

Optimal Monetary Policy with Uncertain Private Sector Foresight

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Motivation

- ▶ **Backdrop:** Persistent post-pandemic inflation revived fears of unanchored long-run expectations (“inflation scares”).
- ▶ **Key question:** How should a central bank set policy when it is **uncertain** about expectation formation and faces the risk of inflation scares?
- ▶ **Punchline:** Optimal policy augments Clarida–Galí–Gertler (1999) *lean-against-the-wind* logic.
 - ▶ adds a **forward-looking, pre-emptive** term to guard against scares.
 - ▶ uncertainty about expectations makes policy more aggressive on average.

Methodology

- ▶ We study optimal discretionary policy in a finite-horizon planning (FHP) model (Woodford, 2018).
- ▶ **Planning:** Households and firms plan k periods ahead; beyond k , beliefs about continuation values are averages of past outcomes \Rightarrow vulnerability to inflation scares.
- ▶ **Uncertainty:** The population share of short vs. long-horizon agents is time-varying; greater uncertainty **worsens** the inflation–output tradeoff frontier.
- ▶ **Policy problem:** Consider optimal policy for dual-mandate CB; has a risk-management motive to pre-empt drift in LR inflation expectations.

Why We Use the NK-FHP Model?

- ▶ **Empirical fit.** Estimated NK-FHP fits macro data *and* survey data on expectations well.
 - ▶ Better fit than hybrid NK-RE and other behavioral models.
 - ▶ Replicates long-run inflation-expectations dynamics in the 1970s–early 1980s.
 - ▶ Explains “underreaction then overreaction” in survey inflation forecasts (Angeletos et al., 2020).
- ▶ **Mechanism.** Finite planning horizons with longer-run learning \Rightarrow disciplined approach for modeling drift in longer-run expectations.
- ▶ **Uncertainty in expectations formation.** Time-varying mix of short- vs. long-horizon agents captures shifts in beliefs and the risk of inflation scares.
- ▶ **Policy relevance.** Uncertainty in planning horizons requires an aggressive policy response to preempt drift in LR beliefs.

FHP and Bounded Rationality

- ▶ **Within horizon** ($t \rightarrow t+k$). Households/firms use all structural relationships to evaluate contingencies over next k periods. *As-if RE* conditional on finite horizon.
- ▶ **Beyond horizon (terminal beliefs)**. Continuation values ($v_{p,t}, v_{h,t}$) are *learned from past data*, not pinned by the full infinite-horizon model \Rightarrow departure from RE.
- ▶ **Beliefs about others**. Agents assume others share their own horizon k (a *common-horizon* belief) \Rightarrow second departure from RE.
- ▶ **Policy perception gap**. Agents *perceive* the CB as having k -period horizon; the CB has infinite horizon-RE \Rightarrow wedge between *perceived* and *implemented* policy.

The NK-FHP Model

Agents and planning

Two planner types $k \in \{k_0, k_1\}$ (short/long). At date t , a type- k agent forms a k -horizon plan over $\tau = t, \dots, t+k$. Use j to denote how far from the end of an agent's planning horizon: $j = 0, 1, \dots, k$ at $\tau = t+k-j$.

Beliefs within horizon for type- k

$$\pi_{\tau}^j = \beta \mathbb{E}_{\tau} \pi_{\tau+1}^{j-1} + \kappa y_{\tau}^j + u_{\tau}, \quad y_{\tau}^j = \mathbb{E}_{\tau} y_{\tau+1}^{j-1} - \sigma (i_{\tau}^j - \mathbb{E}_{\tau} \pi_{\tau+1}^{j-1} - r_{\tau}^e).$$

End-of-Horizon Beliefs (next slide): at $j=0$, (π_{t+k}^0, y_{t+k}^0) are pinned by continuation values v_{pt}, v_{ht} .

Aggregation and uncertainty

$$\pi_t = \omega_t \pi_t^{k_0} + (1-\omega_t) \pi_t^{k_1}, \quad y_t = \omega_t y_t^{k_0} + (1-\omega_t) y_t^{k_1},$$

with $\omega_t = \omega(m_t)$, where $m_t \in \{0, 1\}$ follows a Markov process (time-varying composition of horizons).

