Dynamic Inattention, the Phillips Curve and Forward Guidance

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Monetary policy presumably plays a key role in shaping [inflation] expectations ... by providing guidance about the FOMC's objectives for inflation in the future. Even so, economists' understanding of exactly how and why inflation expectations change over time is limited.

- Objective of forward guidance:
 - affect the economy today through news about future policy.
- Two natural questions:
 - ▶ Do price setters pay attention to the news about future policy?
 - * Yes
 - If so, do their prices respond to such news?
 - * Yes
- More generally, how are price setters' expectations formed and how do they affect inflation dynamics?

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Two types of Phillips curves for inflation dynamics:

Sticky/Noisy information Phillips curves:

$$\pi_{t} = \tilde{\mathbb{E}}_{t-1}[\pi_{t} + \alpha \Delta y_{t}] + \alpha \frac{\lambda}{1-\lambda} y_{t}$$

- ★ criticized for not being forward looking.
- Sticky price models:

$$\pi_{\mathsf{t}} = \beta \mathbb{E}_{\mathsf{t}}[\pi_{\mathsf{t}+1}] + \gamma \mathsf{y}_{\mathsf{t}},$$

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- Firms' flow nominal profit depends on their own price, aggregate price and output:

$$\begin{split} \Pi_{i,t} &= \Pi(P_{i,t}, P_t, Y_t) \\ &\approx - (p_{i,t} - mc_t)^2 + terms \text{ independent of } p_{i,t} \end{split}$$

$$mc_t = p_t + \alpha y_t$$
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- Here:
 - ▶ y_t is output gap.
 - p_t is the aggregate price: $p_t = \int_0^1 p_{i,t} di$.
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- For any t, let S_i^t be i's information set at time t.
- i's pricing problem is

$$\mathbf{L}_{0}^{\mathbf{i}} \equiv \min_{\mathbf{p}_{\mathbf{i},\mathbf{t}}:\mathbf{S}_{\mathbf{i}}^{\mathbf{t}} \rightarrow \mathbb{R}} \mathbb{E} \left[\sum_{\mathbf{t}=0}^{\infty} \beta^{\mathbf{t}} (\mathbf{p}_{\mathbf{i},\mathbf{t}} - \mathbf{mc}_{\mathbf{t}})^{2} |\mathbf{S}_{\mathbf{i}}^{0} \right]$$

Solution

$$p_{i,t}(S_i^t) = \mathbb{E}[mc_t|S_i^t]$$

and

$$L_0^i = \sum_{t=0}^{\infty} \beta^t var(mc_t|S_i^t).$$

Kalman filtering:

$$\Delta \mathsf{p}_\mathsf{i,t} = \mathbb{E}[\Delta \mathsf{mc_t} | \mathsf{S}_\mathsf{i}^\mathsf{t-1}] + \mathsf{k}_\mathsf{t}^\mathsf{i}(\mathsf{s}_\mathsf{i,t} - \mathbb{E}[\mathsf{s}_\mathsf{i,t} | \mathsf{S}_\mathsf{i}^\mathsf{t-1}])$$

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- Firm i wakes up at time t with S_i^{t-1} .
- Chooses $S_{i,t} \subset \mathcal{S}_t$ subject to cost $C(S_{i,t}|S_i^{t-1})$ and forms $S_i^t = S_i^{t-1} \cup S_{i,t}$
- $\bullet \ \, \text{Chooses} \, \, p_{i,t}: S_i^t \to \mathbb{R}. \\$

$$\begin{split} L_t(S_i^{t-1}) &= \underset{S_{i,t} \subset \mathcal{S}_t}{\text{min}} \{ \underbrace{var(mc_t|S_i^t)}_{\text{gain from information}} + \underbrace{C(S_{i,t}|S_i^{t-1})}_{\text{cost of information}} + \beta L_{t+1}(S_i^t) \} \\ s.t. \ S_i^t &= S_i^{t-1} \cup S_{i,t} \end{split}$$

