Materials 31 - Parameterized expectations against value iteration

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The value function problem: the planner chooses $\{i_t\}_{t=t_0}^{\infty}$ to minimize

$$V(\mathbf{x}_t, t) = \min \left\{ (\pi_t^2 + \lambda_x x_t^2) + \beta \, \mathbb{E}_t \, V(\mathbf{x}_{t+1}, t+1) \right\}$$
 (1)

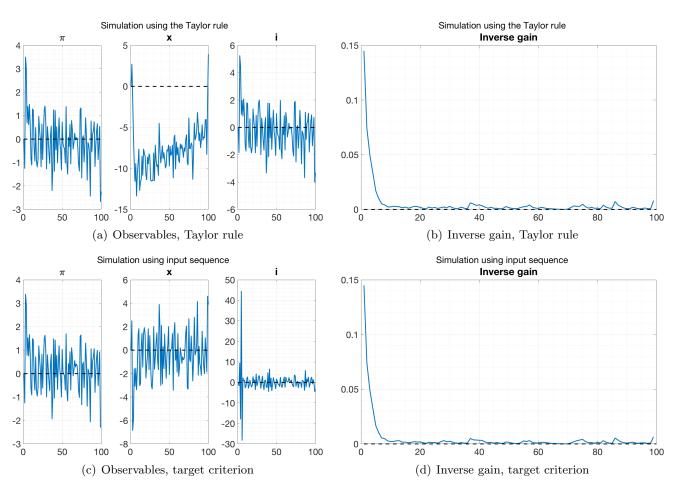
s.t. to model equations,
$$(A.1)$$
- $(A.8)$ (2)

where $\mathbf{x} \equiv (k^{-1}, \bar{\pi}, r^n, u)$. V is thus $\mathcal{R}^4 \to \mathcal{R}$.

Note:

- I'm using a $4 \times 4 \times 2 \times 2$ grid, and cubic spline interpolation for the value function. This takes 260 seconds. I'm sure I could speed things up by doing policy iteration instead.
- Since equations (A.5)-(A.7) involve past shocks, I need to take expectations regarding what node of the state-space we're coming from ("backward expectations").
- How to compare the results? For value iteration, I get decision rules for $i, \bar{\pi}$ and k^{-1} given grid values, while for projection I get a sequence for these given a simulated time series of shocks. I'm thinking that if I'd approximate the policy function instead of the value function, I could pass the same time series of states to that as I obtained with projection and see if I get the same interest rate sequence.

Figure 1: Parameterized expectations solution



A Model summary

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

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

$$i_t = \psi_\pi \pi_t + \psi_x x_t + \bar{i}_t$$
 (if imposed) (A.3)

PLM:
$$\hat{\mathbb{E}}_t z_{t+h} = a_{t-1} + b h_x^{h-1} s_t \quad \forall h \ge 1 \qquad b = g_x h_x$$
 (A.4)

Updating:
$$a_t = a_{t-1} + k_t^{-1} (z_t - (a_{t-1} + bs_{t-1}))$$
 (A.5)

Anchoring function:
$$k_t = k_{t-1} + \mathbf{g}(fe_{t-1}^2)$$
 (A.6)

Forecast error:
$$fe_{t-1} = z_t - (a_{t-1} + bs_{t-1})$$
 (A.7)

LH expectations:
$$f_a(t) = \frac{1}{1 - \alpha \beta} a_{t-1} + b(\mathbb{I}_{nx} - \alpha \beta h)^{-1} s_t$$
 $f_b(t) = \frac{1}{1 - \beta} a_{t-1} + b(\mathbb{I}_{nx} - \beta h)^{-1} s_t$ (A.8)

This notation captures vector learning (z learned) for intercept only. For scalar learning, $a_t = \begin{pmatrix} \bar{a}_t & 0 & 0 \end{pmatrix}'$ and b_1 designates the first row of b. The observables (π, x) are determined as:

$$x_t = -\sigma i_t + \begin{bmatrix} \sigma & 1 - \beta & -\sigma \beta \end{bmatrix} f_b + \sigma \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} (\mathbb{I}_{nx} - \beta h_x)^{-1} s_t$$
 (A.9)

$$\pi_t = \kappa x_t + \begin{bmatrix} (1 - \alpha)\beta & \kappa \alpha \beta & 0 \end{bmatrix} f_a + \begin{bmatrix} 0 & 0 & 1 \end{bmatrix} (\mathbb{I}_{nx} - \alpha \beta h_x)^{-1} s_t$$
 (A.10)

B Target criterion

The target criterion in the simplified model (scalar learning of inflation intercept only, $k_t^{-1} = \mathbf{g}(fe_{t-1})$):

$$\pi_{t} = -\frac{\lambda_{x}}{\kappa} \left\{ x_{t} - \frac{(1-\alpha)\beta}{1-\alpha\beta} \left(k_{t}^{-1} + ((\pi_{t} - \bar{\pi}_{t-1} - b_{1}s_{t-1})) \mathbf{g}_{\pi}(t) \right) \right\}$$

$$\left(\mathbb{E}_{t} \sum_{i=1}^{\infty} x_{t+i} \prod_{j=0}^{i-1} (1 - k_{t+1+j}^{-1} - (\pi_{t+1+j} - \bar{\pi}_{t+j} - b_{1}s_{t+j}) \mathbf{g}_{\bar{\pi}}(t+j)) \right)$$
(B.1)

where I'm using the notation that $\prod_{j=0}^{0} \equiv 1$. For interpretation purposes, let me rewrite this as follows:

$$\pi_{t} = -\frac{\lambda_{x}}{\kappa} x_{t} + \frac{\lambda_{x}}{\kappa} \frac{(1-\alpha)\beta}{1-\alpha\beta} \left(k_{t}^{-1} + f e_{t|t-1}^{eve} \mathbf{g}_{\pi}(t) \right) \mathbb{E}_{t} \sum_{i=1}^{\infty} x_{t+i}$$

$$-\frac{\lambda_{x}}{\kappa} \frac{(1-\alpha)\beta}{1-\alpha\beta} \left(k_{t}^{-1} + f e_{t|t-1}^{eve} \mathbf{g}_{\pi}(t) \right) \left(\mathbb{E}_{t} \sum_{i=1}^{\infty} x_{t+i} \prod_{j=0}^{i-1} (k_{t+1+j}^{-1} + f e_{t+1+j|t+j}^{eve}) \mathbf{g}_{\pi}(t+j) \right)$$
(B.2)

Interpretation: tradeoffs from discretion in RE + effect of current level and change of the gain on future tradeoffs + effect of future expected levels and changes of the gain on future tradeoffs