# Materials 9 - Still the overshooting

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#### 1 Model summary with interest rate smoothing $\rho i_{t-1}$

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

$$i_t = \psi_\pi \pi_t + \psi_x x_t + \rho i_{t-1} + \bar{i}_t \tag{3}$$

$$\hat{\mathbb{E}}_t z_{t+h} = \begin{bmatrix} \bar{\pi}_{t-1} \\ 0 \\ 0 \end{bmatrix} + bhx^{h-1} s_t \quad \forall h \ge 1 \qquad b = gx \ hx \qquad \text{PLM}$$
(4)

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

$$k_t = \mathbb{I} \times (k_{t-1} + 1) + (1 - \mathbb{I}) \times \bar{g}^{-1}$$
 (6)

$$\mathbb{I} = \begin{cases}
1 & \text{if } \theta_t \le \bar{\theta} \\
0 & \text{otherwise.} 
\end{cases} 
\tag{7}$$

$$\theta_t = |\hat{\mathbb{E}}_{t-1}\pi_t - \mathbb{E}_{t-1}\pi_t|/\sigma_s$$
 CEMP criterion for the gain (8)

The alternative criterion for the choice of gain is a recursive variant of the CUSUM-test (Brown, Durbin, Evans 1975):

- 1. Let  $FE_t$  denote the short-run forecast error, and  $\omega_t$  firms' estimate of the FE variance.
- 2. Let  $\kappa \in (0,1)$  and  $\tilde{\theta}$  be the new threshold value for the criterion.
- 3. Then for initial  $(\omega_0, \theta_0)$ , firms in every period estimate the criterion and the FEV as:

$$\omega_t = \omega_{t-1} + \kappa k_{t-1}^{-1} (FE_t^2 - \omega_{t-1}) \tag{9}$$

$$\theta_t = \theta_{t-1} + \kappa k_{t-1}^{-1} (F E_t^2 / \omega_t - \theta_{t-1})$$
(10)

$$k_t = \mathbb{I} \times (k_{t-1} + 1) + (1 - \mathbb{I}) \times \bar{g}^{-1}$$
 (11)

$$\mathbb{I} = 1 \quad \text{if} \quad \theta_t \le \tilde{\theta} \tag{12}$$

#### 2 Compact notation - with lagged interest rate term in TR

$$z_t = A_p^{RE} \, \mathbb{E}_t \, z_{t+1} + A_s^{RE} s_t \tag{13}$$

$$z_t = A_a^{LH} f_a(t) + A_b^{LH} f_b(t) + A_s^{LH} s_t (14)$$

$$s_t = Ps_{t-1} + \epsilon_t \qquad \rightarrow \quad s'_t = hx \ s'_{t-1} + \epsilon'_t \tag{15}$$

where 
$$s'_{t} \equiv \begin{pmatrix} r_{t}^{n} \\ \bar{i}_{t} \\ u_{t} \\ i_{t-1} \end{pmatrix}$$
  $hx \equiv \begin{pmatrix} \rho_{r} & 0 & 0 & 0 \\ 0 & \rho_{i} & 0 & 0 \\ 0 & 0 & \rho_{u} & 0 \\ gx_{3,1} & gx_{3,2} & gx_{3,3} & gx_{3,4} \end{pmatrix}$   $\epsilon'_{t} \equiv \begin{pmatrix} \varepsilon_{t}^{r} \\ \varepsilon_{t}^{i} \\ \varepsilon_{t}^{u} \\ 0 \end{pmatrix}$  and  $\Sigma' = \begin{pmatrix} \sigma_{r} & 0 & 0 & 0 \\ 0 & \sigma_{i} & 0 & 0 \\ 0 & 0 & \sigma_{u} & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$  (16)

 $i_{t-1}$  is an endogenous state and breaks the link that previously had P = hx; now this is no longer true. In particular, using Matlabby notation, P = hx(1:3,1:3).

And the  $A_s^{RE}$  and  $A_s^{LH}$  are given by:

$$A_s^{RE} = \begin{pmatrix} \frac{\kappa \sigma}{w} & -\frac{\kappa \sigma}{w} & 1 - \frac{\kappa \sigma \psi_{\pi}}{w} & 0\\ \frac{\sigma}{w} & -\frac{\sigma}{w} & -\frac{\sigma \psi_{\pi}}{w} & 0\\ \psi_x(\frac{\sigma}{w}) + \psi_{\pi}(\frac{\kappa \sigma}{w}) & \psi_x(-\frac{\sigma}{w}) + \psi_{\pi}(-\frac{\kappa \sigma}{w}) + 1 & \psi_x(-\frac{\sigma \psi_{\pi}}{w}) + \psi_{\pi}(1 - \frac{\kappa \sigma \psi_{\pi}}{w}) & \rho \end{pmatrix}$$
(17)

