# Materials 13 - Still looking for a version of the model w/o overshooting

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# Overview

| 1 | Model summary                                | 1 |
|---|--|---|
| 2 | Ideas  | 2 |
| 3 | Details on the Markov-switching setup        | 4 |
| 4 | An issue again with those endogenous states! | 6 |

# 1 Model summary

$$x_{t} = -\sigma i_{t} + \hat{\mathbb{E}}_{t} \sum_{T=t}^{\infty} \beta^{T-t} \left( (1 - \beta) x_{T+1} - \sigma(\beta i_{T+1} - \pi_{T+1}) + \sigma r_{T}^{n} \right)$$
 (1)

$$\pi_t = \kappa x_t + \hat{\mathbb{E}}_t \sum_{T=t}^{\infty} (\alpha \beta)^{T-t} \left( \kappa \alpha \beta x_{T+1} + (1-\alpha) \beta \pi_{T+1} + u_T \right)$$
 (2)

$$i_t = \psi_\pi \pi_t + \psi_x x_t + \bar{i}_t \tag{3}$$

$$\hat{\mathbb{E}}_t z_{t+h} = \begin{bmatrix} \bar{\pi}_{t-1} \\ 0 \ (\bar{x}_{t-1}) \\ 0 \ (\bar{i}_{t-1}) \end{bmatrix} + b h_x^{h-1} s_t \quad \forall h \ge 1 \qquad b = g_x \ h_x \qquad \text{PLM}$$

$$(4)$$

$$\bar{\pi}_t = \bar{\pi}_{t-1} + k_t^{-1} \underbrace{\left(\pi_t - (\bar{\pi}_{t-1} + b_1 s_{t-1})\right)}_{\text{fest error using (4)}} \qquad (b_1 \text{ is the first row of } b)$$
 (5)

$$k_t = \begin{cases} k_{t-1} + 1 & \text{for decreasing gain learning} \\ \bar{g}^{-1} & \text{for constant gain learning.} \end{cases}$$
 (6)

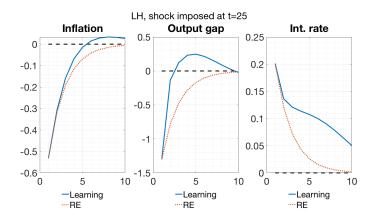
LH, shock imposed at t=25 Inflation **Output gap** Int. rate Inflation Output gap Int. rate 0.2 0.8 0.6 -0.1 0.6 0.4 -0.2 -0.5 -0.2 0.4 -0.3 0.2 -0.4 -0.4 0.2 0 -1.5 -0.5 -0.6 -0.2 \_ 0 0 -2<sub>0</sub> -0.6 0 20 20 20 10 20 20 10 20 10 10 10 10 -Learning ---RE -Learning Learning -Learning -Learning Learning (a) Learning constant only (b) Learning slope and constant

Figure 1: Reference: baseline model

# 2 Ideas

1. Check  $\psi_{\pi}$  above but close to 1, or lower  $\beta$ 

 $\rightarrow$  works but only quantitatively; qualitatively, the overshooting is still there, likely because this only cancels out one of the two channels through which  $\mathbb{E}\pi$  affects  $x_t$  negatively.



**Figure 2:** Baseline,  $\beta = 0.96$  instead of 0.99

#### 2. Fix shock for simulation

Indeed the issue was that for learning, I accidentally scaled down the shock by  $\sigma_i < 1$ , while for RE I had maintained  $\sigma_i = 1$ .

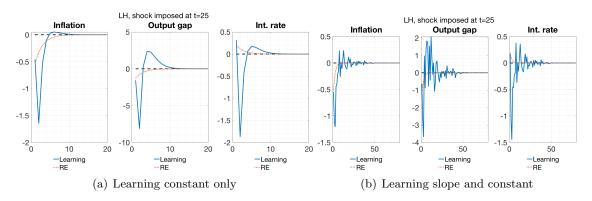
3. Interest rate smoothing as  $i_t = \rho i_{t-1} + (1-\rho)(\psi_\pi \pi_t + \psi_x x_t) + \bar{i}_t$ Doesn't work either - it doesn't change the model except reduces  $\psi_\pi$ .

#### 4. Indexation in NKPC

Doesn't work either - same model dynamics.

## 5. Learn $h_x$

Figure 3: Learning  $h_x$ , baseline



Like learning the Taylor rule b/c agents initially don't know if the shock will continue.

## 6. Central bank's $\mathbb{E} \pi_{t+1}$ in TR?

Done a correction for  $\mathbb{E}\pi_{t+1}$  in TR, now both are stable, but overshooting is still there in both. Not so dissimilar to baseline except that the periods are shifted.

#### 7. Initialize beliefs away from RE somehow

Slobodyan & Wouters do this, but in an estimation context, which I think is necessary because you need pre-sample data to condition priors on.

