The dangers of using Seasonal Adjustment and other filters in Econometrics

Some economic and environmental examples

Antonio García-Ferrer ¹ Marcos Bujosa ²

¹Dpto. de Análisis Económico: Economía Cuantitativa. Universidad Autónoma de Madrid

²Dpto. de Análisis Económico y Economía Cuantitativa. Universidad Complutense de Madrid

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1 Introduction

- When using seasonally unadjusted data, how can we decide what is the optimal seasonal adjustment to use?
 - Not theoretical point of view
- Do we have sensible statistical tools to discriminate among the different available alternatives?
- Knowing that the estimated components are not observable, is it enough to pay attention to just the component of interest and forget about the remaining ones?
- Is the ideal property of orthogonality among the different component reasonably fulfilled?
- How potential outliers and other variants of intervention analysis affect final estimated components?

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2 Traditional approach

$$y_t = T_t + C_t + S_t + \varepsilon_t$$

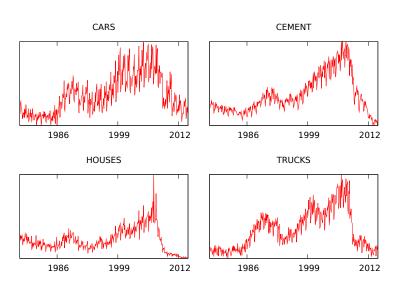
3 Small empirical exercise

Four monthly time series pertaining to the Spanish economic CLI used in: http://uam-ucm-economic-indicators.es/

- CAR REGISTRATIONS
- HOUSING STARTS
- CEMENT CONSUMPTION
- TRUCKS

From 1978M01 to 2013M12

4 Small empirical exercise



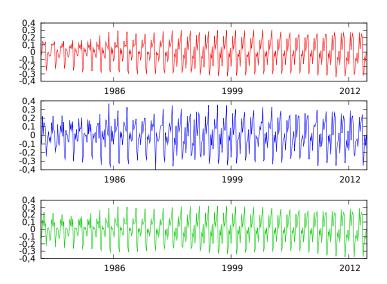
5 Several signal extraction methodologies

Using several model-based signal extraction methodologies, namely

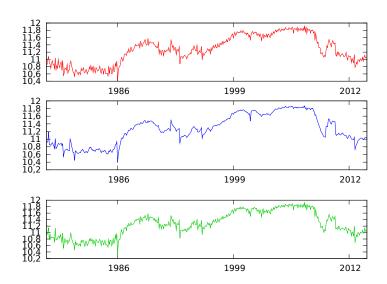
- SEATS-TRAMO
- X-12 ARIMA
- Linear Dynamic Harmonic Regression (Bujosa et al., 2007)

Disclaimer and explanation of the posterior empirical results

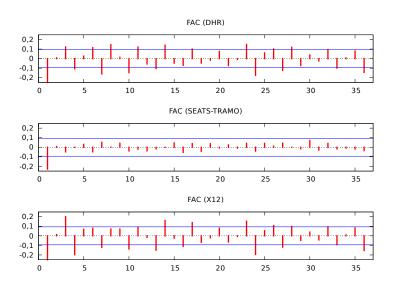
6 Car registrations Seasonal Factors: DHR, ST, X12



7 Seasonally adjusted Car registrations: DHR, ST, X12



8 FAC – First Difference of Seasonally adjusted Car registrations



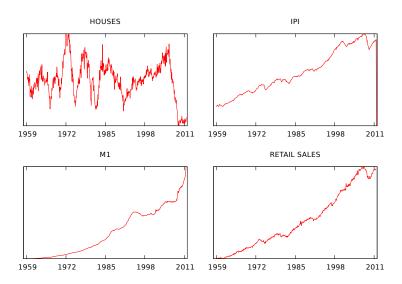
 Outlier detection plus other interventions as easter effects and calendar effects are crucial in the estimation of unobserved components models

 As a matter of fact when you don't use this option in SEATS-TRAMO there is evidence of seasonality in the SA series

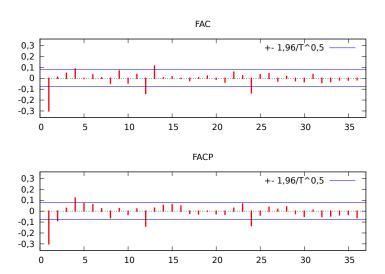
 Using outlier detection plus easter and calendar effects produce considerable reduction in the estimated residual variances ranging from 21% to 31% 10 Results from a Stock & Watson data base

- Housing starts
- IPI
- Money supply M1
- Retail sales

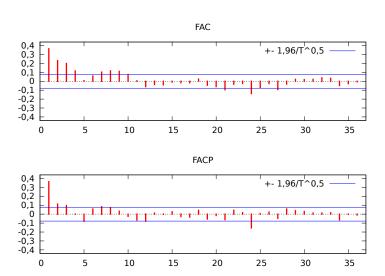
11 Results from a Stock & Watson data base



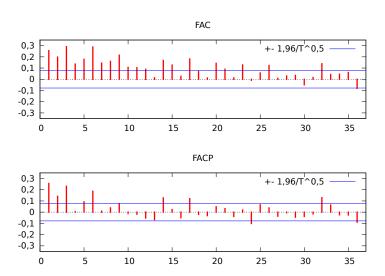
12 Results from a Stock & Watson data base: Housing starts



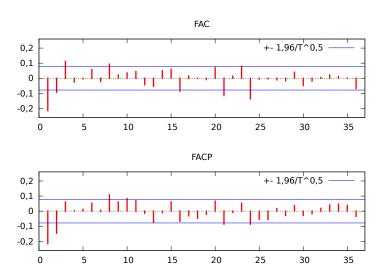
13 Results from a Stock & Watson data base: IPI



