Monetary Policy & Anchored Expectations An Endogenous Gain Learning Model

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Policymakers came out of the Great Inflation era with a clear understanding that it was essential to anchor inflation expectations at some low level.

Jerome Powell, Chairman of the Federal Reserve ¹

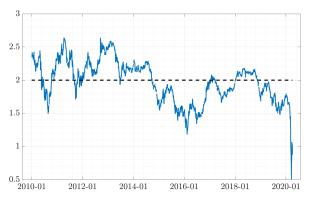


Figure: Market-based inflation expectations, 10 year, average, %

¹Federal Reserve "Challenges for Monetary Policy," August 23, 2019.

This project

• Estimation of the anchoring function: when do expectations become unanchored?

 Model of anchoring expectation formation as an endogenous gain adaptive learning scheme

→ How to conduct optimal monetary policy in interaction with the anchoring expectation formation?

Preview of results

• A 1%-point forecast error unanchors expectations

• Optimal monetary policy responsiveness time-varying

 \hookrightarrow Unanchored expectations introduce an intertemporal volatility tradeoff

→ Illustrate analytically in special case: target criterion

Related literature

 Optimal monetary policy in New Keynesian models Clarida, Gali & Gertler (1999), Woodford (2003)

• Econometric learning

Evans & Honkapohja (2001, 2006), Preston (2005, 2008), Molnár & Santoro (2014), Ferrero (2007), Eusepi & Preston (2011), Milani (2007, 2014), Lubik & Matthes (2018)

• Anchoring and the Phillips curve

Sargent (1999), Svensson (2015), Hooper et al (2019), Afrouzi & Yang (2020), Gobbi et al (2019), Carvalho et al (2019)

Structure of talk

1. Model

2. Solving the Ramsey problem

3. Implications

Households: standard up to $\hat{\mathbb{E}}$

Maximize lifetime expected utility

$$\hat{\mathbb{E}}_t \sum_{T=t}^{\infty} \beta^{T-t} \left[U(C_T^i) - \int_0^1 v(h_T^i(j)) dj \right]$$
 (1)

Budget constraint

$$B_t^i \le (1 + i_{t-1})B_{t-1}^i + \int_0^1 w_t(j)h_t^i(j) + \Pi_t^i(j)dj - T_t - P_tC_t^i$$
 (2)



Firms: standard up to $\hat{\mathbb{E}}$

Maximize present value of profits

$$\hat{\mathbb{E}}_t \sum_{T=t}^{\infty} \alpha^{T-t} Q_{t,T} \left[\Pi_t^j(p_t(j)) \right]$$
 (3)

subject to demand

$$y_t(j) = Y_t \left(\frac{p_t(j)}{P_t}\right)^{-\theta} \tag{4}$$

▶ Profits, stochastic discount factor

Expectations: $\hat{\mathbb{E}}$ instead of \mathbb{E}

• If use \mathbb{E} (rational expectations, RE)

Model solution

$$s_t = hs_{t-1} + \epsilon_t \qquad \epsilon_t \sim \mathcal{N}(\mathbf{0}, \Sigma)$$

$$y_t = gs_t$$
(5)

$$s_t \equiv (r_t^n, u_t)'$$
 (states)
 $y_t \equiv (\pi_t, x_t, i_t)'$ (jumps)

- If use Ê → private sector does not know g
 → estimate using observed states & knowledge of (5)
- Households and firms don't know they are identical

Adaptive learning

- Estimate *g* using recursive least squares (RLS)
 - \rightarrow nonrational expectations:

$$\hat{\mathbb{E}}_t y_{t+1} = \phi_{t-1} \begin{bmatrix} \mathbf{1} \\ s_t \end{bmatrix} \tag{7}$$

• Note: misspecified

Can write:

$$\hat{\mathbb{E}}_t y_{t+1} = a_{t-1} + b_{t-1} s_t \tag{8}$$

In RE,
$$a_{t-1} = (0, 0, 0)'$$
, $b_{t-1} = g h \quad \forall t$

Recursive least squares

Special case: learn only intercept of inflation:

$$a_{t-1} = (\bar{\pi}_{t-1}, 0, 0)', \quad b_{t-1} = g h \quad \forall t$$
 (9)

$$\rightarrow$$
 RLS

$$\bar{\pi}_t = \bar{\pi}_{t-1} + k_t \underbrace{\left(\pi_t - (\bar{\pi}_{t-1} + b_1 s_{t-1})\right)}_{\equiv fe_{t|t-1}, \text{ forecast error}}$$
(10)