$$A_s^{LH} = \begin{pmatrix} g_{\pi s} & & & \\ g_{xs} & & & \\ \psi_{\pi} g_{\pi s} + \psi_x g_{xs} + \begin{bmatrix} 0 & 1 & 0 & \rho \end{bmatrix} \end{pmatrix}$$
 (18)

$$g_{\pi s} = (1 - \frac{\kappa \sigma \psi_{\pi}}{w}) \begin{bmatrix} 0 & 0 & 1 & 0 \end{bmatrix} (I_4 - \alpha \beta hx)^{-1} - \frac{\kappa \sigma}{w} \begin{bmatrix} -1 & 1 & 0 & \rho \end{bmatrix} (I_4 - \beta hx)^{-1}$$
(19)

$$g_{xs} = \frac{-\sigma\psi_{\pi}}{w} \begin{bmatrix} 0 & 0 & 1 & 0 \end{bmatrix} (I_4 - \alpha\beta hx)^{-1} - \frac{\sigma}{w} \begin{bmatrix} -1 & 1 & 0 & \rho \end{bmatrix} (I_4 - \beta hx)^{-1}$$
 (20)

### 3 Recap of timing

Define some objects: (I usually let t denote the time in which the variable is formed.)

$$f_t^j = \hat{\mathbb{E}}_t(z_{t+1})$$
 one-period-ahead forecast formed at time  $t, j = m, e$  (morning or evening) (21)

$$FE_t = z_{t+1} - f_t$$
 one-period-ahead forecast error realized at time  $t+1$  (22)

$$= ALM(t+1) - PLM(t) \tag{23}$$

$$\theta_t = \hat{\mathbb{E}}_{t-1}(z_t) - \mathbb{E}_{t-1}(z_t)$$
 CEMP's criterion (24)

$$= PLM(t-1) - \mathbb{E}_{t-1} ALM(t) \tag{25}$$

$$PLM(t): \hat{\mathbb{E}}_t z_{t+1} = \bar{z}_{t-1} + bs_t$$

**Morning**: morning of time t available:  $\mathcal{I}_t^m = \{\bar{z}_{t-1}, s_t, k_{t-1}, FE_{t-2}\}$ 

- 1. Form all future expectations using PLM(t) (morning forecast)  $\to z_t$  realized,  $\to FE_{t-1}$  realized
- 2. Form  $\theta_t \to k_t$  realized
- 3. **Evening**: Update  $\bar{z}_t = \bar{z}_{t-1} + k_t^{-1}(FE_{t-1}^e)$

where  $FE_{t-1}^e = z_t - f_{t-1}^e = z_t - (\bar{z}_{t-1} + bs_{t-1})$  is the most recent realized FE, so:

$$\bar{z}_t = \bar{z}_{t-1} + k_t^{-1}(z_t - (\bar{z}_{t-1} + bs_{t-1}))$$

 $\rightarrow$  evening of time t available:  $\mathcal{I}^e_t = \{\bar{z}_t, s_t, k_t, FE_{t-1}\}$ 

## 4 Current set of baseline parameters

$\beta$	0.99	stochastic discount factor	standard (Woodford 2003/2011)
$\sigma$	1	IES	consistent with balanced growth
$\alpha$	0.5	Calvo probability of not adjusting	match 6-month duration of prices (can increase to 0.75)
$\overline{\psi_{\pi}}$	1.5	coefficient of inflation in Taylor rule	Taylor
$\overline{\psi_x}$	0	coefficient of output gap in Taylor rule	focus on $\pi$
$\bar{g}$	0.145	value of the constant gain	CEMP
$ar{ heta}$	1	threshold deviation between $\hat{\mathbb{E}}~\&~\mathbb{E}$	CEMP: 0.029
$ ho_r$	0	persistence of natural rate shock	n.a.
$ ho_i$	0.6	persistence of monetary policy shock	CEMP: 0.877 (can increase to 0.78 if $\alpha = 0.75$ )
$ ho_u$	0	persistence of cost-push shock	CEMP
$\sigma_r$	0.1	standard deviation of natural rate shock	n.a.
$\sigma_i$	0.359	standard deviation of mon. policy shock	CEMP
$\sigma_u$	0.277	standard deviation of cost-push shock	CEMP
$\theta$	10	price elasticity of demand	Woodford 2003/2011, Chari, Kehoe & McGrattan 2000
$\omega$	1.25	elasticity of marginal cost to output	Woodford 2003/2011, Chari, Kehoe & McGrattan 2000

