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# PRICE SETTING FREQUENCY AND THE PHILLIPS CURVE<sup>1</sup>

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# OUTLINE OF THE PRESENTATION

- 1 Motivation
- 2 An extended NK model
- 3 The asymmetric Phillips curve
- 4 Fitting micro and macro data with a small NK model
- 5 Conclusion

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# NK MODELS AND THE SHIFT IN PHILLIPS CURVE I

1. Well established fact that NK models struggle to fit the shifts in the Phillips curve:

$$\hat{\pi}_t = \beta(\chi) \mathbb{E}_t \hat{\pi}_{t+1} + \kappa(\theta, \chi, \cdot) \hat{y}_t + \chi \hat{\pi}_{t-1} + \varepsilon_t^s$$

2. Solving the missing deflation and inflation in NK models:

- ▶ higher Calvo  $\Rightarrow$  stickier prices / flatter NKPC (Del Negro et al., 2015);

- ▶  $\lambda \rightarrow 0$

- ▶ large autocorrelated cost-push shocks and indexation (Fratto and Uhlig, 2020; King and Watson, 2012);

- ▶  $\lambda \rightarrow 1$  and  $\lambda \rightarrow 0$

3. We end up explaining inflation with shocks on inflation ....

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# NK MODELS AND THE SHIFT IN PHILLIPS CURVE II

- ▶ Literature focuses on:
  - ▶ **non-linear** effects (Harding et al., 2022);
  - ▶ exogenous **shift in price stickiness** (Davig, 2016; Fernández-Villaverde and Rubio-Ramírez, 2007);
  - ▶ **change in price updating behaviour** (Del Negro et al., 2020; Costain et al., 2022).
- ▶ Point of departure, a combination of all of that:
  - ▶  $\Rightarrow$  **endogenous time-varying price-setting frequency  $\theta_t$ .**

# MOTIVATION FOR TIME VARIATION IN THE CALVO I

- ▶ The Calvo probability  $0 < \theta < 1$  can be interpreted as the **exogenous share of unchanged prices** at one period.
  - ▶ It is assumed to be a structural parameter (Fernández-Villaverde and Rubio-Ramírez, 2007), yet the estimated value has moved from  $\theta \simeq 0.75$  to  $\theta \simeq 0.9$  with post 2008 samples?
- ▶ **Micro-data contradicts the static Calvo** assumption (Blinder et al., 1998; Klenow and Kryvtsov, 2008; Nakamura et al., 2018).
- ▶ Pure state dependent pricing models struggle with empirical money non-neutrality (Nakamura and Steinsson, 2010; Costain et al., 2022).

# MOTIVATION FOR TIME VARIATION IN THE CALVO II

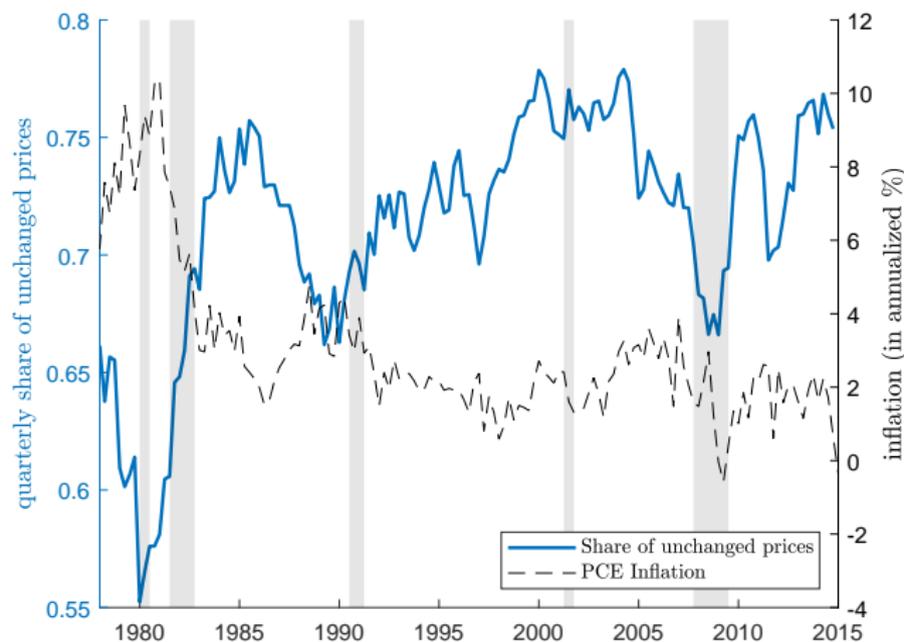


FIGURE 1: Seasonally adjusted share of unchanged prices,  $\theta_t$ , in the US from price tags data changes weighted according to the 2000 household consumption basket based on Nakamura et al. (2018).

