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RJDemetra: an R interface to JDemetra+

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Introduction to seasonal adjustment

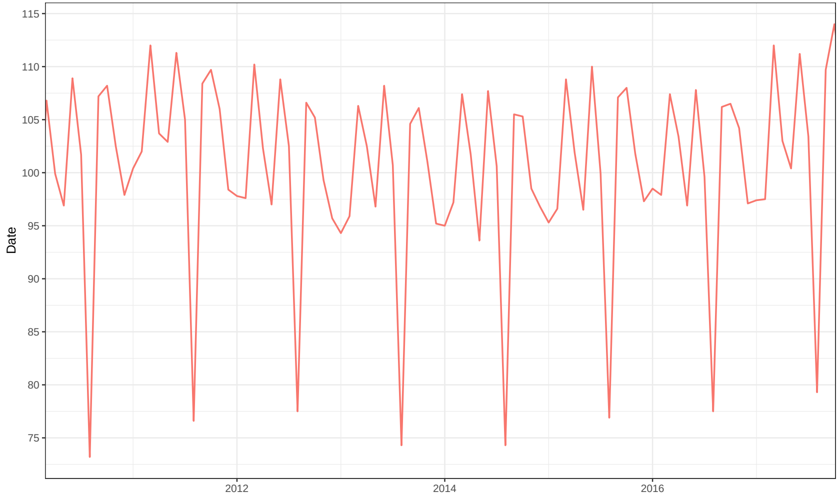

Figure 1: Industrial production index in France

Introduction to seasonal adjustment (2/3)

Purpose of seasonal adjustment:

- Time comparison (outlook, short-term evolution. . .)
- Spatial comparison

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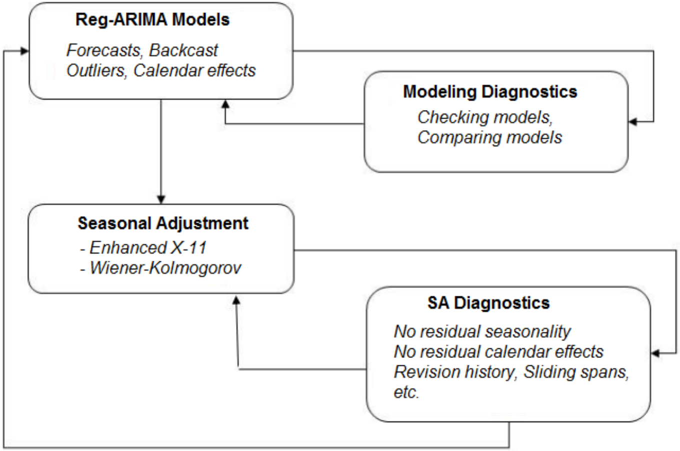
$\rightarrow$ proceed in two steps

Introduction to seasonal adjustment (3/3)

1. Pre-adjusting the series of deterministics effects with a RegARIMA model
2. Decomposition: to extract seasonal component



What's JDemetra+ ?

Time Series Software for Official Statistics TRAMO/SEATS+ and X-13ARIMA-SEATS are implemented in JDemetra+

Software officially recommended by Eurostat and the ECB for seasonal and calendar adjustment of official statistics

→ RJDemetra is an interface to JDemetra+ based on the libraries of JD+

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Current status

- RegARIMA, TRAMO-SEATS and X-13-ARIMA:
  - pre-defined and user-defined specifications
  - S3 classes with plot, summary, print methods
- Manipulate JD+ workspaces:
  - Import JD+ workspace to get: input raw series or SA model
  - Export R models created via RJDemetra
- Include a dataset: industrial production indices in manufacturing in the European Union