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- A forward looking firm cares about mc_t , mc_{t+1} , mc_{t+2} , ...
- These are subject to shocks that might not have been realized at time t.
- \bullet So if $\mathbb{E}_t^f[.]$ captures availability of infromation at t, firms can learn about

$$\mathsf{mc_t}, \; \mathbb{E}_\mathsf{t}^\mathsf{f}[\mathsf{mc_{t+1}}], \; \mathbb{E}_\mathsf{t}^\mathsf{f}[\mathsf{mc_{t+2}}], \; ...$$

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- Data Processing Inequality (DPI) in information theory:
 - ▶ for $\{s_1, s_2\} \subset \mathcal{S}_t$, seeing a combination of them is less costly than seeing both

$$\mathsf{C}(\mathsf{a}\mathsf{s}_1 + \mathsf{b}\mathsf{s}_2|\mathsf{S}^{\mathsf{t}-1}) \leq \mathsf{C}(\mathsf{s}_1,\mathsf{s}_2|\mathsf{S}^{\mathsf{t}-1})$$

Proposition

Every firm observes only one signal at any time.

- Intuition:
 - ▶ Price is a linear combination of signals.
 - So instead of seeing signals separately and paying a high cost, the firm would like to see the combination.

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 \bullet The marginal cost of learning more about any $\mathbb{E}_t^f[\mathsf{mc}_{t+\tau}]$ is increasing.

Proposition

Optimal signals are forward looking $(b_{\tau>0} \neq 0)$

$$\mathbf{s}_{\mathsf{i},\mathsf{t}} = \sum_{\mathsf{i}=0}^{\infty} \beta^{\mathsf{j}} \mathbf{b}_{\mathsf{j}} \mathbb{E}_{\mathsf{t}}^{\mathsf{f}} [\mathsf{mc}_{\mathsf{t}+\mathsf{j}}] + \sigma_{\mathsf{s}}^{\mathsf{i}} \mathbf{e}_{\mathsf{i},\mathsf{t}}$$

• The agent is forward looking and wants to know about

$$\mathsf{mc}_\mathsf{t}, \mathbb{E}_\mathsf{r}^\mathsf{f}[\mathsf{mc}_\mathsf{t+1}], \mathbb{E}_\mathsf{r}^\mathsf{f}[\mathsf{mc}_\mathsf{t+2}], \dots$$

Marginal benefit is decreasing with horizon while marginal cost is increasing with precision \Rightarrow Information smoothing.

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 \bullet The marginal cost of learning more about any $\mathbb{E}_t^f[\mathsf{mc}_{t+\tau}]$ is increasing.

Proposition

Optimal signals are forward looking $(b_{\tau>0} \neq 0)$

$$\mathbf{s}_{\mathsf{i},\mathsf{t}} = \sum_{\mathsf{i}=0}^{\infty} \beta^{\mathsf{j}} \mathbf{b}_{\mathsf{j}} \mathbb{E}_{\mathsf{t}}^{\mathsf{f}} [\mathsf{mc}_{\mathsf{t}+\mathsf{j}}] + \sigma_{\mathsf{s}}^{\mathsf{i}} \mathbf{e}_{\mathsf{i},\mathsf{t}}$$

• The agent is forward looking and wants to know about

$$\mathsf{mc}_\mathsf{t}, \mathbb{E}_\mathsf{t}^\mathsf{f}[\mathsf{mc}_{\mathsf{t}+1}], \mathbb{E}_\mathsf{t}^\mathsf{f}[\mathsf{mc}_{\mathsf{t}+2}], \dots$$

Marginal benefit is decreasing with horizon while marginal cost is increasing with precision \Rightarrow Information smoothing.