# WHAT WE DO

1. Implement a **time-varying price-setting frequency** in a NK model via the *Calvo law of motion*:
  - ▶ update or not  $\mapsto$  **discrete choice process**;
  - ▶ decision is based on the **present values** of updating;
  - ▶ Time-dependent pricing with a flavour of state-dependence  
 $\Rightarrow$  **highly tractable!**
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- ▶ Our extended model:
  1. generates an **asymmetric Phillips curve** which is:
    - steep during boom;
    - flat during bust;
  2. is consistent with **micro and macro**-data;
  3. can **explain the shifts in the Phillips Curve** without large cost-push shocks, high indexation or very sticky prices;

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=> How to approximate the resetting problem?

- ▶ Our **innovation**: discrete choice model à la Brock and Hommes (1997), McFadden (2001) or Matějka and McKay (2015):

$$\theta_t = \frac{\exp(\omega U_t^f)}{\exp(\omega U_t^f) + \exp(\omega (U_t^* - \tau + \varepsilon_t^\theta))}, \quad (1)$$

- ▶  $\theta_t$ : Share of non resetting firms;
- ▶  $*$  is the index for the optimal resetting price;
- ▶  $f$  is the index for the average old price;
- ▶  $U_t^f, U_t^*$ : Present values of the pricing decisions;
- ▶  $\omega, \tau$ : Intensity of choice and fixed cost of updating;
- ▶  $\varepsilon_t^\theta$ : AR(1) shock explaining the residual variation.

⇒ Consistent with **state-dependent pricing models**.

- ▶ Calvo aggregation:

$$P_t = \left( \theta_t P_{t-1}^{1-\epsilon} + (1 - \theta_t) P_t^*{}^{1-\epsilon} \right)^{\frac{1}{1-\epsilon}} \quad (2)$$

- ▶ Firm maximization problem (w/ linear production technology):

$$\begin{aligned} \max_{P_t^*} \mathbb{E}_t \sum_{j=0}^{\infty} \mathcal{D}_{t,t+j} \left( \prod_{k=0}^j \theta_{t+k} \right) \theta_t^{-1} \left[ \frac{P_t^*}{P_{t+j}} - \frac{\Gamma'_{t+j}}{P_{t+j}} \right] Y_{i,t+j} \\ \text{s.t. } Y_{i,t+j} = \left( \frac{P_t^*}{P_{t+j}} \right)^{-\epsilon} Y_{t+j} \end{aligned}$$

- ▶ Firm's FOC:

$$P_t^* = \frac{\epsilon}{\epsilon - 1} \frac{\mathbb{E}_t \sum_{j=0}^{\infty} \left( \prod_{k=0}^j \theta_{t+k} \right) \theta_t^{-1} \mathcal{D}_{t,t+j} \Pi_{t+1,t+j}^{\epsilon} Y_{t+j} w_{t+j}}{\mathbb{E}_t \sum_{j=0}^{\infty} \left( \prod_{k=0}^j \theta_{t+k} \right) \theta_t^{-1} \mathcal{D}_{t,t+j} \Pi_{t+1,t+j}^{\epsilon-1} Y_{t+j}} \quad (3)$$

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# NON-LINEAR DYNAMICS (FAIR AND TAYLOR, 1983)

