Monetary Policy and Inflation Scares¹

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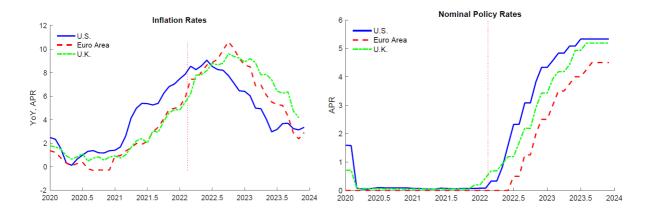
Cleveland Fed - ECB Conference: Inflation: Drivers and Dynamics

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¹The views expressed herein are those of the authors and should not be attributed to the IMF, its Executive Board, or its management.

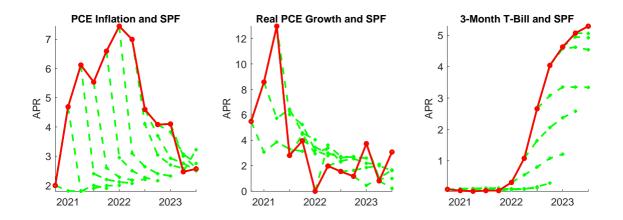
Motivation: The Inflation Surge

• Major CBs attempted to "look through" a transient inflation surge.



Motivation: U.S. Data

• Also professional forecasters underestimated inflationary pressures and growth momentum



Questions

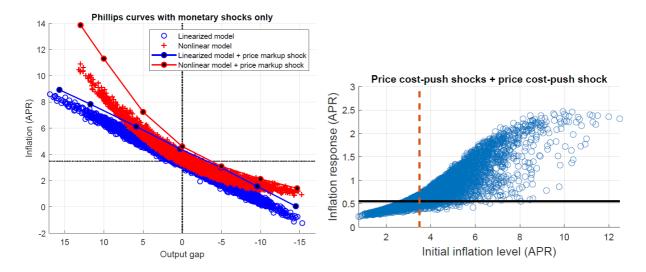
- How can we account for the surge in observed inflation *and* the dynamics of SPF inflation forecasts?
- What are the risks of "looking through" large supply shocks, especially in a hot economy?
- How do Phillips Curve nonlinearities affect the costs of disinflation?
- What are the lessons of post-Covid inflation dynamics for the conduct of monetary policy going forward?

Outline

- Motivation & Questions \checkmark
- Model Overview
- Model-Data Comparison
- Transmission of Adverse Cost-Push Shocks
- Conclusions

Model Overview

- Starting point: model in Harding, Lindé and Trabandt (2022, 2023).
 - Nonlinear New Keynesian model with sticky prices and sticky wages (Erceg, Henderson and Levin, 2000) plus Kimball (1995) aggregation.
 - Nonlinear price and wage inflation Phillips curves.
- Three new model features:
- Learning about the nature of cost-push shock.
- **2** Forecast-targeting based Taylor rule with smoothing.
- **③** State-dependent price and wage indexation (rule-of-thumb price/wage setting).



Cost-push Shock: Unobserved Components Representation

- Cost-push shock, a_t , consists of an iid part $a_{T,t}$ and a persistent part $a_{P,t}$.
 - Agents can observe a_t but not $a_{T,t}$ or $a_{P,t}$.
- Similar to e.g. Erceg and Levin (2003) and Edge, Laubach and Williams (2007).
- State space system:

$$a_t = egin{bmatrix} 1 & 1 \end{bmatrix} egin{bmatrix} a_t^P \ a_t^T \end{bmatrix} \ \begin{bmatrix} a_{t+1}^P \ a_{t+1}^T \end{bmatrix} = egin{bmatrix}
ho_P & 0 \ 0 & 0 \end{bmatrix} egin{bmatrix} a_t^P \ a_t^T \end{bmatrix} + egin{bmatrix} \sigma_P & 0 \ 0 & \sigma_T \end{bmatrix} egin{bmatrix} arepsilon_{t+1}^P \ arepsilon_{t+1}^P \end{bmatrix}$$

• Agents use Kalman filter to predict $a_{T,t|t}$ and $a_{P,t|t}$ observing a_t for t = 0, 1, ..., assume $\sigma_T > \sigma_P$.

Forecast-targeting Taylor Rule

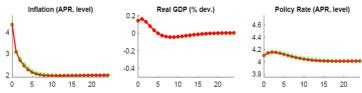
• Forecast-targeting Taylor rule:

$$\frac{R_t}{R} = \left\{\frac{R_{t-1}}{R}\right\}^{\rho} \left\{\frac{\mathsf{E}_t \Pi_{t+4}}{\Pi}\right\}^{(1-\rho)\gamma_{\pi}} \left\{\frac{y_t}{y_t^{pot}}\right\}^{(1-\rho)\gamma_{\chi}}$$

- Baseline: Four-quarter ahead expected inflation (qoq), i.e. $E_t \Pi_{t+4} = E_t \frac{P_{t+4}}{P_{t+3}}$.
- Variations: Π_t or $\mathsf{E}_t \Pi_{t+8}$.