RegARIMA examples (1/3)

```
library(RJDemetra)
ipi_fr <- ipi_c_eu[,"FR"]
regarima_model <- regarima_def_x13(ipi_fr, spec = "RG4c")
regarima_model
```

```
## y = regression model + arima (2, 1, 1, 0, 1, 1)
## Log-transformation: no
## Coefficients:
##           Estimate Std. Error
## Phi(1)      0.3358      0.171
## Phi(2)      0.2060      0.096
## Theta(1)   -0.2450      0.173
## BTheta(1)  -0.5112      0.050
##
##              Estimate Std. Error
## Easter [1]     -1.133      0.337
## LS (11-2008)   -8.000      1.283
## LS (1-2009)    -7.551      1.283
## LS (5-2008)    -5.069      1.234
##
##
## Residual standard error: 1.696 on 323 degrees of freedom
## Log likelihood =  -631, aic =  1280 aicc =  1281, bic(corrected for length) =
```

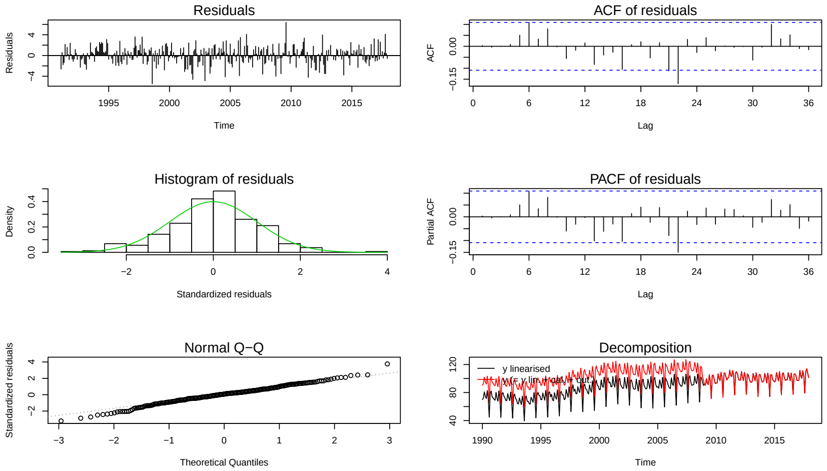
RegARIMA examples (2/3)

```
summary(regarima_model)
```

```
## y = regression model + arima (2, 1, 1, 0, 1, 1)
##
## Model: RegARIMA - X13
## Estimation span: from 1-1990 to 12-2017
## Log-transformation: no
## Regression model: no mean, no trading days effect, no leap year effect, Easte
##
## Coefficients:
## ARIMA:
##            Estimate Std. Error  T-stat Pr(>|t|)
## Phi(1)      0.33579    0.17106   1.963   0.0505 .
## Phi(2)      0.20600    0.09643   2.136   0.0334 *
## Theta(1)   -0.24498    0.17272  -1.418   0.1571
## BTheta(1) -0.51123    0.05004 -10.216   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Regression model:
##               Estimate Std. Error T-stat Pr(>|t|)
## Easter [1]     -1.1332     0.3373 -3.359 0.000875 ***
## LS (11-2008)   -7.9997     1.2831 -6.235 1.42e-09 ***
```

RegARIMA examples (3/3)

```
layout(matrix(1:6, 3, 2));plot(regarima_model, ask = FALSE)
```



Seasonal adjustment examples (1/7)

A `SA` object is a `list()` of 5 elements:

```
SA
 ├─ regarima (≠ X-13 and TRAMO-SEAT)
 │  ├─ specification
 │  └─ ...
 ├─ decomposition (≠ X-13 and TRAMO-SEAT)
 │  ├─ specification
 │  └─ ...
 ├─ final
 │  ├─ series
 │  └─ forecasts
 ├─ diagnostics
 │  ├─ variance_decomposition
 │  ├─ combined_test
 │  └─ ...
 └─ user_defined
```

Seasonal adjustment examples (2/7)

