

Empirical Investigation of a Sufficient Statistic for Monetary Policy Shocks

Fernando Alvarez, Hervé Le Bihan, Andrea Ferrara, Erwan Gautier, Francesco Lippi

A Analytics of the generalized random menu cost model

We describe the price setting problem for a firm in steady state using the random menu cost model of Caballero and Engel (1999, 2007), which covers a vast class of sticky-price models.

The firm's problem. The firm maximizes the expected discounted value of profits and chooses the optimal times and size of price adjustment as a function of its state x , as encoded in the value function $v(x)$ defined below. A second order approximation of the profit function at the optimal price gives a quadratic period return Bx^2 , where the constant term B relates to the curvature of the profit function.⁴³ The firm's value function solves the following HJB equation

$$r v(x) = \min \left\{ Bx^2 + v'(x)\mu + \frac{\sigma^2}{2}v''(x) + \kappa \int_0^\Psi \min \left\{ \psi + \min_z v(z) - v(x), 0 \right\} dG(\psi), r \left(\Psi + \min_z v(z) \right) \right\}$$

The first argument in the curly bracket represents the continuation value with a flow cost Bx^2 and the usual expected change in the value function, which includes the possibility to adjust if the firm draws a sufficiently small menu cost. For a firm with a gap x that draws a menu cost ψ the net cost effect of adjusting is $\psi + \min_z v(z) - v(x)$, which is optimally chosen by the firm only when it is smaller than zero. The second argument in the curly bracket represents the firms' option to reset the gap at any moment by paying the fixed cost Ψ . The value function features the smooth pasting conditions: $v'(\underline{X}) = v'(\bar{X}) = v'(x^*) = 0$ where $x^* = \arg \min_{\tilde{x}} v(\tilde{x})$ is the optimal price gap chosen by a firm that adjusts and \underline{X} and \bar{X} delimit the state space so that $x \in [\underline{X}, \bar{X}]$, and value matching conditions $v(\underline{X}) = v(\bar{X}) = v(x^*) + \Psi$.

Mapping the model to observables. The density f solves the following Kolmogorov forward equation:

$$f(x)\Lambda(x) = -\mu f'(x) + \frac{\sigma^2}{2}f''(x) \text{ for all } x \in (\underline{X}, \bar{X}), x \neq x^* \quad (16)$$

with boundary conditions: $\lim_{x \downarrow x^*} f(x) = \lim_{x \uparrow x^*} f(x)$; $1 = \int_{\underline{X}}^{\bar{X}} f(x)dx$ and $\lim_{x \rightarrow \bar{X}} f(x) = \lim_{x \rightarrow \underline{X}} f(x) = 0$.

Notice how the cross sectional distribution of price gaps $f(x)$ is fully determined by the generalized hazard function: Also, note that density is zero at the boundaries of the domain.⁴⁴

⁴³See Appendix B in Alvarez and Lippi (2014) for a detailed derivation of the approximation.

⁴⁴This is an implication of \underline{X} and \bar{X} being exit points, for every model where μ/σ^2 is finite. In the case where the domain of x is unbounded, then the zero density is a requirement for integrability.

The distribution of price changes has density q and two mass points at the boundaries of the inaction region:

$$\begin{aligned} q(x^* - x) &= \frac{f(x)\Lambda(x)}{N} \text{ for all } x \in [\underline{X}, \bar{X}] \\ dQ(x^* - \bar{X}) &= -\frac{\sigma^2 f'(\bar{X})}{2} \frac{1}{N} \quad , \quad dQ(x^* - \underline{X}) = +\frac{\sigma^2 f'(\underline{X})}{2} \frac{1}{N} \end{aligned} \quad (17)$$

where the number of price changes satisfies:

$$N(\mu) = \int_{\underline{X}}^{\bar{X}} \Lambda(x)f(x)dx + \frac{\sigma^2}{2} f'(\underline{X}) - \frac{\sigma^2}{2} f'(\bar{X}) \quad (18)$$

and the notation emphasises the dependence of the frequency on the rate of steady state inflation μ .

Computation of CIR for the zero drift case ($\mu = 0$). The contribution to the cumulative impulse response of a firm with price gap x is

$$m(x) = -\mathbb{E} \left[\int_0^\tau e^{-\Lambda(x)t} x(t) dt \mid x(0) = x \right] \quad (19)$$

where τ is the stopping time defined as the first time the price gap hits the barriers $\pm\bar{X}$. In words, $m(x)$ is the expected (cumulative) price gap of a firm that starts with a gap x .⁴⁵

The expectation in the right hand side of [equation \(19\)](#) is with respect to the process for x , a jump-diffusion with jump intensity $\Lambda(x)$, diffusion variance σ^2 , and zero drift. The function $m : [-X, X] \rightarrow \mathbb{R}$ is once continuously differentiable, antisymmetric around $x = 0$, and satisfies:

$$m(x)\Lambda(x) = -x + \frac{\sigma^2}{2} m''(x) \text{ for all } x \text{ at which } \Lambda \text{ is continuous} \quad (20)$$

$$0 = m(X) \text{ if } X < \infty \text{ and } \lim_{x \rightarrow \infty} \frac{|m(x)|}{x} \leq \frac{1}{\inf_y \Lambda(y)} \text{ if } X = \infty \quad . \quad (21)$$

Now we can define the cumulative impulse response to a monetary shock of size δ as

$$CIR(\delta) = \int_{-\bar{X}}^{\bar{X}} m(x)f(x + \delta)dx \quad . \quad (22)$$

The proof of sufficient statistic result, [equation \(6\)](#). First note that the identity

$$N \cdot Var = \sigma^2 \quad (23)$$

holds in the model. Let $x(0) = 0$. Consider the process $z(t) \equiv x(t)^2 - \sigma^2 t$ for $t \geq 0$. Using Ito's lemma we can verify that the drift of x^2 is σ^2 , and hence $z(t)$ is a Martingale. Let τ be a stopping time, i.e. an instant where a price adjustment occurs (anywhere in the state space, including the boundaries),

⁴⁵The definition above uses the steady state decision rule $\Lambda(x)$, thus ignoring the general equilibrium feedback effect of the shock on the firm's decision. In Proposition 7 of [Alvarez and Lippi \(2014\)](#) it is shown that, given a combination of the general equilibrium setup in [Golosov and Lucas \(2007\)](#) and the lack of the strategic complementarities, these general equilibrium effects are of second order. In addition, we use the fact that after the first price change the expected contribution to output of each firm is zero, since positive and negative output contributions are equally likely, so $m(0) = 0$. This allows us to characterize the propagation of the monetary shocks without tracking the time evolution of the whole price gap distribution.

so that x is reset at $x(0) = 0$. By the optional sampling theorem $z(\tau)$, the process stopped at τ , is also a martingale. Then $\mathbb{E}[z(\tau) \mid x(0)] = \mathbb{E}[x(\tau)^2 \mid x(0)] - \sigma^2 \mathbb{E}[\tau \mid x(0)] = x(0) = 0$. Since $N = 1/\mathbb{E}[\tau \mid x(0)]$ and $Var = \mathbb{E}[x(\tau)^2 \mid x(0)]$ we get the identity in [equation \(23\)](#).

For simplicity, we focus next on the case with unbounded support $\bar{X} \rightarrow \infty$ (the logic for the case with bounded support is identical but the equations are slightly more cumbersome). Using the definition of the density of price changes in [equation \(17\)](#) we can rewrite the identity as

$$\int_{-\infty}^{\infty} x^2 \Lambda(x) f(x) dx = \sigma^2 \tag{24}$$

it is then straightforward to write the formula for kurtosis over $6N$ as:

$$\frac{Kurt}{6N} = \frac{\int_{-\infty}^{\infty} x^4 \Lambda(x) f(x) dx}{6 \left(\int_{-\infty}^{\infty} x^2 \Lambda(x) f(x) dx \right)^2} = \frac{\int_{-\infty}^{\infty} x^4 \Lambda(x) f(x) dx}{6\sigma^4}$$

where the last passage uses [equation \(24\)](#). Using the Kolmogorov forward equation,

$$\int_{-\infty}^{\infty} x^4 \Lambda(x) f(x) dx = \frac{\sigma^2}{2} \int_{-\infty}^{\infty} x^4 f''(x) dx$$

Integrating by parts twice gives

$$\int_{-\infty}^{\infty} x^4 \Lambda(x) f(x) dx = 6\sigma^2 \int_{-\infty}^{\infty} x^2 f(x) dx$$

This allows us to write

$$\frac{Kurt}{6N} = \frac{\int_{-\infty}^{\infty} x^2 f(x) dx}{\sigma^2} \tag{25}$$

Recall that we have a system of two equations:

$$\Lambda(x) f(x) = \frac{\sigma^2}{2} f''(x) \quad , \quad \Lambda(x) m(x) = \frac{\sigma^2}{2} m''(x) - x$$

Eliminate Λ to get:

$$\frac{\sigma^2}{2} \frac{m(x) f''(x)}{f(x)} = -x + \frac{\sigma^2}{2} m''(x)$$

Multiply both sides by $f(x)x$ and rearrange:

$$\frac{\sigma^2}{2} [m(x) f''(x) - m''(x) f(x)] x = -x^2 f(x)$$

Integrate both sides from 0 to ∞ :

$$\frac{\sigma^2}{2} \int_0^{\infty} [m(x) f''(x) - m''(x) f(x)] x dx = - \int_0^{\infty} x^2 f(x) dx$$

