

A pairwise approach to model and forecast a large set of disaggregates with common trends

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- **General Objective:** Model and forecast all the components of a macro or business aggregate.
- **Main contribution:** cases with large number of components, when multivariate approaches are not feasible.

Motivation:

① Disaggregating is relevant

- **Own interest** for economic and business decision making.
- **Diagnosis and fore. the agg:** increasing agreement that deeper knowledge of the components may lead to better understanding of the agg. (Espasa et.al, 2002; GG, 2004; HH, 2005, 2011; EA, 2007; EMB, 2013)
- **Comparative analysis** (e.g relative prices)

- ## ② PW strategy (EMB, 2013) as a procedure to deal with the **Estimation uncertainty vs informational losses trade-off**, and the problem of non-pervasive common factors in DFM.

Outline

- 1 Description and asymptotic properties ($T \rightarrow \infty$)
- 2 (X) Simulation results
- 3 (X) Outliers and breaks
- 4 Forecasting strategy and application
 - A note about the forecasting equations
 - Forecasting results for the US CPI
- 5 Conclusions

How the *pairwise* procedure (restricted to common trends) works...

- 1 Perform **Johansen cointegration tests** between all the $N(N - 1)/2$ **pairs** (4950 for $N = 100$).
- 2 Find the largest subset in which every series is cointegrated with all the others (maximum *clique* - *fully cointegrated*).
- 3 Continue looking for the second largest and so on...
- 4 For ease of exposition just focus on the largest.

Asymptotic properties...

- The Pairwise strategy inherits the asy. props. of the Johansen test
- However; two specific features deserve special attention:
 - ① Partial model estimation
 - ② **Multiple testing**; are we inflating the false rejection probability? **NO**

Theorem (for pairs inside the FC set):

Given a set of n_1 pairwise cointegrated series, the prob. of finding the same result in all the $n_1(n_1 - 1)/2$ Johansen's tests $\rightarrow 1$ as $T \rightarrow \infty$

- $\Rightarrow FWER = \alpha$. Even if the hypothesis of interest is the *universal* one (*FWER*), **p-values should not be corrected**.
- For the other pairs \Rightarrow bounds for error probabilities already 'tolerable'

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Multiple testing: analytical results...

In summary:

Size corrections are not needed; multiple testing does not occur (FC series), or bounds provide already tolerable error probabilities"

Simulation experiments:

- Confirm the results for the pairs in the FC set
- Bounds are quite loose
- The PW strategy dominates a DFM alternative in terms of *Gauge* and *Potency*

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The effects of alternative normalizations for β

- The CI VAR remains the same if we change $\alpha\beta'$ by $\alpha H^{-1}H\beta'$. Let:

$$\alpha = \begin{bmatrix} 0 & 0 & 0 \\ -0.2 & 0 & 0 \\ 0 & -0.2 & 0 \\ 0 & 0 & -0.2 \end{bmatrix}, \quad \beta' = \begin{bmatrix} -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix}, \quad W = \begin{bmatrix} 0.25 \\ 0.25 \\ 0.25 \\ 0.25 \end{bmatrix}$$

- Define H in order to express CI rels as deviations form the aggregate

$$\left(\begin{array}{l} \text{equivalence condition;} \\ \alpha^* = \tilde{\alpha}(I_r - \beta_{nf}^* W_{nf})^{-1} \end{array} \right), \quad \alpha H^{-1} = \begin{bmatrix} 0 & 0 & 0 \\ -0.4 & -0.2 & -0.2 \\ -0.2 & -0.4 & -0.2 \\ -0.2 & -0.2 & -0.4 \end{bmatrix},$$

- N° of relevant CI rels.** in each equation (except for the first one) **increased** form 1 to 3.

A note about the forecasting equations

The normalization may not be innocuous for the forecasting stage

- N fore single eq models including CI rels. and other regressors:

$$\Delta x_{i,t} = c + \alpha_{i1} CR_{1i,t-1} + \dots + \alpha_{ir} CR_{ri,t-1} + \text{other regressors} + \epsilon_{i,t}$$

- N° of relevant CI rels. may change depending on β 's normalization.
- \Rightarrow Possible **'free' reduction in the number of parameters** \Rightarrow 'free' estimation uncertainty reduction \Rightarrow **forecasting accuracy improvement.**
- **Strategy:** try different normalizations and select over CI rels and the other regressors using *Autometrics*

Forecasting: exercise design

Regressors are selected using *Autometrics* in the following initial GUMs:

- **Cointegrated series:**

$$\Delta x_{i,t} = c + \alpha_{i1} CR_{1i,t-1} + \dots + \alpha_{ir} CR_{ri,t-1} + \sum_{k_1=1}^{K_1} \phi_{k_1} \Delta x_{i,t-k_1} + \dots + \sum_{k_2=1}^{K_2} \theta_{k_2} \Delta SubAgg_{i,t-k_2} + \sum_{k_3=1}^{K_3} \delta_{k_3} \Delta CPI_{t-k_3} + \sum_{i=1}^{11} \gamma_i S_{i,t} + \epsilon_{i,t}$$

- **Rest of the series:**

Restricted such that the forecast for the sub-aggregate 'rest' is the one coming from a uni-equational model.

- In both cases forecasts are produced with and without IIS.

Forecasting results: RMSE ($\Delta_{12}\log(CPI)$)

- Recursive forecasting 2011.1 - 2013.12, from $H = 1$ to $H = 12$ steps ahead.
- **Baseline: univariate model for the CPI .**
- Form Diebold-Mariano: **PW dominates the baseline in horizons 4 to 8.**
- Form Capistrán: considering jointly **horizons 1 to 6 PW dominates the baseline** (this is not the case when using all horizons, but few observations)

Conclusions

- The pairwise procedure allows to **discover blocks** of series that share a unique common trend in a large set of series.
- **Statistical properties** were studied analytically, and confirmed by Monte Carlo.
- A **small samples correction** procedure was designed and its properties studied by simulations.
- **Comparison with DFM**: PW clearly dominates when n_1 is not large.
- Forecasting accuracy may be improved by changing β 's normalization.
- PW strategy may improve CI test power
- A strategy for dealing with **outliers and breaks** was designed.
- **Application US CPI**: The PW procedure improves **forecasting accuracy** wrt the baseline at least for some horizons.

Thanks for your attention!