Like in JD+ users can defined their own specification or use a pre-defined one:

```
x13_usr_spec <- x13_spec_def(spec = c("RSA5c"),
                             usrdef.outliersEnabled = TRUE,
                             usrdef.outliersType = c("LS", "AO"),
                             usrdef.outliersDate = c("2008-10-01",
                                                     "2002-01-01"),
                             usrdef.outliersCoef = c(36, 14),
                             transform.function = "None")
x13_mod <- x13(ipi_fr, x13_usr_spec)
ts_mod <- tramoseats_def(ipi_fr, spec = "RSAfull")
```

Seasonal adjustment examples (3/7): decomposition

```
x13_mod$decomposition
```

```
##  Monitoring and Quality Assessment Statistics:
##       M stats
## M(1)    0.055
## M(2)    0.041
## M(3)    0.926
## M(4)    0.621
## M(5)    0.724
## M(6)    0.215
## M(7)    0.074
## M(8)    0.208
## M(9)    0.056
## M(10)   0.158
## M(11)   0.146
## Q       0.297
## Q-M2    0.329
##
## Final filters:
## Seasonal filter:  3x5
## Trend filter:  13 terms Henderson moving average
```

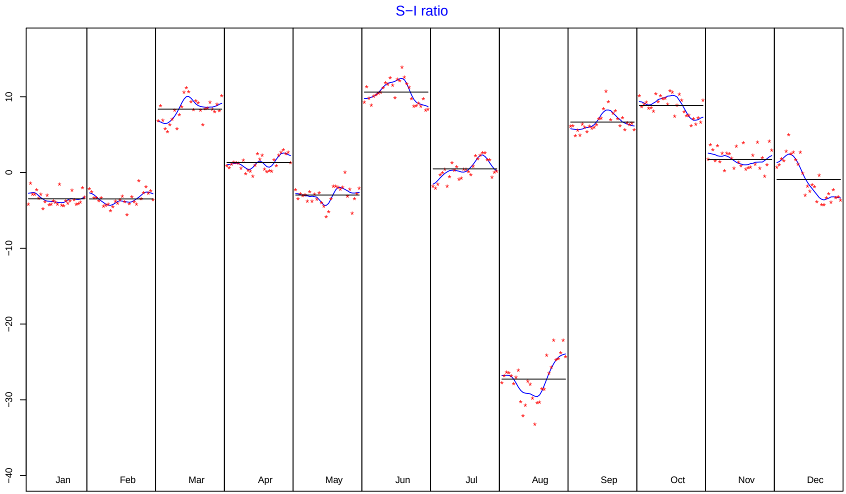
Seasonal adjustment examples (4/7): decomposition

```
ts_mod$decomposition
```

```
## Model
## AR :  1 + 0.352498 B + 0.133616 B^2
## D :  1 - B - B^12 + B^13
## MA :  1 - 0.186819 B - 0.610856 B^12 + 0.114119 B^13
##
##
## SA
## D :  1 - 2.000000 B + B^2
## MA :  1 - 1.314459 B + 0.340427 B^2
## Innovation variance:  0.4669153
##
## Trend
## D :  1 - 2.000000 B + B^2
## MA :  1 + 0.040206 B - 0.959794 B^2
## Innovation variance:  0.04869563
##
## Seasonal
## AR :  1 + 0.352498 B + 0.133616 B^2
## D :  1 + B + B^2 + B^3 + B^4 + B^5 + B^6 + B^7 + B^8 + B^9 + B^10 + B^11
## MA :  1 + 0.717848 B + 0.460721 B^2 + 0.310085 B^3 + 0.132447 B^4 - 0.049053
## Innovation variance:  0.1601924
```