14 Results from a Stock & Watson data base: Money supply



15 Results from a Stock & Watson data base: Retail sales



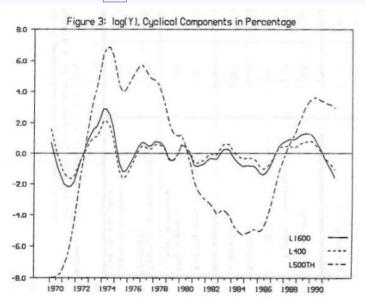
16 Hodrick–Prescott filter

$$y_t = \tau_t + c_t + \epsilon_t$$

Given a positive λ , there is a trend component τ that solves

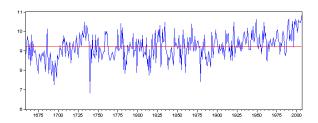
$$\min_{\tau} \left(\sum_{t=1}^{T} (y_t - \tau_t)^2 + \lambda \sum_{t=2}^{T-1} [(\tau_{t+1} - \tau_t) - (\tau_t - \tau_{t-1})]^2 \right)$$

17 Hodrick–Prescott filter

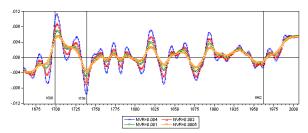


roduction Exercise Methodologies S&W data HP Filter Climate 1 Climate 2 Covid CET References

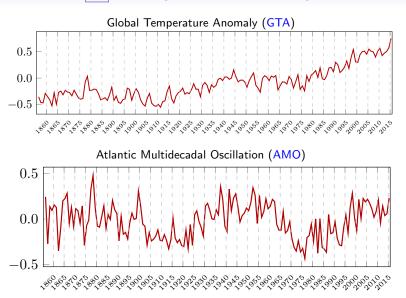
18 The Central England Temperature (CET)



Alternative Temperature Cycles and Bayesian Turning Points



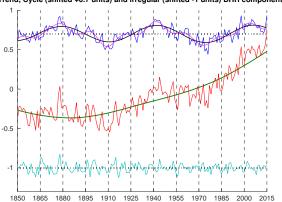
19 Modelling of Global Climate Change



20 Have AMO and GTA a common 63-years cycle?

DHR components for GTA

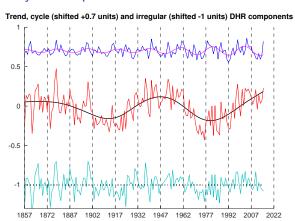
Trend, Cycle (shifted +0.7 units) and irregular (shifted -1 units) DHR components



$$GTA = T + S^{63} + S^{21} + \sum (\text{other harmonics}) + Irreg$$

21 Have AMO and GTA a common 63-years cycle?

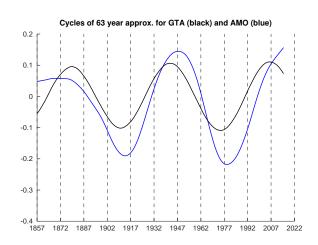
DHR Trend-cycle component for AMO



$$AMO = T + S^{21} + \sum (\text{other harmonics}) + Irreg$$



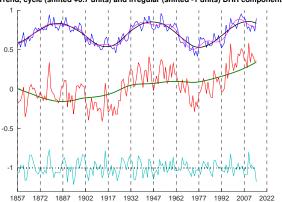
Not clear GTA has a periodic cycle, but not AMO



23 Have original AMO and GTA a common 63-years cycle?

DHR components for "original" AMO data

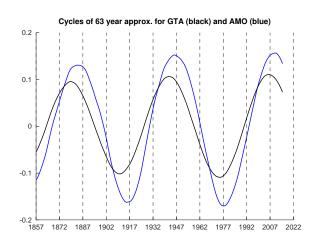
Trend, cycle (shifted +0.7 units) and irregular (shifted -1 units) DHR components



$$AMO_{\rm with\ trend} = T + S^{63} + S^{21} + \sum ({\rm other\ harmonics}) + Irreg$$

24 Have the "original" AMO and GTA a common cycle?

They seem to have a common cycle (as suggested in Professor Young's article)



25 Number of confirmed cases at 3/22/2020

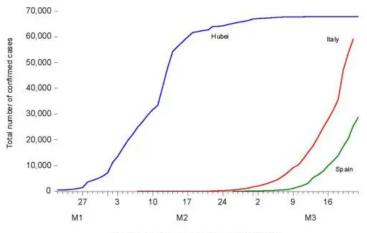


Figure 1: Number of confirmed cases at 3/22/2020

26 Observed contagions and forecasts in Spain

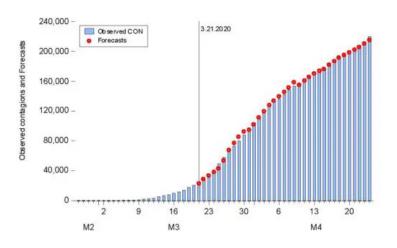


Figure 2: Observed contagions and Forecasts in Spain

27 Observed deaths and forecasts in Spain

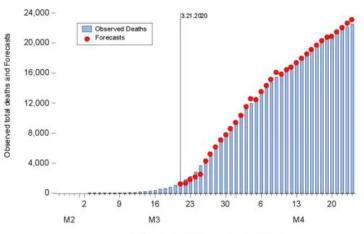


Figure 3: Observed Deaths and Forecasts in Spain

28 Mean Central England Temperature (Degrees Celsius)

Decidir que meter aquí

Bujosa, M., García-Ferrer, A., and Young, P. C. (2007). Linear dynamic harmonic regression. *Comput. Stat. Data Anal.*, **52**(2), 999–1024. ISSN 0167-9473.