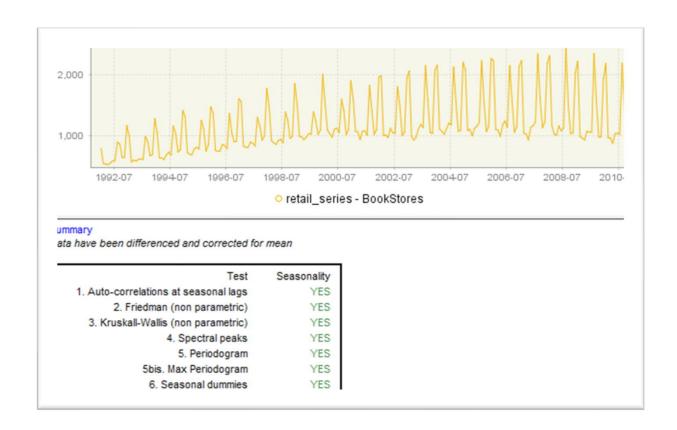


Seasonality and trading days tests in JD+

ESTP training

Overview





Seasonality tests

Non parametric tests

- Friedman
 - ANOVA-type. The test uses the rankings of the observations within each year. It does not require distributional assumptions.
 - H0: all periods can be treated equally (= no seasonality), H1: series is seasonal
 - P-value < 0.01 ⇒ H0 is rejected
 - Applied on series without trend (for instance (log) differenced series)
- Kruskall-Wallis
 - The test uses the rankings of all the observations. It does not require distributional assumptions.
 - H0: All periods have the same mean(average ranking), H1: series is seasonal
 - Applied on series without trend (could work on any series)

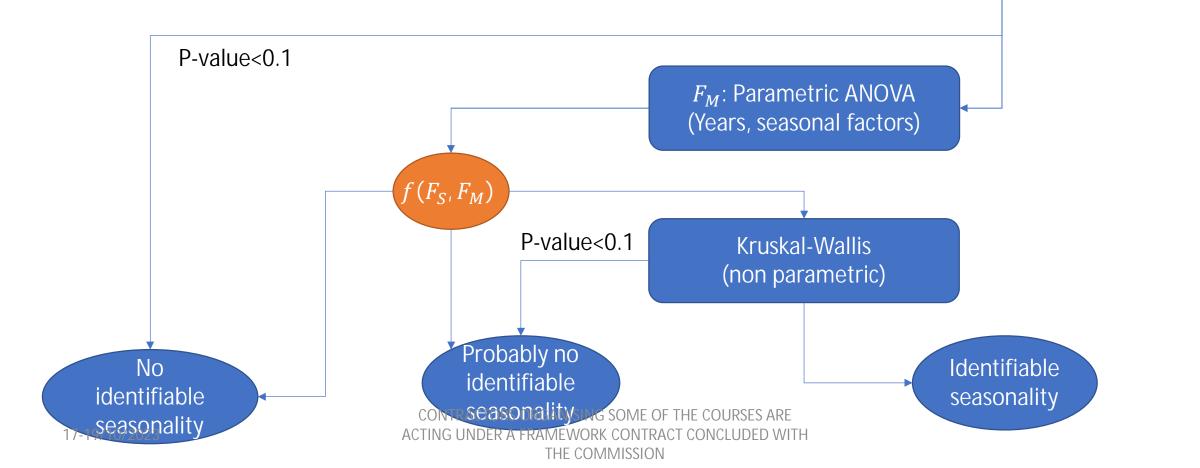
Seasonality tests (cont.)

- QS test
 - Ljung-box on seasonal auto-correlations
 - H0: No correlation (no seasonality), H1: seasonality
 - Applied on (log) differenced series (no trend), corrected for mean effect
- F test on seasonal dummies
 - Regression test with seasonal dummies $(D_{s,t})$ and different models
 - $y_t y_{t-1} = D_{s,t}\beta + \varepsilon_t + \theta \varepsilon_{t-1}$ (GUI) • $y_t = \alpha + \gamma y_{t-1} + D_{s,t}\beta + \varepsilon_t$ (diagnostics) • $y_t - y_{t-1} = \alpha + D_{s,t}\beta + \varepsilon_t$ (diagnostics)
 - H0: coefficients (β) are equal to 0, H1: coefficients jointly differ from 0
 - Applied on (log) series. Should be applied on parts of the series

Seasonality tests (cont.)

