Monetary Policy & Anchored Expectations An Endogenous Gain Learning Model

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Inflation that runs below its desired level can lead to an unwelcome fall in longer-term inflation expectations, which, in turn, can pull actual inflation even lower, resulting in an adverse cycle of ever-lower inflation and inflation expectations. [...] Well-anchored inflation expectations are critical[.]

Jerome Powell, Chairman of the Federal Reserve ¹ (Emphases added.)

¹"New Economic Challenges and the Fed's Monetary Policy Review," August 27, 2020.

Long-run expectations: capturing responsiveness to short-run conditions

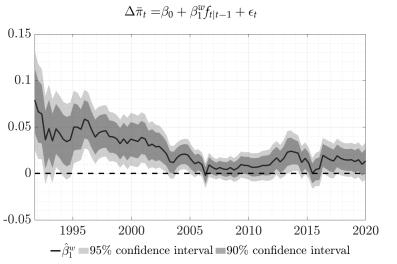
Individual SPF forecasts: for 1991-Q4 onward, estimate rolling regression

$$\Delta \bar{\pi}_t = \beta_0 + \beta_1^w f_{t|t-1} + \epsilon_t \tag{1}$$

where w indexes windows of 20 quarters length,

 $f_{t|t-1} \equiv \pi_t - \mathbb{E}_{t-1} \, \pi_t$ individual one-year-ahead forecast error

Time-varying responsiveness



This project

 How to conduct monetary policy when expectations can become unanchored?

- Estimate how unanchoring takes place in data

 → quantify novel anchoring channel

Preview of results

1. Estimation

Larger mistakes unanchor more

2. Optimal policy

 Responds aggressively to inflation when unanchored, accommodates inflation when anchored

3. Taylor rule

Less aggressive on inflation than under rational expectations

Related literature

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

Adaptive learning

Evans & Honkapohja (2001, 2006), Sargent (1999), Primiceri (2006), Lubik & Matthes (2018), Bullard & Mitra (2002), Preston (2005, 2008), Ferrero (2007), Molnár & Santoro (2014), Mele et al (2019), Eusepi & Preston (2011), Milani (2007, 2014), Marcet & Nicolini (2003)

• Anchoring and the Phillips curve

Goodfriend (1993), Svensson (2015), Hooper et al (2019), Afrouzi & Yang (2020), Reis (2020), Hebden et al 2020, Gobbi et al (2019), Carvalho et al (2019)

Structure of talk

- 1. Model of anchoring expectations
- 2. Quantification of learning channel
- 3. Solving the Ramsey problem
- 4. Implementing optimal policy
- 5. Approximating optimal policy with a Taylor rule

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

Maximize lifetime expected utility

$$\hat{\mathbb{E}}_t \sum_{T=t}^{\infty} \beta^{T-t} \left[U(C_T^i) - \int_0^1 v(h_T^i(j)) dj \right]$$
 (2)

Budget constraint

$$B_t^i \le (1 + i_{t-1})B_{t-1}^i + \int_0^1 w_t(j)h_t^i(j)dj + \Pi_t^i(j)dj - T_t - P_tC_t^i$$
 (3)

▶ Consumption, price level

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

Maximize present value of profits

$$\hat{\mathbb{E}}_{t} \sum_{T=t}^{\infty} \alpha^{T-t} Q_{t,T} \left[\Pi_{t}^{j}(p_{t}(j)) \right]$$
(4)

subject to demand

$$y_t(j) = Y_t \left(\frac{p_t(j)}{P_t}\right)^{-\theta} \tag{5}$$

▶ Profits, stochastic discount factor

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

• Model implies mapping between exogenous states s_t and observables $y_t \equiv (\pi, x, i)'$

$$y_t = gs_t \tag{6}$$

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Under rational expectations (RE), private sector knows model
 → knows (6)

$$\mathbb{E}_t \, y_{t+1} = g \, \mathbb{E} \, s_{t+1} \tag{7}$$

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• $\hat{\mathbb{E}}$: agents do not internalize that identical \to do not know aggregate model \to do not know (6)