CALIBRATION INTUITION

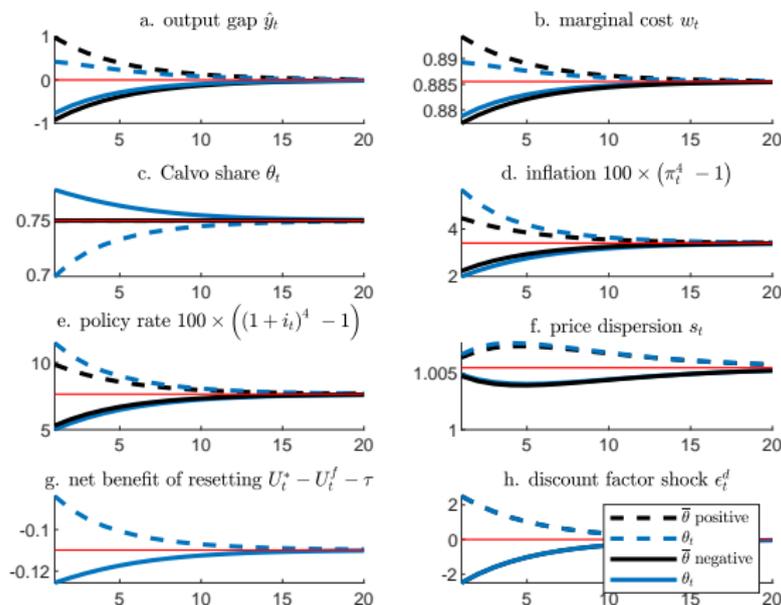
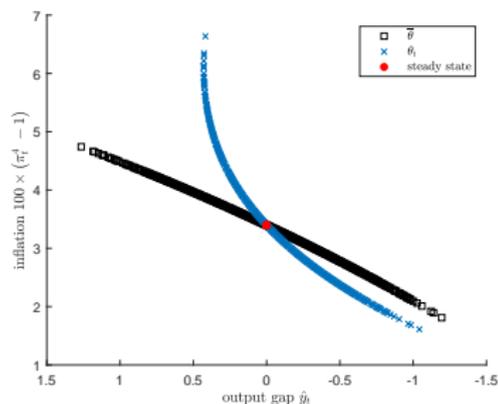
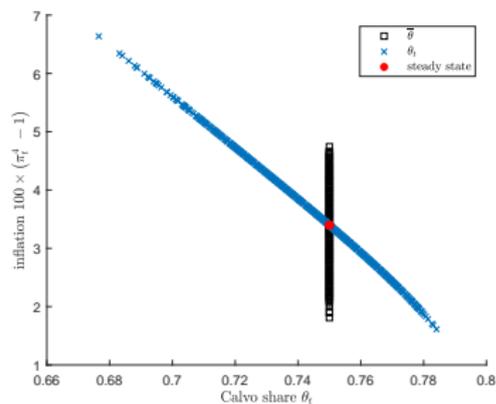


FIGURE 2: Asymmetric impulse responses to a positive or negative demand shock in the small-scale NK model. The shock is a  $\pm 2.5\%$  shock at the discount factor.

# THE NON-LINEAR NKPC (FAIR AND TAYLOR, 1983)



(a) Inflation and Output



(b) Inflation and the Calvo

FIGURE 3: Simulated moments of the non-linear model under discount factor shocks.

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# ESTIMATION RESULTS OF THE SMALL NK MODEL

## POSTERIORS

- ▶ **Objective:** demonstrate the **quantitative** relevance of the mechanism.
- ▶ We estimate the model using data for the US (GDPC1, PCE, FEDFUNDS) from 1964 to 2019.
- ▶ Measurement equations are

$$y_t^{obs} = \hat{y}_t$$

$$\pi_t^{obs} = 100 \times \ln(\bar{\pi}) + \hat{\pi}_t, \quad \text{where} \quad \bar{\pi} = 1 + \gamma_\pi / 100$$

$$r_t^{obs} = 100 \times ((\bar{\pi} / \beta) - 1) + \hat{i}_t$$

$$\theta_t^{obs} = \theta_t,$$

- ▶ Key novelty: Nakamura et al. (2018) **micro**-data for the last equation.