Seasonal adjustment examples (5/7)

```
plot(x13_mod$decomposition)
```



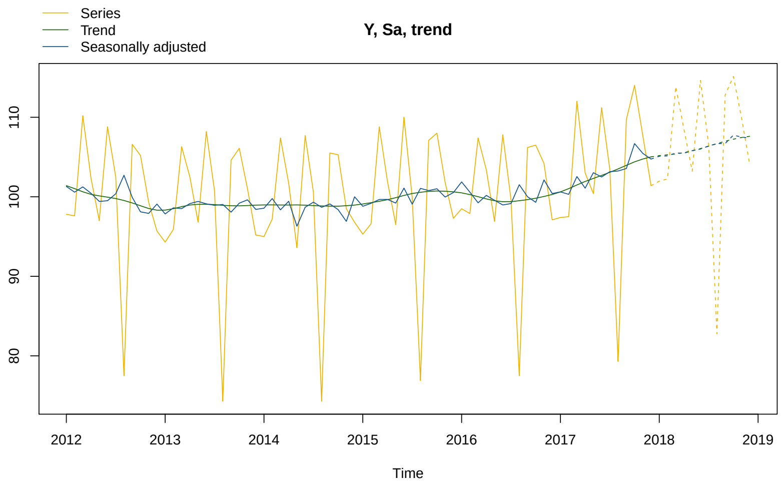
Seasonal adjustment examples (6/7)

```
x13_mod$final
```

```
## Last observed values
##                   y       sa        t          s            i
## Jan 2017  97.4 100.6172 100.6174  -3.2172329 -0.0001992082
## Feb 2017  97.5 100.3127 101.0283  -2.8126932 -0.7155966863
## Mar 2017 112.0 102.5469 101.4894   9.4530696  1.0575376567
## Apr 2017 103.0 101.0897 101.9282   1.9103111 -0.8385432983
## May 2017 100.4 103.0319 102.3136  -2.6318733  0.7182480125
## Jun 2017 111.2 102.4926 102.6921   8.7074293 -0.1994894034
## Jul 2017 103.4 103.1596 103.0816   0.2404277  0.0779236963
## Aug 2017  79.3 103.2483 103.5055 -23.9483256 -0.2572170473
## Sep 2017 109.7 103.5536 103.9555   6.1464361 -0.4019376040
## Oct 2017 114.0 106.6886 104.3955   7.3113786  2.2931579296
## Nov 2017 107.7 105.4631 104.7505   2.2369236  0.7125546908
## Dec 2017 101.4 104.7490 105.0214  -3.3490189 -0.2723590878
##
## Forecasts:
##                     y_f      sa_f        t_f          s_f          i_f
## Jan 2018 101.96630 105.0963 105.1795  -3.1299775 -0.083200162
## Feb 2018 102.23632 105.1464 105.2838  -2.9100563 -0.137428535
## Mar 2018 113.85794 105.5026 105.3966   8.3553336  0.105971540
## Apr 2018 108.47477 105.4896 105.5573   2.9851827 -0.067754048
```

Seasonal adjustment examples (7/7)

```
plot(x13_mod$final, first_date = 2012, type_chart = "sa-trend")
```



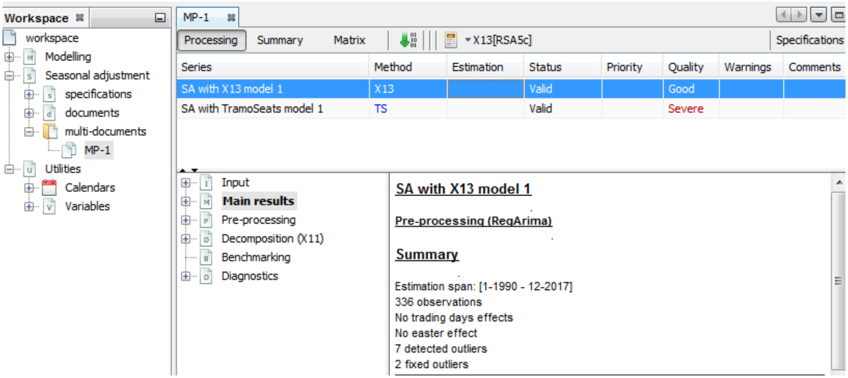
Seasonal adjustment examples (8/7)

```
x13_mod$diagnostics
```

```
##  Relative contribution of the components to the stationary
##  portion of the variance in the original series,
##  after the removal of the long term trend
##  Trend computed by Hodrick-Prescott filter (cycle length = 8.0 years)
##            Component
##  Cycle         1.557
##  Seasonal     39.219
##  Irregular     0.362
##  TD & Hol.     0.018
##  Others       61.971
##  Total       103.128
##
##  Combined test in the entire series
##  Non parametric tests for stable seasonality
##                                                           P.value
##    Kruskall-Wallis test                                     0.000
##    Test for the presence of seasonality assuming stability  0.000
##    Evolutive seasonality test                               0.032
##
##  Identifiable seasonality present
##
```