• Agents know exogenous evolution of states

$$s_t = h s_{t-1} + \epsilon_t \qquad \quad \epsilon_t \sim \mathcal{N}(\mathbf{0}, \Sigma)$$
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 $a \rightarrow$ concept of long-run expectations in the model

• Estimate *a*, *b* using recursive least squares (RLS) using observed states and knowledge of (8)

Recursive least squares (RLS)

Observables are: $(\pi, x, i)'$

Assumption: learn only intercept of inflation:

$$a_{t-1} = (\bar{\pi}_{t-1}, 0, 0)', \quad b_{t-1} = g h \quad \forall t$$
 (10)

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 $\bar{\pi}_{t-1}$: long-run inflation expectations \rightarrow anchoring

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 $\bar{\pi}_{t-1}$: long-run inflation expectations \rightarrow anchoring

$$\rightarrow$$
 RLS

$$\bar{\pi}_t = \bar{\pi}_{t-1} + k_t \underbrace{\left(\pi_t - (\bar{\pi}_{t-1} + b_1 s_{t-1})\right)}_{\equiv f_{t|t-1}, \text{ forecast error}}$$
(11)

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



1. Decreasing gain:

$$\bar{\pi}_t = \bar{\pi}_{t-1} + \frac{1}{t} f_{t|t-1} \tag{12}$$

2. Constant gain:

$$\bar{\pi}_t = \bar{\pi}_{t-1} + k f_{t|t-1} \tag{13}$$

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Model summary

• New Keynesian core: IS and Phillips curves

$$x_{t} = -\sigma i_{t} + \hat{\mathbb{E}}_{t} \sum_{t=0}^{\infty} \beta^{T-t} ((1-\beta)x_{T+1} - \sigma(\beta i_{T+1} - \pi_{T+1}) + \sigma r_{T}^{n})$$
 (15)

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

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 \rightarrow How should $\{i_t\}$ be set?

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Estimating form of gain function

Calibrate parameters of New Keynesian core to literature

 Estimate flexible form of expectations process via simulated method of moments
 (Duffie & Singleton 1990, Lee & Ingram 1991, Smith 1993)

$$\bar{\pi}_t = \bar{\pi}_{t-1} + \mathbf{g}(f_{t|t-1}) f_{t|t-1}$$
 (18)

• Moments: autocovariances of inflation, output gap, federal funds rate and 1-year ahead SPF inflation expectations at lags $0, \ldots, 4$



Calibration - parameters from the literature

β	0.98	stochastic discount factor
σ	1	intertemporal elasticity of substitution
α	0.5	Calvo probability of not adjusting prices
κ	0.0842	slope of the Phillips curve
ψ_{π}	1.5	coefficient of inflation in Taylor rule
ψ_x	0.3	coefficient of the output gap in Taylor rule
σ_r	0.01	standard deviation, natural rate shock
σ_i	0.01	standard deviation, monetary policy shock
σ_u	0.5	standard deviation, cost-push shock
Ē	0.145	initial value of the gain

Main sources: Chari et al 2000, Woodford 2003, Nakamura & Steinsson 2008, Carvalho et al 2019

Estimated expectations process

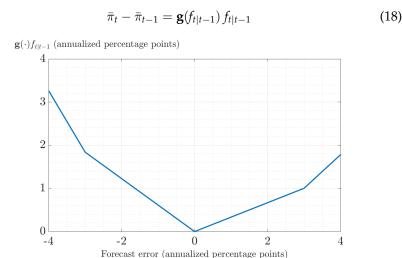


Figure: Changes in long-run inflation expectations as a function of forecast errors