Implication. Multiplicative uncertainty from ω_t breaks certainty equivalence.

Continuation Value Functions, Learning, and Inflation Scores

Terminal beliefs at horizon $t+k$

$$\pi_{t+k}^0 = \beta v_{p,t} + \kappa y_{t+k}^0 + u_{t+k}, \quad y_{t+k}^0 = v_{h,t} - \sigma (i_{t+k}^0 - r_{t+k}^e).$$

Updating of Continuation Value Functions

$$\begin{aligned} v_{p,t+1} &= (1-\gamma_p) v_{p,t} + \gamma_p v_{p,t}^e, & v_{p,t}^e &= \pi_t, \\ v_{h,t+1} &= (1-\gamma_h) v_{h,t} + \gamma_h v_{h,t}^e, & v_{h,t}^e &= y_t + \sigma \pi_t. \end{aligned}$$

Interpretation: $v_{p,t}$ is firm's (marginal) continuation value function; $v_{p,t}^e$ represents new estimate coming from firm optimization; Through adaptive learning, persistent high $\pi_t \Rightarrow v_{p,t}$ drifts up \Rightarrow **inflation score**.

Perceived Monetary Policy

Perceived objective

- ▶ Private-sector agents believe the central bank minimizes discounted expected losses over a k -period horizon.
- ▶ In short: *as-if* time-consistent optimal policy, but only over $t \rightarrow t+k$.

Implied targeting rule

$$y_{t+k-j}^j = -\frac{\kappa}{\lambda} \pi_{t+k-j}^j, \quad j = 0, \dots, k-1.$$

Agents forecast a *lean-against-the-wind (LAW)* targeting rule à la CGG.

Perceived shock management

- ▶ **Demand shocks:** fully offset—outcomes remain on target.
- ▶ **Supply (cost-push) shocks:** contract demand ($y < 0$) to offset inflationary pressure ($\pi > 0$).

Implemented Policy: Modified LAW

Optimal targeting rule (actual CB policy)

$$\pi_t + \gamma_p \beta E_t W_{p,t+1} = -\frac{\lambda}{\kappa} y_t.$$

What is $W_{p,t}$? (incentive to anchor longer-run inflation beliefs)

$W_{p,t} :=$ marginal change in the CB's discounted loss from a unit increase in $v_{p,t}$,

$$W_{p,t} = \beta a_{p,t} \pi_t + \beta(1 - \gamma_p + \gamma_p \beta a_{p,t}) E_t W_{p,t+1},$$

$$a_{p,t} := \frac{\partial E_t \pi_{t+1}}{\partial v_{p,t}} \quad (\text{sensitivity of near-term inflation to long-run belief}).$$

Interpretation

Preemption: If $E_t W_{p,t+1} > 0$, then CB expects belief drift to cause persistent future deviation of inflation. The CB will contract demand ($y_t < 0$) to dampen belief drift and re-anchor expectations.

Estimation Strategy and Policy Analysis

Data & estimation

- ▶ **Sample & observables.** U.S. data on output growth, inflation, and the policy rate over 1966Q1-2007Q4.
- ▶ **Method.** Bayesian estimation with time-varying parameter, ω_t , fraction of short-horizon planners.

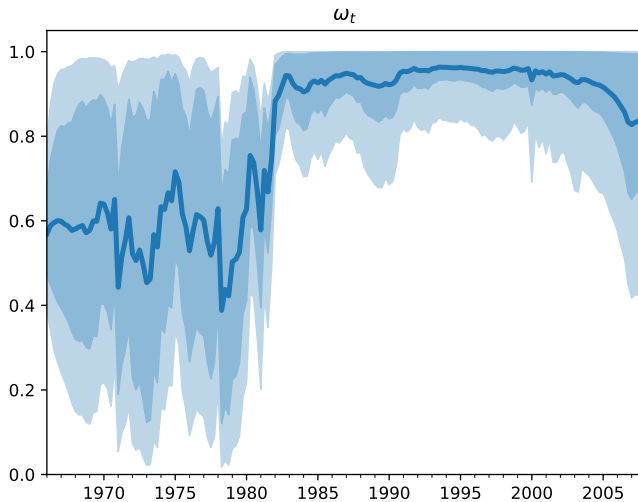
Time-varying planning horizon

- ▶ Two planner types with fixed horizons: $k_0 = 4$, $k_1 = 32$ (1y vs. 8y).
- ▶ Fluctuations in ω_t shift the economy-wide effective planning horizon over time.

