

Identifying Taylor Rules in macro-finance models

David Backus (NYU), Mikhail Chernov (UCLA),
and Stanley Zin (NYU)

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Identifying the Taylor rule

- Long-term goal
 - Integrate models of bond pricing and monetary policy
- Open question
 - Can we identify the monetary policy parameter(s)?
 - Can we distinguish systematic policy from shocks to it?

Identifying the Taylor rule

- Common views
 - Macro: not identified
 - Finance: can't extract policy component from affine model
- What we do
 - Describe conditions for identification
 - Revisit earlier work on dynamic rational expectations models

Outline

- Two examples of the problem
- Rational expectations solutions and identification
- More complex models
- What if you don't see the state?

Setup

- State

$$x_{t+1} = Ax_t + Cw_{t+1}$$

- Shocks

$$s_{it} = d_j^\top x_t$$

- Identification issue: we observe state x , but not shock s_i

Cochrane's example

- Model

$$i_t = r + E_t \pi_{t+1} \quad (\text{Fisher equation})$$

$$i_t = r + \tau \pi_t + s_t \quad (\text{Taylor rule})$$

Cochrane's example

- Model

$$i_t = r + E_t \pi_{t+1} \quad (\text{Fisher equation})$$

$$i_t = r + \tau \pi_t + s_t \quad (\text{Taylor rule})$$

- Expectational difference equation

$$E_t \pi_{t+1} = \tau \pi_t + s_t$$

- Solution: $\pi_t = b^\top x_t$ with $b^\top = -d^\top (\tau I - A)^{-1}$

- Identification problem: any τ works for some d

$$b^\top A = \tau b^\top + d^\top$$

Affine example

- Model

$$m_{t+1}^{\$} = -\delta^{\top} x_t - \lambda^{\top} \lambda / 2 + \lambda^{\top} w_{t+1} \quad ((\log) \text{ Pricing kernel})$$

$$i_t = -\log E_t \exp(m_{t+1}^{\$}) = \delta^{\top} x_t \quad (\text{Interest rate})$$

Affine example

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- Expectational difference equation

$$E_t \pi_{t+1} = \tau \pi_t + s_t$$

- Is interest rate equation a Taylor rule or the Fisher equation?

$$\delta = \tau b^{\top} + d^{\top} \text{ or } \delta = b^{\top} A$$

Questions

- Would an extra shock help?

$$i_t = E_t \pi_{t+1} + s_{1t} \quad (\text{Fisher equation})$$

$$i_t = \tau \pi_t + s_{2t} \quad (\text{Taylor rule})$$

- If shocks are independent, can use s_{1t} as an instrument
- Would long rates help?
 - May span state, but we see state anyway

A representative-agent model

- Equations

$$\begin{aligned}i_t &= -\log E_t \exp(m_{t+1} - \pi_{t+1}) && \text{(Euler/Fisher equation)} \\m_t &= -\rho - \alpha g_t && \text{((log) Pricing kernel)} \\g_t &= g + s_{1t} && \text{(Consumption growth)} \\i_t &= \tau \pi_t + s_{2t} && \text{(Taylor rule)}\end{aligned}$$

- Guess $i_t = a^\top x_t$, $\pi_t = b^\top x_t$
- Solution: $b^\top = (\alpha d_1^\top - d_2^\top)(\tau I - A)^{-1}$, $a^\top = \tau b^\top + d_2^\top$
- Estimate a and b from OLS of i_t and π_t on x_t
- We have n knowns a and $n+1$ unknowns (τ, d_2)
- One exclusion restriction is sufficient to identify τ e.g., independence of shocks

An exponential-affine model

- Equations

$$i_t = -\log E_t \exp(m_{t+1} - \pi_{t+1}) \quad (\text{Euler/Fisher equation})$$

$$m_{t+1} = -\rho - s_{1t} + \lambda^\top w_{t+1} \quad ((\log) \text{ Pricing kernel})$$

$$i_t = \tau \pi_t + s_{2t} \quad (\text{Taylor rule})$$

- Guess $i_t = a^\top x_t$, $\pi_t = b^\top x_t$
- Solution: $b^\top = (\delta^\top - d^\top)(\tau I - A)^{-1}$, $a^\top = \tau b^\top + d^\top$
- Estimate a and b from OLS of i_t and π_t on x_t
- We have n knowns a and $n+1$ unknowns (τ, d)
- One exclusion restriction is sufficient to identify τ , e.g., $d_{2i} = 0$.

Observations

- OLS of i_t on π_t does not recover τ
- One restriction on MP shock suffices to identify the Taylor rule
- It does not matter whether they are shocks in the Euler equation or not
- This success reflects what Hansen and Sargent call the hallmark of rational expectations models: that cross-equation restrictions connect the parameters in one equation to those in the others
- What about more elaborate models?

