



# Discussion of Çakmakli and Altug: “Constructing Coincident Economic Indicators for Emerging Economies”

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## Disclaimer

**The views expressed are those of the authors and do not necessarily reflect the official view of the European Commission.**

## What is the paper after?

- ▶ Constructing a coincident economic indicator for Turkey
- ▶ Using a dynamic latent factor model
  - ▶ There is a latent driver of all cyclical activity (a common factor)
  - ▶ Every relevant macro variable is some function of it
  - ▶ It can be coincident, or allow for leads and lags

## Extra features: specifics of emerging markets

- ▶ Using mixed frequency data
  - ▶ Quarterly and monthly frequencies, sometimes even a change in this respect
- ▶ Using an “unbalanced panel” of variables
  - ▶ Indicators become available sequentially
- ▶ Changing volatilities, extreme observations
  - ▶ Allow time variation in the mean and the variance of the innovation of the indicator
  - ▶ By using a Bayesian semiparametric estimation
- ▶ Explicit modeling of seasonality
- ▶ Overall, quite an involved and complex estimation framework!

## Empirical strategy 1: factor dynamics

- ▶ An AR(p) process  $\Phi(L) C_t = \eta_{C,t}$
- ▶ The innovation term follows an infinite Markov mixture model
  - ▶ K-state Markov chain, with a state dependent mean and variance ( $\theta$ ), and constant transition probabilities
  - ▶ The transition probabilities out of state  $i$  are Dirichlet ( $\alpha\beta$ ), where  $\beta$  is Dirichlet ( $\gamma/K$ )
    - ▶ This allows for some dependence between transitions out of different states
    - ▶ The Dirichlet process parametrizes the conditional distribution of  $\theta_t$  given  $\theta_{t-1}$
- ▶  $K \rightarrow \infty$  makes the finite model infinite

## Advantages

- ▶ The support of  $\theta$  is not bounded
- ▶ We obtain a density estimation of  $\theta$ , not just parameter values
- ▶ Changes in parameters are discrete, compatible with low frequency macro data

## Empirical strategy 2: Seasonality and temporal aggregation

- ▶ Two seasonal components (quarterly and monthly), explicitly modelled

$$S_{1,t} = - \sum_{s=1}^{11} S_{1,t-s} + \eta S_{1,t}$$

$$S_{2,t} = -S_{2,t-3} - S_{2,t-6} - S_{2,t-9} + \eta S_{2,t}$$

$$X_t^i = \lambda_1^i C_t + \lambda_2^i S_{1,t} + \lambda_3^i S_{2,t} + \varepsilon_t^i$$

- ▶ Temporal aggregation: the factor always evolves at a monthly frequency, but some  $X$  variables are missing
  - ▶ Flow variables correspond to the sum of the unobserved parts

## Questions: clarifications

- ▶ Detrended series?
  - ▶ What is the detrending procedure?
  - ▶ Can/should you incorporate into your DGP directly (like done by Canova and Ferroni)?
  - ▶ Are we somewhat “cutting first with an axe and then doing laser surgery”?
- ▶ Are the variables which start later indeed used in the analysis? Is there an issue with this?
- ▶ How do we get the time variation in the factor loadings?
  - ▶ Is it for the variables where data frequency varies?
  - ▶ Could we have time variation in others?
- ▶ Not sure if I understand how and why the procedure is immune to extreme observations



## Comments and questions: methodology

- ▶ Role of the priors and identification restrictions
  - ▶ Are they standard?
  - ▶ Do they make a difference?
  - ▶ The identification assumption matters – elaborate on that?
- ▶ Role of the modeling choice for seasonality
  - ▶ Is it standard – both the form, and also the coexistence of a quarterly and a monthly pattern?
  - ▶ (How much) does it matter?
  - ▶ Can we compare the seasonal patterns to the outcome of a “standard Tramo-Seats” estimate?
- ▶ Can one add to the priors the information that earlier data is likely to be of lower quality?
  - ▶ Or can we recover/confirm this information?

## Comments and questions: results and applications

- ▶ What/how much do we gain relative to other methodologies and/or existing indicators?
- ▶ How much do the specific assumptions bring to the (better?) performance of the CEI relative to other papers?
- ▶ Can we relate changes in the mean and volatility of the CEI innovation to events/developments in a formal/systematic way?
  - ▶ How would a constant  $\theta$  CEI compare to this one?
  - ▶ Can we identify a shift in  $\theta$  in real time?

## Conclusions

- ▶ A very nice semiparametric/flexible approach to further extend the latent dynamic factor methodology
- ▶ Tailored at emerging market features
  - ▶ Not sure if this has been explained and played out completely
- ▶ Would be interesting to see how much the methodology “matters”
  - ▶ Would it indeed not matter in developed countries?
  - ▶ How much does it matter in Turkey, and in other emerging economies?
  - ▶ In terms of “sheer estimates”, but even more in terms of conclusions/applications