

Understanding Growth-at-Risk A Markov-Switching Approach

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**11th ECB Conference on Forecasting Techniques:
Macroeconomic Forecasting in Abnormal Times**

June 15th, 2021

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Motivation

- **Risk management** is an important consideration for policy decisions
 - Evans (2019): “a very important consideration in charting the course of monetary policy” in low r^* environment
 - Adrian *et al.* (2019): can reduce probability of a future financial crisis
- **Growth-at-Risk Modeling:**
 - Goal: Measure uncertainty and risks around forecast.
 - Key result: (Conditional) mean and volatility are negatively correlated.
 - High mean - Low volatility: Normal state
 - Low mean - High volatility: Large downside risks → **Growth-at-Risk!**

Novelty of Approach

- **Standard approach** to measure risk: Quantile regressions (QR).
- **Our conjecture:** Markov-switching (MS) models should work well.
- **This paper:** **Markov-Switching** model of the **entire** distribution of future real GDP growth **conditional on macroeconomic and financial indicators**.
 - Transition probabilities depend on macroeconomic and financial conditions
 - Parsimonious model to capture features of “growth-at-risk”
- **Advantages of MS model:**
 - Explicit about GAR mechanism
 - Semi-structural interpretation → link to non-linear DSGE
 - Well-established parametric approach
 - **Scenario analysis!**

Novelty of Approach

A Markov-Switching Model of GAR

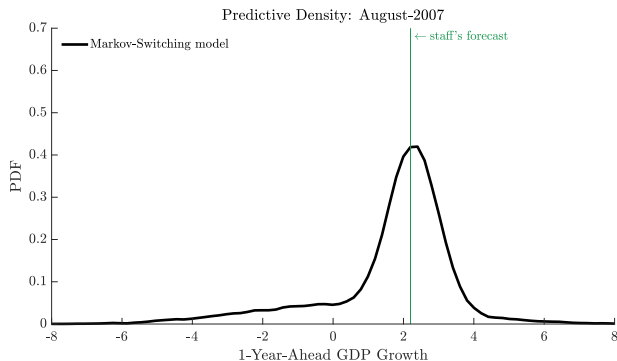
$$\underbrace{\bar{\Delta}y_{t+1,t+12}}_{\text{1-Year-Ahead Avg. Growth}} = \alpha_y(s_t) + \beta_y(s_t)f_t + \gamma_y(s_t)m_t + \sigma_y(s_t)\varepsilon_t^y$$

- f_t and m_t : financial and macroeconomic indicators from **DFM model**
- **Two regimes:** $s_t = 1$: Normal regime, $s_t = 2$: Bad regime
- **Three ingredients:**
 1. Regime specific mean and volatility
 2. Regime specific sensitivity to fundamentals
 - Akin to non-linear dynamics of DSGE models (Gertler *et al.*, 2019; Fernandez-Villaverde *et al.*, 2019; Aruoba *et al.*, 2020)
 3. Financial and macroeconomic conditions influence regime probabilities

An Illustration

*“Financial market conditions have deteriorated, and tighter credit conditions and increased uncertainty have the **potential to restrain economic growth going forward**. In these circumstances, although recent data suggest that the economy has continued to expand at a moderate pace, the Federal Open Market Committee judges that the **downside risks to growth have increased appreciably**.”*

August 17, 2007 FOMC statement

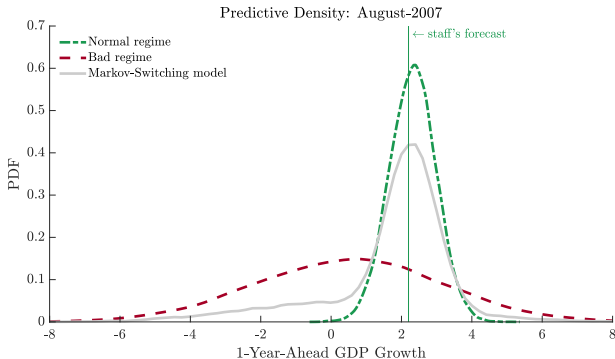


- Optimistic forecast but concern about **downside risk** → MS model left tail

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- MS model: endogenously weights on **normal** and **bad** regimes

First Key Result

Financial and Macro Conditions Both Affect Downside Risk

$$\bar{\Delta}y_{t+1,t+12} = \alpha_y(s_t) + \beta_y(s_t)f_t + \gamma_y(s_t)m_t + \sigma_y(s_t)\varepsilon_t^y$$

1. Negative correlation between mean and volatility

	Bad Regime		Normal Regime	
$\alpha_y(s_t)$	-0.94	[-1.26,-0.59]	0.60	[0.57, 0.63]
$\sigma_y(s_t)$	2.70	[2.48, 2.98]	0.66	[0.60, 0.72]

2. Asymmetry of sensitivity to fundamentals

	Bad Regime		Normal Regime	
$\beta_y(s_t)$	-0.27	[-0.47,-0.07]	-0.07	[-0.13,-0.01]
$\gamma_y(s_t)$	0.65	[0.36, 0.95]	0.21	[0.11, 0.31]

Note: Numbers in brackets represent 95% confidence intervals.

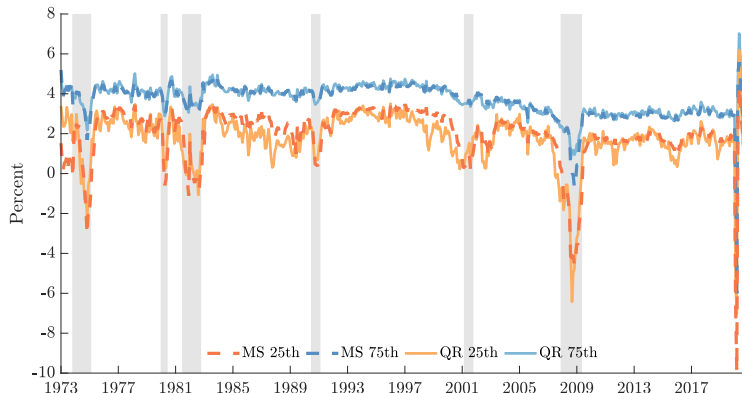
3. Asymmetry in regime transition probabilities

Second Key Result

MS and QR Both Capture Growth-at-Risk

- Follow QR framework of Adrian *et al.* (2019)

$$\widehat{Q}_\tau(\bar{\Delta}y_{t+1,t+12}|x_t) = \hat{\alpha}_\tau + \hat{\beta}_\tau f_t + \hat{\gamma}_\tau m_t$$



- MS and QR as complementary tools for risk assessment**

Thank You

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