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

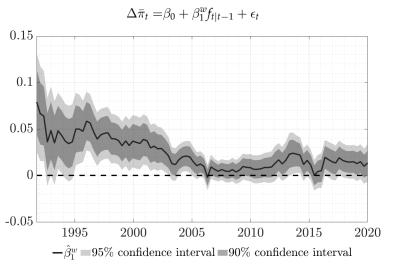
Individual SPF forecasts: for 1991-Q4 onward, estimate rolling regression

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

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

Analyze monetary policy

 → analytically and numerically using novel model disciplined by data

Preview of results

1. Estimation

- · Larger mistakes unanchor more
- Overestimating inflation unanchors more than underestimating it (Hebden et al 2020)

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

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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

$$s_t = hs_{t-1} + \epsilon_t \qquad \quad \epsilon_t \sim \mathcal{N}(\mathbf{0}, \Sigma)$$
 (8)

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$$s_t = h s_{t-1} + \epsilon_t \qquad \epsilon_t \sim \mathcal{N}(\mathbf{0}, \Sigma)$$
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• Postulate linear functional relationship instead of (6):

$$\hat{\mathbb{E}}_t y_{t+1} = a_{t-1} + b_{t-1} s_t \tag{9}$$

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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:

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

$$\pi_t = \kappa x_t + \hat{\mathbb{E}}_t \sum_{T=1}^{\infty} (\alpha \beta)^{T-t} \left(\kappa \alpha \beta x_{T+1} + (1-\alpha) \beta \pi_{T+1} + u_T \right)$$
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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 $0, \dots, 4$



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

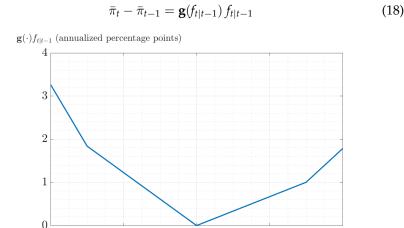


Figure: Changes in long-run inflation expectations as a function of forecast errors

Forecast error (annualized percentage points)

-2

The expectation process over time

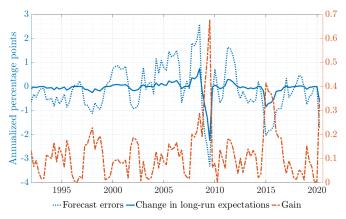


Figure: Time series of forecast errors, changes in long-run expectations and implied gain

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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



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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

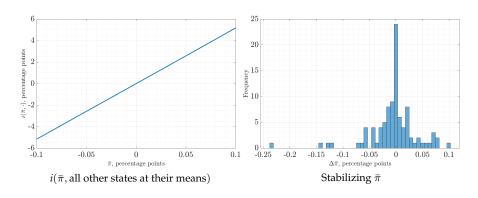
Solve system of model equations + target criterion

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

 \hookrightarrow solve using parameterized expectations algorithm

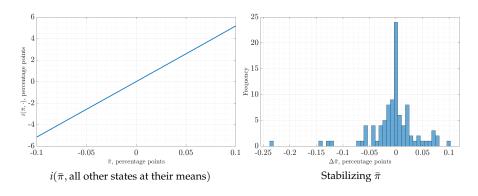
 \hookrightarrow obtain a cubic spline approximation to optimal policy function

Optimal policy - responding to unanchoring



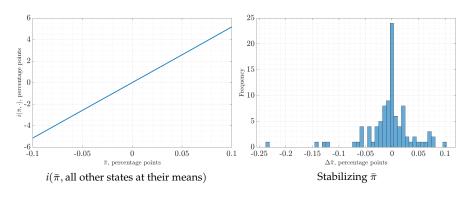
5 bp movement in $\bar{\pi} \rightarrow$ 250 bp movement in i

Optimal policy - responding to unanchoring



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

Optimal policy - responding to unanchoring



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

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

Unanchoring causes volatility

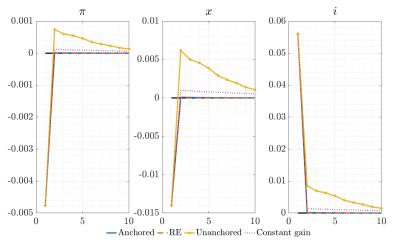


Figure: Impulse responses after a contractionary monetary policy shock when policy follows a Taylor rule

Volatility comes from endogenous gain

• Constant gain:

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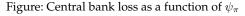
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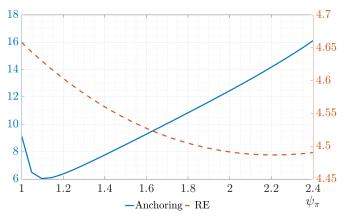
Shocks raise the gain → central bank needs to anchor

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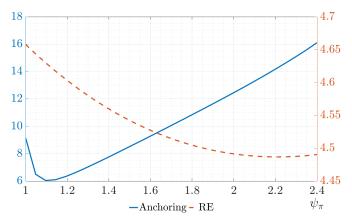
Optimal Taylor-coefficient on inflation





Optimal Taylor-coefficient on inflation

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



Anchoring-optimal coefficient: $\psi_{\pi}^{A} = 1.09$ RE-optimal coefficient: $\psi_{\pi}^{RE} = 2.21$

Respond but not too much

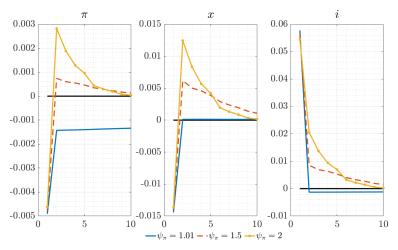


Figure: Impulse responses for unanchored expectations for various values of ψ_π

Why less aggressive? Future interest rate expectations

IS- and Phillips curve:

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

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

• Current interest rate *i*_t: one channel of policy

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- Current interest rate *i*_t: one channel of policy
- Expected future interest rates: additional channel of policy
- Only if policy reaction function internalized

Conclusion

First theory of monetary policy for potentially unanchored expectations

Estimation of novel unanchoring channel

- Large and negative surprises unanchor more
- Estimated gain time series: on average, people only use the last 8 quarters
 of data

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

- → 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

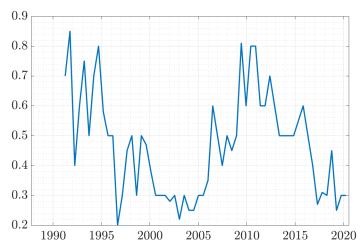
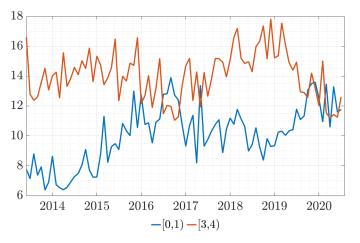




Figure: New York Fed Survey of Consumers: Percent of respondents indicating 3-year ahead inflation will be in a particular range





Oscillatory dynamics in adaptive learning

Consider a stylized adaptive learning model in two equations:

 $\lim_{k\to 1} = -1$

$$\pi_t = \beta f_t + u_t \tag{19}$$

$$f_t = f_{t-1} + k(\pi_t - f_{t-1})$$
 (20)

Solve for the time series of expectations f_t

$$f_{t} = \underbrace{\frac{1 - k^{-1}}{1 - k^{-1}\beta}}_{\approx 1} f_{t-1} + \frac{k^{-1}}{1 - k^{-1}\beta} u_{t}$$
 (21)

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

$$f_t = \underbrace{-\frac{1-\beta}{1-k\beta}}_{t-1} f_{t-1} + \frac{1}{1-k\beta} u_t$$
 (22)

Functional forms for g in the literature

• Smooth anchoring function (Gobbi et al, 2019)

$$p = h(y_{t-1}) = A + \frac{BCe^{-Dy_{t-1}}}{(Ce^{-Dy_{t-1}} + 1)^2}$$
 (23)

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

• Kinked anchoring function (Carvalho et al, 2019)

$$k_t = \begin{cases} \frac{1}{t} & \text{when } \theta_t < \bar{\theta} \\ k & \text{otherwise.} \end{cases}$$
 (24)

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



Choices for criterion θ_t

• Carvalho et al. (2019)'s criterion

$$\theta_t^{CEMP} = \max |\Sigma^{-1}(\phi_{t-1} - T(\phi_{t-1}))|$$
 (25)

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

CUSUM-criterion

$$\omega_t = \omega_{t-1} + \kappa k_{t-1} (f_{t|t-1} f'_{t|t-1} - \omega_{t-1})$$
(26)

$$\theta_t^{CUSUM} = \theta_{t-1} + \kappa k_{t-1} (f'_{t|t-1} \omega_t^{-1} f_{t|t-1} - \theta_{t-1})$$
 (27)

