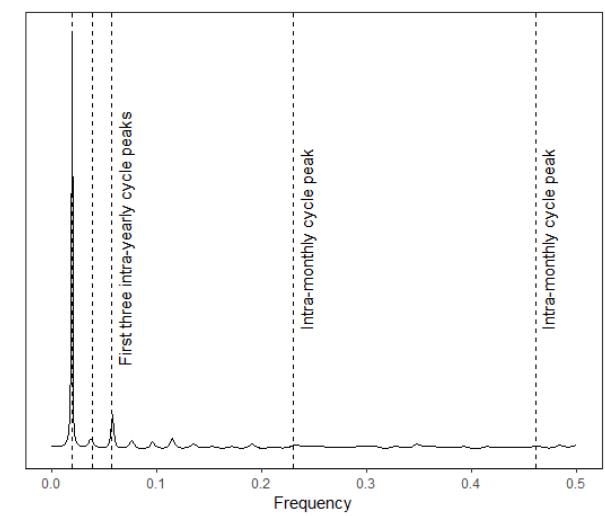
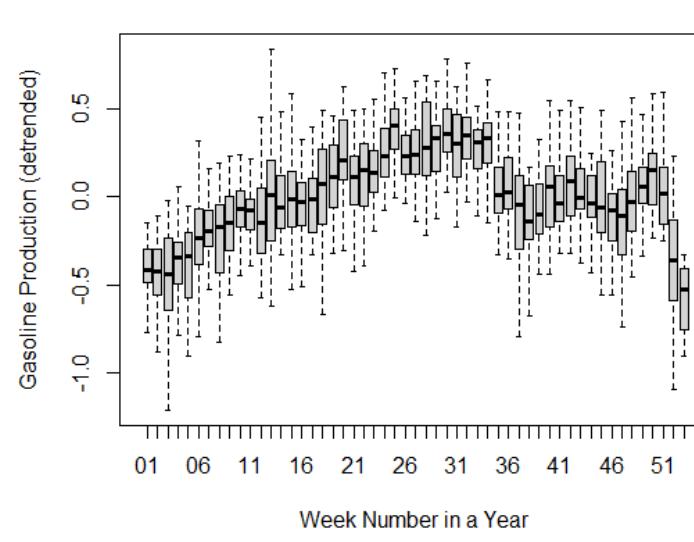
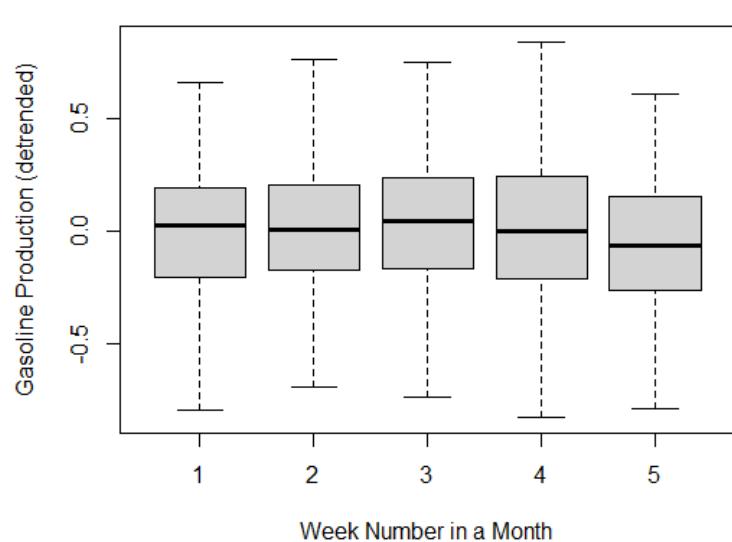
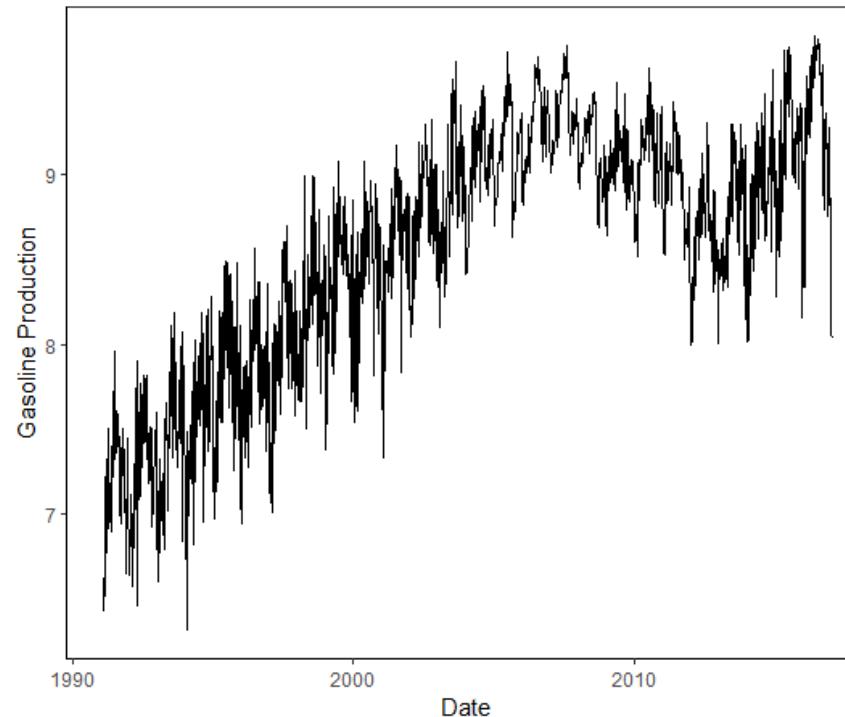
# Weekly Number of Initial Registrations in the Israeli Employment Service