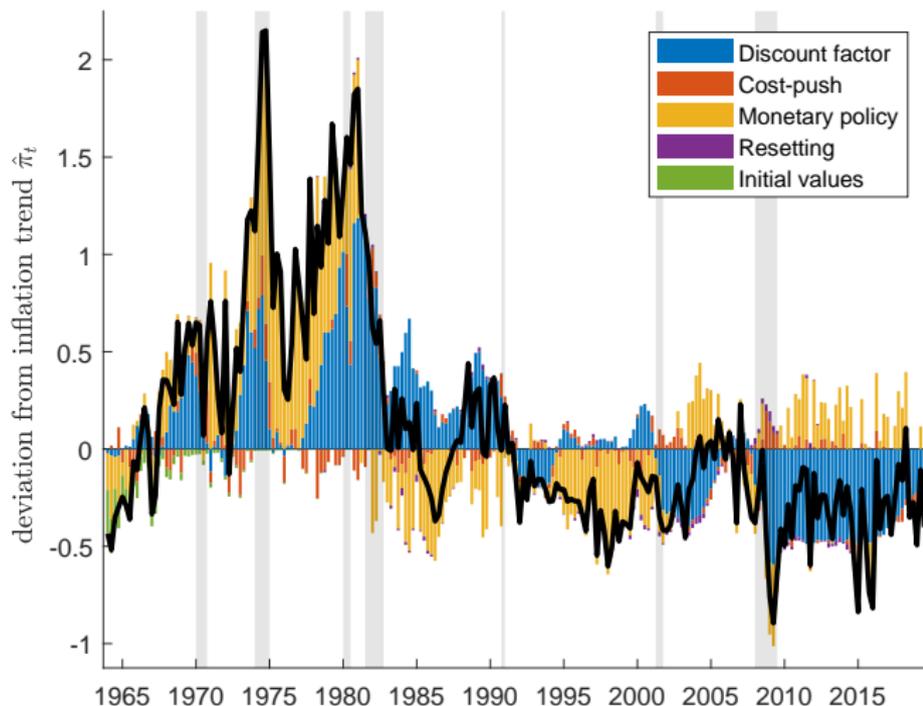


FIGURE 4: Historical decomposition, observed inflation, US data (1964-2019).

# RELEVANCE OF THE ENDOGENOUS CALVO MODEL

## DETAILED MOMENTS

<i>(1964-2019 (full sample))</i>		Filtered model	$\theta_t = \bar{\theta} \forall t$	$\epsilon_t^\theta = 0 \forall t$
$\pi_t$	mean	3.3665	3.2370	3.3926
	median	2.6056	2.6782	2.6595
	variance	5.3527	3.8370	5.4351
	skewness	1.3271	0.8472	1.3343
$\text{corr}(\pi_t, \theta_t)$		-0.8443	0	-0.9844
$\text{corr}(\pi_t, \hat{y}_t)$		0.0839	0.1442	0.0734

TABLE 1: Inflation moments and related statistics, filtered non-linear model and counter-factuals.

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# CONCLUSION

1. Assuming a **static Calvo share** has **limitations**;
2. We provide a model that approximates well the **aggregate** variation in price resetting;
3. The model is consistent with **micro-data** and **macro-data** dynamic;
4. The endogenous price resetting variation drives the non-linearity in the Phillips Curve;
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Thank you for your attention.

Questions? Comments?

- In a simple linear production NK economy we have :

$$\begin{aligned}
 U_t^x &= \mathbb{E}_t \sum_{k=0}^{\infty} \mathcal{D}_{t,t+k} \left( \prod_{j=0}^k \theta_{t+j} \right) \theta_t^{-1} \\
 &\left[ Y_{t+k} \left( \frac{p_t^x}{(\Pi_{t,t+k-1}) \Pi_t^{-1}} \right)^{1-\epsilon} - Y_{t+k} w_{t+k} \left( \frac{p_t^x}{(\Pi_{t,t+k-1}) \Pi_t^{-1}} \right)^{-\epsilon} \right] \\
 &= \left( p_t^{x^{1-\epsilon}} \phi_t - p_t^{x^{-\epsilon}} \psi_t \right) Y_t^\sigma,
 \end{aligned}$$

- $\theta_t$ : Share of non resetting firms;
- $p_t^x$ : Relative price ;
- $w_t$ : real wage;
- $\Pi_t$ : is the cumulated inflation;
- $\epsilon$ : elasticity of substitution among goods;
- $\phi_t$  and  $\psi_t$ : numerator and denominator of the FOC of the optimal price decision.