Perform integration by parts in the left-hand side using the fact that $[m(x) f'(x) - m'(x) f(x)]' = m(x) f''(x) -$

$m''(x)f(x)$:

$$\begin{aligned} \frac{\sigma^2}{2} \int_0^\infty [m(x)f''(x) - m''(x)f(x)]x dx &= \frac{\sigma^2}{2} \left([m(x)f'(x) - m'(x)f(x)]x \Big|_0^\infty - \int_0^\infty [m(x)f'(x) - m'(x)f(x)]dx \right) \\ &= -\sigma^2 \int_0^\infty m(x)f'(x)dx \end{aligned}$$

where the last equality uses integration by parts again. We used $\mathbb{E}[m(x)] < \infty$ and $m(\cdot)$ being almost linear at infinity to justify setting $f'(x)m(x)x$ and $f(x)m'(x)x$ at infinity to 0. Hence, we have

$$\sigma^2 \int_0^\infty m(x)f'(x)dx = \int_0^\infty x^2 f(x)dx$$

Plugging this result in [equation \(25\)](#) we have

$$\frac{Kurt}{6N} = \int_{-\infty}^\infty m(x)f'(x)dx$$

It appears from the CIR definition in [equation \(22\)](#) that the right hand side is just the first derivative of the CIR with respect to δ , evaluated at $\delta = 0$, or $CIR'(0)$. This completes the proof. \square

Aggregation across heterogenous firms. We briefly discuss how the above results can be applied to economies composed of heterogenous firms. Assume that there are J groups of firms with different parameters, each with an expenditure weight $e_j > 0$, N_j price changes per unit of time, and a distribution of price changes with kurtosis Kur_j . In this case, after repeating the arguments above for each group and aggregating, we obtain that the area under the IRF of *aggregate* output for a small monetary shock δ is

$$CIR(\delta) = \frac{\delta}{6\epsilon} \sum_{j \in J} e_j \frac{Kur_j}{N_j} + o(\delta^2) = \frac{\delta}{6\epsilon} D \sum_{j \in J} d_j Kur_j + o(\delta^2) \quad (26)$$

where D is the expenditure-weighted average duration of prices $D \equiv \sum_{j \in J} \frac{e_j}{N_j}$, and $d_j \equiv \frac{e_j}{N_j D}$ are weights that take into account both relative expenditures and durations. When all groups have the same durations, then $d_j = e_j$ and the *CIR* is proportional to the average of the kurtosis of the sectors.⁴⁶ However, if groups are heterogenous in duration (or expenditures), then the kurtoses of the groups with longer duration (or higher expenditures) receive a higher weight in the computation of the *CIR*.⁴⁷

⁴⁶The effect of heterogeneous N is well known for the Calvo model: due to Jensen's inequality D differs from the (reciprocal of) the average of N 's, see for example [Carvalho \(2006\)](#) and [Nakamura and Steinsson \(2010\)](#).

⁴⁷Suppose for instance that a fraction of firms have flexible prices (zero duration in our model, or infinitely many price changes per unit of time), as in [Dotsey and Wolman \(2020\)](#). The above formula implies that the group of the flexible price firms are excluded (zero duration yields a zero weight), and that the cumulative impulse response (CIR) is computed on the mass of firms with sticky prices. Notice that this is different from computing the CIR as the ratio of the cross-sectional average kurtosis and the average frequency. Since the latter is diverging because of the firms with flexible prices, the CIR computed this way would be zero, while obviously it is not.

B FAVAR estimation

The Factor Augmented Vector Autoregression (FAVAR) was originally developed by [Bernanke, Boivin, and Elias](#) (2005) and by [Boivin, Giannoni, and Mihov](#) (2009). [Stock and Watson](#) (2016) provide also a clear explanation of the model.

Let i_t be a vector of observable economic variables with dimension $M \times 1$, $M \geq 1$, and let \tilde{F}_t be a vector of unobserved factors with dimension $K \times 1$, $K \geq 1$.⁴⁸ Assume that the dynamics of the economy is driven by (Y_t', \tilde{F}_t') which follows the transition equation:

$$\begin{bmatrix} \tilde{F}_t \\ i_t \end{bmatrix} = \Phi(L) \begin{bmatrix} \tilde{F}_{t-1} \\ i_{t-1} \end{bmatrix} + v_t \quad (27)$$

where $\Phi(L)$ is a lag polynomial of finite order and v_t is an error term with zero mean and covariance matrix Q . While [equation \(27\)](#) has a VAR form, given that F_t is unobserved we cannot directly estimate [equation \(27\)](#). However, the factors \tilde{F}_t are interpreted as representing forces that potentially affect many economic variables from which we can estimate the factors. Indeed, assume that a large number of time series X_t , called informational time series, are related to the observed variables i_t and to the unobservable factors \tilde{F}_t by the following equation:

$$X_t = \Lambda F_t + e_t \quad (28)$$

where $F_t \equiv [\tilde{F}_t' i_t']'$ and e_t is a vector $N \times 1$ of error terms with zero mean.⁴⁹ Notice that the number of informational time series, N , must be large which means N is much larger than the number of variables that drives the economy (F_t and i_t), i.e. $N > K + M$, and potentially N can be larger than the number of observations in the time dimension, T . Moreover, notice that F_t can always capture arbitrary lags of fundamental factors, thus it is not restrictive to assume that X_t depends only on the current values of the factors⁵⁰.

Under the above assumptions, it is possible to estimate the model, using a two-step approach.⁵¹ In the first step, the common factors are estimated extracting the first K principal components, $\hat{C}^{(0)}$, from the information variables, X_t . Indeed, as shown by [Stock and Watson](#) (2002), for N large enough and if the number of principal components used is as least as large as the true number of factors, the principal components of X_t span the space generated by the factors \tilde{F} and the observable variables i_t ; thus, the principal components represent independent but arbitrary linear combinations of \tilde{F}_t and i_t . However, we want that these combinations do not depend on i_t and that they are only independent combinations of the factors. For this reason, the factors are estimated as follow. Regress X_t on $\hat{C}^{(0)}$ and i_t to obtain $\hat{B}_r^{(0)}$, the coefficient of i_t . Then compute $\tilde{X}_t^{(0)} = X_t - \hat{B}_r^{(0)} i_t$ and estimate $\hat{C}^{(1)}$ as the first K principal components of $\tilde{X}_t^{(0)}$. Iterate until convergence of $\hat{B}_r^{(i)}$ to obtain the desired estimated factors, $\hat{\tilde{F}}_t$. The second step consists in estimating [equation \(27\)](#) as a structural VAR⁵², replacing F_t with their estimated counterpart $\hat{F}_t \equiv [\hat{\tilde{F}}_t' i_t']'$. Indeed, we can rewrite [equation \(27\)](#) as

$$\hat{F}_t = \Phi(L)\hat{F}_t - 1 + v_t \quad (29)$$

where $\hat{F}_t^+ \equiv [\hat{\tilde{F}}_t' i_t']'$. Assuming $v_t = H\epsilon_t$, it is clear that [equation \(29\)](#) can be treated as a structural VAR.

⁴⁸We adopt the notation ' i_t ' as in our application the observable factor reduces to the interest rate.

⁴⁹If factors are estimated using a principal components analysis, errors can display a small amount of cross-correlation that must vanish as N goes to infinity. See [Stock and Watson](#) (2002) for a detailed discussion.

⁵⁰For this reason [Stock and Watson](#) (1999) refer to [equation \(28\)](#) as a dynamic factor model.

⁵¹An alternative to estimate the model is to use a single-step Bayesian likelihood approach.

⁵²In our application, we estimate the structural VAR using a Cholesky decomposition. However, any other approach can be used.

The final step we are interested in is to estimate the IRFs of X_t . Consider again [equation \(29\)](#) and assume that the MA representation exists. Denoting the MA coefficient with $\Psi(L)$, we obtain

$$\hat{F}_t = \Psi(L)H\epsilon_t \quad (30)$$

Moreover, using \hat{F}_t instead of F_t in [equation \(28\)](#) and replacing in this equation [equation \(30\)](#), we get

$$X_t = \Lambda\Psi(L)^{-1}H\epsilon_t + e_t \quad (31)$$

[Equation \(31\)](#) links the information variables, X_t , to the shocks and provides the theoretical framework to retrieve the IRFs of X_t . However, in practice, the IRFs of X_t are not estimated using the MA representation and, thus, [equation \(31\)](#). Indeed, let $\widehat{IRF}(A)$ be the estimated IRFs of the time series A_t to a given shock. The IRFs of X_t is calculated as

$$\widehat{IRF}(X) = \hat{\beta} * \widehat{IRF}(\hat{F}) \quad (32)$$

where $\widehat{IRF}(\hat{F})$ is the VAR estimated IRF of \hat{F}_t and $\hat{\beta}$ is the estimated coefficient of the regression of X_t on \hat{F}_t .