- Recall, $mc_t = p_t + \alpha y_t$.
- In sticky/noisy information models:

$$\pi_{\mathsf{t}} = \tilde{\mathbb{E}}_{\mathsf{t}-1}[\pi_{\mathsf{t}} + \alpha \Delta \mathsf{y}_{\mathsf{t}}] + \alpha \frac{\lambda}{1-\lambda} \mathsf{y}_{\mathsf{t}}$$

- sticky information: λ is the fraction that update their information.
- ightharpoonup noisy information: λ is the Kalman gain.
- Under dynamic inattention:

$$\begin{split} \pi_{t} &= \tilde{\mathbb{E}}_{t-1}[\pi_{t} + \alpha \Delta \mathbf{y}_{t}] + \alpha \delta_{0} \mathbf{y}_{t} \\ &- \sum_{\tau=1}^{\infty} \beta^{\tau} \delta_{\tau} \tilde{\mathbb{FE}}_{t}[\pi_{t+\tau} + \alpha \Delta \mathbf{y}_{t+\tau}] \end{split}$$

where $\tilde{\mathbb{FE}}[x] \equiv \tilde{\mathbb{E}}_t[x] - \mathbb{E}_t^f[x]$ is the forecast error of firms relative to a fully informed agent.

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Remarks:

- lacktriangle This imbeds the noisy information Phillips curve when eta=0.
- Inflation is affected by expectations about future, but in a different way than sticky price models:
 - $\pi_t = \beta \mathbb{E}_t[\pi_{t+1}] + \kappa y_t$: inflation is increasing in expected inflation.
 - 2 dynamic inattention: inflation is decreasing in forecast errors.

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• Estimate the Phillips curve using GMM

$$\begin{split} \pi_{\mathbf{t}} &= \tilde{\mathbb{E}}_{\mathbf{t}-1}[\pi_{\mathbf{t}} + \alpha \Delta \mathbf{y}_{\mathbf{t}}] + \alpha \delta_{0} \mathbf{y}_{\mathbf{t}} \\ &- \sum_{\tau=1}^{\mathsf{T}} \beta^{\tau} \delta_{\tau} \tilde{\mathbb{F}} \tilde{\mathbb{E}}_{\mathbf{t}}[\pi_{\mathbf{t}+\tau} + \alpha \Delta \mathbf{y}_{\mathbf{t}+\tau}] + \xi_{\mathbf{t}} \end{split}$$

- Use Survey of Professional Forecasters as proxy for firms' forecasts.
- Instrument forecast revisions for forecast errors. (Coibion Gorodnichenko (2015))
- ① Null hypothesis: $\delta_{\tau} \neq 0$ for $\tau > 0$.

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| | π_{t} | |
|--|--------------------------|-----------------|
| | (1) | (2) |
| | GDP Deflator (72Q1-16Q4) | CPI (81Q3-16Q4) |
| $\tilde{\mathbb{E}}_{t-1}[\pi_{t} + \alpha \Delta y_{t}]$ | 1.00 *** | 1.01 *** |
| | (0.01) | (0.14) |
| $lpha \mathbf{y_t}$ | I.28 ** | 0.67 *** |
| | (0.50) | (0.10) |
| $\tilde{\mathbb{FE}}_{t}[\pi_{t+1} + \alpha \Delta y_{t+1}]$ | 0.42 *** | 0.16 *** |
| | (0.05) | (0.03) |
| $\tilde{\mathbb{FE}}_{t}[\pi_{t+2} + \alpha \Delta y_{t+2}]$ | 0.21 *** | -0.31 *** |
| | (0.05) | (0.03) |
| $\tilde{\mathbb{FE}}_{\mathbf{t}}[\pi_{\mathbf{t}+3} + \alpha \Delta \mathbf{y}_{\mathbf{t}+3}]$ | 0.11 *** | -0.17 *** |
| | (0.02) | (0.04) |

Newey-West robust standard errors in parentheses **** p<0.01, *** p<0.05, * p<0.1

