#### SACE MEETING #5













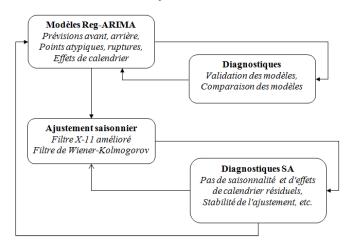
# Stability of Reg-ARIMA estimates: the case of leap year and outliers

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#### Introduction to the seasonal adjustment procedure

#### X-13ARIMA-SEATS et TRAMO-SEATS

Procédure d'ajustement saisonnier



#### Mathematical Writing of Reg-ARIMA

Mathematical writing of the Reg-ARIMA Model in Seasonal Adjustment:

Additive: 
$$Y_t$$
Multiplicative:  $log(Y_t)$  =  $\underbrace{\beta_0 L Y_t + \beta_1 W D_t}_{WD \text{ regressors}} + \underbrace{\sum_i \gamma_i O_{i,t}}_{outliers} + \underbrace{\varepsilon_t}_{\sim ARIMA}$ 

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The goal of the study: illustrate some problems of instability of the estimates with examples on:

- the leap year adjustment
- · the outliers estimates
- the identification of the ARIMA model

#### Sommaire

- 1. The leap year adjustment
- 1.1 How and when carry out the leap year adjustment?
- 1.2 Methodology of the study
- 1.3 Examples
- 1.4 Results
- 2. Outliers adjustment
- 3. Identification of the ARIMA mode
- 4. Conclusion and recommendations

#### Quand faut-il le corriger ?

A leap year: one additional day in February  $\simeq$  every 4 years

 $\rightarrow$  takes into account the "length of the month" effect: it is a calendar effect

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Study of European IPI (1330 series): the leap year effect exists (but not always measurable due to collection)

1. With the Reg-ARIMA model:

$$LY_t = \begin{cases} 0.75 & \text{in February during leap years} \\ -0.25 & \text{in February during non leap years} \\ 0 & \text{for all other months} \end{cases}$$

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2. Correcting values prior to modelling multiplying by a fixed proportion:

$$\begin{cases} \frac{28.25}{29} \simeq 0.974 & \text{in February during leap years} \\ \frac{28.25}{28} \simeq 1.009 & \text{in February during non leap years} \\ 1 & \text{for all other months} \end{cases}$$

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 $\rightarrow$  Study of estimates of the 1st method

#### Methodology used

Methodology: identified the model throughout the entire sample (ARIMA, outliers, etc.) and re-estimate month-by-month the past values setting the first estimated date

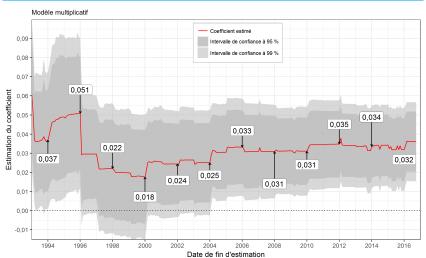
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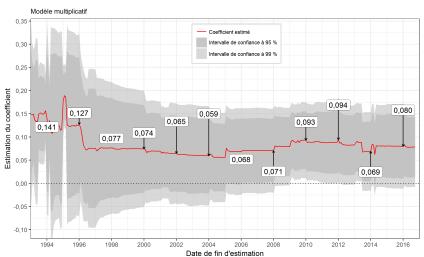
We considered that the estimate has converged whe the estimated coefficient remains:

- positive
- not significantly different from last estimation
- significant: stability of the choice to adjust the leap year effect
- ⇒ European IPI: 410 converge

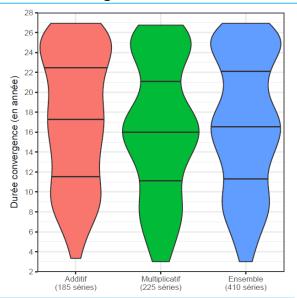
# Examples (1/2): IPI FR-0610 (extraction of crude petroleum)



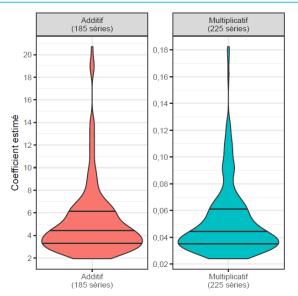
# Examples (2/2): IPI FR-1391 (manufacture of knitted and crocheted fabrics)



### A rather slow convergence...



# ... Towards a value not always coherent



Stability of Reg-ARIMA estimates: the case of leap year and outliers

#### Comparison of the two correction methods

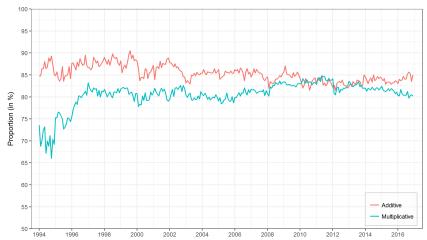


Figure 1: Percentage of series for which the AICC of the 2nd method (LY pre-adjustment) is lower than the AICC of the 1st method (LY regressor)