The expectation process over time

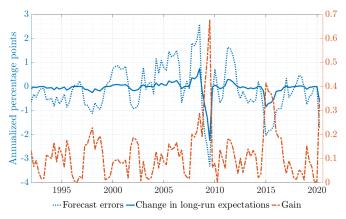


Figure: Time series of forecast errors, changes in long-run expectations and implied gain

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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
- s.t. evolution of expectations

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

Proposition

Let $\mathbf{g}_{z,t} \equiv \frac{\partial \mathbf{g}}{\partial z}$ at t and define b_1 as the first row of b. Then monetary policy optimally brings about the following target relationship between inflation and the output gap

$$\pi_t = -\frac{\lambda_x}{\kappa} x_t$$

RE (discretion): move π_t and x_t to offset cost-push shocks

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$$\pi_t - \Gamma(k) \mathbb{E}_t \sum_{i=1}^{\infty} x_{t+i} = -\frac{\lambda_x}{\kappa} x_t$$

Adaptive learning: can move $\mathbb{E}_t x_{t+i}$ too if k > 0



Proposition

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$$\pi_t - \Omega\left(k_t + f_{t|t-1}\mathbf{g}_{\pi,t}\right) \left(\mathbb{E}_t \sum_{i=1}^{\infty} x_{t+i} \prod_{j=0}^{i-1} (1 - k_{t+1+j} - f_{t+1+j|t+j}\mathbf{g}_{\bar{\pi},t+j})\right) = -\frac{\lambda_x}{\kappa} x_t$$

Endeogenous gain: ability to move $\mathbb{E}_t x_{t+i}$ depends on present and future degree of unanchoring



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Numerical solution procedure

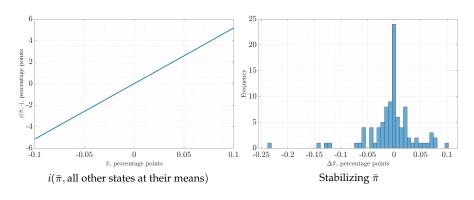
Solve system of model equations + target criterion

For calibrated model with $\lambda_x = 0.05$ (Rotemberg & Woodford 1997),

 \hookrightarrow solve using parameterized expectations algorithm

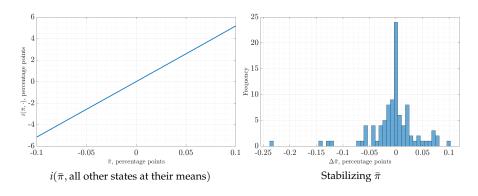
 \hookrightarrow obtain a cubic spline approximation to optimal policy function

Optimal policy - responding to unanchoring



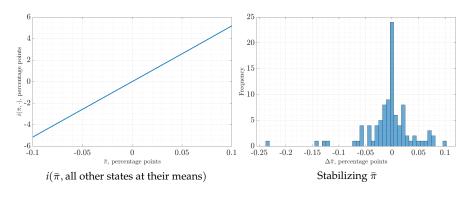
5 bp movement in $\bar{\pi} \rightarrow$ 250 bp movement in *i*

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5 bp movement in $\bar{\pi} \rightarrow$ 250 bp movement in i (1 standard dev. change in $\bar{\pi}$)

Optimal policy - responding to unanchoring



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Mode: 0.3 bp movement in $\bar{\pi} \rightarrow 15$ bp movement in i

Unanchoring causes volatility

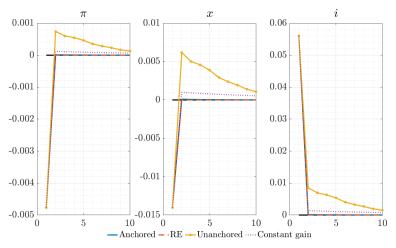


Figure: Impulse responses after a contractionary monetary policy shock when policy follows a Taylor rule

Volatility comes from endogenous gain

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Shocks raise the gain → central bank needs to anchor