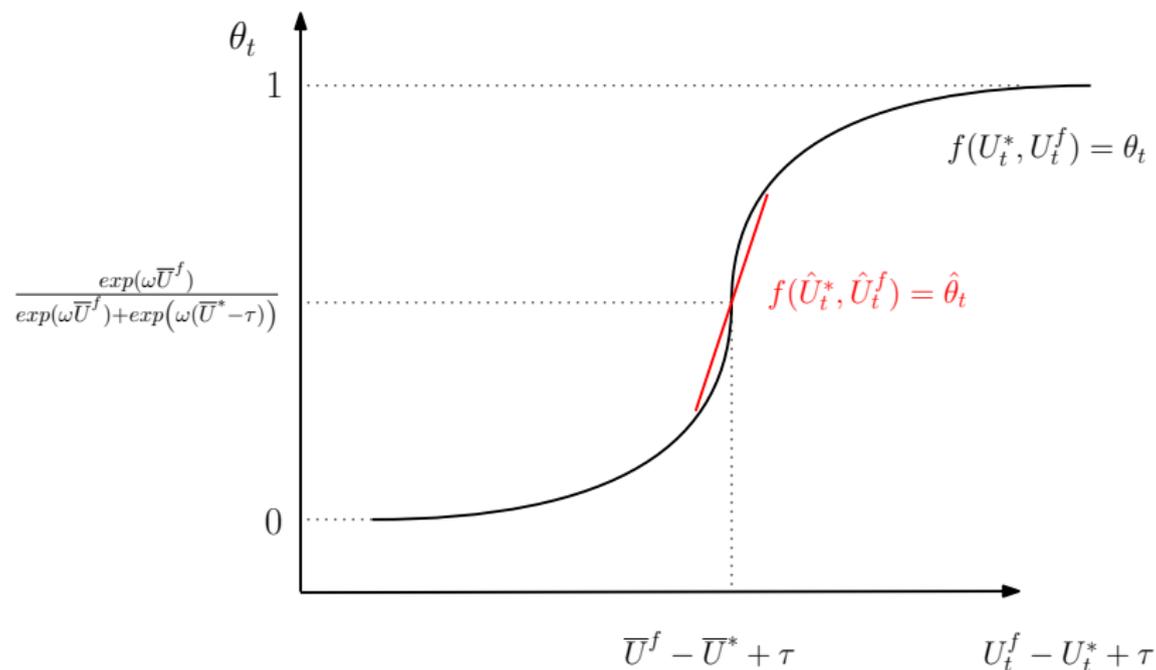


FIGURE 5: The Calvo law of motion (black). The y-axis is the level of  $\theta$  and the x-axis is the difference between the expected profit of not updating and updating the price.

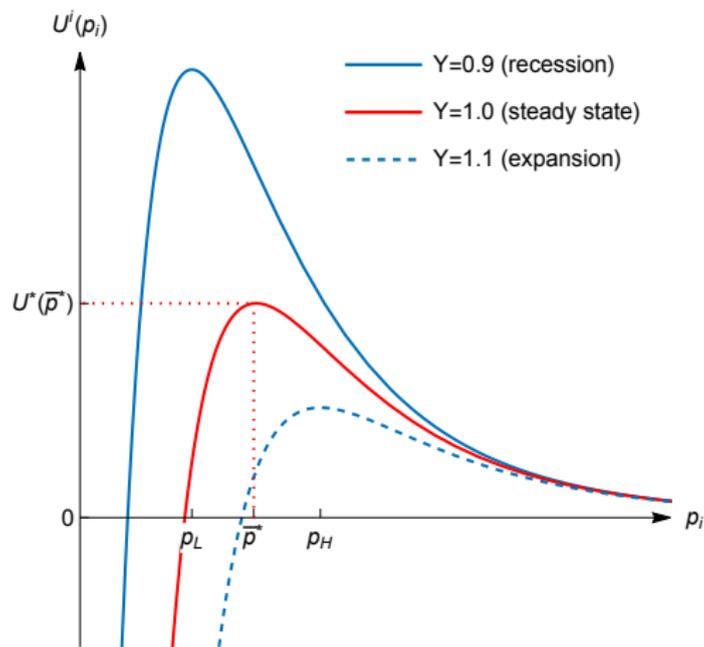


FIGURE 6: Comparative statics: present value of real profits as function of relative price at different levels of output.

The negative relation between inflation and realized/expected Calvo (non-price resetting) share:

$$\hat{\pi}_t = \alpha_1 \hat{y}_t + \alpha_2 \mathbb{E}_t \hat{\pi}_{t+1} + \alpha_3 \mathbb{E}_t \hat{\phi}_{t+1} + \alpha_4 \hat{\theta}_t + \alpha_5 \mathbb{E}_t \hat{\theta}_{t+1} + \varepsilon_t^s, \quad (4)$$

with  $\alpha_1, \alpha_2, \alpha_3, \alpha_5 > 0 > \alpha_4$ .