#### Sommaire

- 1. The leap year adjustment
- 2. Outliers adjustment
- 2.1 Usuals outliers
- 2.2 Methodology of the study
- 2.3 Example
- 2.4 Résultats des simulations
- 3. Identification of the ARIMA mode
- 4. Conclusion and recommendations

#### Usuals outliers

 $\textbf{Additive outlier} \; (AO)$ 

Level Shift (LS)

Seasonal Outlier (SO)

**Transitory Change (TC)** 

#### Methodology used

On European IPI: 1. identification and estimation of the model over 12 years

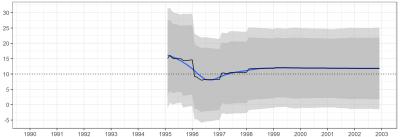
- 2. The raw series is corrected for outliers in the year of the introduction of the break and it is rebased to 100 at the month of the introduction of the break
- simulation of a break, 5 years after the beginning of the time series date, of level 10 for a additive model
- 4. The regressor related to the simulated break is added to the Reg-ARIMA model and its coefficient it is estimated by freezing the estimates of all other parameters and setting the first estimated date

We considered that the series has converged when:

$$\left| rac{ ext{estimated value}}{ ext{last estimated value}} - 1 
ight| < 5 \%$$

# Example of a AO for IPI IT-1413 (manufacture of other outerwear)





### A rather slow convergence..



# ... But not always to the right value

	Minimum	25 %	50 %	75 %	Maximum
Additive models					
Additive outlier (AO)	-11.6	7.8	11.1	14.2	36.9
Level Shift (LS)	-11.4	5.6	9.3	12.7	49.8
Seasonal outlier (SO)	-5.8	7.3	8.8	11.0	31.1
Transitory Change (TC)	-17.4	6.5	10.2	14.1	47.2

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#### Identification of two equivalent models

We use the same model in two different forms mathematically equivalent:

- 1. The leap year regressor is added as a working days regressor
- 2. The leap year regressor is added as an external regressos
- $\rightarrow$  study of the automatic models

#### Different automatic models

Le régresseur LY est dans les effets de calendrier	Le régresseur LY est dans les régresseurs externes			
Summary	Summary			
Estimation span: [1-1990 - 11-2016] 323 observations Trading days effects (7 variables) 3 detected outliers  Arima model [(2,0,0)(0,1,1)]	Estimation span: [1-1990 - 11-2016] 323 observations No trading days effects 8 detected outliers  Arima model [(0,1,1)(0,1,1)]			
Coefficients         T-Stat         P[ T  > t]           Phi(1)         -0,5256         -9,46         0,0000           Phi(2)         -0,2878         -5,17         0,0000           BTheta(1)         -0,7913         -20,56         0,0000	Coefficients         T-Stat         P[I∏ > t]           Theta(1)         -0,5051         -10,08         0,0000           BTheta(1)         -0,7533         -18,80         0,0000			
Correlation of the estimates	Correlation of the estimates  Theta(1) BTheta(1)			
Phi(1)         Phi(2)         BTheta(1)           Phi(1)         1,0000         -0,7388         -0,0184           Phi(2)         -0,7388         1,0000         0,0489           BTheta(1)         -0,0184         0,0489         1,0000	Theta(1) 1,0000 0,0280 BTheta(1) 0,0280 1,0000			
Coefficients         T-Stat         P[ T  > t]           Leap year         4,3861         2,65         0,0085	User variables  Coefficients T-Stat P[ T  > t]  Leap year 4,5569 2,92 0,0038			

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Simulations are questionables and can be improved but highlight a potential instability of Reg-ARIMA models often used as black boxes

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Instabilities have a limited effect on the SA-WD series. . . but have an impact on the short term history and on revisions!

Automatic algorithms use in X-13ARIMA-SEATS and TRAMO-SEATS are important and very useful

### Conclusion and recommendations (2/2)

Specify the model beforehand at the level of the series series:

- base selection procedures on economic reasoning (pay attention to long series)
- do not use methods like black boxes. . .

# Conclusion and recommendations (2/2)

Specify the model **beforehand** at the level of the series **series**:

- base selection procedures on economic reasoning (pay attention to long series)
- do not use methods like black boxes. . . Otherwise, you will be like this statistician who. . .

#### Thank you for your attention

"He uses statistics as a drunk man uses lamp-posts: for support rather than for illumination.

Quote widely attributed to Andrew Lang (1844-1912)