Export a workspace

```
wk <- new_workspace()
new_multiprocessing(wk, name = "MP-1")
add_sa_item(wk, multiprocessing = "MP-1",
            sa_obj = x13_mod, name =  "SA with X13 model 1 ")
add_sa_item(wk, multiprocessing =  "MP-1",
            sa_obj = ts_mod, name = "SA with TramoSeats model 1")
save_workspace(wk, "workspace.xml")
```



Series	Method	Estimation	Status	Priority	Quality	Warnings	Comments
SA with X13 model 1	X13		Valid		Good		
SA with TramoSeats model 1	TS		Valid		Severe		

SA with X13 model 1

Pre-processing (RegArima)

Summary

Estimation span: [1-1990 - 12-2017]
336 observations
No trading days effects
No easter effect
7 detected outliers
2 fixed outliers

Import a workspace (1/3)

```r
wk <- load_workspace("workspace.xml")
get_ts(wk)
```

```
## $`MP-1`
## $`MP-1`$`SA with X13 model 1 `
##        Jan   Feb   Mar   Apr   May   Jun   Jul   Aug   Sep   Oct   Nov
## 1990  90.5  92.6 101.9  95.2  92.1 103.3  91.8  65.5  99.0 102.8  94.3
## 1991  90.9  89.6  99.9  93.3  88.3 103.0  89.7  65.1  98.2 100.8  95.8
## 1992  89.4  89.0  99.5  93.0  89.1 101.3  89.4  64.1  94.9  98.6  92.2
## 1993  85.3  84.3  93.2  87.8  83.5  95.4  86.2  60.1  92.1  95.8  88.1
## 1994  84.9  84.0  94.1  90.1  86.8 100.4  90.8  64.5  96.8 101.0  96.6
## 1995  90.4  90.5 100.4  94.5  89.7 103.7  93.8  65.5  99.7 101.8  94.6
## 1996  90.3  88.8 100.7  93.8  91.2 104.4  92.3  67.2 100.2 102.3  96.9
## 1997  90.5  91.6 104.0  99.7  93.9 108.8  98.2  73.4 105.8 111.8 102.4
## 1998  99.2  99.0 109.4 103.0 100.7 114.8 104.9  73.3 109.6 112.7 105.9
## 1999 100.5  98.6 111.8 104.3 101.3 117.4 106.6  74.9 113.4 118.2 110.9
## 2000 104.8 104.9 118.9 110.2 108.0 122.5 111.8  80.5 117.5 121.7 114.3
## 2001 108.8 109.2 123.7 111.8 108.4 124.7 111.1  84.2 117.8 121.0 111.6
## 2002 106.6 107.0 121.4 112.8 106.4 122.2 109.7  82.3 117.1 118.7 113.0
## 2003 105.4 105.7 120.1 111.1 102.8 118.3 108.8  78.7 115.9 119.9 110.8
## 2004 105.8 107.0 120.0 112.1 105.8 123.6 112.0  78.4 120.0 122.0 112.0
## 2005 109.1 106.7 117.9 113.5 106.8 122.3 110.3  80.0 121.4 118.4 115.2
## 2006 107.3 106.3 121.9 112.5 110.8 126.7 112.5  82.5 122.2 121.9 113.7
                                                                  24.
                                                                  .6 3
```

