# Modelling yields at the lower bound through regime shifts<sup>1</sup>

#### Peter Hördahl (Bank for International Settlements) Oreste Tristani (European Central Bank)

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<sup>1</sup>The views expressed do not necessarily reflect those of the BIS or the ECB.

#### Short-term interest rates and U.S. yields



• Lessons from recent effective lower bound (ELB) experiences

- ELB is not zero, due to cash storage costs; its exact level is uncertain
- ELB spells can be extremely persistent; and expected to be persistent
- US experience suggests very slow pace of normalisation after exiting ELB
- Shadow rate models (e.g. Bauer&Rudebusch 2015; Wu&Xia 2016) empirically successful (and parsimonious), but:
  - they impose a hard constraint, arguably too strong an assumption
  - the state vector dynamics are the same at the ELB as in normal times
  - this suggests fast pace of normalisation after exiting ELB

- We study an alternative model of yields at the ELB. Two regimes: Normal (*N*) and Lower bound (*L*), with stochastic switches
  - allows for different dynamics conditional on regime
- Regime-switching probabilities are state dependent: the probability of switching to *L* is high when the policy rate is close to 0; the prob. of switching to *N* increases as the short rate rises
- Benefit: explicit account of state nonlinearity at the ELB; allows ELB episodes to be very persistent; bond prices reflect these features – also after exiting
- Cost: more parameters  $\rightarrow$  use solely observable state variables

- Application to US term structure using yield factors
- Good fit; clear identification of regimes
- The RS model rules out a deeply negative policy rate (but allows it to dip below the estimated LB)
- The model implies a slow pace of policy rate normalisation in coming years
- Compared to an affine model: higher term premia in recent years / lower average expected policy rates
- Regime shift risk is priced by investors, but magnitude is small

#### Regime-switching model

State vector

$$x_{t+1} = \mu^j + \Phi^j x_t + \Sigma^j \varepsilon_{t+1}$$

with j = N, L and  $x_t = \begin{bmatrix} c_t & s_t & r_t \end{bmatrix}'$ ; i.e. curvature, slope, r.

- By assumption, under *L* the policy rate:
  - is expected to remain constant;
  - does not affect the other factors

$$\begin{aligned} r_{t+1} &= & \mu_r^L + \sigma^L \varepsilon_{r,t+1}, \\ \Phi^L &= & \left[ \begin{array}{cc} \phi_{cc}^L & \phi_{cs}^L & 0 \\ \phi_{sc}^L & \phi_{ss}^L & 0 \\ 0 & 0 & 0 \end{array} \right]. \end{aligned}$$

- RS probabilities are state-dependent:
  - general intuition: the lower r, the more likely a switch to L; the higher r, the more likely a switch to N.
- Specifically:

$$\pi_t^{\mathbb{P},NL} = \int^{\theta_r} \frac{1}{\sigma_r^N \sqrt{(2\pi)^2}} \exp\left(-\frac{1}{2}\left(\frac{r-\mu_{t+1}^{N,r}}{\sigma_r^N}\right)^2\right) dr.$$

and  $\pi_t^{\mathbb{P},LN} = 1 - \pi_t^{\mathbb{P},NL}$ .

• Assume constant Q-RS probabilities,  $\pi^{Q,NL}$  and  $\pi^{Q,LN}$ .

# Adding yields

- Pricing according to Dai, Singleton and Yang (2007) and Bansal and Zhou (2002)
- Recall state vector

$$x_{t+1} = \mu^j + \Phi^j x_t + \Sigma^j \varepsilon_{t+1}$$

• Risk-aversion induces

$$x_{t+1} = \mu^{\mathbb{Q}j} + \Phi^{\mathbb{Q}j}x_t + \Sigma^j \varepsilon_{t+1}$$

and, using a conditionally log-normal approximation

$$y_{t,n} = \frac{A_n^j}{n} + \frac{B_n^j}{n} x_t$$

where

$$\begin{aligned} \mathcal{A}_{n}^{j} &= \sum_{k=1}^{S} \pi^{\mathbb{Q}_{jk}} \left( \delta_{0}^{j} + \mathcal{A}_{n-1}^{k} + \mathcal{B}_{n-1}^{k} \mu^{\mathbb{Q}_{j}} - \frac{1}{2} \mathcal{B}_{n-1}^{k} \Sigma^{j} \Sigma^{j} \left( \mathcal{B}_{n-1}^{k} \right)^{\prime} \right) \\ \mathcal{B}_{n}^{j} &= \sum_{k=1}^{S} \pi^{\mathbb{Q}_{jk}} \left( \delta_{x}^{\prime} + \mathcal{B}_{n-1}^{k} \Phi^{\mathbb{Q}_{j}} \right) \end{aligned}$$