 ω_t estimated forecast-error variance



Recursive least squares algorithm

$$\phi_t = \left(\phi'_{t-1} + k_t R_t^{-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \left(y_t - \phi_{t-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \right)' \right)' \tag{28}$$

$$R_{t} = R_{t-1} + k_{t} \begin{pmatrix} 1 \\ s_{t-1} \end{pmatrix} \begin{bmatrix} 1 & s_{t-1} \end{bmatrix} - R_{t-1}$$
 (29)



Actual laws of motion

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

$$s_{t} = hs_{t-1} + \epsilon_{t}$$
(30)

where

$$y_t \equiv \begin{pmatrix} \pi_t \\ x_t \\ i_t \end{pmatrix} \qquad s_t \equiv \begin{pmatrix} r_t^n \\ u_t \end{pmatrix} \tag{32}$$

(30)

and

$$f_{a,t} \equiv \hat{\mathbb{E}}_t \sum_{T-t}^{\infty} (\alpha \beta)^{T-t} y_{T+1} \qquad f_{b,t} \equiv \hat{\mathbb{E}}_t \sum_{T-t}^{\infty} (\beta)^{T-t} y_{T+1}$$
 (33)

Piecewise linear approximation to gain function

$$\mathbf{g}(f_{t|t-1}) = \sum_{i} \gamma_i b_i (f_{t|t-1})$$
(34)

- $b_i(f_{t|t-1})$ = 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

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

$$2\pi_t + 2\frac{\lambda}{\kappa}x_t - \varphi_{2,t}(k_t + \mathbf{g}_{\pi}(\pi_t - f_{t-1})) = 0$$
 (35)

$$-2\beta \frac{\lambda}{\kappa} x_t + \varphi_{2,t} - \varphi_{2,t+1} (1 - k_{t+1} - \mathbf{g_f}(\pi_{t+1} - f_t)) = 0$$
 (36)



Target criterion system for anchoring function as changes of the gain

$$\varphi_{6,t} = -cf_{t|t-1}x_{t+1} + \left(1 + \frac{f_{t|t-1}}{f_{t+1|t}}(1 - k_{t+1}) - f_{t|t-1}\mathbf{g}_{\bar{\pi},t}\right)\varphi_{6,t+1} - \frac{f_{t|t-1}}{f_{t+1|t}}(1 - k_{t+1})\varphi_{6,t+2}$$
(37)

$$0 = 2\pi_t + 2\frac{\lambda_x}{\kappa} x_t - \left(\frac{k_t}{f_{t|t-1}} + \mathbf{g}_{\pi,t}\right) \varphi_{6,t} + \frac{k_t}{f_{t|t-1}} \varphi_{6,t+1}$$
(38)

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

The solution to (38) is given by:

$$\varphi_{6,t} = -2 \, \mathbb{E}_t \sum_{i=0}^{\infty} (\pi_{t+i} + \frac{\lambda_x}{\kappa} x_{t+i}) \prod_{j=0}^{i-1} \frac{\frac{k_{t+j}}{f_{t+j|t+j-1}}}{\frac{k_{t+j}}{f_{t+j|t+j-1}} + \mathbf{g}_{\pi,t+j}}$$
(39)



Details on households and firms

Consumption:

$$C_t^i = \left[\int_0^1 c_t^i(j)^{\frac{\theta - 1}{\theta}} dj \right]^{\frac{\sigma}{\theta - 1}} \tag{40}$$

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

Aggregate price level:

$$P_t = \left[\int_0^1 p_t(j)^{1-\theta} dj \right]^{\frac{1}{\theta-1}} \tag{41}$$

Profits:

$$\Pi_t^j = p_t(j)y_t(j) - w_t(j)f^{-1}(y_t(j)/A_t)$$
(42)

Stochastic discount factor

$$Q_{t,T} = \beta^{T-t} \frac{P_t U_c(C_T)}{P_T U_c(C_t)}$$
(43)



Derivations

Household FOCs

$$\hat{C}_t^i = \hat{\mathbb{E}}_t^i \hat{C}_{t+1}^i - \sigma(\hat{i}_t - \hat{\mathbb{E}}_t^i \hat{\pi}_{t+1}) \tag{44}$$

$$\hat{\mathbb{E}}_t^i \sum_{s=0}^{\infty} \beta^s \hat{C}_t^i = \omega_t^i + \hat{\mathbb{E}}_t^i \sum_{s=0}^{\infty} \beta^s \hat{Y}_t^i$$
(45)

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

- 1. Solve (44) backward to some date *t*, take expectations at *t*
- 2. Sub in (45)
- 3. Aggregate over households *i*
- \rightarrow Obtain (15)