# 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) 

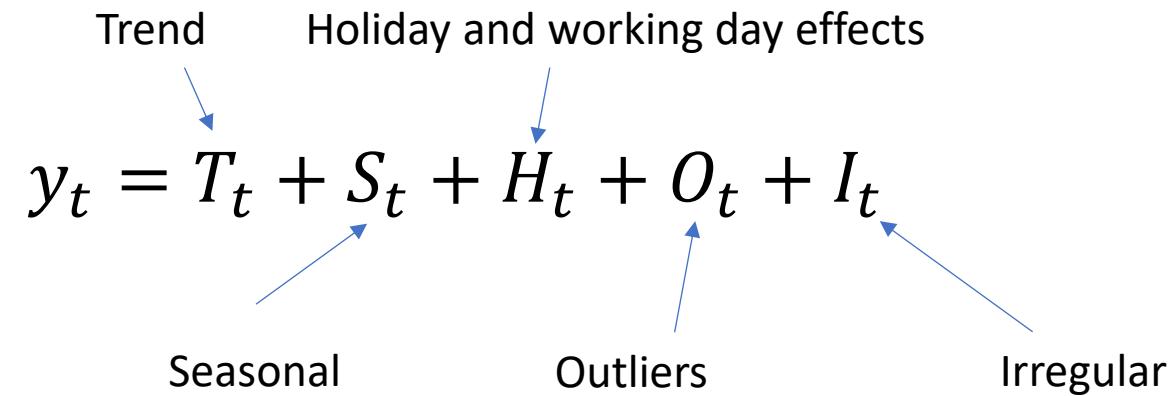
# Software tools for the seasonal adjustment of weekly data

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

# Methodology

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

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.