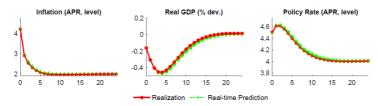
Transient Supply Shock: Actual Vs. Forward-looking Rule

- Forward-looking rule implies "looking trough" transient cost-push ε_t^T .
- Rule responding to π_t implies $y_t \downarrow$ for transient ε_t^T but small inflation gain.



• Taylor Rule with Actual Inflation π_t

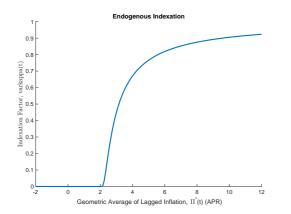
• Forecast-based Taylor Rule with $E_t \pi_{t+4}$



State-Dependent Indexation

• Non-optimizing (rule-of-thumb) firms set $P_{i,t} = \tilde{\Pi}_t P_{i,t-1}$ where $\tilde{\Pi}_t = \Pi^{1-\varkappa_t} \Pi_{t-1}^{\varkappa_t}$ and

$$arkappa_t = e^{-rac{arrho}{\max(\Pi_t^* - \Pi_t \, 0.0001)}}$$
, $\Pi_t^* = \left(\Pi_{t-1}^*
ight)^\omega \left(\Pi_{t-1}
ight)^{1-\omega}$



Parameters:
$$\varrho = 0.002$$
, $\omega = 0.8$, $\Pi = 1.005$.

Note: state-dependent indexation disappears upon log-linearization.

Similar setup for wage indexation.

Parameters I

П Steady state gross inflation rate 1.005

- Net price markup in steady state 0.1
- $heta_p \\ ilde{\xi}_p$ 2/3 Calvo price stickiness parameter
- ψ_p Parameter Kimball aggregator prices -12
 - 0.002 Curvature parameter endogenous indexation 0
- 0.8 Parameter in endogenous. indexation ω
- Inflation indexation parameter in linear model 0 \mathcal{X}
- θ_w 0.1 Net wage markup in steady state
- ξ_w Calvo wage stickiness parameter 0.75
- -6 Parameter Kimball aggregator wages ψ_w

Parameters II

ρ	0.85	Taylor rule: interest rate smoothing
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- γ_{π} 1.5 Taylor rule: coef. on expected inflation
- $\gamma_x = 0.125$ Taylor rule: coef. on output gap
- β 0.995 Household discount factor
- h 0.7 Household consumption habit
- χ 0 Inverse Frisch elasticity of labor supply
- $ho_P = 0.9$ AR(1) persistent markup shock
- $\rho_T = 0$ AR(1) transitory markup shock
- σ_P 1 Standard deviation persistent markup shock
- σ_T 10 Standard deviation transitory markup shock
- ρ_{δ} 0.9 AR(1) discount factor shock

Solution Algorithm

- Solve the nonlinear model with a sequential Fair-Taylor (1983) algorithm.
- Stochastic simulation under certainty equivalence teases out difference between linear and nonlinear solutions.

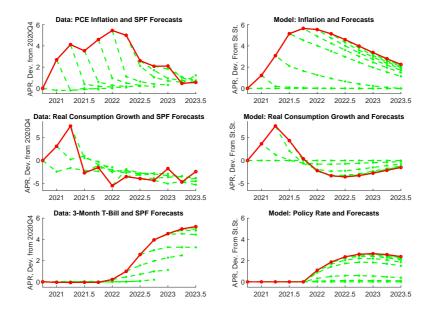
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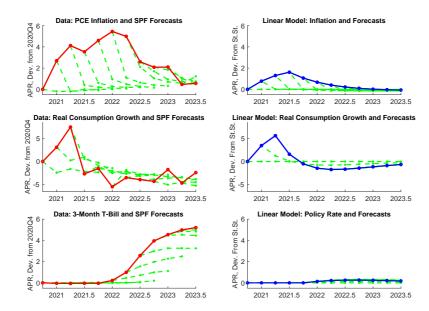
Model-Data Comparison: Model Setup

- Add *Large* shocks ε_t^P to persistent cost-push component, $a_{P,t}$.
- Allow for *Small* positive demand shocks (negative discount factor shocks).
- Phase in shocks over 2-3 quarters.
 - Agents use Kalman filtering for cost-push shocks, demand shocks observed.