**Aggregate demand:**  $Y_t^{-\sigma} \exp(\epsilon_t^d) = \beta \mathbb{E}_t \left\{ \frac{(1+i_t)}{\pi_{t+1}} Y_{t+1}^{-\sigma} \exp(\epsilon_{t+1}^d) \right\}$

**Labor supply:**  $w_t = \exp(\epsilon_t^s) \chi N_t^\varphi Y_t^\sigma,$

**Price setting freq. :**  $\theta_t = \frac{\exp(\omega U_t^f)}{\exp(\omega U_t^f) + \exp(\omega(U_t^* - \tau + \epsilon_t^\theta))},$

**Value of firm:**  $U_t^x = (p_t^{x^{1-\epsilon}} \phi_t - p_t^{x^{-\epsilon}} \psi_t) Y_t^\sigma \quad \text{for } x \in \{*, f\}$

**Opt. relative price:**  $p_t^* = \frac{\epsilon}{\epsilon - 1} \frac{\psi_t}{\phi_t}$

$$\psi_t = w_t Y_t^{1-\sigma} + \mathbb{E}_t \beta \theta_{t+1} \pi_{t+1}^\epsilon \psi_{t+1}$$

$$\phi_t = Y_t^{1-\sigma} + \mathbb{E}_t \beta \theta_{t+1} \pi_{t+1}^{\epsilon-1} \phi_{t+1}$$

**Av. relative old price:**  $p_t^f = 1/\pi_t$

**Inflation:**  $1 = (\theta_t \pi_t^{\epsilon-1} + (1 - \theta_t) p_t^{*1-\epsilon})^{\frac{1}{1-\epsilon}}$

**Price dispersion:**  $s_t = (1 - \theta_t) p_t^{*-\epsilon} + \theta_t \pi_t^\epsilon s_{t-1}$

**Aggregate output:**  $Y_t = N_t / s_t.$

**Monetary policy:** 
$$\left( \frac{1 + i_t}{1 + \bar{i}} \right) = \left( \frac{1 + i_{t-1}}{1 + \bar{i}} \right)^\rho \left( \left( \frac{\pi_t}{\bar{\pi}} \right)^{\phi_\pi} \left( \frac{Y_t}{\bar{Y}} \right)^{\phi_y} \right)^{(1-\rho)} \exp(\epsilon_t^r),$$

**Cost-push shock:**  $\epsilon_t^s = \rho_s \epsilon_{t-1}^s - \mu_s u_{\epsilon^s, t-1} + u_{\epsilon^s, t}$

**Other shocks:**  $\epsilon_t^j = \rho_j \epsilon_{t-1}^j + u_{\epsilon^j, t},$

where  $j \in \{d, r, \theta\},$

with  $0 \leq \rho_j, \rho_s < 1, 0 \leq \mu_s < 1$  and  $u_{\epsilon^j, t}, u_{\epsilon^s, t} \sim \text{iid } \mathcal{N}(0, \sigma_j^2).$

<i>Price setting</i>		Value	Source
$\omega$	Intensity of choice	10	-
$\bar{\theta}$	Calvo share	0.75	Galí (2015)
<i>Monetary authority</i>			
$\phi_\pi$	MP. stance, $\pi_t$	1.5	Galí (2015)
$\phi_y$	MP. stance, $Y_t$	0.125	Galí (2015)
$\rho$	Interest-rate smoothing	0	-
$\bar{\pi}$	Gross inflation trend	1.008387	Average log growth of PCE implicit price deflator, 1964-2019
<i>Preferences and technology</i>			
$\beta$	Discount factor	0.99	Galí (2015)
$\sigma$	Relative risk aversion	1	Galí (2015)
$\varphi$	Inverse of Frisch elasticity	0	Ascari and Ropele (2009)
$\epsilon$	Price elasticity of demand	9	Galí (2015)
<i>Exogenous processes</i>			
$\rho_d$	Discount factor shock, AR(1)	0.8	illustrative purpose
$\rho_r$	MP shock, AR(1)	0.8	illustrative purpose

TABLE 2: Calibrated parameters (Galí, 2015) for dynamic simulations (quarterly basis)