C A Filter for the Euribor

For the purpose of our empirical test, we want the empirical monetary shock to capture a monetary policy shocks, characterized by a transient impact on inflation and output. We filter the 3-month Euribor so as to ensure this property, as with unfiltered data it is not fulfilled. One possible reason why it is not fulfilled (unlike in typical VAR) is because in our sample period, on the euro area, this variable is not stationary, as depicted in [Figure A](#). We exploit two criteria to choose the value of the value of the HP filter, λ^{HP} . Both criterion are based on the behavior of the IRFs of PPI and CPI time series as λ^{HP} varies. We estimated the FAVAR model, for alternative values of filtered Euribor rates letting λ^{HP} vary from 6 to 10^5 .

One first criterion consists in considering the number of negative IRFs of PPI after two or three years, since our strong prior is that after a contractionary monetary shock, prices should decline as compared to the no-shock baseline. Thus, we are interested in estimating a FAVAR that is in line with this prediction. The top panel of [Figure B](#) shows our finding: the number of negative IRFs is maximized around the value of λ^{HP} of 500. The number of PPI sectors included in the analysis are 118 and for $\lambda^{HP} = 500$ we have around 100 sectors with negative IRF. Moreover, this curve is very flat around 500, as for any value of λ^{HP} in between 200 and 3000, more than the 60% of the sectors have a negative IRF after two years or three years.

As a second criterion to guide our choice of λ^{HP} , we consider the value of the *aggregate* IRF of PPI (or CPI) to the contractionary monetary shock as a function of λ^{HP} . For PPI, these responses are reported in the bottom panel of [Figure B](#). This panel shows four different lines: we consider the response of an aggregate time series of PPI after 24 or 36 months; and that of the arithmetic average of the sectoral response of PPIs in addition to the response of the aggregate price index. In all cases, the minimum response is found λ^{HP} equal to a value around 1000. Developing the same criteria for CPI, we obtain very similar results, as depicted by [Figure C](#). The only difference is that in the bottom panel for value of λ^{HP} bigger than 10^4 , the aggregate responses after two or three years become positive, implying that large values of λ^{HP} would defeat our purpose. Overall, based on the above results, we select as our benchmark to filter the 3-month Euribor an HP filter with $\lambda = 1000$.

Some additional figures: Filtering of interest rate, Identification of monetary policy shock

Figure A: 3-month Euribor: period 2005-2019

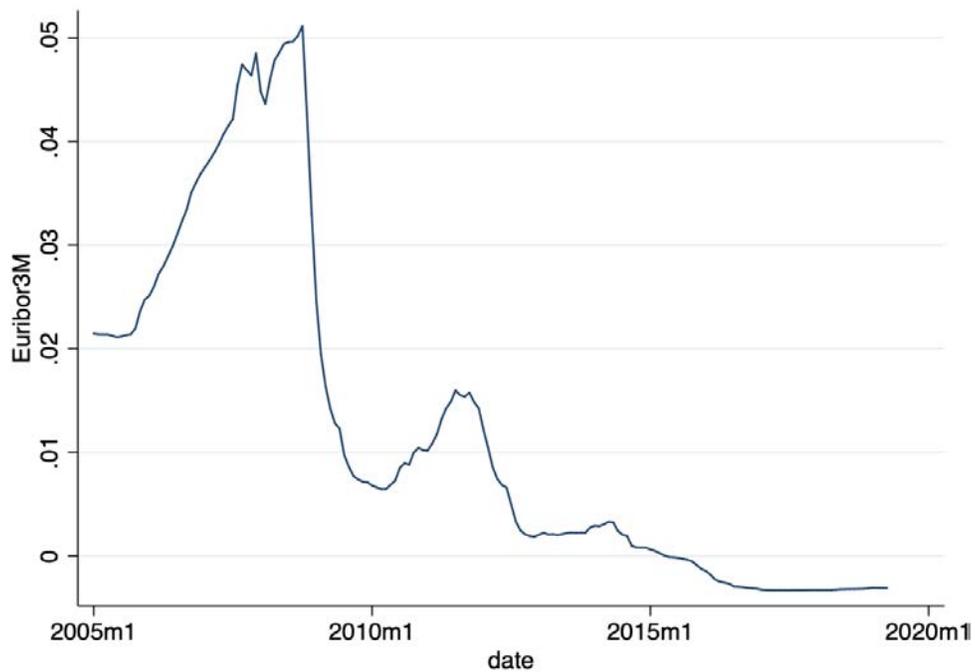
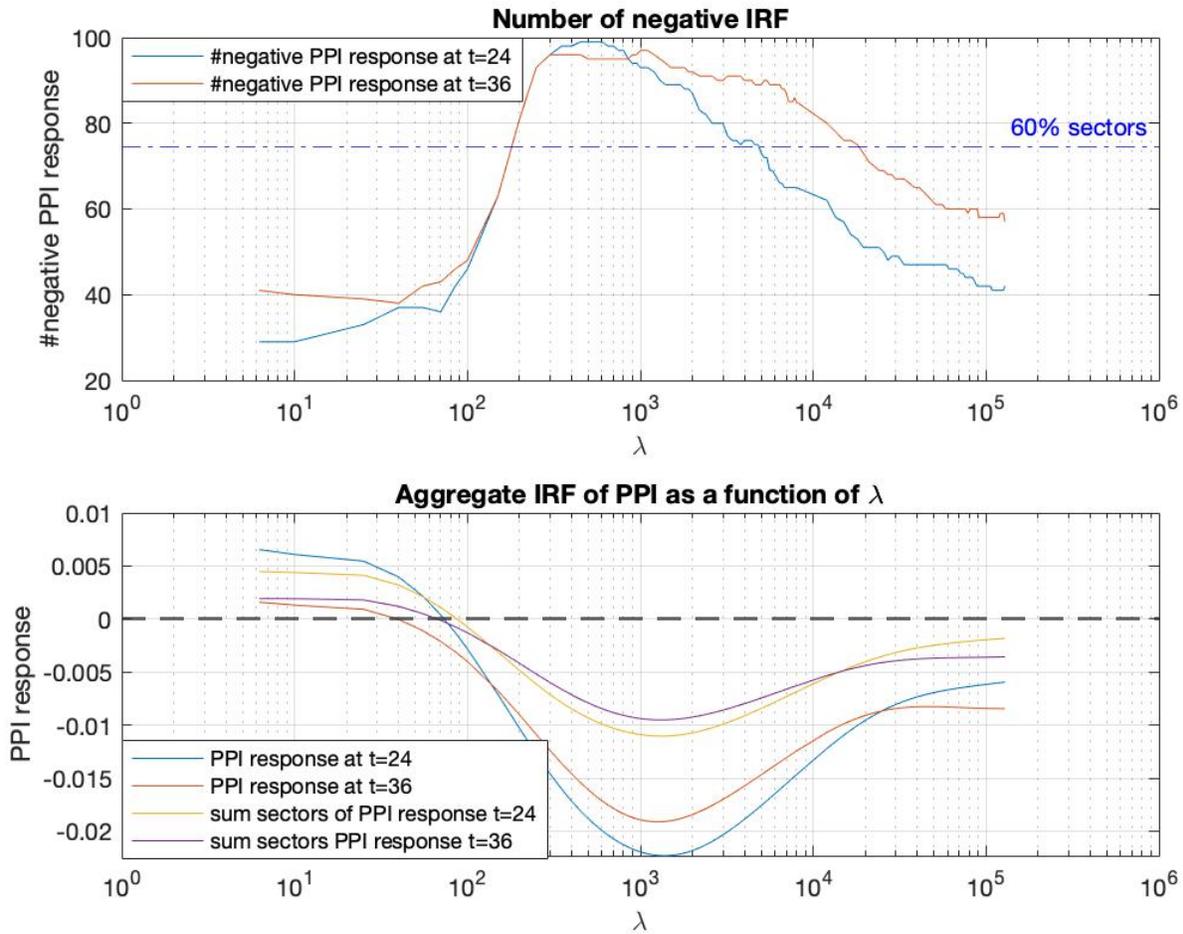
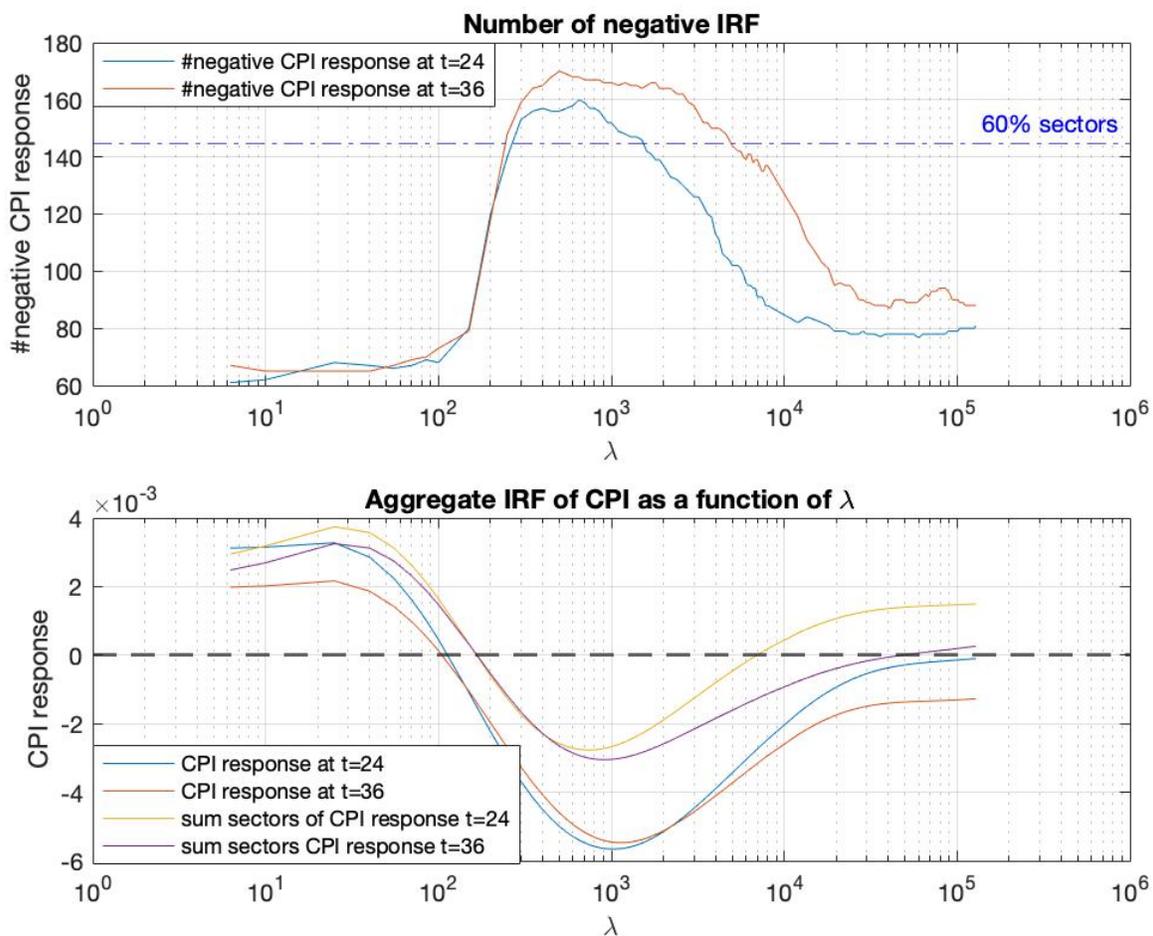