 $k_t \in (0,1)$ gain b_1 first row of b



Anchoring mechanism: endogenous gain

Gain in literature usually exogenous:

$$k_t = \begin{cases} \frac{1}{t} & \text{decreasing} \\ k & \text{constant} \end{cases}$$

Here instead

$$k_t = k_{t-1} + \mathbf{g}(fe_{t|t-1})$$
 (11)

▶ Functional forms

Model summary

• IS- and Phillips curve:

$$x_{t} = -\sigma i_{t} + \hat{\mathbb{E}}_{t} \sum_{T=t}^{\infty} \beta^{T-t} \left((1 - \beta) x_{T+1} - \sigma(\beta i_{T+1} - \pi_{T+1}) + \sigma r_{T}^{n} \right)$$
(12)

$$\pi_t = \kappa x_t + \hat{\mathbb{E}}_t \sum_{T=0}^{T} (\alpha \beta)^{T-t} \left(\kappa \alpha \beta x_{T+1} + (1-\alpha) \beta \pi_{T+1} + u_T \right)$$
 (13)

▶ Derivations

- Expectations evolve according to RLS with the endogenous gain given by (11)
- \rightarrow How should $\{i_t\}$ be set?

Structure of talk

1. Model

2. Solving the Ramsey problem

3. Implications

Ramsey problem

$$\min_{\{y_t, \bar{\pi}_{t-1}, k_t\}_{t=t_0}^{\infty}} \mathbb{E}_{t_0} \sum_{t=t_0}^{\infty} \beta^{t-t_0} (\pi_t^2 + \lambda_x x_t^2)$$

s.t. model equations

- E is the central bank's (CB) expectation
- Assumption: CB observes private expectations and knows the model

Special case

- Only inflation intercept learned
- Anchoring function simplified to

$$k_t = \mathbf{g}(fe_{t|t-1}) \tag{14}$$

Target criterion for special case

Result

In the simplified model with anchoring, monetary policy optimally brings about the following target relationship between inflation and the output gap

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

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

where
$$\mathbf{g}_{z,t} \equiv \frac{\partial \mathbf{g}}{\partial z}$$
 at t , $\prod_{i=0}^{0} \equiv 1$ and b_1 is the first row of b .

Two layers of intertemporal tradeoffs

$$\begin{split} \pi_t &= -\frac{\lambda_x}{\kappa} x_t + \frac{\lambda_x}{\kappa} \frac{(1-\alpha)\beta}{1-\alpha\beta} \left(k_t + f e_{t|t-1} \mathbf{g}_{\pi,t} \right) \mathbb{E}_t \sum_{i=1}^{\infty} x_{t+i} \\ &- \frac{\lambda_x}{\kappa} \frac{(1-\alpha)\beta}{1-\alpha\beta} \left(k_t + f e_{t|t-1} \mathbf{g}_{\pi,t} \right) \mathbb{E}_t \sum_{i=1}^{\infty} x_{t+i} \prod_{j=0}^{i-1} (k_{t+1+j} + f e_{t+1+j|t+j} \mathbf{g}_{\pi,t+j}) \end{split}$$

Intratemporal tradeoffs in RE (discretion)

Intertemporal tradeoff: current level and change of the gain

Intertemporal tradeoff: future expected levels and changes of the gain

Corollary

Optimal policy under adaptive learning is time-inconsistent.

Lemma

The commitment solution of the Ramsey problem does not exist under adaptive learning.

➤ Why no commitment?

Structure of talk

1. Model

2. Solving the Ramsey problem

3. Implications

Implementation: feedback rule

- Related issue under RE: optimal interest rate sequence implies indeterminate equilibrium
- ⇒ Reaction function stabilizes expectations

Recall IS-curve:

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

• E.g. Taylor rule disciplines expectations:

$$\hat{\mathbb{E}}_t i_T = \psi_\pi \hat{\mathbb{E}}_t \pi_T + \psi_x \hat{\mathbb{E}}_t x_T$$

Next steps: form of reaction function

• Model suggests $i_t = \mathbf{f}(\pi_t, k_t, \bar{\pi}_{t-1}; t)$ nonlinear

• Explains deviations from Taylor rule

 However: no commitment makes Taylor rule more viable than under RE as a rough approximation of optimal feedback rule

