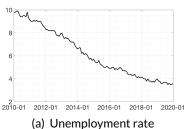
## Monetary Policy & Anchored Expectations An Endogenous Gain Learning Model

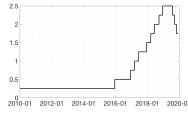
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#### Puzzling US business cycle fall 2019





(b) Fed funds rate target, upper limit



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

#### This project

Model anchored expectations as an endogenous gain learning scheme

 $\rightarrow\,$  How to conduct optimal monetary policy in interaction with the anchoring expectation formation?

#### Preview of results

 intertemporal tradeoff: short-run costs vs. long-run benefits of anchoring expectations

optimal monetary policy time-inconsistent

→ illustrate 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), Preston (2005), Molnár & Santoro (2014)

Anchoring / endogenous gain

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

#### STRUCTURE OF TALK

1 Model

2 Ramsey problem for special case

## 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** 

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

► Consumption, price level

## Firms - Standard up to $\hat{\mathbb{E}}$

Maximize present value of profits

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

$$s_t = hs_{t-1} + \epsilon_t \tag{5}$$

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

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

• If use  $\hat{\mathbb{E}} \to \operatorname{don't} \operatorname{know} g$  $\to \operatorname{estimate} \operatorname{using} \operatorname{observed} \operatorname{states} \& \operatorname{knowledge} \operatorname{of} (5)$ 

#### Adaptive Learning

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

$$\hat{\mathbb{E}}_{t} \mathbf{y}_{t+1} = \phi_{t-1} \begin{bmatrix} \mathbf{1} \\ \mathbf{s}_{t} \end{bmatrix} \tag{7}$$

Note: misspecified

Can write:

$$\hat{\mathbb{E}}_{t} y_{t+1} = a_{t-1} + b_{t-1} s_{t}$$
 (8)

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

#### Anchoring - endogenous gain

Special case: learn only intercept of inflation:

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

 $\rightarrow \mathsf{RLS}$ 

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

➤ General RLS algorithm

Gain in literature usually exogenous:

$$k_t = egin{cases} rac{1}{t} & ext{decreasing} \\ k & ext{constant} \end{cases}$$

Here instead

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

#### ANCHORING FUNCTION - INTERPRETATION

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

FIGURE: U Michigan inflation expectations (%)



- If gain nondecreasing,  $\bar{\pi}$  changes  $\rightarrow$  unanchored expectations
- If gain decreasing,  $\bar{\pi}$  stable o anchored expectations

#### Model Summary

• IS- and Phillips curve:

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

$$(12)$$

$$\pi_t = \kappa \mathbf{x}_t + \hat{\mathbb{E}}_t \sum_{T-t}^{\infty} (\alpha \beta)^{T-t} \left( \kappa \alpha \beta \mathbf{x}_{T+1} + (1-\alpha) \beta \pi_{T+1} + \mathbf{u}_T \right)$$
 (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 Ramsey problem for special case

#### Ramsey Problem

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

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

$$\pi_t = -\frac{\lambda_x}{\kappa} \bigg\{ x_t - \frac{(1-\alpha)\beta}{1-\alpha\beta} \bigg( k_t + ((\pi_t - \bar{\pi}_{t-1} - b_1 s_{t-1})) \mathbf{g}_{\pi,t} \bigg)$$

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

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.

#### Interpretation - Intertemporal Tradeoffs

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

#### 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

#### LEMMA

The commitment solution of the Ramsey problem does not exist.

Let 
$$k_t \to 0$$
,  $g_{z,t} \to 0$ .

Target criterion becomes

$$\pi_t = -\frac{\lambda_x}{\kappa} \mathbf{x}_t \tag{15}$$

= target criterion under RE discretion

#### COROLLARY

#### Optimal policy is time-inconsistent.

#### Already true for exogenous gain learning!

#### Constant gain specification:

- $\bullet$   $k_t = k$
- $g_{z,t} = 0$  (still)

$$\pi_t = -\frac{\lambda_x}{\kappa} x_t + \frac{\lambda_x}{\kappa} \frac{(1-\alpha)\beta}{1-\alpha\beta} k \left( \sum_{i=1}^{\infty} x_{t+i} (1-k)^i \right)$$
 (16)

 $\rightarrow$  A first intertemporal tradeoff

# Anchoring as a Second Intertemporal Tradeoff

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

- + first intertemporal tradeoff from stance of learning
- + second intertemporal tradeoff from stance of anchoring

#### CONCLUSION

- Interaction between monetary policy and anchoring
- Optimal policy conditions on
  - stance of expectations
  - stance of anchoring and expected future anchoring
- Optimal policy trades off short-run costs with future benefits of anchoring expectations
- Can explain departures from the Taylor rule like US, fall 2019

Thank you!

## DETAILS ON HOUSEHOLDS AND FIRMS

Consumption:

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

 $\theta > 1$ : 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:

$$\Pi_t^j = p$$

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

(18)

(17)

$$Q_{t,T} = eta^{\mathsf{T}-t} rac{P_t U_c(C_T)}{P_T U_c(C_t)}$$

### RECURSIVE LEAST SQUARES ALGORITHM

$$\phi_{t} = \left(\phi'_{t-1} + k_{t}^{-1} R_{t}^{-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \left(y_{t} - \phi_{t-1} \begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix}\right)'\right)'$$

$$R_{t} = R_{t-1} + k_{t}^{-1} \left(\begin{bmatrix} 1 \\ s_{t-1} \end{bmatrix} \begin{bmatrix} 1 & s_{t-1} \end{bmatrix} - R_{t-1} \right)$$

$$(21)$$

#### DERIVATIONS

#### Household FOCs

$$\hat{C}_t^i = \hat{\mathbb{E}}_t^i \hat{C}_{t+1}^i - \sigma(\hat{\mathbf{i}}_t - \hat{\mathbb{E}}_t^i \hat{\pi}_{t+1})$$
 (23)

$$\hat{\mathbb{E}}_{t}^{i} \sum_{s=0}^{\infty} \beta^{s} \hat{C}_{t}^{i} = \omega_{t}^{i} + \hat{\mathbb{E}}_{t}^{i} \sum_{s=0}^{\infty} \beta^{s} \hat{Y}_{t}^{i}$$
(24)

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

- $lue{}$  Solve (23) backward to some date t, take expectations at t
- Sub in (24)
- Aggregate over households i
- $\rightarrow$  Obtain (12)



Compact notation

$$z_t = A_1 f_{a,t} + A_2 f_{b,t} + A_3 s_t$$
  $s_t = h s_{t-1} + \epsilon_t$ 

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

and

$$z_t \equiv \begin{pmatrix} \pi_t \\ x_t \\ i_t \end{pmatrix} \qquad s_t \equiv \begin{pmatrix} r_t^n \\ \overline{i}_t \\ i_t \end{pmatrix}$$

 $\equiv \begin{pmatrix} \frac{r_t^n}{i_t} \\ u_t \end{pmatrix}$ 



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(25)

(26)