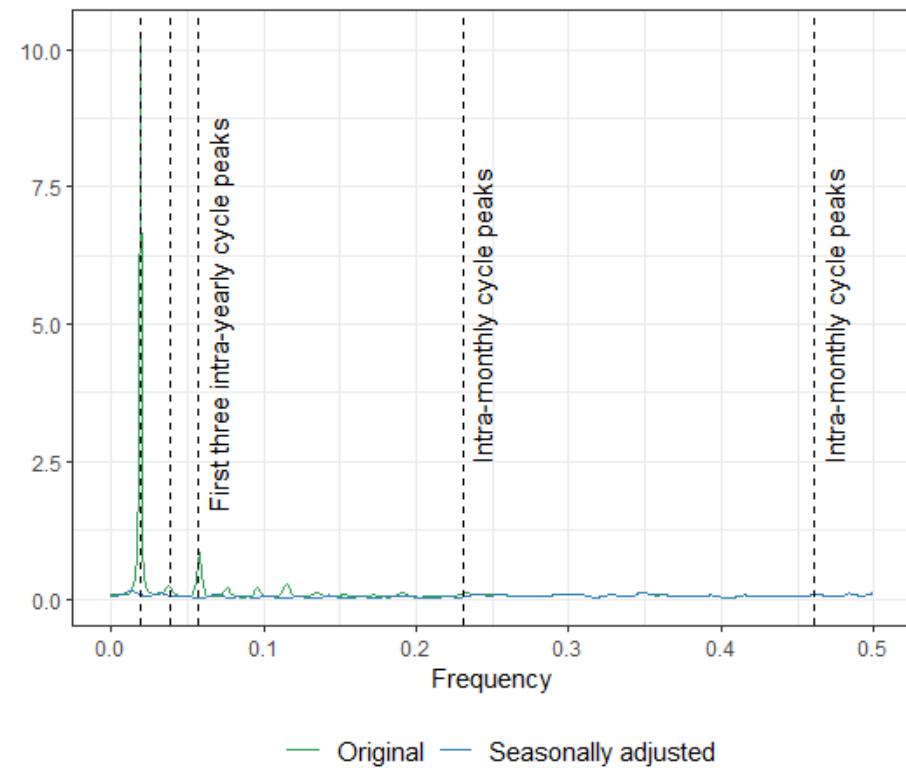
# Methodology

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:

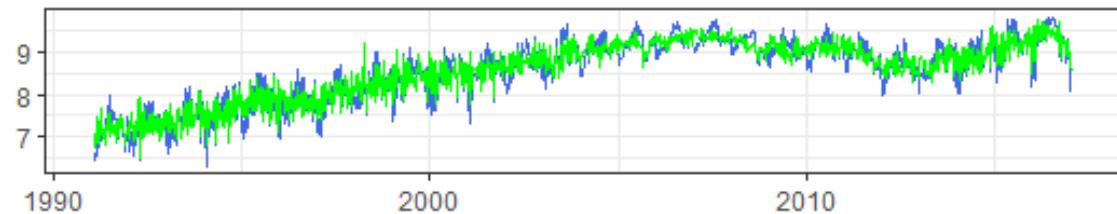
1. Estimation of trend using `stats::supsmu()`.
2. Estimation of the Seasonal-Irregular component; Searching for additive outliers; Identifying the optimal number of trigonometric variables.
3. Calculation of seasonal factors.
4. Estimation of trend from seasonally and outlier adjusted series.
5. Estimation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

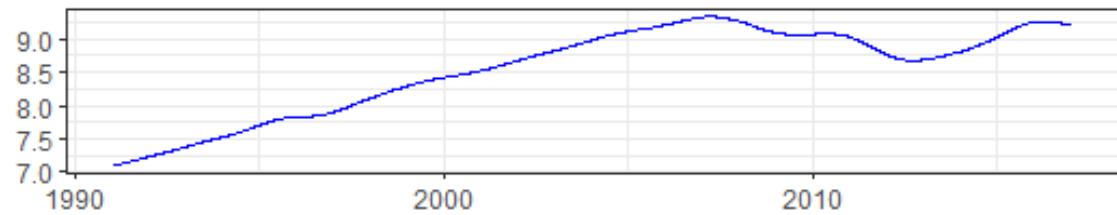
```
```{r}
res=boiwsa(x=gasoline.data$y,
            dates=gasoline.data$date)
```
```



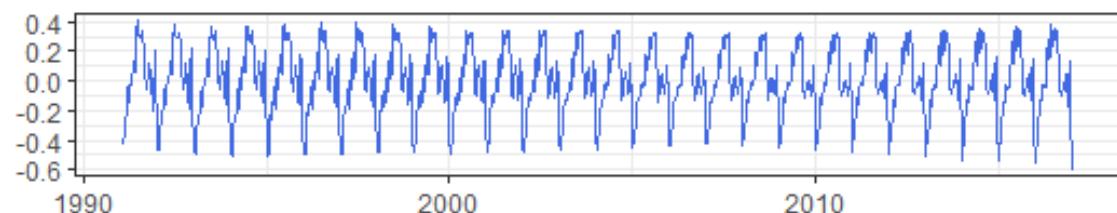
Original (blue) and Seasonally adjusted (green)