• If Taylor rule, model prefers being less aggressive on inflation

Conclusion

• Interaction between monetary policy and anchoring

 Optimal policy conditions on stance of current and expected future anchoring

 $\hookrightarrow determine\ intertemporal\ tradeoffs$

• Explain departures from the Taylor rule like US, fall 2019

• If Fed acted to anchor expectations, then Missing Deflation and Inflation are not "missing"

Short-run costs, long-run benefits

Assume Taylor rule and no concern for output gap stabilization

$$i_t = \psi_\pi \pi_t \qquad \lambda_x = 0$$

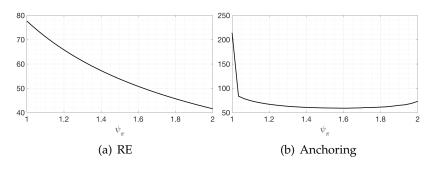


Figure: Central bank loss as a function of ψ_{π}



Functional forms for g

• Smooth anchoring function

$$k_t = k_{t-1} - c + df e_{t|t-1}^2 (15)$$

c, d > 0

• Kinked anchoring function

$$k_t = \begin{cases} \frac{1}{t} & \text{when } \theta_t < \bar{\theta} \\ k & \text{otherwise.} \end{cases}$$
 (16)

 θ_t criterion, $\bar{\theta}$ threshold value



Choices for criterion θ_t

• Carvalho et al. (2019)'s criterion

$$\theta_t^{CEMP} = \max |\Sigma^{-1}(\phi_{t-1} - T(\phi_{t-1}))|$$
 (17)

 Σ variance-covariance matrix of shocks $T(\phi)$ mapping from PLM to ALM

CUSUM-criterion

$$\omega_t = \omega_{t-1} + \kappa k_{t-1} (f e_{t|t-1} f e'_{t|t-1} - \omega_{t-1})$$
(18)

$$\theta_t^{CUSUM} = \theta_{t-1} + \kappa k_{t-1} (f e'_{t|t-1} \omega_t^{-1} f e_{t|t-1} - \theta_{t-1})$$
 (19)

 ω_t estimated forecast-error variance



Recursive least squares algorithm

$$\phi_t = \left(\phi'_{t-1} + k_t R_t^{-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \left(y_t - \phi_{t-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \right)' \right)' \tag{20}$$

$$R_t = R_{t-1} + k_t \left(\begin{bmatrix} 1 \\ S_{t-1} \end{bmatrix} \begin{bmatrix} 1 & S_{t-1} \end{bmatrix} - R_{t-1} \right)$$
 (21)



where

and

 $f_{a,t} \equiv \hat{\mathbb{E}}_t \sum_{t=0}^{\infty} (\alpha \beta)^{T-t} y_{T+1}$

Compact notation

$$s_t = h s_{t-1} + \epsilon_t$$

 $y_t = A_1 f_{a,t} + A_2 f_{b,t} + A_3 s_t$

$$s_t \equiv$$

$$s_t \equiv$$

$$y_t \equiv \begin{pmatrix} \pi_t \\ x_t \\ i_t \end{pmatrix} \qquad s_t \equiv \begin{pmatrix} r_t^n \\ \bar{i}_t \\ u_t \end{pmatrix}$$

 $f_{b,t} \equiv \hat{\mathbb{E}}_t \sum_{t=0}^{\infty} (\beta)^{T-t} y_{T+1}$

$$\begin{pmatrix} r_t^n \\ \overline{i}_t \\ u_t \end{pmatrix}$$

(22)

(23)

(24)

(25)

No commitment - no lagged multipliers

Simplified version of the model: planner chooses $\{\pi_t, x_t, f_t, k_t\}_{t=t_0}^{\infty}$ to minimize

$$\mathcal{L} = \mathbb{E}_{t_0} \sum_{t=t_0}^{\infty} \beta^{t-t_0} \left\{ \pi_t^2 + \lambda x_t^2 + \varphi_{1,t} (\pi_t - \kappa x_t - \beta f_t + u_t) + \varphi_{2,t} (f_t - f_{t-1} - k_t (\pi_t - f_{t-1})) + \varphi_{3,t} (k_t - \mathbf{g}(\pi_t - f_{t-1})) \right\}$$

$$2\pi_t + 2\frac{\lambda}{\kappa}x_t - \varphi_{2,t}(k_t + \mathbf{g}_{\pi}(\pi_t - f_{t-1})) = 0$$
 (26)

$$-2\beta \frac{\lambda}{\kappa} x_t + \varphi_{2,t} - \varphi_{2,t+1} (1 - k_{t+1} - \mathbf{g_f}(\pi_{t+1} - f_t)) = 0$$
 (27)