Figure B: Response of sectoral PPI as a function of λ



Top panel: number of PPI sectors with a negative IRF after two or three years to a contractionary monetary shock of 25 bp as a function of the HP filter parameter, λ^{HP} . Bottom panel: sectoral IRF of production prices to a contractionary monetary shock of 25 bp as a function of the HP filter parameter, λ^{HP} ; blue and red lines represent the IRF of the aggregate production price index after two and three years, respectively. Yellow and purple lines show the IRF of the arithmetic average of all the production price sectors after two and three years, respectively.

Figure C: Response of sectoral CPI as a function of λ

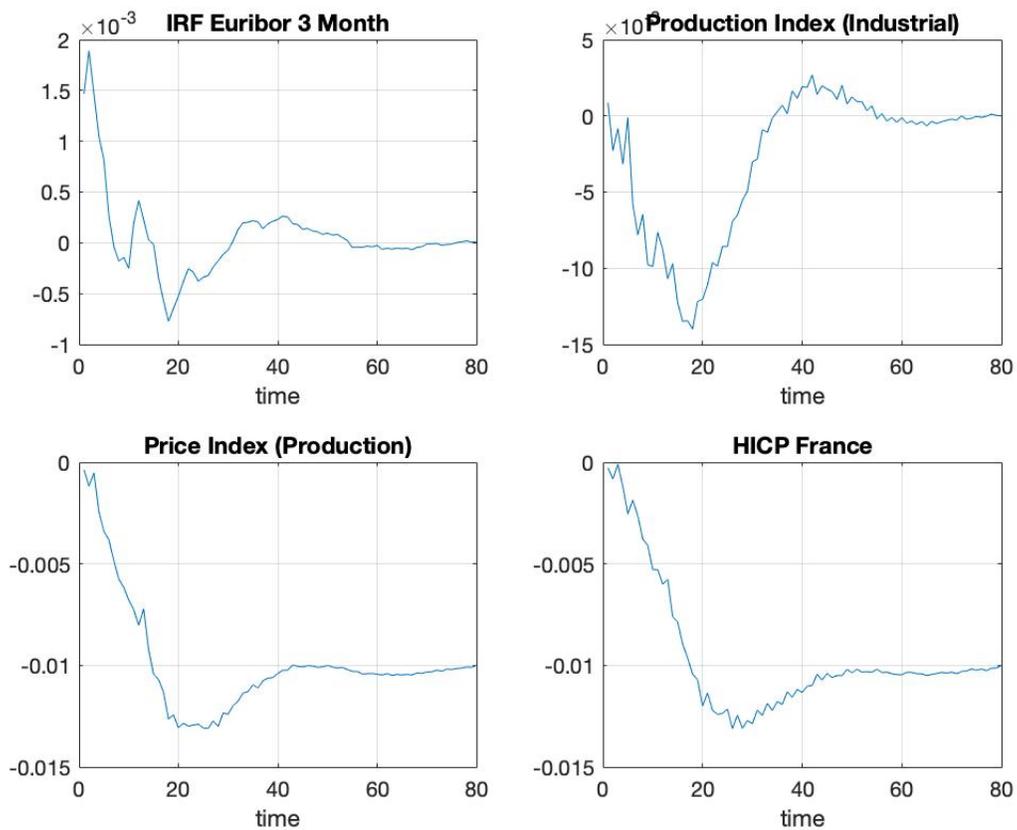


Top panel: number of CPI sectors with a negative IRF after two or three years to a contractionary monetary shock of 25 bp as a function of the HP filter parameter, λ^{HP} . Bottom panel: sectoral IRF of consumer prices to a contractionary monetary shock of 25 bp as a function of the HP filter parameter, λ^{HP} ; blue and red lines represent the IRF of the harmonized index of consumer prices after two and three years, respectively. Yellow and purple lines show the IRF of the arithmetic average of all the consumer price sectors after two and three years, respectively.

D Additional FAVAR Results

Cholesky - Long-run restriction

Figure D: Aggregate response to a contractionary monetary policy shock

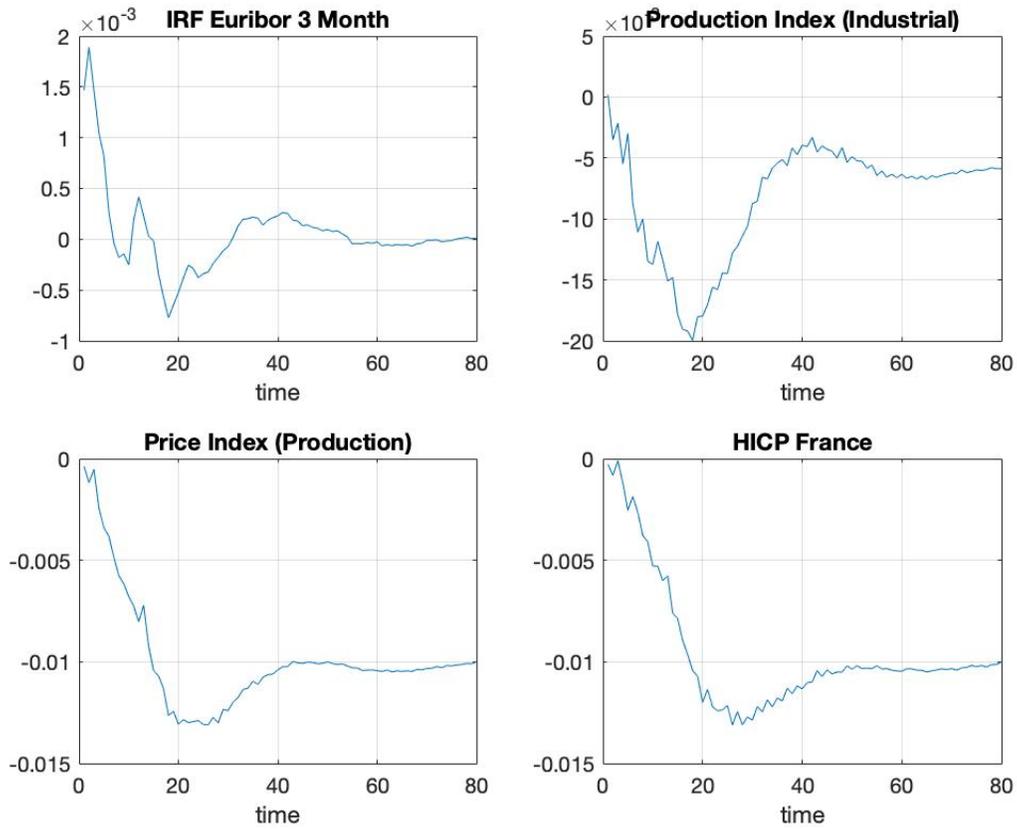


y-axis: log points in deviation from the "steady state".