Example: One Period Ahead News

- Suppose $mc_t = mc_{t-1} + u_{t-1}$
- Shocks are announced one period ahead.
- How much do agents pay attention to this news and react?
- Under myopic inattention($\beta = 0$):

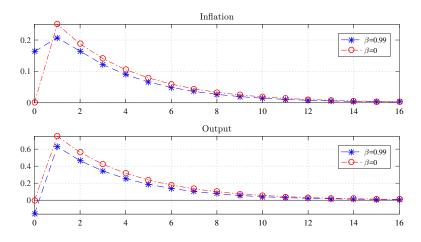
$$s_{i,t} = mc_t + e_{i,t} \\$$

- Notice that in this case $u_t \perp S_i^t$: myopic firms completely ignore news about future.
- Under dynamic inattention($\beta > 0$):

$$\mathbf{s}_{\mathrm{i,t}} = \mathbf{m} \mathbf{c}_{\mathrm{t}} + \gamma \mathbf{m} \mathbf{c}_{\mathrm{t+1}} + \mathbf{e}_{\mathrm{i,t}}$$

Example: One Period Ahead News

- Under dynamic inattention inflation responds to the news shock.
- Output falls on impact because marginal cost is fixed by assumption, which is relaxed in GE.



A Three Equation Model

Dynamic Phillips curve:

$$\begin{split} \pi_{\mathbf{t}} &= \tilde{\mathbb{E}}_{\mathbf{t}-1}[\pi_{\mathbf{t}} + \alpha \Delta \mathbf{y}_{\mathbf{t}}] + \alpha \delta_{0} \mathbf{y}_{\mathbf{t}} \\ &- \sum_{\tau=1}^{\infty} \beta^{\tau} \delta_{\tau} \tilde{\mathbb{FE}}_{\mathbf{t}}[\pi_{\mathbf{t}+\tau} + \alpha \Delta \mathbf{y}_{\mathbf{t}+\tau}] \end{split}$$

Dynamic IS curve:

$$\mathbf{y_t} = \mathbb{E}_{\mathbf{t}}^{\mathbf{f}}[\mathbf{y_{t+1}}] - \sigma^{-1}(\mathbf{i_t} - \mathbb{E}_{\mathbf{t}}^{\mathbf{f}}[\pi_{\mathbf{t+1}}])$$

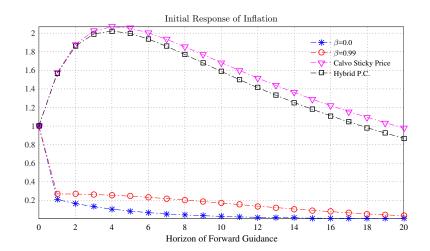
Taylor rule:

$$\mathbf{i_t} = \rho \mathbf{i_{t-1}} + (1 - \rho) \left(\phi_{\pi} \pi_{\mathsf{t}} + \phi_{\mathsf{y}} \mathbf{y_t} \right) + \mathbf{u_{t-k}}$$

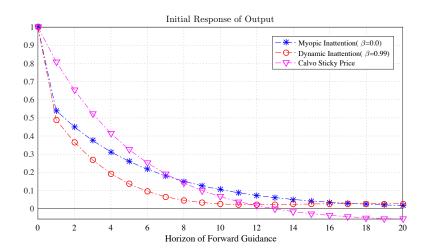
where k is the horizon of forward guidance.

Forward Guidance Puzzle

• Impact response of inflation is decreasing in horizon of forward guidance



4-period ahead Forward Guidance Shock



Conclusion

- Showed that firms have information smoothing incentives:
 - they pay attention to news about future,
 - and incorporate such news in their current prices.
- Derived and estimated a new micro founded Phillips curve:
 - inflation is forward looking in contrast to other models of information rigidity.
 - no forward guidance puzzle despite inflation being forward looking.

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- Showed that firms have information smoothing incentives:
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4-period ahead Forward Guidance Shock