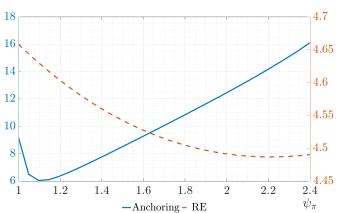
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Optimal Taylor-coefficient on inflation

$$i_t = \psi_\pi \pi_t + \psi_x x_t \tag{19}$$

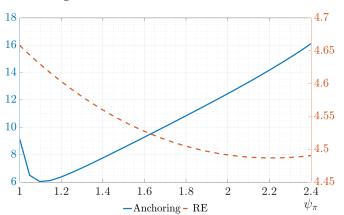
Figure: Central bank loss as a function of ψ_{π}



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



Anchoring-optimal coefficient: $\psi_{\pi}^{A} = 1.1$

RE-optimal coefficient: $\psi_{\pi}^{RE} = 2.2$

Why less aggressive? Future interest rate expectations

IS curve:

$$x_t = -\sigma \mathbf{i}_t + \hat{\mathbb{E}}_t \sum_{T=t}^{\infty} \beta^{T-t} \left((1-\beta)x_{T+1} - \sigma(\beta \mathbf{i}_{T+1} - \pi_{T+1}) + \sigma r_T^n \right)$$

• Current interest rate i_t : one channel of policy



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- Current interest rate *i*_t: one channel of policy
- Taylor rule implies interest rate expectation

$$\hat{\mathbb{E}}_t i_{t+k} = \psi_\pi \hat{\mathbb{E}}_t \pi_{t+k} + \hat{\mathbb{E}}_t \psi_x x_{t+k}$$
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• If private sector understands and believes Taylor rule, expected future interest rates additional channel of policy



Conclusion

First theory of monetary policy for potentially unanchored expectations

Estimation of novel unanchoring channel

• Large and negative surprises unanchor more

Monetary policy

- Expectations unanchoring makes smoothing shocks over time possible
- · Optimal policy aggressive when unanchored, accommodates otherwise
- Taylor rule less aggressive than under rational expectations

Future work

- \hookrightarrow How to anchor at zero-lower bound?
- \hookrightarrow Other applications: currency crises



Breakeven inflation

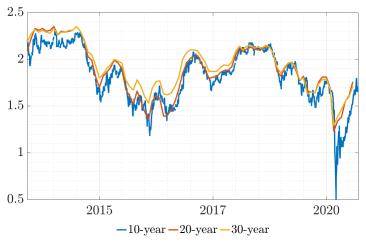


Figure: Market-based inflation expectations, various horizons, %



Correcting the TIPS from liquidity risk

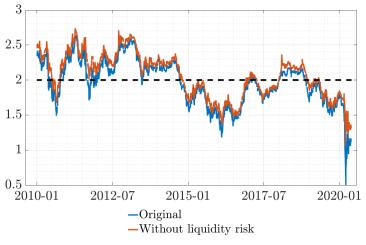


Figure: Market-based inflation expectations, 10 year, %



Further evidence

Figure: Livingston Survey of Firms: Interquartile range of 10-year ahead inflation expectations

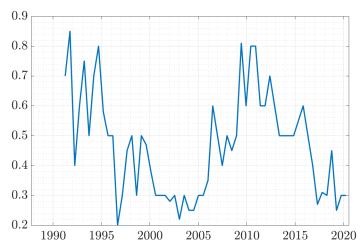
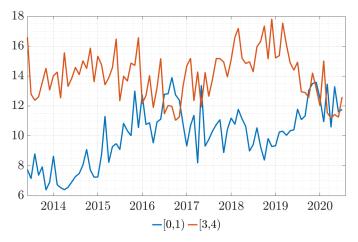




Figure: New York Fed Survey of Consumers: Percent of respondents indicating 3-year ahead inflation will be in a particular range





Oscillatory dynamics in adaptive learning

Consider a stylized adaptive learning model in two equations:

$$\pi_t = \beta f_t + u_t \tag{21}$$

$$f_t = f_{t-1} + k(\pi_t - f_{t-1})$$
 (22)

