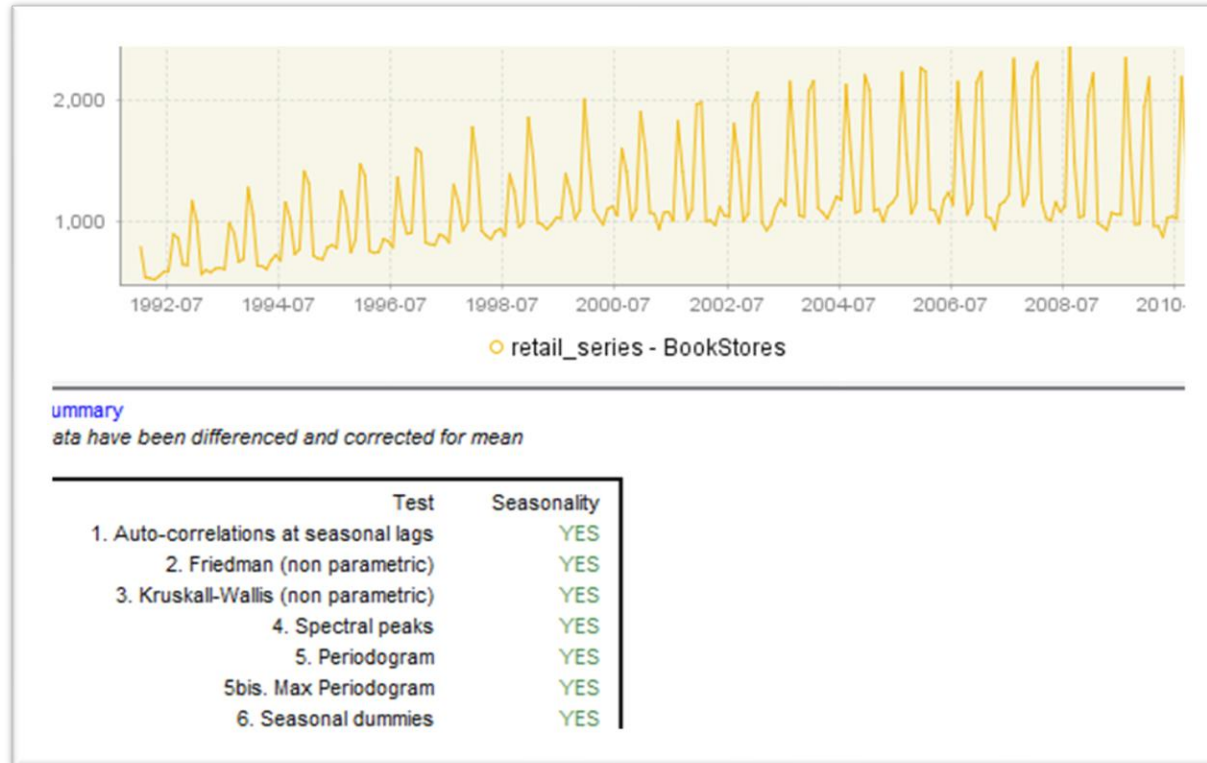


# Seasonality and trading days tests in JD+

## 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.
    - $H_0$ : all periods can be treated equally (= no seasonality),  $H_1$ : series is seasonal
    - $P\text{-value} < 0.01 \Rightarrow H_0$  is rejected
    - Applied on series without trend (for instance (log) differenced series)

- **Kruskal-Wallis**

- The test uses the rankings of all the observations. It does not require distributional assumptions.
    - $H_0$ : All periods have the same mean(average ranking),  $H_1$ : series is seasonal
    - Applied on series without trend (for instance (log) differenced series)

