

Discussion of
Monfort, Pegoraro, Renne, and Roussellet

“Staying at Zero with Affine Processes:
A New Dynamic Term Structure Model”

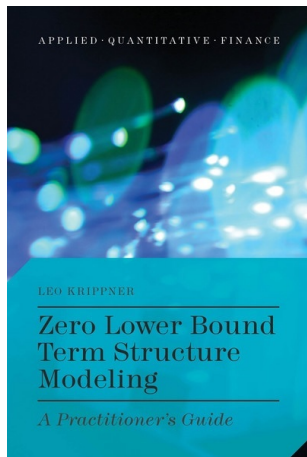
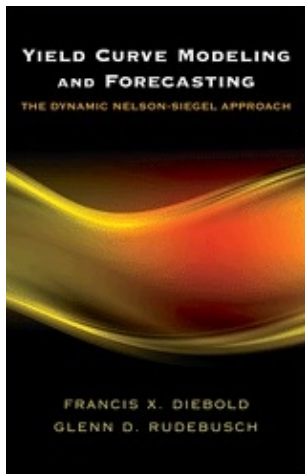
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(With special thanks to Leo Krippner and Minchul Shin)

Volatility Institute
Stern School, NYU
Seventh Annual Conference



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Favorite Books



Lots of Constrained Series in Finance

“Soft” barriers:

- ▶ Exchange rate target zones
- ▶ Inflation corridors

“Hard” barriers:

- ▶ Volatilities: e.g., asset returns
- ▶ Durations: e.g., intertrade
- ▶ Rare-event counts: e.g., bankruptcies
- ▶ Nominal bond yields

Lots of Associated Constrained Processes in Financial Econometrics

- ▶ Vols: GARCH, stochastic volatility, and more
- ▶ Durations: ACD and more
- ▶ GAS and MEM
(Creal, Koopman, and Lucas, 2013; Harvey, 2013)

What About Bond Yields?

Duffie-Kan (1996) Gaussian affine term structure model (GATSM):

State x_t is an affine diffusion under the risk-neutral measure:

$$dx_t = K(\theta - x_t)dt + \Sigma dW_t$$

Instantaneous risk-free rate r_t is affine in x_t :

$$r_t = \rho_0 + \rho_1' x_t$$

Duffie-Kan arbitrage-free result:

$$y_t(\tau) = -\frac{1}{\tau} B(\tau)' x_t - \frac{1}{\tau} C(\tau)$$

- Arbitrage-free
- Simple (closed-form)
- But fails to respect the ZLB

Constrained Processes for Bonds

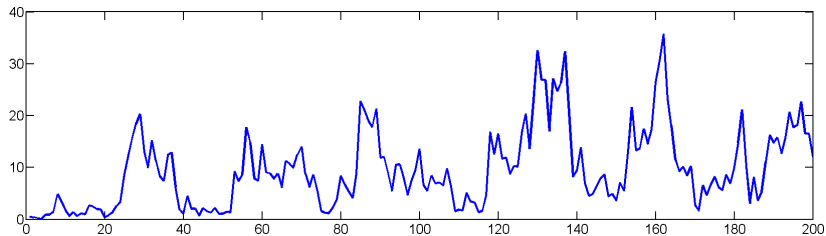
- ▶ Square root: $dx_t = k(\theta - x_t) dt + \sigma\sqrt{x_t} dW_t$
(Cox, Ingersol and Ross, 1976)
- ▶ Others: lognormal, quadratic
- ▶ Autoregressive gamma ($ARG(1)$)
(Gourieroux and Jasiak, 2006)
- ▶ $ARG0(1)$
(MPRR, 2015)

ARG(1)

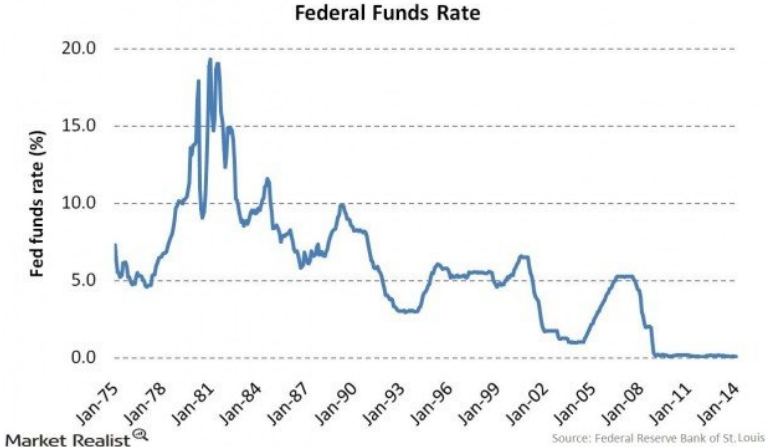
x_t is an ARG(1) process if
 $x_t | x_{t-1}$ is distributed non-central gamma with:

- ▶ Non-centrality parameter βx_{t-1}
- ▶ Scale parameter $c > 0$
- ▶ Degree of freedom parameter $\delta > 0$
 - Non-negative (obvious)
 - Diffusion limit is CIR (not obvious)

Simulated $ARG(1)$ Realization



But Alas...