Import a workspace (2/3)

```r
compute(wk) # Important to get the Sa model
models <- get_model(wk) # A progress bar is printed by default
```

```
## Multiprocessing 1 on 1:
##
  |
  |                                                                  |   0%
  |
  |=================================                                 |  50%
  |
  |==================================================================| 100%
```

```r
# To extract only one model
mp <- get_object(wk, 1)
count(mp)
```

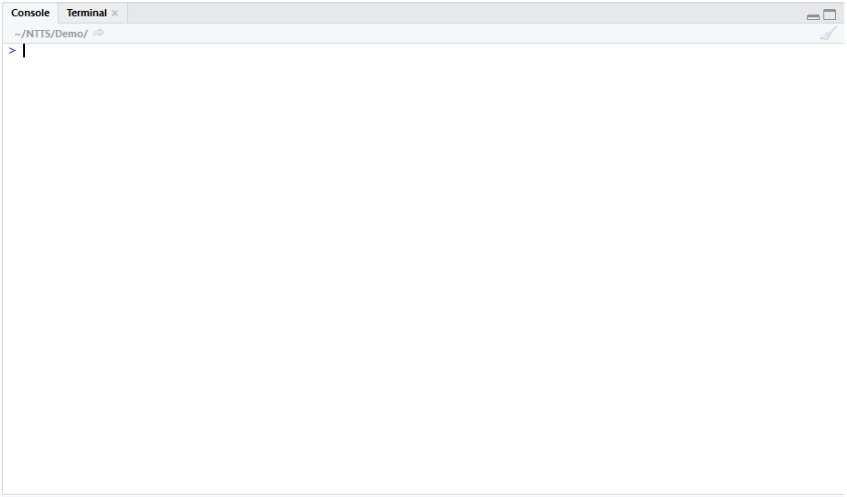
```
## [1] 2
```

```r
sa2 <- get_object(mp,2)
get_name(sa2)
```

```
## [1] "SA with TramoSeats model 1"
```

```r
mod <- get_model(wk, sa2)
```

Import a workspace (2/3)



```
>
```

How to install the package?

The package is available on GitHub: https://github.com/jdemetra/rjdemetra

It has also it's own website: https://jdemetra.github.io/rjdemetra/

It package can be installed from CRAN:

```r
install.packages("RJDemetra")
```

Or from github (development version):

```r
devtools::install_github("jdemetra/rjdemetra")
```

ℹ To install it you need Java8: in case you don't, install a portable version of Java8 and set the `JAVA_HOME` path.

What's next? (1/2)

Documentation:

- Vignette/article for the Journal of Statistical Software
- Guide to install the package with portable version of Java (when you don't have administrator rights)
- Cheat sheet

What's next? (2/2)

Package:

- Get only the Java object of a SA (to reduce computation/customize the output)
- Possibility to used user-defined calendar regressors (currently: only user-defined regressors)
- Function to "refresh" the model (JD+ 3.0.0)

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Examples of current use of RJDemetra

- rjdqa (experimental, no documentation): package to help quality assessment (dashboard and quality report matrix)

https://github.com/AQLT/rjdqa

- persephone: enable easy processing during production of SA series (interactive plots, dashboards. . .)

https://github.com/statistikat/persephone

- Non explore topics: direct vs indirect adjustment (persephone), analyse of revisions, etc.
- Carry out studies on SA: Ladiray D., Quartier-la-Tente A., “(In)Stability of Reg-ARIMA Models for Seasonal Adjustment” $\rightarrow$ STS05 in room MANS

Thank you for your attention

jdemetra/rjdemetra
@JdemetraPlus

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