		Prior			Posterior		
		Shape	Mean	STD	Mean	5%	95%
<i>Price setting</i>							
$\omega$	Intensity of choice	$\mathcal{N}$	10	.5	8.3664	7.5543	9.1891
$\bar{\theta}$	Calvo share	$\mathcal{B}$	.5	.1	0.7105	0.6984	0.7231
<i>Monetary authority</i>							
$\phi_\pi$	MP. stance, $\pi_t$	$\mathcal{N}$	1.5	.15	2.4311	2.2542	2.6162
$\phi_y$	MP. stance, $Y_t$	$\mathcal{N}$	.12	.05	0.2499	0.1886	0.3101
$\rho$	Interest-rate smoothing	$\mathcal{B}$	.75	.1	0.1585	0.1006	0.2151
$\gamma_\pi$	Quarterly inflation trend	$\mathcal{G}$	.839	.1	0.7486	0.6610	0.8351
<i>Preferences and technology</i>							
$100((\bar{\pi}/\beta) - 1)$	Natural interest rate	$\mathcal{G}$	1.292	.1	1.1861	1.0507	1.3224
$\sigma$	Relative risk aversion	$\mathcal{N}$	1.5	.25	1.6180	1.2940	1.9398
$\varphi$	Inverse of Frisch elasticity	$\mathcal{N}$	2	.37	1.9044	1.3785	2.4297
<i>Exogenous processes</i>							
$\sigma_d$	Discount factor shock, std.	$\mathcal{IG}$	.1	2	0.0255	0.0183	0.0320
$\sigma_s$	Cost-push shock, std.	$\mathcal{IG}$	.1	2	0.0322	0.0272	0.0371
$\sigma_r$	MP shock, std.	$\mathcal{IG}$	.1	2	0.0079	0.0072	0.0086
$\sigma_\theta$	Resetting shock, std.	$\mathcal{IG}$	.1	2	0.0139	0.0121	0.0155
$\rho_d$	Discount factor shock, AR(1)	$\mathcal{B}$	.5	.1	0.9362	0.9173	0.9552
$\rho_s$	Cost-push shock, AR(1)	$\mathcal{B}$	.5	.1	0.9779	0.9676	0.9889
$\mu_s$	Cost-push shock, MA(1)	$\mathcal{B}$	.5	.1	0.1732	0.1195	0.2265
$\rho_r$	MP shock, AR(1)	$\mathcal{B}$	.5	.1	0.5271	0.4789	0.5770
$\rho_\theta$	Resetting shock, AR(1)	$\mathcal{B}$	.5	.1	0.7749	0.7071	0.8427
<i>Log-likelihood</i>						-74.6242	

TABLE 3: Estimated parameters of the augmented small-scale NK model (US: 1964-2019).  $\mathcal{B}$ ,  $\mathcal{G}$ ,  $\mathcal{IG}$ ,  $\mathcal{N}$  denote beta, gamma, inverse gamma and normal distributions, respectively.

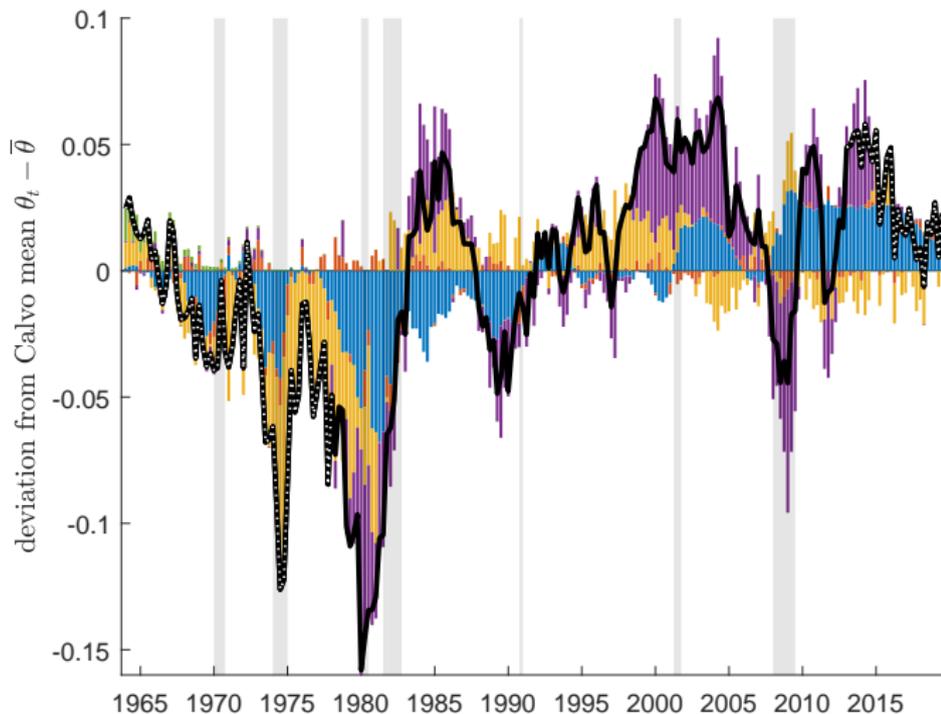


FIGURE 7: Historical decomposition, observed Calvo share, US data (1964-2019).