Top panel: 3-month Euribor IRF. Top right panel: production index IRF. Bottom left panel: production price index IRF. Bottom right panel: IRF of the harmonized index of consumer prices

Cholesky - No Long-run restriction

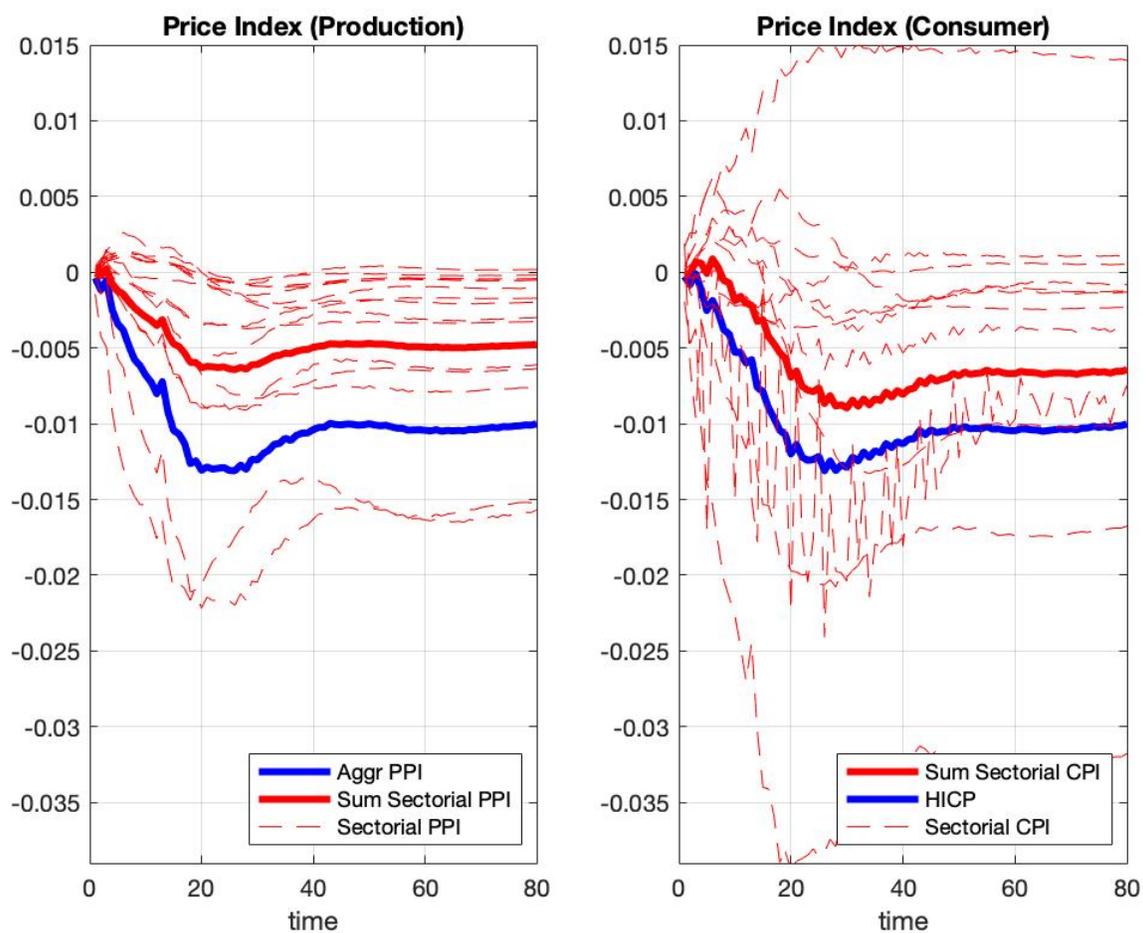
Figure E: Aggregate response to a contractionary monetary policy shock



y-axis: log points in deviation from the "steady state".

Top panel: 3-month Euribor IRF. Top right panel: production index IRF. Bottom left panel: production price index IRF. Bottom right panel: IRF of the harmonized index of consumer prices

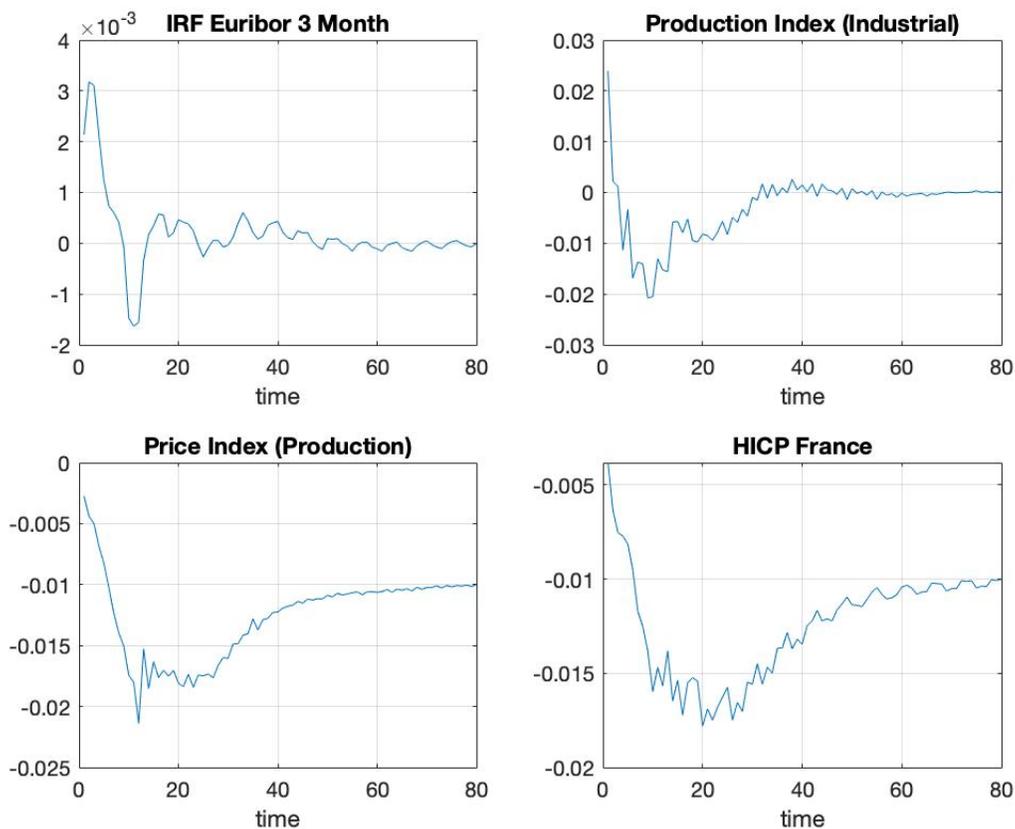
Figure F: Sectoral Responses of PPI and CPI to a Contractionary Monetary Shock



y-axis: log points in deviation from the "steady state". Left panel sectoral IRFs of PPI, right panel sectoral IRFs of CPI. In both panel: blue line IRF of aggregate time series, dashed red lines sectoral IRFs, thick red line arithmetic average of sectoral IRFs.

High-Freq. IV - Long-run restriction

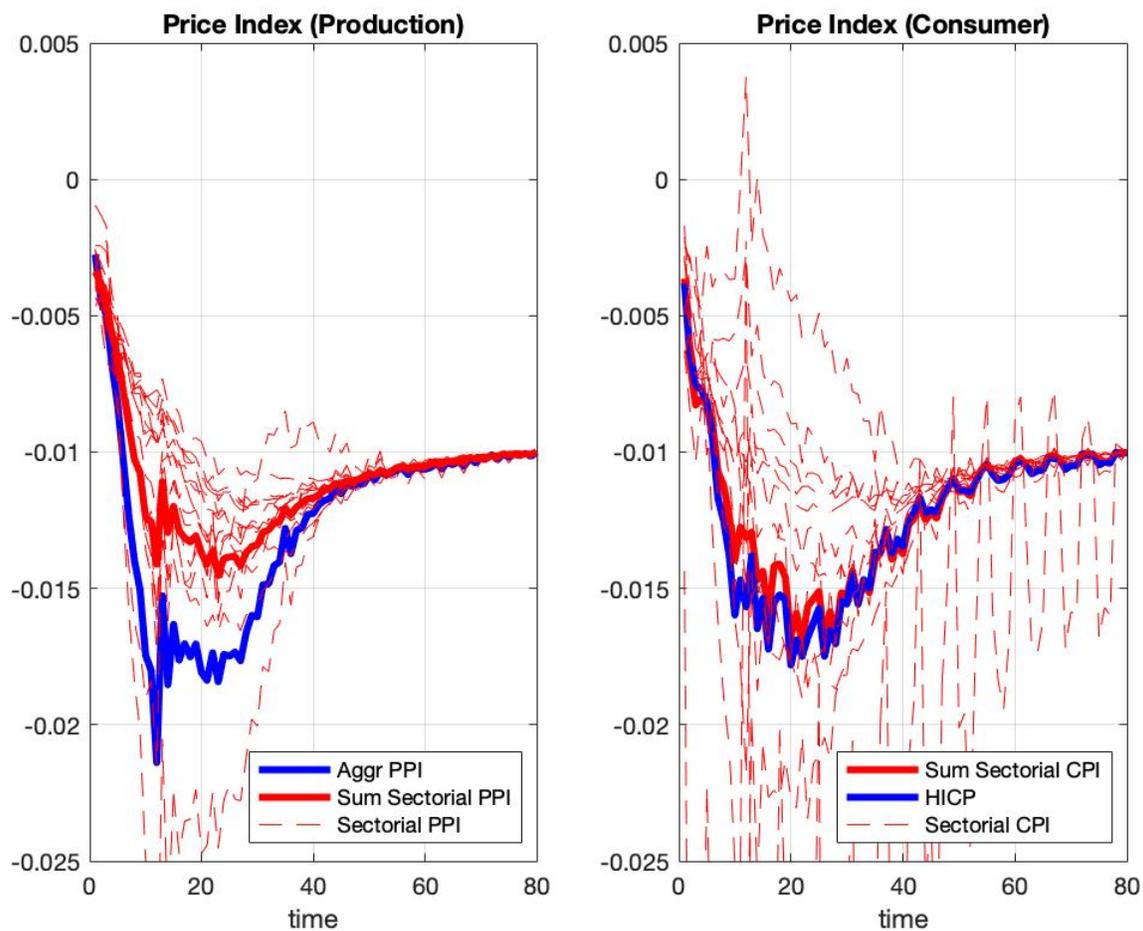
Figure G: Aggregate response to a contractionary monetary policy shock



y-axis: log points in deviation from the "steady state".