Trend



Seasonal



# Weekly number of initial registrations in the Israeli Employment Service

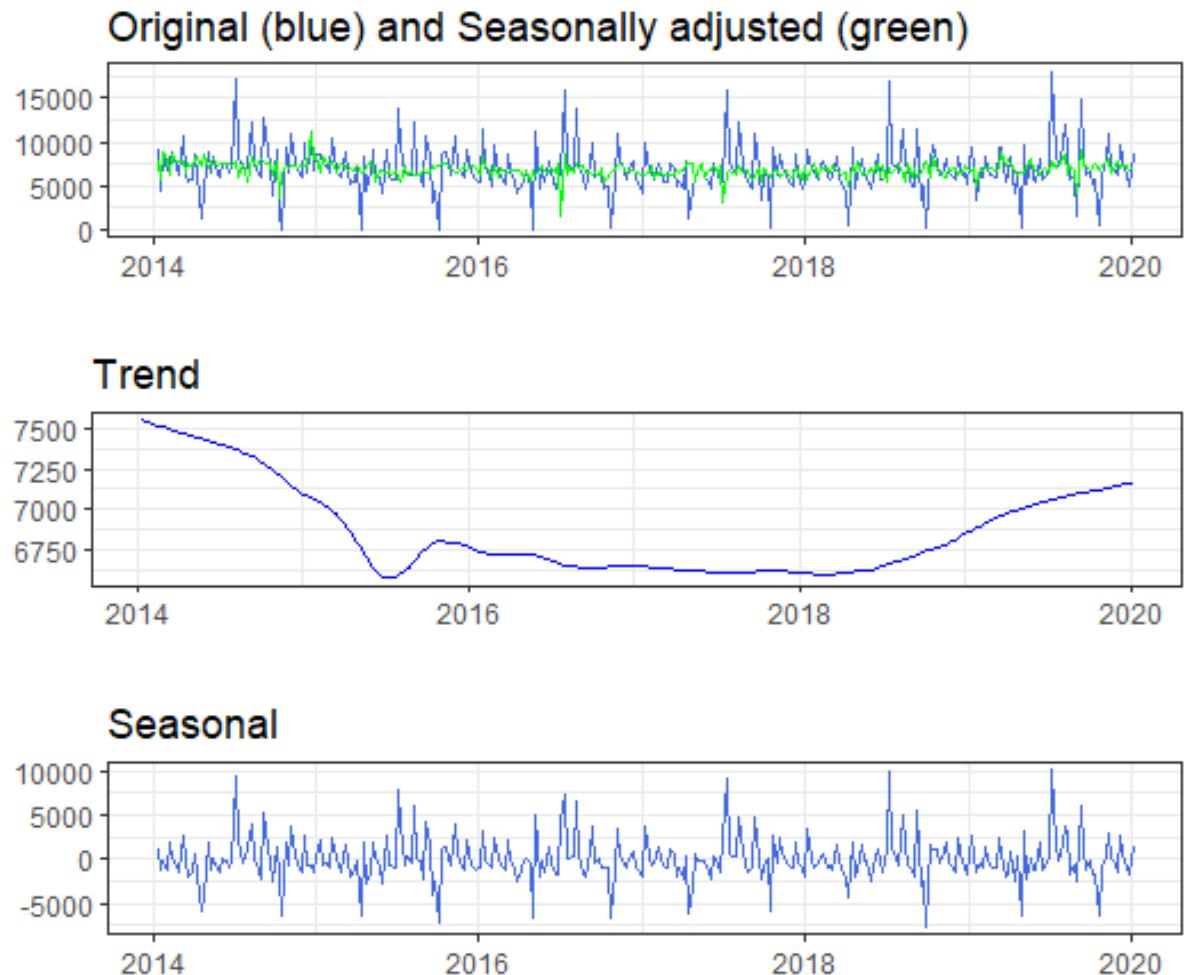
```
```{r}
# creating an input for simple_td
dates_il %>%
  dplyr::select(DATE_VALUE, ISR_WORKING_DAY_PART) %>%
  `colnames<-` (c("date", "WORKING_DAY_PART")) %>%
  dplyr::mutate(date=as.Date(date)) -> df_td
# creating a matrix with a working day variable
td=simple_td(dates = lbm$date, df_td = df_td)