Model-Data Comparison: Nonlinear Model



Model-Data Comparison: Linearized Model



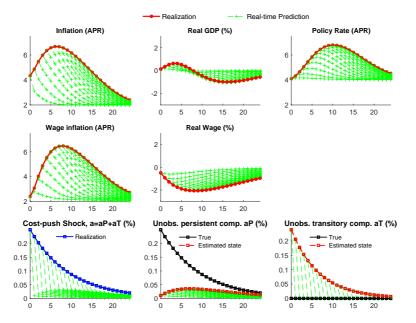
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Transmission of Cost-Push Shocks: Model Setup

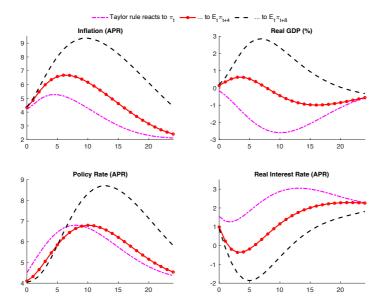
- Now study single large innovation ε_t^P to persistent cost-push shock $a_{P,t}$.
 - As before, agents only observe a_t , use Kalman filtering to calculate $a_{P,t|t}$ and $a_{T,t|t}$.

Transmission of an Adverse Cost-Push Shock



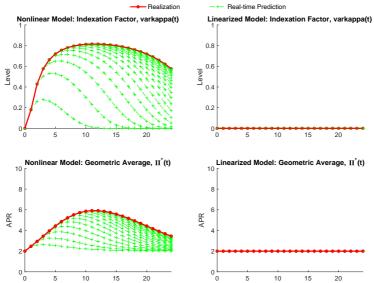
Role of Inflation Variable in the Policy Rule

• Transmission of the same cost-push shock under alternative policy rules.



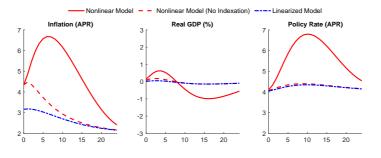
Intrinsic Endogenous Indexation Quantitatively Important

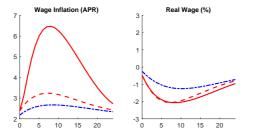
• Amplification via increasing endogenous indexation $\varkappa_t > 0$ in nonlinear model.



Impact of Nonlinearities and Indexation

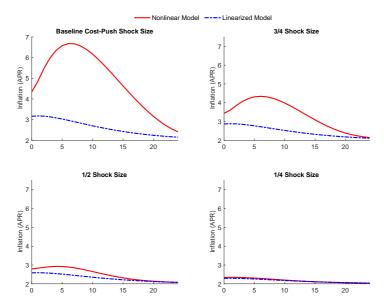
• Transmission of same ε_t^p in baseline (nonlin), nonlin with $\varkappa_t = 0$, and linearized models.





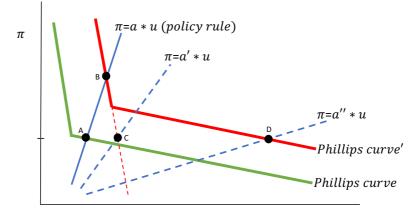
Impact of Shock Size

• Transmission of different realizations of ε_t^p in nonlinear and linearized solutions.



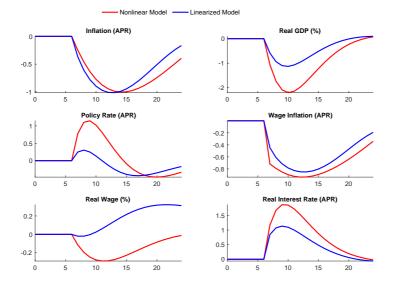
How Costly is the Last Mile?

• Adverse shock shifts economy from A to B, but more aggressive policy response may bring economy to point D rather than C.



Quantifying the Last Mile – More Monetary Tightening

• PC slope in linear model set to imply same initial $\Delta \pi / \Delta y$ as in nonlinear model.



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Conclusions - Modeling

- Macromodel with 1. Nonlinear Price and Wage Phillips curves, 2. Learning about adverse cost-push shock, 3. Endogenous intrinsic indexation, accounts for persistent inflation surge without severe contraction in economic activity.
 - **Key**: central bank follow forecast-based policy rule and misjudges inflationary pressures, so lower real interest rates boost economic activity initially even as inflation is rising.
 - Model less reliant on large demand shocks to explain the 2021-23 inflation scare.
 - Model implies that sign restricted SVARs likely misleading, at least for this episode.
- Under identical assumptions, a linearized formulation of our model cannot generate inflation cycle.
- Large shocks key, little difference between nonlinear and linearized model for small shocks.