Top panel: 3-month Euribor IRF. Top right panel: production index IRF. Bottom left panel: production price index IRF. Bottom right panel: IRF of the harmonized index of consumer prices

Figure H: Sectoral Responses of PPI and CPI to a Contractionary Monetary Shock



y-axis: log points in deviation from the "steady state". Left panel sectoral IRFs of PPI, right panel sectoral IRFs of CPI. In both panel: blue line IRF of aggregate time series, dashed red lines sectoral IRFs, thick red line arithmetic average of sectoral IRFs.

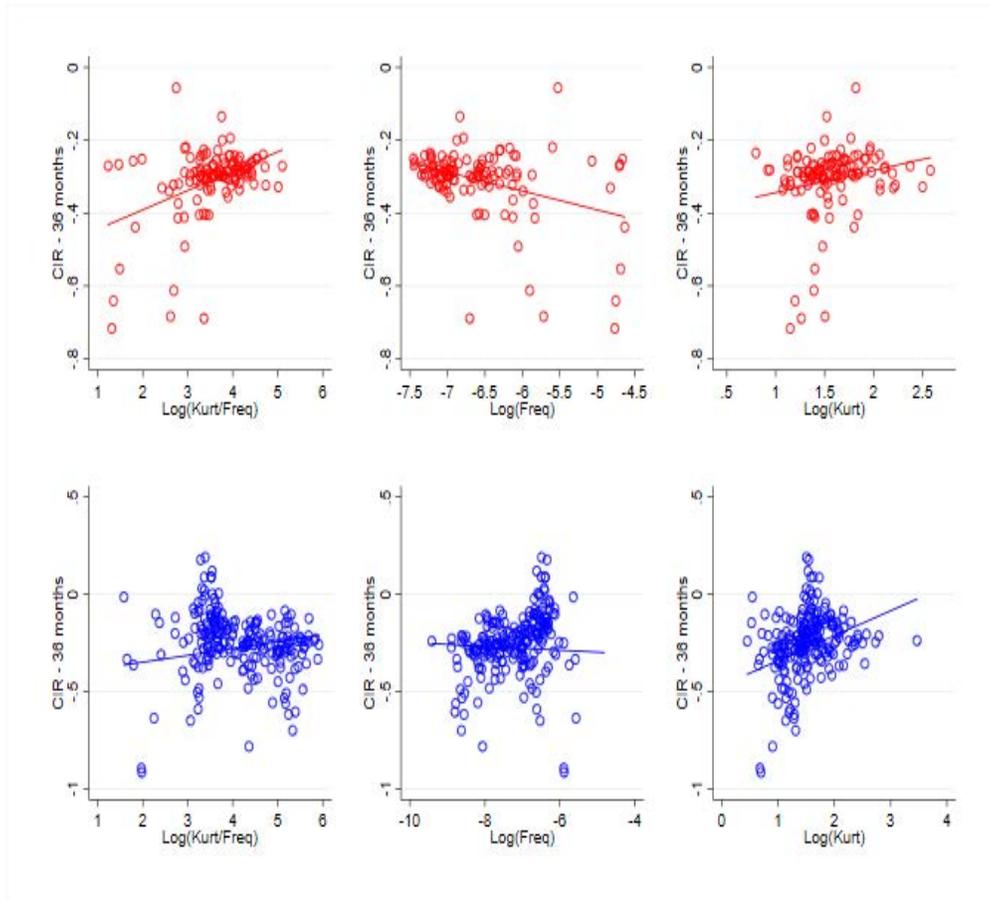
Table A: Product-specific CIR^P : Descriptive Statistics

	<i>Moments of the CIR distribution</i>								
	Mean	Std. Dev.	min	5%	25%	50%	75%	95%	max
<i>PANEL A: PRODUCER PRICES</i>									
<i>Cholesky - Long-run restriction</i>									
24 months	-0.18	0.09	-0.64	-0.39	-0.18	-0.16	-0.14	-0.09	0.12
36 months	-0.31	0.10	-0.72	-0.55	-0.32	-0.29	-0.27	-0.22	-0.06
<i>Cholesky - No long-run restriction</i>									
24 months	-0.09	0.27	-1.73	-0.71	-0.07	-0.01	0.02	0.10	0.17
36 months	-0.16	0.41	-2.77	-0.82	-0.16	-0.03	0.01	0.13	0.27
<i>High Frequency Instrument - Long-run restriction</i>									
24 months	-0.26	0.15	-1.21	-0.50	-0.27	-0.21	-0.19	-0.11	-0.01
36 months	-0.42	0.20	-1.58	-0.80	-0.44	-0.36	-0.32	-0.23	0.11
<i>PANEL B: CONSUMER PRICES</i>									
<i>Cholesky - Long-run restriction</i>									
24 months	-0.13	0.22	-1.92	-0.38	-0.20	-0.12	-0.04	0.15	0.49
36 months	-0.28	0.25	-2.44	-0.61	-0.32	-0.24	-0.16	-0.01	0.19
<i>Cholesky - No long-run restriction</i>									
24 months	-0.07	0.50	-4.76	-0.42	-0.09	0.01	0.09	0.24	0.91
36 months	-0.17	0.80	-7.83	-0.74	-0.22	-0.04	0.10	0.32	1.27
<i>High Frequency Instrument - Long-run restriction</i>									
24 months	-0.30	0.29	-2.50	-0.84	-0.32	-0.23	-0.17	-0.07	0.02
36 months	-0.48	0.38	-3.28	-1.22	-0.51	-0.39	-0.30	-0.16	-0.03

Note: this table reports descriptive statistics on the distribution of the product-specific CIR for the different specifications and at two horizons (24 and 36 months). These statistics are computed over 118 products for PPI and 223 products for CPI.

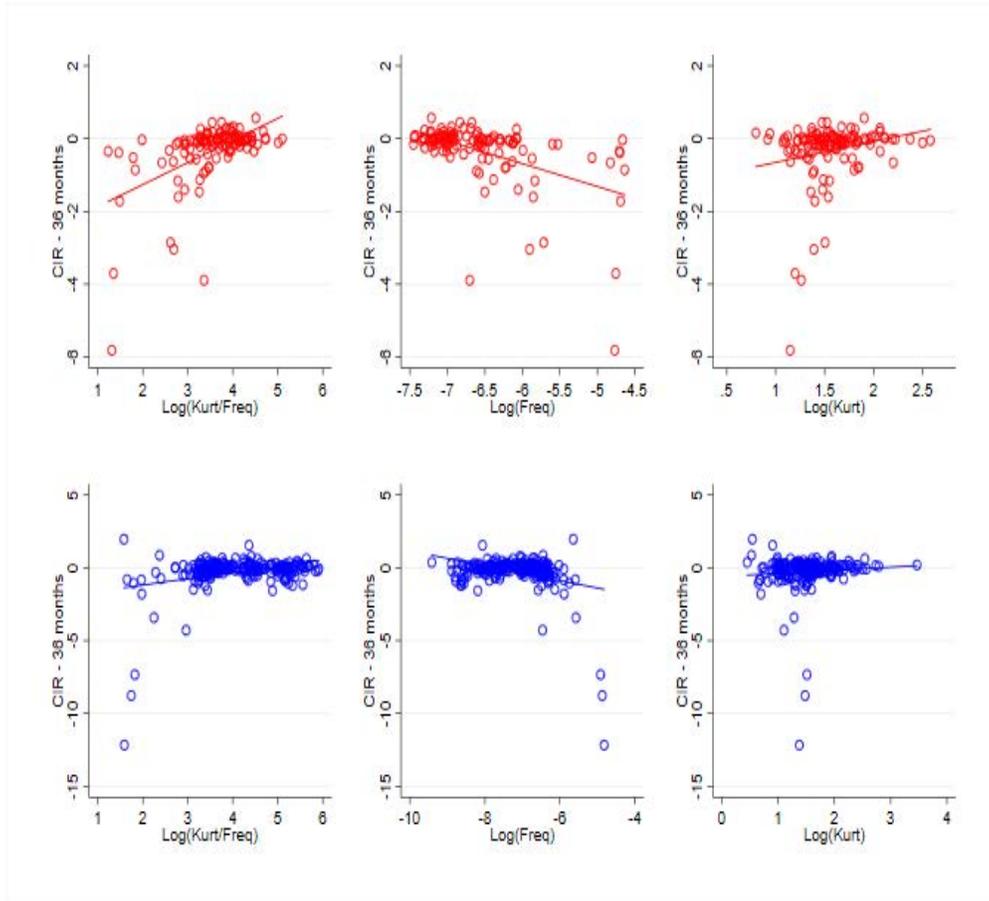
Additional Scatter Plots CIR^P - moments