A Phillips curve

- Equations

$$i_t = -\log E_t \exp(m_{t+1} - \pi_{t+1}) \quad (\text{Euler equation})$$

$$m_t = -\rho - \alpha g_t \quad (\text{Pricing kernel})$$

$$\pi_t = \beta E_t \pi_{t+1} + \kappa g_t + s_{1t} \quad (\text{Phillips curve})$$

$$i_t = \tau_1 \pi_t + \tau_2 g_t + s_{2t} \quad (\text{Taylor rule})$$

- Guess $i_t = a^\top x_t$, $\pi_t = b^\top x_t$ and $g_t = c^\top x_t$

- Solution:

$$b^\top = \beta A b^\top + \kappa c^\top + d_1^\top$$

$$a^\top = \tau_1 b^\top + \tau_2 c^\top + d_2^\top$$

- Estimate a , b and c from OLS of i_t , π_t and g_t on x_t
- In the 2nd eq-n we have n knowns a and $n+2$ unknowns (τ_1, τ_2, d_2)
- Two exclusion restrictions is sufficient to identify τ_1 (and τ_2)

Unobserved state

- Until now we have assumed observed state
 - This assumption allowed us to run the OLS critical for identification
- What changes when the state has to be estimated?
 - Nothing
- Have to discuss identification and estimation of the state

Identifying the state

- The state can be “rotated”, $\hat{x}_t = T x_t$, without affecting implications for observed variables
- The new state is:

$$\hat{x}_{t+1} = T A T^{-1} \hat{x}_t + T C w_{t+1} = \hat{A} \hat{x}_t + \hat{C} w_{t+1}.$$

- The interest rate is $i_t = r + a^\top T^{-1} \hat{x}_t = r + \hat{a}^\top \hat{x}_t$
- Similarly, $\hat{b}^\top = b^\top T^{-1}$
- The Taylor rule implies

$$\hat{a}^\top = \tau \hat{b}^\top + \hat{d}_2^\top,$$

- Identification of τ is exactly the same as before
- Have to impose restrictions to fix a state: $C = I$ and lower-triangular A is a common choice

Transformation-invariant restrictions

- The impact on restrictions: $d_2^\top e = 0$ translates into $\hat{d}_2^\top \hat{e} = 0$ with $\hat{e} = Te$.
- If T is unknown, can we deduce \hat{e} ?

Transformation-invariant restrictions

- The impact on restrictions: $d_2^\top e = 0$ translates into $\hat{d}_2^\top \hat{e} = 0$ with $\hat{e} = Te$.
- If T is unknown, can we deduce \hat{e} ?
- Ex. 1: The TR shock is uncorrelated with the other (Euler equation) shock – standard in the New-Keynesian literature

$$\hat{d}_2^\top E(\hat{x}\hat{x}^\top)\hat{d}_1 = (d_2^\top T^{-1})(TV_x T^\top)(d_1^\top T^{-1})^\top = d_2^\top V_x d_1$$

- Ex. 2: In the optimal monetary policy setting all variables are affected by s_{1t} , thus $s_{2t} = ks_{1t}$
 - The implied restriction is $d_2^\top - kd_1^\top = 0$
 - In terms of the transformed state:

$$\hat{d}_2^\top - k\hat{d}_1^\top = d_2^\top T^{-1} - kd_1^\top T^{-1} = (d_2^\top - kd_1^\top)T^{-1} = 0$$

Estimating the state

- Introduce a measurement equation:

$$y_t = Gx_t + Hv_t.$$

- The Kalman filter, $x_{t|t} = E(x_t|y^t)$, recovers all states when (A, G) is observable:

$$\text{rank} \begin{bmatrix} G \\ GA \\ \vdots \\ GA^{n-1} \end{bmatrix} = n$$

- As a result, $i_t = a^\top x_{t|t} + a^\top \varepsilon_t$, $\varepsilon_t \perp y^t$
- All the earlier logic applies still by replacing x_t with $x_{t|t}$

Term Structure and Survey Forecasts

- Financial and survey data are rich sources of information about the unobserved state
- In our models forward rates f_t^h and forecasts $F_t(\cdot)$ for horizon h are

$$\begin{aligned}f_t^h &= a^\top A^h x_t + v_t \\F_t(\pi_{t+h}) &= E_t(\pi_{t+h}) + v_t = b^\top A^h x_t + v_t \\F_t(g_{t+h}) &= E_t(g_{t+h}) + v_t = c^\top A^h x_t + v_t\end{aligned}$$

- Usually x_t is low-dimensional and $\dim(y_t)$ is large. In these cases, researchers attach measurement errors to all observables y_t and use the Kalman filter to estimate the state.

Conclusion

- Brute-force identification of the Taylor rule is impossible
- Rational expectations framework brings information from the whole system to bear on the TR coefficients
- We offer a constructive approach towards identification
- In general, one needs exclusion restrictions on MP shocks to identify TR; typical models impose more than what's required