

Discussion of

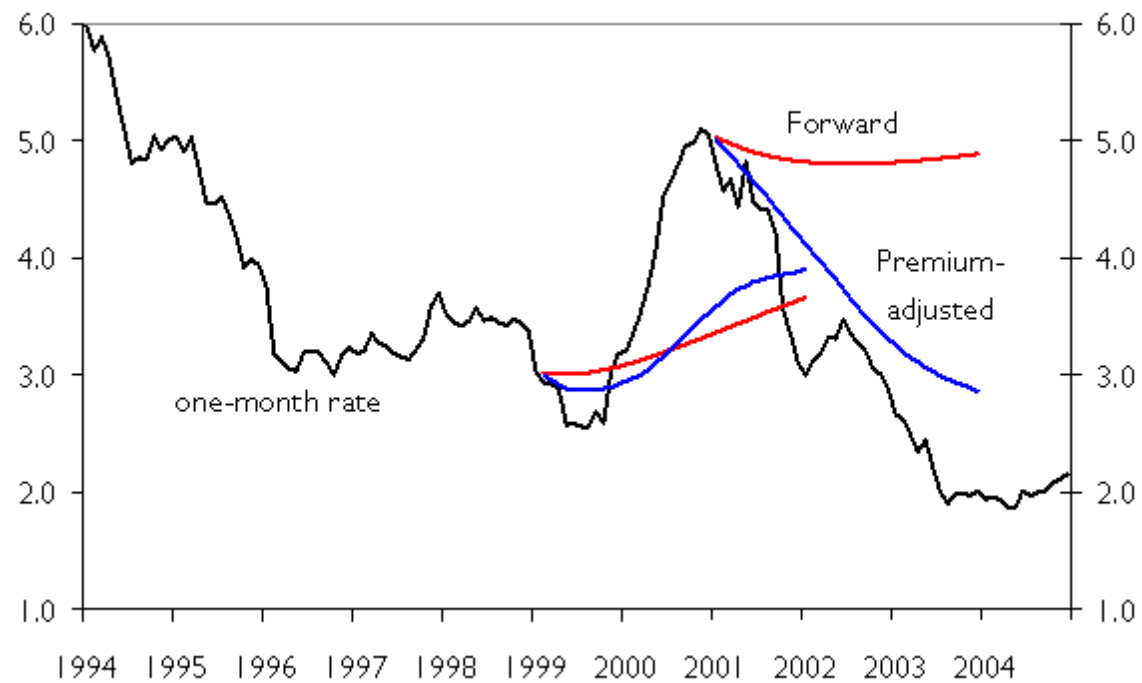
Learning, macroeconomic dynamics and the term structure: A Bayesian analysis

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policy and financial markets

The issue: time-varying premia or model uncertainty/limited information?



An summary of the paper

- Construct and estimate a new-Keynesian model; add constant-gain learning; add risk premia and bond yields and run a horse-race. Results:
 - all in all, the macro-finance model (with constant prices of risk!) beats the competition ...
 - ... but the learning model predicts better macro variables.

Three main comments

- One possible way to interpret the paper
- Discussion of how the various ingredients are combined.
- Comment on the empirical assessment of the model.

One interpretation: perturbation I

- In a micro-founded model with a representative agent, bond prices are $B_{t,t+1} = E_t \left[\beta \frac{1}{\Pi_{t+1}} \frac{\Lambda_{t+1}}{\Lambda_t} \right]$; first order conditions can be collected in a vector function such that $E_t f(z_{t+1}, z_t, x_{t+1}, x_t) = 0$.
- Exact solution in general unknown. Consider approximation via perturbation methods. Standard "log-linearisation" yields

$$\begin{aligned}\hat{b}_{t,t+1} &= b_1 \hat{x}_t & \hat{\pi}_t &= \pi_1 \hat{x}_t & \hat{\lambda}_t &= \lambda_1 \hat{x}_t \\ \hat{x}_{t+1} &= c_1 \hat{x}_t + \sigma \varepsilon_{t+1} \\ \xi &= 0\end{aligned}$$

Perturbation II

- In the scalar- x_t case, II-order approximation (HTV, 2005)

$$\hat{b}_{t,t+1} = b_1 \hat{x}_t + \frac{1}{2} b_2 \hat{x}_t^2 + \frac{1}{2} b_0 \sigma^2$$

$$\hat{x}_{t+1} = c_1 \hat{x}_t + \frac{1}{2} c_2 \hat{x}_t^2 + \frac{1}{2} c_0 \sigma^2 + \sigma \varepsilon_{t+1}$$

$$\xi = (\pi_1 - \lambda_1) \sigma$$

Perturbation III

- In the scalar- x_t case, III-order approximation (Ravenna and Seppala, 2005)

$$\hat{b}_{t,t+1} = b_1 \hat{x}_t + \frac{1}{2} b_2 \hat{x}_t^2 + \frac{1}{2} b_0 \sigma^2 + \frac{1}{6} b_3 \hat{x}_t^3 + \frac{1}{2} b_4 \hat{x}_t \sigma^2$$

$$\hat{x}_{t+1} = c_1 \hat{x}_t + \frac{1}{2} c_2 \hat{x}_t^2 + \frac{1}{2} c_0 \sigma^2 + \frac{1}{6} c_3 \hat{x}_t^3 + \frac{1}{2} c_4 \hat{x}_t \sigma^2 + \sigma \varepsilon_{t+1}$$