Figure I: Correlation CIR^P - Log ratio $\frac{Kurt}{Freq}$ - Cholesky Long-run restriction



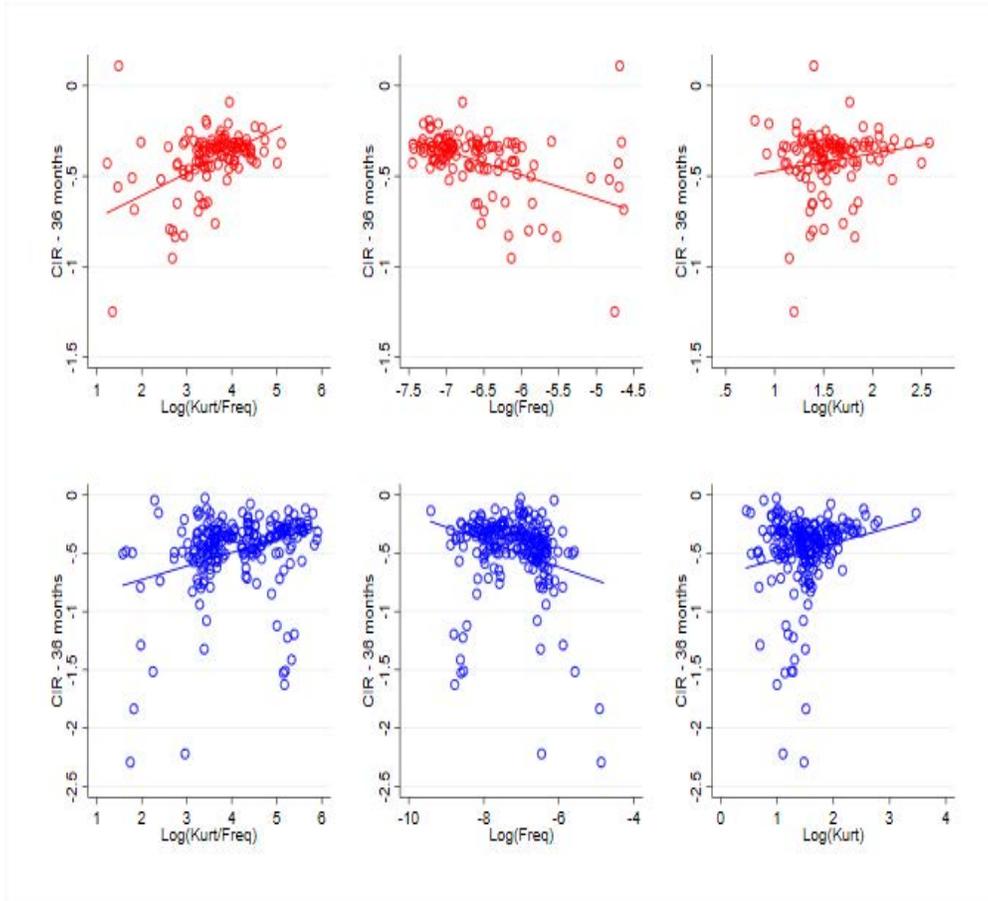
Note: the figure plots the product-specific CIR (at the horizon 36 months) obtained in the FAVAR specification using a Cholesky decomposition and imposing a long-run restriction and the log of the ratio kurtosis over frequency of price changes (left panel), the log of frequency of price changes (center panel), the log of kurtosis of price changes (right panel). The top panel reports results for PPI products whereas the bottom panel reports results for CPI products.

Figure J: Correlation CIR^P - Log ratio $\frac{Kurt}{Freq}$ - Cholesky No Long-run restriction



Note: the figure plots the product-specific CIR (at the horizon 36 months) obtained in the FAVAR specification using a Cholesky decomposition without imposing any long-run restriction and the log of the ratio kurtosis over frequency of price changes (left panel), the log of frequency of price changes (center panel), the log of kurtosis of price changes (right panel). The top panel reports results for PPI products whereas the bottom panel reports results for CPI products.

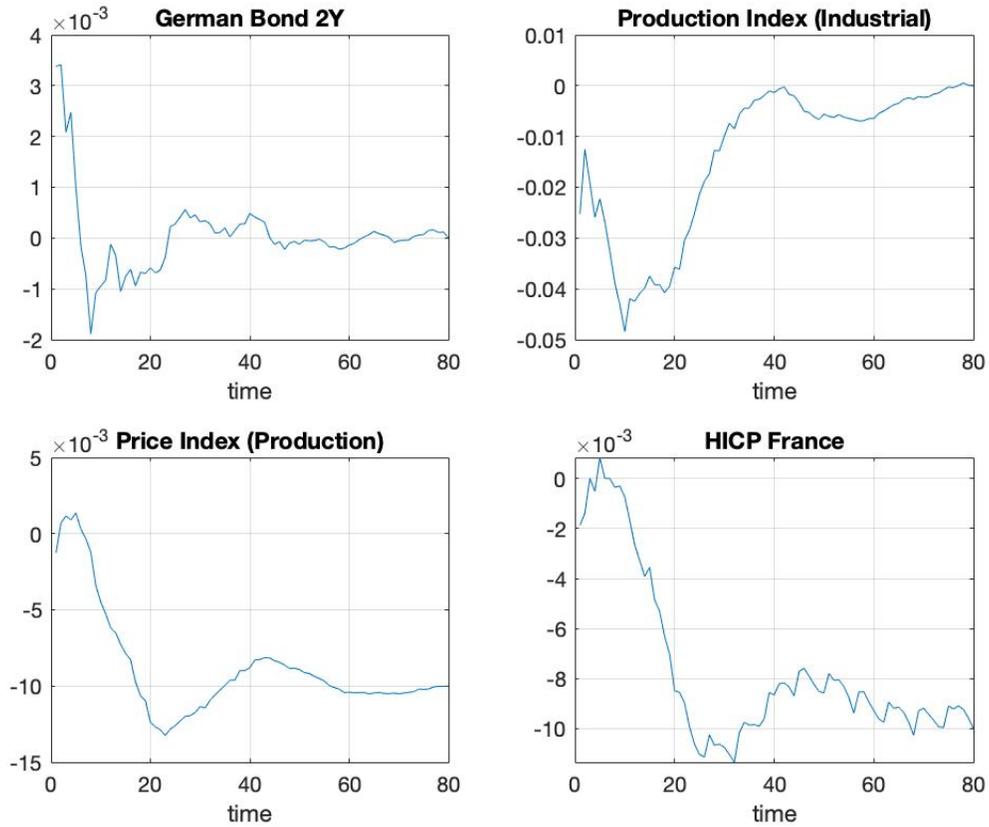
Figure K: Correlation CIR^P - Log ratio $\frac{Kurt}{Freq}$ - HFI Long-run restriction



Note: the figure plots the product-specific CIR (at the horizon 36 months) obtained in the FAVAR specification using a high-frequency instrument variable and imposing a long-run restriction and the log of the ratio kurtosis over frequency of price changes (left panel), the log of frequency of price changes (center panel), the log of kurtosis of price changes (right panel). The top panel reports results for PPI products whereas the bottom panel reports results for CPI products.