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

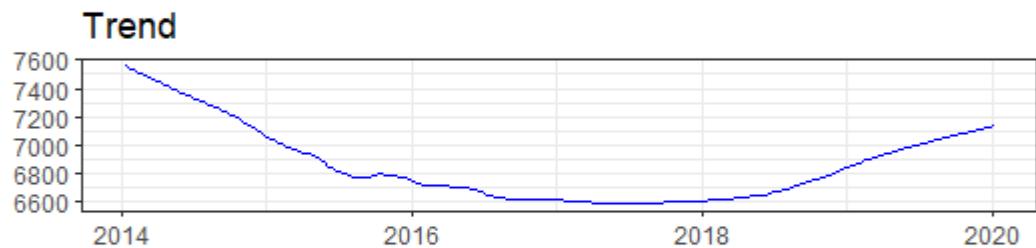
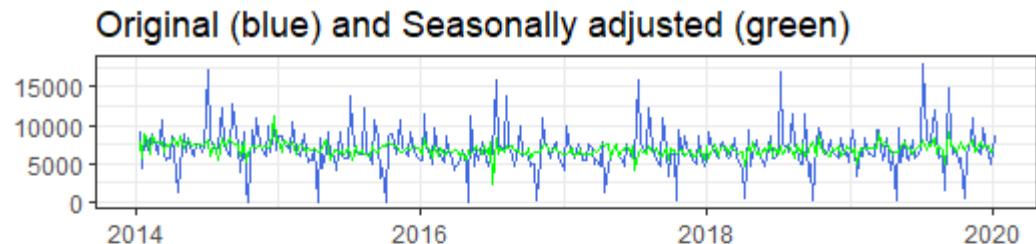
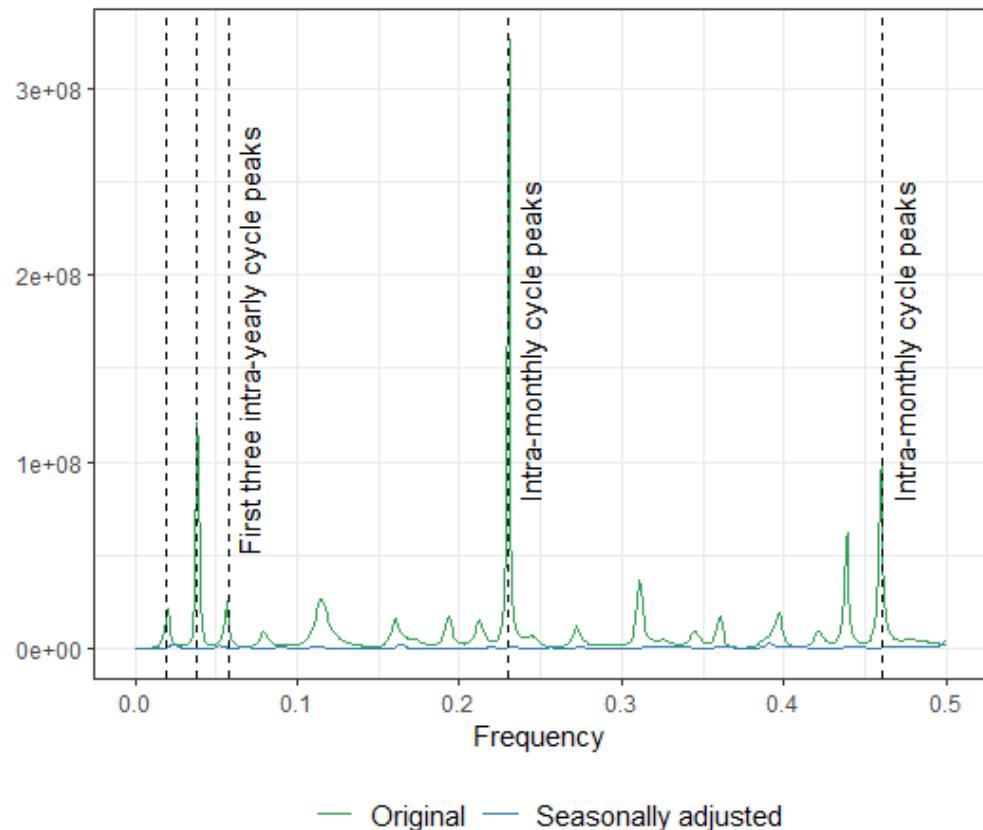
# combining our working day and moving holiday variables
H=as.matrix(cbind(rosh[,-1],pesach[,-1],td[,-1]))
# running seasonal adjustment routine
res=boiwsa(x=lbm$IES_IN_W_ADJ,
            dates = lbm$date,
            H=H,
            out.threshold = 3.8)
...```

```



# Weekly number of initial registrations in the Israeli Employment Service

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