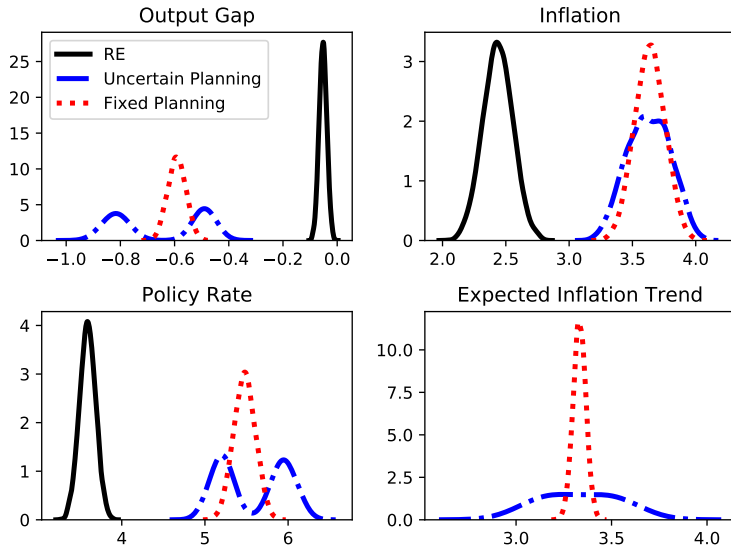
Policy analysis setup

- ▶ **Scenario.** March 2023 FOMC SEP: analyze “inflation-scare” risk when it was a concern for policymakers.
- ▶ **Design.** Stochastic simulations around shocks chosen to reproduce the SEP medians.

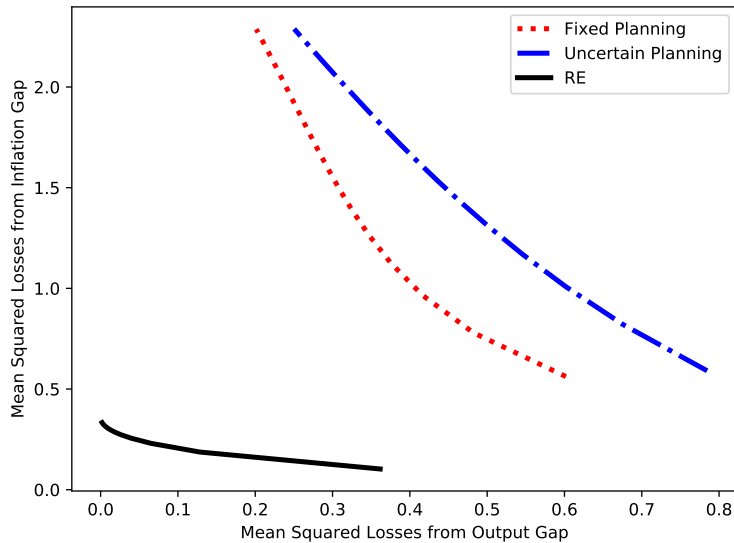
Estimated Fraction of Short-Horizon Planners, ω_t



Distribution of Outcomes under Optimal Discretion



Inflation–Output Variance Frontier



Takeaways

1. FHP + Long-Run Learning: longer-run inflation expectations can drift with realized inflation.
2. Optimal discretion adds a **preemptive term** to LAW of CGG (1999): strengthens anti-inflation stance when scares are plausible.
3. Planning-horizon uncertainty worsens the policy frontier and skews outcome distributions.
4. In the 2023 context, guarding against scares delivers materially lower *expected* losses for CB with dual-mandate objective.