

Discussion of  
“Short-Term Forecasting of Business Cycle Turning  
Points: a Mixed-Frequency Markov-Switching Dynamic  
Factor Analysis”

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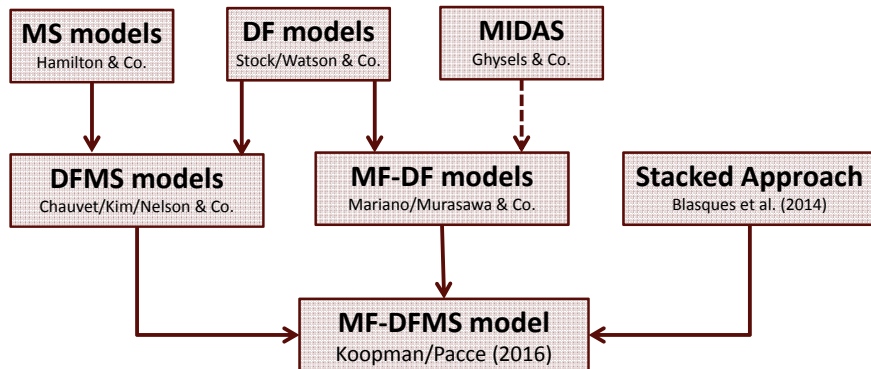
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# What Are the Paper's Contributions?

- Enriching the “stacked” approach for mixed frequency models by Blasques et al. (2014) with a Markov switching feature.
- Demonstration of how the obtained model can be estimated by Bayesian methods.
- Evaluation of MF-DFMS model in terms of ...
  - ... its ability to identify U.S. recessions and expansions in-sample;
  - ... its ability to anticipate business cycle turning points in real-time;
  - ... its ability to make good nowcasts for real GDP growth.

# The MF-DFMS Model vs. Other Models



# What I Am Going to Talk about

- Summary of the MF-DFMS model.
- Timing vs. strength of signals.
- Benchmarks for comparing GDP growth forecasts.
- Minor issues.

- Key idea taken from Blasques et al. (2014): switch from monthly to quarterly frequency.
- Model is given by

$$\begin{bmatrix} y_t \\ x_t^q \end{bmatrix} = \begin{bmatrix} \beta_y & \beta_y & \beta_y \\ \beta_x & 0 & 0 \\ 0 & \beta_x & 0 \\ 0 & 0 & \beta_x \end{bmatrix} f_t^q + \varepsilon_t$$
$$f_t^q = M_{s_t} + \begin{bmatrix} 0 & 0 & \phi_f \\ 0 & 0 & \phi_f^2 \\ 0 & 0 & \phi_f^3 \end{bmatrix} f_{t-1}^q + \xi_t$$

with  $x_t^q = [x_{t,1}^q \ x_{t,2}^q \ x_{t,3}^q]'$  and  $f_t^q = [f_{t,1}^q \ f_{t,2}^q \ f_{t,3}^q]'$ .

- $M_{s_t}$  takes care of the regime-switching means.

- Paper focuses very strongly on the following two timing issues:
  - ▶ Ability to match exactly the NBER dating.
  - ▶ Ability to generate early signals about turning points in real-time.
- ⇒ Not many differences across the models that are considered in the paper.
- Not explicitly discussed: clearness/strength of signals.
- Both timing and the strength of signals are important for policy makers!

# Timing vs. Strength of Signals

## Evaluating Models Based on Their QPSs

- The quadratic probability score (QPS) is computed as follows:

$$QPS = 1/T \sum_{t=1}^T (f_t - x_t)^2,$$

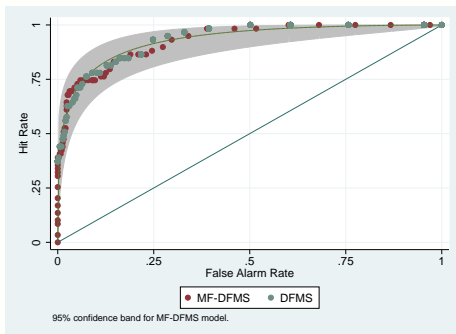
where  $f_t$  denotes a probability forecast and  $x_t$  is the realization of the event.

- Based on the predictive probabilities in Fig. 3, I obtain values of 0.058 for both the MF-DFMS and the DFMS model for the full sample.
- Looking only at recessions, I obtain values of 0.29 for the MF-DFMS model and 0.33 for the DFMS model.
- Looking only at expansions, I obtain values of 0.022 for the MF-DFMS model and 0.016 for the DFMS model.

# Timing vs. Strength of Signals

## Receiver Operating Characteristic (ROC)

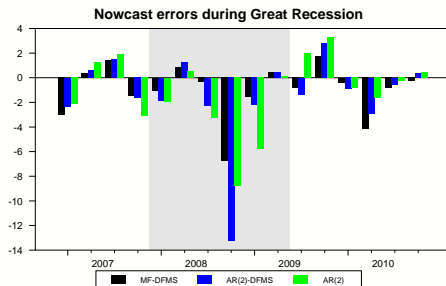
- Are recessions rare events?  $\Rightarrow$  look, e.g., at ROC instead of QPS.
- Idea: compare “hit rate” and “false alarm rate” for different thresholds.
  - ▶ Note the difference to “complex” threshold used in paper.
  - ▶ Authors could show robustness of results with respect to different decision rules.





# Benchmarks for Comparing GDP Growth Forecasts

- Paper presents only assessment vs. an AR(2) model.
  - ▶ What about tougher benchmarks?
- Add information about difference between MF-DFMS and AR(2)-DFMS.
  - ▶ Look at AR(2)-DFMS also for month 1 and 2.
  - ▶ AR(2)-DFMS model based on unbalanced panel approach?



- Maybe also look at  $h=2, 3,$  and  $4$ ?

## Minor Issues/Questions

- Reversing the two parts of the paper title?
- Two different notations used in the paper (Section 2.2.1 vs. Section 2.2.2).
- Could you make the loading coefficients regime-dependent?  
Identification problem?
- First argument against ML estimation (“inference”) not really important in the context of this forecasting paper.
- How do you demean in the real-time out-of-sample analysis? No explicit information given in paper.
- Maybe add table with overview about the timing of the publication process.
- Explicit explanation for why you refer to smoothed probabilities in Fig. 3?