# Seasonality tests (cont.)

- **QS test**

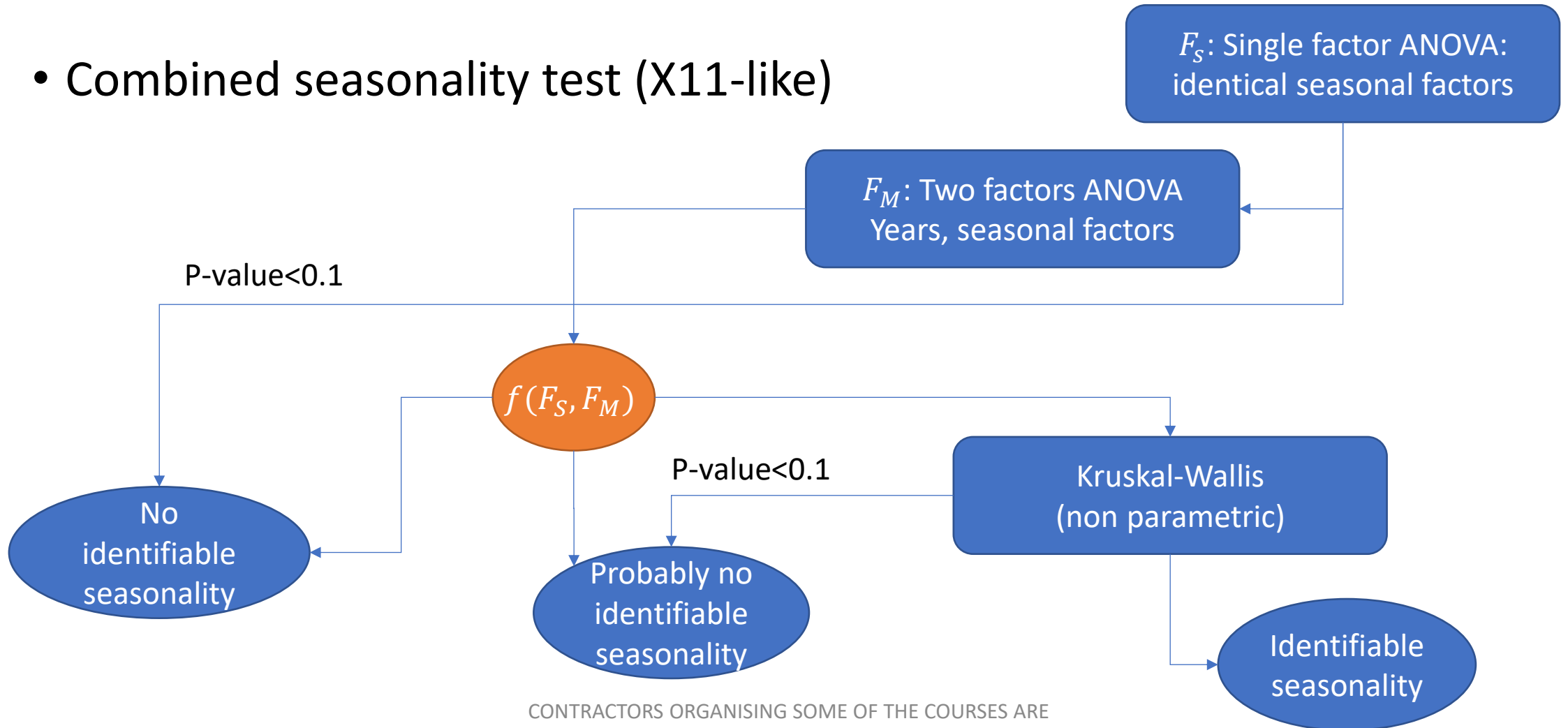
- Modified Ljung-box on seasonal auto-correlations
- H0: No correlation (no seasonality), H1: seasonality
- Applied on (log) differenced series (no trend), corrected for mean effect

- **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 ( $\beta$ ) 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)



# 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.
      - maximum of the periodogram on or around the seasonal frequencies
      - sum 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.

# Seasonality tests (cont.)

- **Canova-Hansen test (R only)**

- test the stability of the coefficients of seasonal regressors (model quite similar to the F-test), using a Lagrange multiplier test.
- $H_0$ : stable seasonal pattern (no seasonality  $\equiv$  stable 0 coefficients).
- $H_1$ : seasonal unit roots.
- P-values computed numerically for small samples.

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



# Trading days tests(cont.)

- **Canova-Hansen test** on trading days variables **(R only)**
  - test the stability of the coefficients of trading days, using a Lagrange multiplier test.
- Likelihood test on time varying trading days coefficients **(R only)**
  - rjd3toolkit >= 3.2.5

# 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<-rjd3toolkit::ABS$X0.2.20.10.M
s<-rjd3toolkit::retail$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})

  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")
# 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 - rsafull")
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")
```