Short-run costs from oscillatory dynamics

Consider a stylized adaptive learning model in two equations:

$$\pi_t = \beta f_t + u_t \tag{28}$$

$$f_t = f_{t-1} + k(\pi_t - f_{t-1}) \tag{29}$$

Solve for the time series of expectations f_t

$$f_t = \underbrace{\frac{1 - k^{-1}}{1 - k^{-1}\beta}}_{\approx 1} f_{t-1} + \frac{k^{-1}}{1 - k^{-1}\beta} u_t \tag{30}$$

Solve for forecast error $fe_t \equiv \pi_t - f_{t-1}$:

$$fe_{t} = \underbrace{-\frac{1-\beta}{1-k\beta}}_{\lim_{k\to 1}=-1} f_{t-1} + \frac{1}{1-k\beta} u_{t}$$
 (31)

Target criterion system for anchoring function as changes of the gain

$$\varphi_{6,t} = -cfe_{t|t-1}x_{t+1} + \left(1 + \frac{fe_{t|t-1}}{fe_{t+1|t}}(1 - k_{t+1}) - fe_{t|t-1}\mathbf{g}_{\bar{\pi},t}\right)\varphi_{6,t+1} - \frac{fe_{t|t-1}}{fe_{t+1|t}}(1 - k_{t+1})\varphi_{6,t+2}$$
(32)

$$0 = 2\pi_t + 2\frac{\lambda_x}{\kappa}x_t - \left(\frac{k_t}{fe_{t|t-1}} + \mathbf{g}_{\pi,t}\right)\varphi_{6,t} + \frac{k_t}{fe_{t|t-1}}\varphi_{6,t+1}$$
(33)

 $\varphi_{6,t}$ Lagrange multiplier on anchoring function

The solution to (33) is given by:

$$\varphi_{6,t} = -2 \,\mathbb{E}_t \sum_{i=0}^{\infty} (\pi_{t+i} + \frac{\lambda_x}{\kappa} x_{t+i}) \prod_{j=0}^{i-1} \frac{\frac{k_{t+j}}{f e_{t+j|t+j-1}}}{\frac{k_{t+j}}{f e_{t+j|t+j-1}}} + \mathbf{g}_{\pi,t+j}$$
(34)



Details on households and firms

Consumption:

$$C_t^i = \left[\int_0^1 c_t^i(j)^{\frac{\theta - 1}{\theta}} dj \right]^{\frac{\circ}{\theta - 1}}$$

 $\theta > 1$: elasticity of substitution between varieties

Aggregate price level:

$$P_t = \left[\int_0^1 p_t(j)^{1-\theta} dj \right]^{\frac{1}{\theta-1}}$$

Profits:

$$\Pi_t^J = p_t(j)y_t(j) - w_t(j)f^{-1}(y_t(j)/A_t)$$

$$Q_{t,T} = \beta^{T-t} \frac{P_t U_c(C_T)}{P_T U_c(C_t)}$$

(38)

(36)

(35)



Derivations

Household FOCs

$$\hat{C}_t^i = \hat{\mathbb{E}}_t^i \hat{C}_{t+1}^i - \sigma(\hat{i}_t - \hat{\mathbb{E}}_t^i \hat{\pi}_{t+1})$$
(39)

$$\hat{\mathbb{E}}_t^i \sum_{s=0}^{\infty} \beta^s \hat{C}_t^i = \omega_t^i + \hat{\mathbb{E}}_t^i \sum_{s=0}^{\infty} \beta^s \hat{Y}_t^i$$
(40)

where 'hats' denote log-linear approximation and $\omega_t^i \equiv \frac{(1+i_{t-1})B_{t-1}^i}{P_tY^*}$.

- 1. Solve (39) backward to some date *t*, take expectations at *t*
- 2. Sub in (40)
- 3. Aggregate over households *i*
- \rightarrow Obtain (12)

