

# Monetary Policy & Anchored Expectations

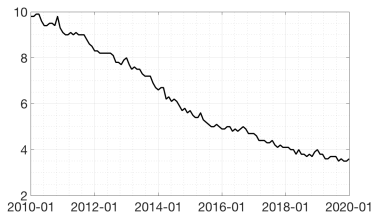
## An Endogenous Gain Learning Model

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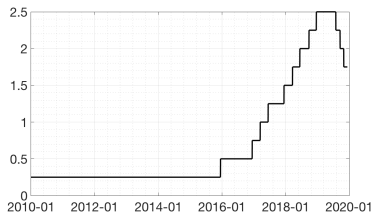
Boston College

April 15, 2020

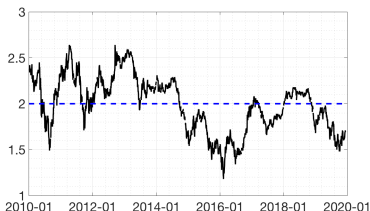
# Puzzling Fed behavior fall 2019



(a) Unemployment rate, %



(b) Fed funds rate target, upper limit, %



(c) Market-based inflation expectations, 10 year, % average

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# Structure of talk

## 1. Model

# Preview of results

1. Two layers of new intertemporal tradeoffs
2. Optimal monetary policy time-inconsistent  
→ Illustrate analytically in special case: target criterion
3. Not today: short-run costs vs. long-run benefits of anchoring expectations

## Related literature

- **Optimal monetary policy in New Keynesian models**

Clarida, Gali & Gertler (1999), Woodford (2003)

- **Econometric learning**

Evans & Honkapohja (2001), Preston (2005), Molnár & Santoro (2014)

- **Anchoring / endogenous gain**

Carvalho et al (2019), Svensson (2015), Hooper et al (2019), Milani (2014)

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

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

Model solution

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

$$y_t = g s_t \quad (2)$$

$$s_t \equiv (r_t^n, u_t)' \quad (\text{states})$$

$$y_t \equiv (\pi_t, x_t, i_t)' \quad (\text{jumps})$$

- If use  $\hat{\mathbb{E}} \rightarrow$  private sector does not know  $g$   
 $\rightarrow$  estimate using observed states & knowledge of (1)
- Households and firms don't know they are identical

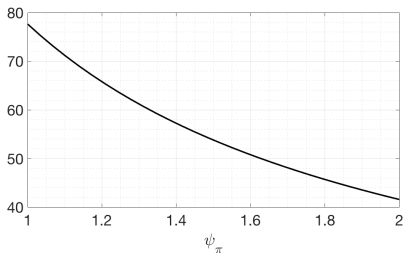
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# Conclusion

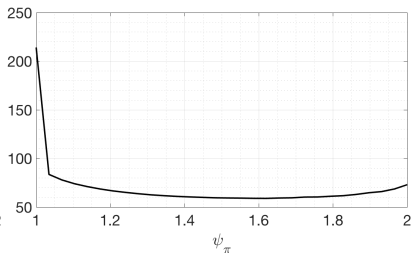
# Short-run costs, long-run benefits

Assume Taylor rule and no concern for output gap stabilization

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



(d) RE



(e) Anchoring

Figure: Central bank loss as a function of  $\psi_\pi$



# Functional forms for $\mathbf{g}$

- Smooth anchoring function

$$k_t = k_{t-1} - c + dfe_{t|t-1}^2 \quad (3)$$

$$c, d > 0$$

- Kinked anchoring function

$$k_t = \begin{cases} \frac{1}{t} & \text{when } \theta_t < \bar{\theta} \\ k & \text{otherwise.} \end{cases} \quad (4)$$

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

# Choices for criterion $\theta_t$

- Carvalho et al. (2019)'s criterion

$$\theta_t^{CEMP} = \max |\Sigma^{-1}(\phi_{t-1} - T(\phi_{t-1}))| \quad (5)$$

$\Sigma$  variance-covariance matrix of shocks

$T(\phi)$  mapping from PLM to ALM

- CUSUM-criterion

$$\omega_t = \omega_{t-1} + \kappa k_{t-1} (fe_{t|t-1} fe'_{t|t-1} - \omega_{t-1}) \quad (6)$$

$$\theta_t^{CUSUM} = \theta_{t-1} + \kappa k_{t-1} (fe'_{t|t-1} \omega_t^{-1} fe_{t|t-1} - \theta_{t-1}) \quad (7)$$

$\omega_t$  estimated forecast-error variance