ARG0(1)

If $x_t \sim ARG(1)$, then

$$x_t | z_t \sim \text{Gamma}(\delta + z_t, c)$$

$$z_t | x_{t-1} \sim \text{Poisson}(\beta x_{t-1})$$

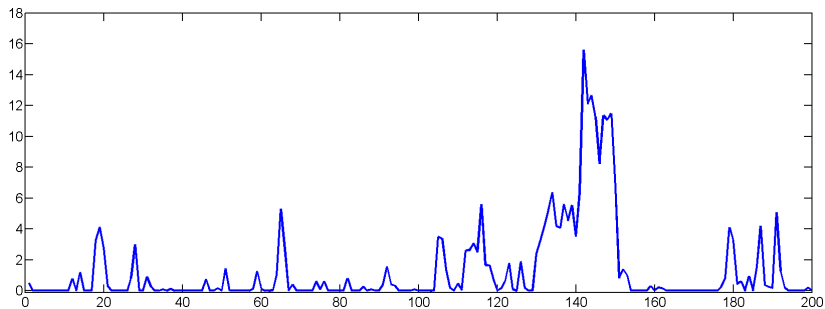
If $x_t \sim ARG0(1)$, then

$$x_t | z_t \sim \text{Gamma}(z_t, c)$$

$$z_t | x_{t-1} \sim \text{Poisson}(\alpha + \beta x_{t-1})$$

- ▶ ARG0 takes $\delta \rightarrow 0$, which makes $x_t = 0$ a mass point.
(As $\delta \rightarrow 0$, $G(\delta, c) \rightarrow \text{Dirac's delta.}$)
- ▶ Introduces α , which governs probability of escaping the ZLB.
(Note that $\alpha = 0 \implies x_t = 0$ is an absorbing state.)

Simulated $ARGO(1)$ Realization



ARGO Approach

$$x_t | z_t \sim \text{Gamma}(z_t, c)$$

$$z_t | x_{t-1} \sim \text{Poisson}(\alpha + \beta x_{t-1})$$

1. Arbitrage-free
2. Simple (closed-form)
3. Respects the ZLB

End of story?

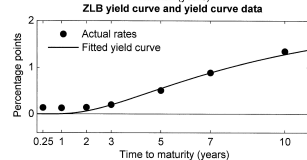
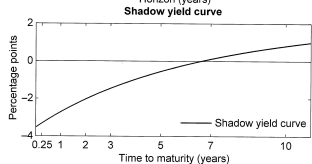
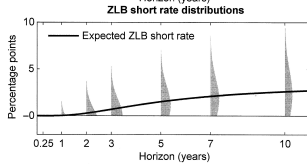
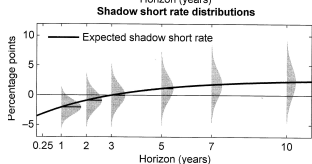
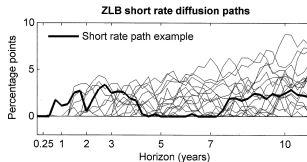
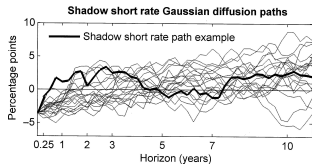
Shadow-Rate Approach (Shadow/ZLB GATSM)

$$x_{s,t} = \mu(1 - \rho) + \rho x_{s,t-1} + \varepsilon_t$$

$$x_t = \max(x_{s,t}, 0)$$

1. Arbitrage-free
2. Simple (simulation)
3. Respects the ZLB

Shadow Rates and ZLB Rates



Shadow-Rate Approach (Shadow/ZLB GATSM)

$$x_{s,t} = \mu(1 - \rho) + \rho x_{s,t-1} + \varepsilon_t$$

$$x_t = \max(x_{s,t}, 0)$$

1. Arbitrage-free
2. Simple (simulation)
3. Respects the ZLB
4. Sample path feature probabilities (e.g., lift-off from ZLB)
5. Sample path integral densities (e.g., effective stimulus)

But MPRR could also do points 4 and 5...

6. Shadow rate path and shadow yield curve

Final Thoughts on Relative Performance

Much boils down to:

- Value of the shadow rate path and shadow yield curve
- Views about “simplicity”

My balance tips slightly toward shadow/ZLB GATSM

Interesting question:

With appropriate constraints on the Gamma and Poisson processes, can MPRR “replicate” a shadow/ZLB GATSM, but without the mechanism of shadow short rates and the shadow yield curve?