#### Cross-sectional IRFs, mon. pol shock only, cgain & dgain only ◀ 5

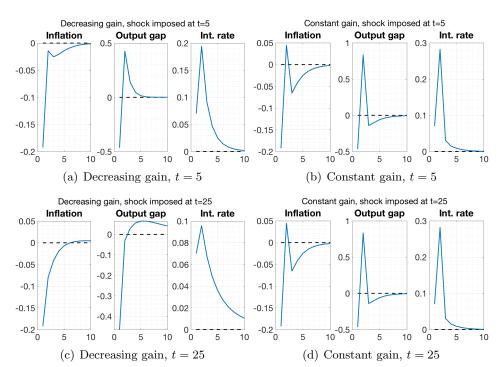


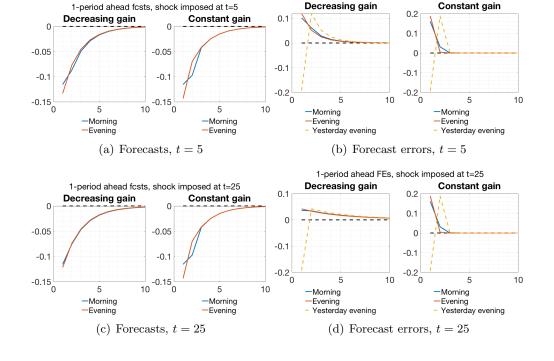
Figure 1: IRF for observables, shock imposed at t

Figure 2: IRF for 1-period ahead forecasts and FEs, together, morning and evening, shock imposed at t

1-period ahead FEs, shock imposed at t=5

Constant gain

Decreasing gain



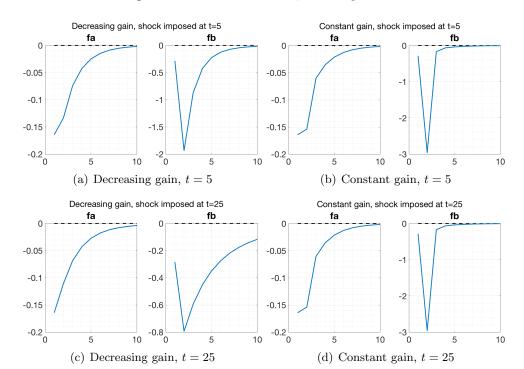


Figure 3: IRF for LH forecasts, shock imposed at t

4 puzzling things, 4 attempts to explain them:

#### 1. Criss-crossing of expectations

If the gain is high enough, expectations overupdate, FE switch signs. (Can make extremely small with lower gain, e.g.  $\bar{g} = 0.02$ , the value of the dgain after 50 periods.)

Note also that  $f_a$ ,  $f_b$  are mainly driven by the intercept, not the slope. The intercept term is  $\frac{\bar{\pi}}{1-\alpha\beta}$  and  $\frac{\bar{\pi}}{1-\beta}$ . So when  $\alpha$  is far from 1,  $f_b$  is much bigger

- both in absolute terms,
- as well as relatively to  $f_a$ .

#### 2. $i \uparrow \text{ at } t = 2 \text{ after shock}$

 $i = \pi \downarrow + \rho_i^{t-1} \delta \uparrow$ . So b/c of correction in  $\pi$  at  $t = 2, |\rho_i^{t-1} \delta| - |\pi|$  becomes larger.

#### 3. Overshooting later than t=2

IR decomposed into two effects pulling opposite ways:  $IR \approx \mathbb{E}(\pi) + \delta$ . Overshooting happens when the expectation effect is larger. As the shock recedes faster than the expectation,  $x, \pi$ 

overshoot. Testable implication: if shock iid, overshooting should happen at t = 2, and indeed it does.

4. Responses to expectations in RE vs learning (recursive vs. LH-horizon)

RE

Here's the stylized representation of how endogenous variables respond to expectations in the two formulations (once you've plugged in the interest rate):

$$RE$$

$$x_{t} = \mathbb{E}(\pi) + \mathbb{E}(x)$$

$$\pi_{t} = \mathbb{E}(\pi) + \mathbb{E}(x)$$

$$Learning$$

$$x_{t} = \mathbb{E}(\pi) + \mathbb{E}(x)$$

$$\pi_{t} = \mathbb{E}(\pi) + \mathbb{E}(x)$$

The difference, marked in red, comes from the fact in learning, expectations are more explicit. Ignoring shocks and setting  $\psi_x = 0$ , so the Taylor rule is just  $i_t = \psi_\pi \pi_t$ , the two systems are

$$x_{t} = -\sigma i_{t} + \mathbb{E}_{t} x_{t+1} + \sigma \mathbb{E}_{t} \pi_{t+1}$$

$$\pi_{t} = \kappa x_{t} + \beta \mathbb{E}_{t} \pi_{t+1}$$

$$Learning$$

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

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

When you plug in the interest rate, you see that the recursive representation hides the negative part of  $x_t$ 's dependance on its own future values into current inflation and its dependance on

future inflation:

$$RE$$

$$x_{t} = -\sigma \psi_{\pi} \pi_{t} + \mathbb{E}_{t} x_{t+1}^{+} + \sigma \mathbb{E}_{t} \pi_{t+1}$$

$$\pi_{t} = \kappa x_{t} + \beta \mathbb{E}_{t} \pi_{t+1}$$

$$Learning$$

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

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

To align with this interpretation, it must be the case that  $\pi_t$ 's dependance on future inflation is stronger under RE than learning. Comparing coefficients for my baseline parameterization, this is true.