(a) 1964-2019 (full sample)		Filtered model	$\epsilon_t^f = 0 \forall t$	$\epsilon_t^* = 0 \forall t$	$\epsilon_t^f = \epsilon_t^* = 0 \forall t$	$\epsilon_t^f = 0 \forall t$	$\epsilon_t^* = 0 \forall t$	$\epsilon_t^f = \epsilon_t^* = 0 \forall t$
$\pi_t$	mean	3.3665	3.3926	3.3725	3.4013	3.3195	3.0151	2.9678
	median	2.6056	2.6595	2.7075	2.7123	2.9865	2.9039	2.9555
	variance	5.3527	5.4351	5.2963	5.3741	3.1133	1.9792	1.025
	skewness	1.3271	1.3343	1.2980	1.3160	1.6055	0.8554	0.5939
corr( $\pi_t, \theta_t$ )		-0.8443	-0.9844	-0.8359	-0.9836	-0.7522	-0.7249	-0.4065
corr( $\pi_t, \hat{y}_t$ )		0.0839	0.0734	-0.0296	-0.0380	-0.0994	0.1762	-0.5882
<hr/>								
(b) 1964-1984								
$\pi_t$	mean	5.3995	5.4256	5.3968	5.4270	4.4178	3.9173	2.9924
	median	5.1631	5.1602	4.9738	4.9894	4.0428	3.4877	2.9997
	variance	6.0894	6.2343	5.8975	6.0505	4.7642	2.0190	0.1814
	skewness	0.4630	0.4876	0.4977	0.5253	0.9690	1.0406	0.3136
corr( $\pi_t, \theta_t$ )		-0.9327	-0.9951	-0.9288	-0.9953	-0.8757	-0.8319	-0.4095
corr( $\pi_t, \hat{y}_t$ )		0.0905	0.0802	-0.0136	-0.0486	-0.0741	0.0891	-0.5886
<hr/>								
(c) 1985-2003								
$\pi_t$	mean	2.3207	2.3314	2.3316	2.3405	2.1526	3.1477	2.9542
	median	2.2032	2.2279	2.1992	2.1889	2.0992	3.0875	2.9703
	variance	0.7802	0.7958	0.8222	0.8333	0.5783	0.7724	0.0441
	skewness	0.7364	0.7155	0.8576	0.8082	0.1354	-0.0328	0.0495
corr( $\pi_t, \theta_t$ )		-0.6005	-0.9820	-0.6236	-0.9848	-0.2971	-0.7056	-0.2960
corr( $\pi_t, \hat{y}_t$ )		0.3484	0.3390	0.1707	0.2207	-0.4339	0.5908	-0.6399
<hr/>								
(d) 2004-2014								
$\pi_t$	mean	2.0393	2.1094	2.0240	2.1017	3.3201	1.7428	2.9709
	median	2.0967	2.0811	2.0509	2.1373	3.4543	1.4219	2.9350
	variance	0.9520	1.0373	1.1295	1.1798	0.7157	0.5455	0.0802
	skewness	-0.4800	-0.2448	-0.7384	-0.5228	-0.5588	0.6189	1.3824
corr( $\pi_t, \theta_t$ )		0.2980	-0.9530	0.3193	-0.9351	0.1237	-0.2042	-0.5880
corr( $\pi_t, \hat{y}_t$ )		0.5540	0.4577	0.4975	0.5322	-0.2105	0.7310	-0.5130
<hr/>								
(e) 2015-2019								
$\pi_t$	mean	1.6207	1.6196	1.6882	1.6916	3.1784	1.3842	2.9028
	median	1.7169	1.7643	1.8971	1.9032	3.4464	1.3859	2.9042
	variance	0.9074	0.9890	0.8298	0.8996	0.7946	0.1362	0.0446
	skewness	-0.5074	-0.6060	-0.5028	-0.5992	-0.1402	-0.0915	0.6438
corr( $\pi_t, \theta_t$ )		-0.7487	-0.9096	-0.7578	-0.9210	-0.8215	-0.7935	-0.4349
corr( $\pi_t, \hat{y}_t$ )		0.1297	0.0907	0.8588	0.8924	-0.2886	0.3788	-0.6304

TABLE 4: Inflation moments and related statistics, filtered non-linear model and counter-factuals.

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