# Weekly Credit Card Purchases

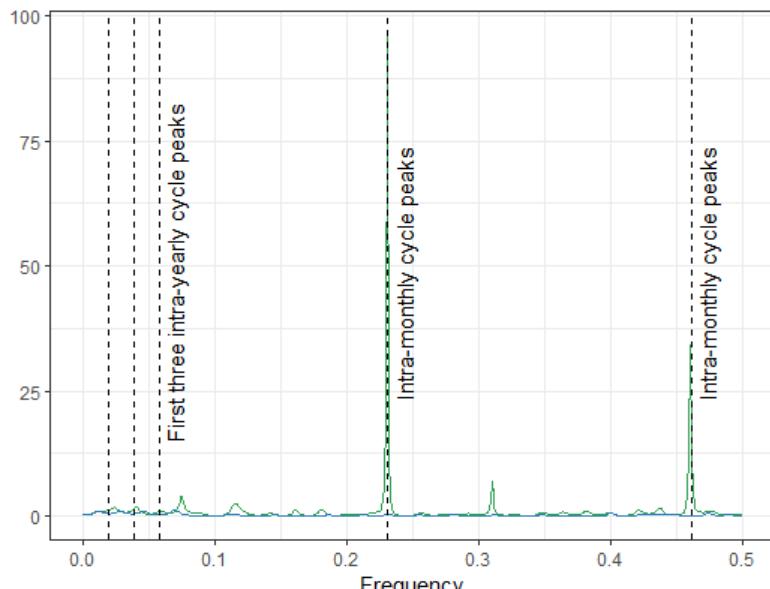
```
```{r}
pesach=genhol(dates = dates,holiday.dates = holiday_dates_il$pesah,start=0, end=15)
rosh=genhol(dates = dates,holiday.dates = holiday_dates_il$rosh,start=4, end=28)

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)

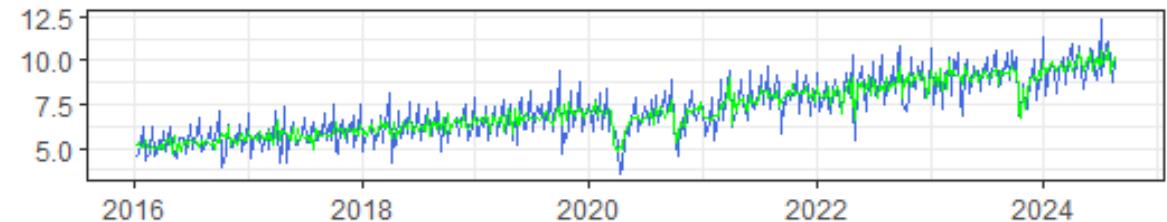
H=cbind(rosh[,-1],pesach[,-1],td[,-1])
colnames(H)=c("rosh","pesach","td")

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



— Original — Seasonally adjusted

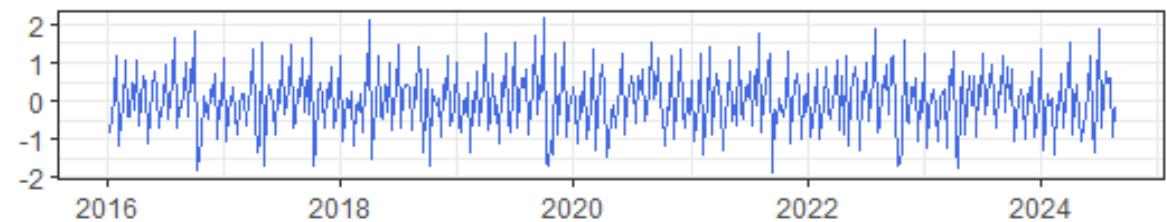
Original (blue) and Seasonally adjusted (green)



Trend



Seasonal



Thank You!