Solve for the time series of expectations f_t

$$f_t = \underbrace{\frac{1 - k^{-1}}{1 - k^{-1}\beta}}_{\approx 1} f_{t-1} + \frac{k^{-1}}{1 - k^{-1}\beta} u_t \tag{23}$$

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

$$f_t = \underbrace{-\frac{1-\beta}{1-k\beta}}_{\lim_{k\to 1}=-1} f_{t-1} + \frac{1}{1-k\beta} u_t$$
 (24)

Functional forms for g in the literature

• Smooth anchoring function (Gobbi et al, 2019)

$$p = h(y_{t-1}) = A + \frac{BCe^{-Dy_{t-1}}}{(Ce^{-Dy_{t-1}} + 1)^2}$$
 (25)

 $p \equiv Prob(\text{liquidity trap regime})$ y_{t-1} output gap

• Kinked anchoring function (Carvalho et al, 2019)

$$k_t = \begin{cases} \frac{1}{t} & \text{when } \theta_t < \bar{\theta} \\ k & \text{otherwise.} \end{cases}$$
 (26)

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



Choices for criterion θ_t

• Carvalho et al. (2019)'s criterion

$$\theta_t^{CEMP} = \max |\Sigma^{-1}(\phi_{t-1} - T(\phi_{t-1}))|$$
 (27)

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

CUSUM-criterion

$$\omega_t = \omega_{t-1} + \kappa k_{t-1} (f_{t|t-1} f'_{t|t-1} - \omega_{t-1})$$
(28)

$$\theta_t^{CUSUM} = \theta_{t-1} + \kappa k_{t-1} (f'_{t|t-1} \omega_t^{-1} f_{t|t-1} - \theta_{t-1})$$
 (29)

 ω_t estimated forecast-error variance



Recursive least squares algorithm

$$\phi_t = \left(\phi'_{t-1} + k_t R_t^{-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \left(y_t - \phi_{t-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \right)' \right)' \tag{30}$$

$$R_t = R_{t-1} + k_t \begin{pmatrix} 1 \\ s_{t-1} \end{pmatrix} \begin{bmatrix} 1 & s_{t-1} \end{bmatrix} - R_{t-1}$$

$$(31)$$



Actual laws of motion

$$y_{t} = A_{1}f_{a,t} + A_{2}f_{b,t} + A_{3}s_{t}$$

$$s_{t} = hs_{t-1} + \epsilon_{t}$$
(32)

where

$$y_t \equiv \begin{pmatrix} \pi_t \\ x_t \\ i_t \end{pmatrix} \qquad s_t \equiv \begin{pmatrix} r_t^n \\ u_t \end{pmatrix} \tag{34}$$

(32)

and

$$f_{a,t} \equiv \hat{\mathbb{E}}_t \sum_{T-t}^{\infty} (\alpha \beta)^{T-t} y_{T+1} \qquad f_{b,t} \equiv \hat{\mathbb{E}}_t \sum_{T-t}^{\infty} (\beta)^{T-t} y_{T+1}$$
 (35)

Piecewise linear approximation to gain function

$$\mathbf{g}(f_{t|t-1}) = \sum_{i} \gamma_i b_i (f_{t|t-1})$$
(36)

- $b_i(f_{t|t-1})$ = piecewise linear basis
- γ_i = approximating coefficient at node i
- \hookrightarrow Estimate $\hat{\gamma}$ via simulated method of moments



Proposition

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

$$\pi_{t} = -\frac{\lambda_{x}}{\kappa} x_{t}$$

$$+ \frac{\lambda_{x}}{\kappa} \frac{(1 - \alpha)\beta}{1 - \alpha\beta} \left(k_{t} + f_{t|t-1} \mathbf{g}_{\pi,t} \right) \left(\mathbb{E}_{t} \sum_{i=1}^{\infty} x_{t+i} \prod_{j=0}^{i-1} (1 - k_{t+1+j} - f_{t+1+j|t+j} \mathbf{g}_{\bar{\pi},t+j}) \right)$$

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

Lemma

The discretion and commitment solutions of the Ramsey problem coincide.