Conclusions - Policy

- Need to rethink standard policy prescription to look-through cost-push shocks.
 - While "Looking through" supply shocks reasonable when inflation is close to target, this policy is risky when inflation rises persistently above target in a unpredictable manner.
- Analysis highlights risks of putting too much weight on "point forecasts" of inflation, especially with large forecast (shock) uncertainty.
- Finally, our analysis implies that the economic costs of "going the last mile" i.e. aggressive tightening aimed at returning inflation quickly to target once it has peaked and started to recede can be considerable.

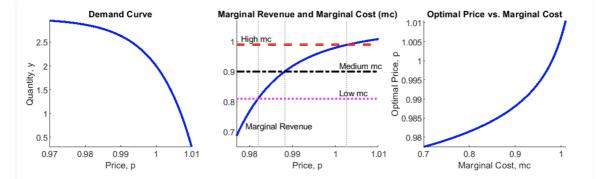
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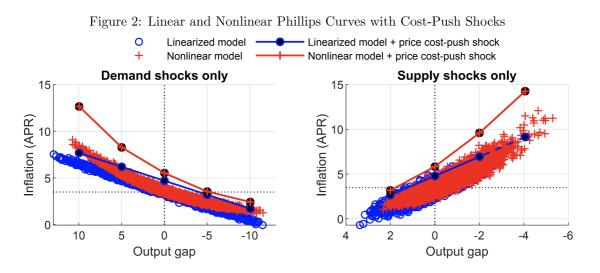
Annex

- Competitive firms aggregate intermediate goods $Y_{i,t}$ into final good Y_t using technology $\int_0^1 G(Y_{i,t}/Y_t) di = 1.$
- Following Dotsey-King (2005) and Levin-Lopez-Salido-Yun (2007):

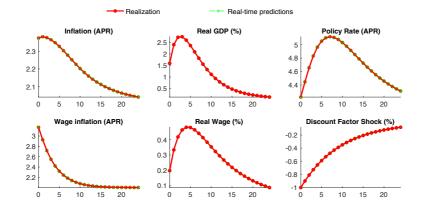
$$G\left(\frac{Y_{i,t}}{Y_t}\right) = \frac{\omega_p}{1+\psi_p} \left[\left(1+\psi_p\right) \left(\frac{Y_{i,t}}{Y_t}\right) - \psi_p \right]^{\frac{1}{\omega_p}} + \frac{1+\psi_p - \omega_p}{1+\psi_p}$$

- $\psi_p < 0$: Kimball (1995), $\psi_p = 0$: Dixit-Stiglitz.
- Kimball aggregator: demand elasticity for intermediate goods increasing function of relative price.
 - Firms increase prices more than they cut prices because of quasi-kinked demand.





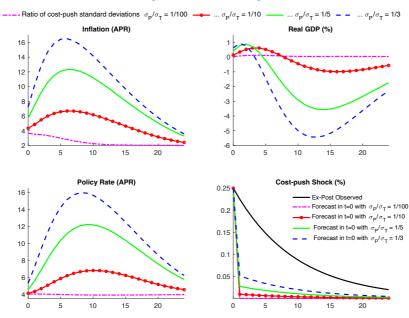
Effects of Discount Factor Shock



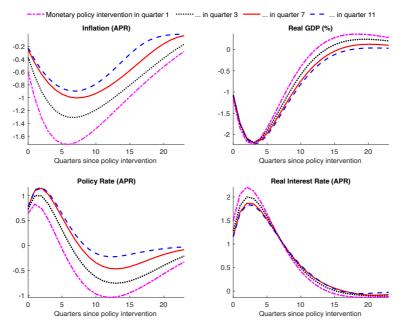
State-Dependent Amplification of Cost-Push Shocks

	e line: Discount	Scenario: Baseline+Same-	State-dependent Effects
	actor Shock	Sized Cost-Push Shock	of Cost-Push Shock
Discount Shock	Inflation Peak	Inflation Peak	∆ Inflation Peak (Scenario-Baseline)
-0.0%	2% (Steady State)	6.7%	4.7%
-0.5%	2.2%	8.0%	5.8%
-1.0%	2.4%	9.4%	7.0%
-1.5%	2.7%	10.8%	8.1%
-2.0%	3.2%	12.2%	9.0%

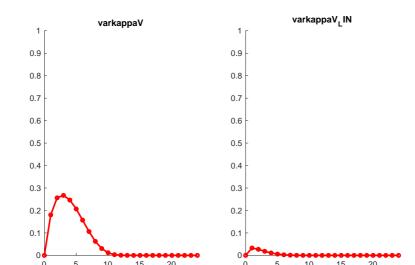
Effects of Unobs. Components Specification



Effects of Timing of Policy Thightening



Cost-Push Shock: Engogenous Indexation Variables in IID Case



Data-Model Comparison

- Macro data for inflation, growth and interest rates, including SPF.
- Evidence on endogenous indexation/rule of thumb firms (in progress).
- Micro data on frequency of price adjustment (in progress).

Linear Model vs. Data