• Combined seasonality test (X11-like)

F_s: Parametric ANOVA (identical seasonal factors)



Seasonality tests (cont.)

- Spectral diagnostics
 - Spectral peaks (X13-like)
 - Based on the spectrum of a long auto-regressive model that fits the series (see X13 documentation)
 - Periodogram
 - Performed on the periodogram (Fourier transformation) of the series at the Fourier frequencies.
 - <u>maximum</u> of the periodogram on or around the seasonal frequencies
 - <u>sum</u> of the values of the periodogram on or around the seasonal frequencies
 - Strictly speaking, only valid against the hypothesis that the (transformed) series is a white noise. As the spectral peaks, they don't perform well for short series.

Trading days tests

- F test on default trading days variables (contrasts)
 - Regression test with trading days contrasts (TD_t) and different models

•
$$y_t = \alpha + \gamma y_{t-1} + TD_t\beta + \varepsilon_t$$
 (2.2.x)
• $y_t - y_{t-1} = \alpha + TD_t\beta + \varepsilon_t$ (2.2.x)
• $\Delta(y_t - TD_t\beta) = \alpha + \varepsilon_t$ (3.0)
• $\Delta(y_t - TD_t\beta) = \varepsilon_t + \theta\varepsilon_{t-1}$ (3.0)
• More generally: $y_t = \alpha + TD_t\beta + arima_t$

- H0: coefficients (β) are equal to 0, H1: coefficients jointly differ from 0
- Applied on (log) series. Should be applied on parts of the series

Seasonality tests in R

```
s<-rjd3toolkit::retail$RetailSalesTotal
ls<-log(s)
st<-rjd3toolkit::do_stationary(ls, 12)
dls<-st$ddata
spec.pgram(dls)
spec.ar(dls)
print(rjd3toolkit::seasonality_qs(dls, 12))
# H0: the series has no seasonality
# pvalue = prob[x>T]
# pvalue nearly 0 -> w reject H0
print(rjd3toolkit::seasonality_kruskalwallis(dls, 12))
print(rjd3toolkit::seasonality_friedman(dls, 12))
print(rjd3toolkit::seasonality_f(ls, 12, "D1"))
```

Trading days tests in R

```
#s<-rid3toolkit::ABS$X0.2.20.10.M
s<-rid3toolkit::retail$RetailSalesTotal
td_all<-function(s, title, len=length(s)/12){
 a<-sapply(6:len, function(j){rjd3toolkit::td_f(s, model = "D1", nyears = j)$pvalue})
 b<-sapply(6:len, function(j){rjd3toolkit::td_f(s, model = "R011", nyears = j)$pvalue})
 c<-sapply(6:len, function(j){rjd3toolkit::td_f(s, model = "R100", nyears = j)$pvalue})</pre>
 matplot(main=title, x=6:len, y=cbind(a,b,c), pch=19, col=c('black', 'blue', 'red'),
          xlab="number of years", ylab="p-value")
# seasonal adjustment with calendar effects correction
sa<-rjd3tramoseats::tramoseats_fast(s, "rsafull")</pre>
# seasonal adjustment without calendar effects correction
sa2<-rjd3tramoseats::tramoseats_fast(log(s), "rsa0")
ssa<-sa$decomposition$stochastics$sa$data
ssa2<-sa2$decomposition$stochastics$sa$data
td_all(ssa, "sa - rasafull")
td_all(ssa2, "sa - rsa0")
sirr<-sa$decomposition$stochastics$i$data
sirr2<-sa2$decomposition$stochastics$i$data
td_all(sirr, "irregular - rsafull")
td_all(sirr2, "irregular - rsa0")
```