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23 May 2017 8 / 26

# Pricing consistency

- Recall  $x_t = \begin{bmatrix} c_t & s_t & r_t \end{bmatrix}'$ . We set  $c_t = r_t + y_{t,120} 2y_{t,36}$  and  $s_t = y_{t,120} r_t$ .
- Hence  $y_{t,120} = s_t + r_t$  and  $y_{t,36} = \frac{1}{2} (s_t c_t) + r_t$ .
- Need to ensure consistency with  $y_{t,120} = \frac{1}{120} \left( A_{120}^j + B_{120}^j x_t \right)$ , so that

• 
$$A_{120}^{j} = 0$$
 and  $B_{120}^{j} = \begin{bmatrix} 0 & 120 & 120 \end{bmatrix}$ 

• and 
$$y_{t,36} = \frac{1}{36} \left( A_{36}^j + B_{36}^j x_t \right)$$
 so that  
•  $A_{36}^j = 0$  and  $B_{36}^j = \begin{bmatrix} -18 & 18 & 36 \end{bmatrix}$ 

- This induces nonlinear constraints on two rows of  $\mu^{{\mathbb Q} j}$  and  $\Phi^{{\mathbb Q} j}$ .
- We impose these constraints in ML estimation.

#### Data

- Monthly US yield data (end-month), January 1987 April 2017.
- Zero-coupon yields from Fed Board (Gürkaynak, Sack, Wright, 2006)
- 1m, 3y, 10y yields used for factors; not included among yields



- We have a large number of parameters, so we need to impose some restrictions.
- We estimate the N parameters on a sub-sample when the economy clearly was in the N regime (1987 2007).
- We estimate the VAR *L* parameters under ℙ on the Dec. 2008 Oct. 2015 sample.
- We set the short rate threshold  $\theta_r$  used in  $\pi_t^{\mathbb{P},NL}$  to the 10th percentile of the distribution of r.
- Remaining parameters are estimated using maximum likelihood.

# Results

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• Mean levels of state variables

Conditional means *N L c* 0.473 0.962 *s* 1.581 2.361 *r* 3.286 0.136

- The lower bound is estimated at 13.6 basis points.
- The steady state short rate in the normal regime is 3.29%.
- Standard deviation of yield measurement errors is  $\sigma^m = 0.124$ .

# Short-term interest rate (zoom in on period 2000 -)

• The RS model effectively combines the N and L dynamics to ensure consistency with actual data.



### Yield fit



# Filtered probabilities of N/L regimes



### Risk premia

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Image: A matrix and a matrix

• RS model implies higher premia than 1-regime affine model recently.



ACM premium source: Adrian, Crump, Moench (2013)

#### Expected short rate and term premia

 Interest rate expectations matter more for yields during normal times; the term premium dominates at the LB.



# Is regime shift risk priced?

• Regime shift premia essentially zero during *N*-regime; small during *L*-regime.



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Yields at the lower bound

23 May 2017 20 / 26

#### Forecasts

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### Yield forecasts at end-2011

• Wide confidence bands, but rate distributions do not include deeply negative values.



#### Yield forecasts at end-April 2017

• Gradual increase in interest rates and yields.



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23 May 2017 23 / 26

### Short rate forecasts at end-April 2017

• Forecast from regime switching model more in line with SPF than shadow rate model (WX) forecasts.



### Regime prob. forecast and short rate forecast distribution

• Highly skewed interest rate forecasts increases the future *L*-regime probability, influencing longer term bond pricing today



23 May 2017 25 / 26

- We propose a dynamic term structure model with regime switches to account for lower bound spells
- Application to US term structure using yield factors: good fit; clear identification of regimes
- The RS model rules out a deeply negative policy rate (but allows it to dip below the estimated LB)
- Compared to an affine model: higher term premia in recent years / lower average expected policy rates
- Regime shift risk is priced by investors, but magnitude is small
- The model implies a slow pace of policy rate normalisation in coming years, in line with SPF forecasts