$$\xi = \xi_0 + \xi_1 \hat{x}_t$$

Macro-finance/learning

$$\begin{aligned}\hat{b}_{t,t+1} &= b_1 \hat{x}_t \boxed{+\frac{1}{2}b_2 \hat{x}_t^2} + \frac{1}{2}b_0 \sigma^2 \boxed{+\frac{1}{6}b_3 \hat{x}_t^3 + \frac{1}{2}b_4 \hat{x}_t \sigma^2} \\ \hat{\pi}_t &= \pi_1 \hat{x}_t \boxed{+\frac{1}{2}\pi_2 \hat{x}_t^2 + \frac{1}{2}\pi_0 \sigma^2 + \frac{1}{6}\pi_3 \hat{x}_t^3 + \frac{1}{2}\pi_4 \hat{x}_t \sigma^2} \\ \hat{x}_{t+1} &= c_1 \hat{x}_t \boxed{+\frac{1}{2}c_2 \hat{x}_t^2 + \frac{1}{2}c_0 \sigma^2 + \frac{1}{6}c_3 \hat{x}_t^3 + \frac{1}{2}c_4 \hat{x}_t \sigma^2} + \sigma \varepsilon_{t+1} \\ \xi &= \xi_0 + \xi_1 \hat{x}_t\end{aligned}$$

- Micro-finance: boxed coefficients equal to zero; ξ_0, ξ_1 unrestricted. Learning: boxed coefficients equal to zero; ξ_0 unrestricted, $\xi_1 = 0$, $b_1, b_0, \pi_1, \pi_0, c_1, c_0$ estimated through a VAR.

Putting the pieces together

- Linearised simple new-Keynesian model for $\mathbf{X}_t = [\pi_t, y_t, i_t]$

$$\mathbf{A}\mathbf{X}_t = \mathbf{C} + \mathbf{B}\mathbf{E}_t\mathbf{X}_{t+1} + \mathbf{D}\mathbf{X}_{t-1} + \mathbf{\Sigma}\boldsymbol{\varepsilon}_t$$

MSV affine in the states, hence PLM

$$\mathbf{X}_{t+1} = \boxed{\hat{\mathbf{C}}(t) + \mathbf{\Phi}(t)\mathbf{X}_t} + \hat{\mathbf{\Sigma}}(t)\mathbf{v}_{t+1}$$

- Note: $\boldsymbol{\varepsilon}_t$ are the unknown structural shocks; \mathbf{v}_t are the observed reduced form shocks. Using expectations ALM

$$\mathbf{X}_{t+1} = \mathbf{F}(t) + [\mathbf{A} - \mathbf{B}\mathbf{\Phi}(t)]^{-1}\mathbf{D}\mathbf{X}_t + [\mathbf{A} - \mathbf{B}\mathbf{\Phi}(t)]^{-1}\mathbf{\Sigma}\boldsymbol{\varepsilon}_{t+1}$$

- At this point, no-arbitrage arguments applied. Correctly, reduced form shocks v_{t+1} are priced. Result

$$\mathbf{Y}_t = \mathbf{A}_y(t) + \mathbf{B}_y(t) \mathbf{X}_t$$

- However, the variance of v_{t+1} is time varying $\hat{\Sigma}(t) \hat{\Sigma}(t)'$. Consistently with the macro literature, time variation is disregarded (anticipated utility). Here, however, intuition is less clear. Some risks are priced, others are not. If agents require a premium to compensate them for fundamental risks, why are they not worried about time-variation in variance?

Putting the pieces together: an alternative

- Linearisation $\mathbf{X}_t = [\pi_t, y_t, i_t, yieldst]$

$$\begin{aligned}\mathbf{A}\mathbf{X}_t &= \mathbf{C} + \mathbf{B}\mathbf{E}_t\mathbf{X}_{t+1} + \mathbf{D}\mathbf{X}_{t-1} + \mathbf{\Sigma}\varepsilon_t \\ \mathbf{X}_{t+1} &= \boxed{\hat{\mathbf{C}}(t) + \mathbf{\Phi}(t)\mathbf{X}_t} + \hat{\mathbf{\Sigma}}(t)\mathbf{v}_{t+1}\end{aligned}$$

- Using expectations ALM

$$\mathbf{Y}_t = \mathbf{F}_y(t) + \overline{[\mathbf{A} - \mathbf{B}\mathbf{\Phi}(t)]^{-1} \mathbf{D}}\mathbf{X}_t$$

- Yields also affine in the states. No further assumptions needed.

Empirical results I

- Which is the most intuitively appealing model?
 - Learning: inflation survey data are matched, but ... announcements have no effects; we throw away expectations effect which can be especially important for asset prices (e.g. "new economy" beliefs); lots of free parameters.
 - Macro-finance models: forecast yields better; but ... agents assumed to have known the Taylor rule and new-Keynesian models in the sixties; inflation survey data are not matched; average std.dev. of target is 1.6%; lots of free parameters.

Empirical results II

- Good to match survey data, but is this a desideratum? Question in SPF: "What do you expect to be the annual average over the next 10 years of the CPI inflation rate?" – large variance, but Ang, Bekaert and Wei (2006).
- "Excessive target volatility puzzle" – what is excessive for a perceived target? Survey data on long run inflation expectations.

Empirical results III

- Macro-finance models win in terms of marginal likelihood. Good! But ... is this result robust?
 - 54 parameters!
 - A bit more information on the estimation: how many MCMC simulations? acceptance rate? prior/posterior distributions?
- Any differences between yields responses in the macro-finance and learning models?

Empirical results IV

- No "bond yields conundrum" in both macro-finance and learning models.
- Most striking result: macro-finance model with constant prices of risk. Do macro-models work better than we think for yields?
- Conjecture: yields and macro variables inherit unit-root behaviour of target and natural rate. Any movement in inflation is permanent and translates in equal movements in long yields. Satisfactory? Impulse responses?

Conclusion

- Very ambitious and stimulating paper.
- Contribution includes theoretical and empirical elements.
- Useful perspective to start answering the question: premia or imperfect knowledge?