8. Slobodyan & Wouters' "VAR-learning": use lagged observables to learn from, not from states.

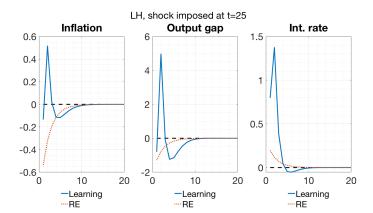
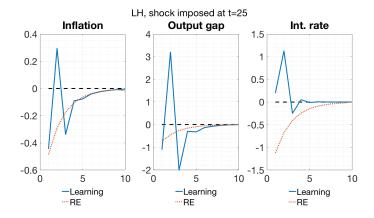


Figure 4: VAR learning, baseline, learning only constant

For learning both slope and constant, not E-stable. Kind of makes sense since I'd think that this amplifies positive feedback.

9. Davig & Leeper-style switching Taylor rule where only long-run Taylor principle holds?

Figure 5: Markov-switching Taylor rule, baseline, learning only constant (slope learning unstable, why?)



10. Some kind of moving average of inflation (or average) in the TR?

A quick question on projection facility: checking eig(phi) when  $\phi$  isn't square? Right now I'm splitting up  $\phi$  as a = phi(:,1), b = phi(:,2:end) and then checking eig(b) and eig(diag(a)).

# 3 Details on the Markov-switching setup

Model equations remain the same, except the Taylor rule now is:

$$i_t = \psi_\pi(r_t)\pi_t + \psi_x x_t + \bar{i}_t \tag{7}$$

$$r_t = \begin{cases} 1 & \text{active regime} \quad \psi_{\pi} = \psi_1 = 1.8 \\ 2 & \text{passive regime} \quad \psi_{\pi} = \psi_2 = 0.89 \end{cases}$$
 (8)

$$r_{t+1} = \begin{cases} p_{11}1 + (1 - p_{11})2 & \text{if} \quad r_t = 1\\ (1 - p_{22})1 + p_{22}2 & \text{if} \quad r_t = 2 \end{cases} \quad \text{where} \quad p_{ji} \equiv Prob(s_{t+1} = j | s_t = i)$$
 (9)

So I solve the RE model by introducing the new jump variables  $\pi_{it}$ ,  $x_{it}$ ,  $i_{it}$ , i = 1, 2 and writing the model equations as

$$x_{it} = (p_{1i} \mathbb{E}_t x_{1t+1} + p_{2i} \mathbb{E}_t x_{2t+1}) - \sigma(i_{it} - (p_{1i} \mathbb{E}_t \pi_{1t+1} + p_{2i} \mathbb{E}_t \pi_{2t+1})) + \sigma r_t^n$$
(10)

$$\pi_{it} = \kappa x_{it} + \beta (p_{1i} \,\mathbb{E}_t \,\pi_{1t+1} + p_{2i} \,\mathbb{E}_t \,\pi_{2t+1})) + u_t \tag{11}$$

$$i_{it} = \psi_i \pi_{it} + \psi_x x_{it} + \bar{i}_t \tag{12}$$

Now I unleash the usual method of solving for the observable and state transition matrix  $g_x, h_x$ . The only difference will be that since the number of jumps now is double the old number,  $g_x$  will be  $2n_y \times n_x$ . Is it correct to interpret  $g_x(1) \equiv g_x(1:n_y,:)$  as pertaining to regime 1, and  $g_x(2) \equiv g_x(n_y+1:end,:)$  to regime 2?

Then, generating an exogenous regime sequence r, I compute RE IRFs as usual for the state block, but depending on the state, I use the corresponding block of  $g_x$ . With  $x_0$  being the impulse, so that  $IR_1^x = x_0$ , I do the following:

$$IR_t^y = \begin{cases} g_x(1)x_t & \text{if } r_t = 1\\ g_x(2)x_t & \text{if } r_t = 2 \end{cases}$$
 (13)

$$IR_{t+1}^x = h_x x_t \tag{14}$$

As for the learning model, the compact notation for the model was:

$$z_t = A_a f_a(t) + A_b f_b(t) + A_s s_t \tag{15}$$

where the A-matrices are functions of the parameters, including  $\psi_{\pi}$ . The LH expectations  $f_a$  and  $f_b$  are only functions of the learning coefficients a, b and of the state transition matrix  $h_x$ . So the only salient difference is that agents react to expectations differently depending on the regime: the A-matrices become state-dependent.

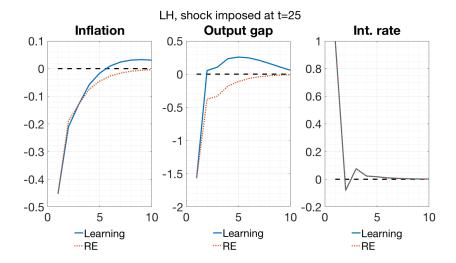
Is it correct then to simulate the model as

$$z_t = A_a(i)f_a(t) + A_b(i)f_b(t) + A_s(i)s_t \quad \text{for} \quad i = 1, 2$$
 (16)

?

# 4 An issue again with those endogenous states!

Figure 6: Lagged inflation in TR, "suboptimal forecasters" info assumption,  $\beta = 0.96$  instead of 0.99



Look at the interest rate: it doesn't move differently from the RE model. The code doesn't realize that the 4th state is the lagged 1st jump; my attempts to do the replacement have only lead to instability.