Corollary

 $Optimal\ policy\ under\ adaptive\ learning\ is\ time-consistent.$

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

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

$$2\pi_t + 2\frac{\lambda}{\kappa}x_t - \varphi_{2,t}(k_t + \mathbf{g}_{\pi}(\pi_t - f_{t-1})) = 0$$
 (37)

$$-2\beta \frac{\lambda}{\kappa} x_t + \varphi_{2,t} - \varphi_{2,t+1} (1 - k_{t+1} - \mathbf{g_f}(\pi_{t+1} - f_t)) = 0$$
 (38)



Target criterion system for anchoring function as changes of the gain

$$\varphi_{6,t} = -cf_{t|t-1}x_{t+1} + \left(1 + \frac{f_{t|t-1}}{f_{t+1|t}}(1 - k_{t+1}) - f_{t|t-1}\mathbf{g}_{\bar{\pi},t}\right)\varphi_{6,t+1} - \frac{f_{t|t-1}}{f_{t+1|t}}(1 - k_{t+1})\varphi_{6,t+2}$$
(39)

$$0 = 2\pi_t + 2\frac{\lambda_x}{\kappa} x_t - \left(\frac{k_t}{f_{t|t-1}} + \mathbf{g}_{\pi,t}\right) \varphi_{6,t} + \frac{k_t}{f_{t|t-1}} \varphi_{6,t+1}$$
 (40)

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

The solution to (40) is given by:

$$\varphi_{6,t} = -2 \, \mathbb{E}_t \sum_{i=0}^{\infty} (\pi_{t+i} + \frac{\lambda_x}{\kappa} x_{t+i}) \prod_{j=0}^{i-1} \frac{\frac{k_{t+j}}{f_{t+j|t+j-1}}}{\frac{k_{t+j}}{f_{t+j|t+j-1}} + \mathbf{g}_{\pi,t+j}}$$
(41)



Respond but not too much

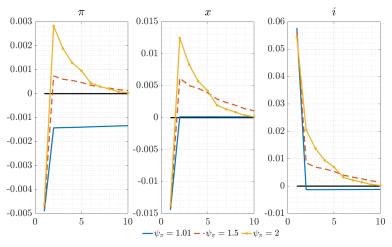


Figure: Impulse responses for unanchored expectations for various values of ψ_π



Details on households and firms

Consumption:

$$C_t^i = \left[\int_0^1 c_t^i(j)^{\frac{\theta - 1}{\theta}} dj \right]^{\frac{\sigma}{\theta - 1}} \tag{42}$$

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

Aggregate price level:

$$P_t = \left[\int_0^1 p_t(j)^{1-\theta} dj \right]^{\frac{1}{\theta-1}}$$
 (43)

Profits:

$$\Pi_t^j = p_t(j)y_t(j) - w_t(j)f^{-1}(y_t(j)/A_t)$$
(44)

Stochastic discount factor

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

$$\tag{45}$$



Derivations

Household FOCs

$$\hat{C}_t^i = \hat{\mathbb{E}}_t^i \hat{C}_{t+1}^i - \sigma(\hat{i}_t - \hat{\mathbb{E}}_t^i \hat{\pi}_{t+1}) \tag{46}$$

$$\hat{\mathbb{E}}_t^i \sum_{s=0}^{\infty} \beta^s \hat{C}_t^i = \omega_t^i + \hat{\mathbb{E}}_t^i \sum_{s=0}^{\infty} \beta^s \hat{Y}_t^i$$
(47)

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

- 1. Solve (46) backward to some date *t*, take expectations at *t*
- 2. Sub in (47)
- 3. Aggregate over households *i*
- \rightarrow Obtain (15)