FAVAR - High-Freq. IV - German bond rate - Long-run restriction

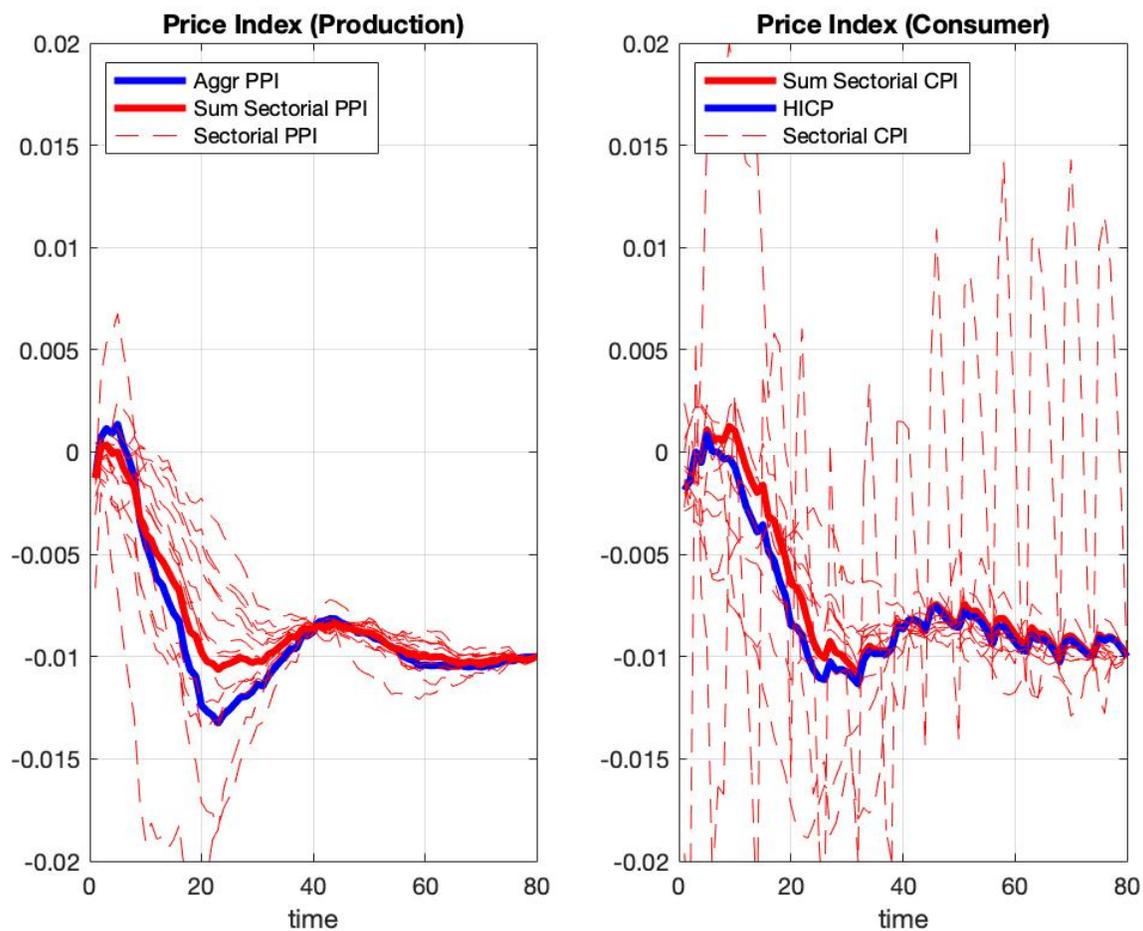
Figure L: Aggregate response to a contractionary monetary policy shock Updated (nothing should change)



y-axis: log points in deviation from the "steady state".

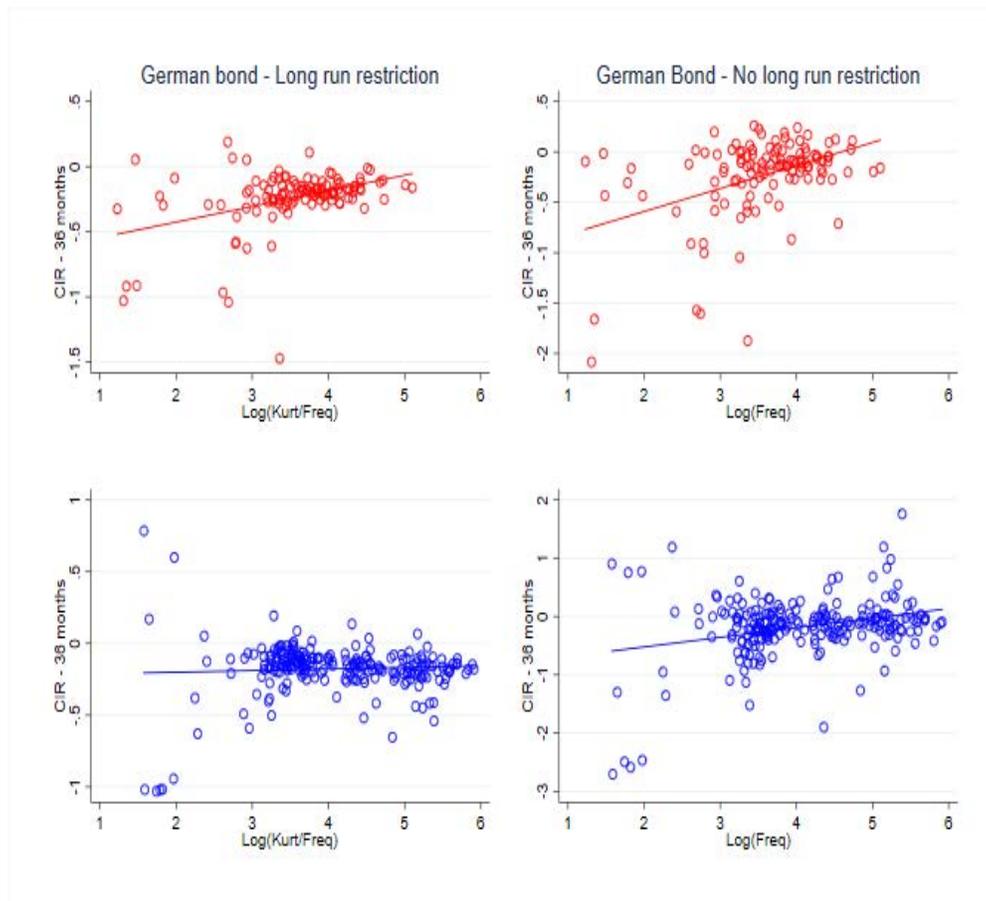
Top panel: 3-month Euribor IRF. Top right panel: production index IRF. Bottom left panel: production price index IRF. Bottom right panel: IRF of the harmonized index of consumer prices

Figure M: Sectoral Responses of PPI and CPI to a Contractionary Monetary Shock



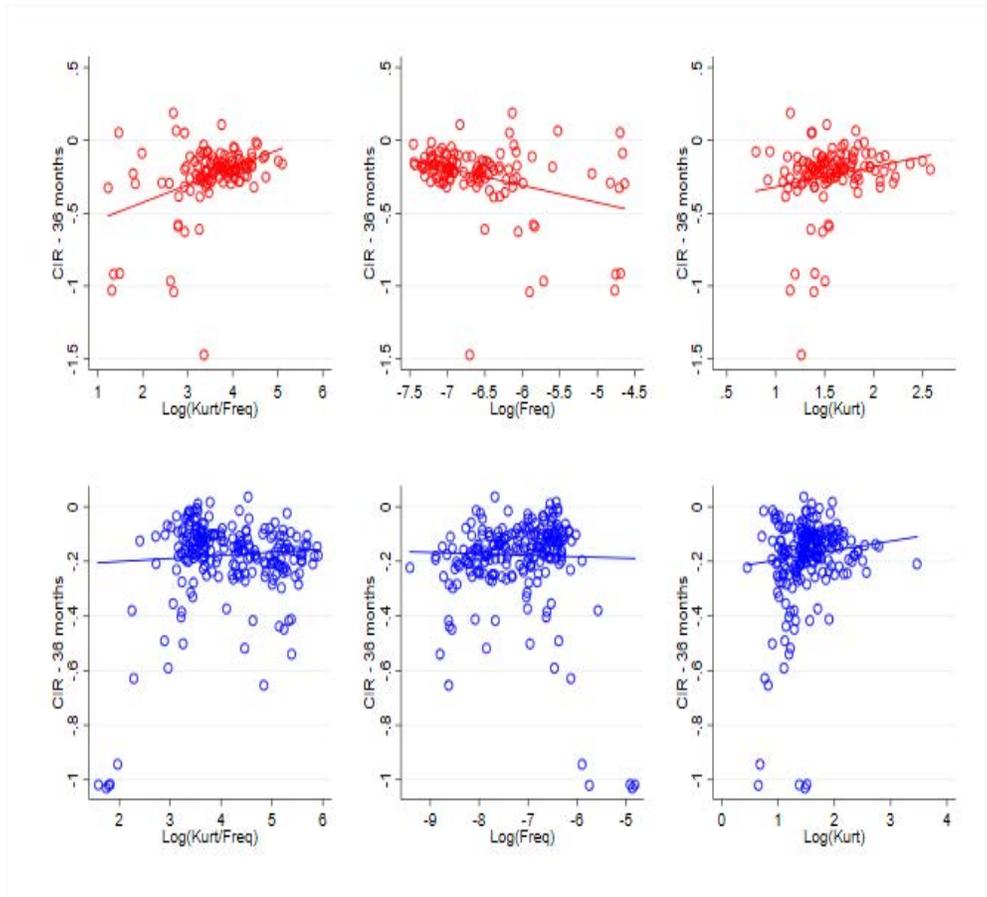
y-axis: log points in deviation from the "steady state". Left panel sectoral IRFs of PPI, right panel sectoral IRFs of CPI. In both panel: blue line IRF of aggregate time series, dashed red lines sectoral IRFs, thick red line arithmetic average of sectoral IRFs.

Figure N: Correlation CIR^P - Log ratio $\frac{Kurt}{Freq}$ - HFI 2-year German Bond Rate



Note: the figure plots for each FAVAR specification the product-specific CIR (at the horizon 36 months) and the log of the ratio kurtosis over frequency of price changes. The top panel reports result for PPI products whereas the bottom panel reports results for CPI products.

Figure O: Correlation CIR^P - Log ratio $\frac{Kurt}{Freq}$, $\log(Kurt)$ and $\log(Freq)$ - HFI 2-year German Bond Rate



Note: the figure plots the product-specific CIR (at the horizon 36 months) obtained in the FAVAR specification using a high-frequency instrument variable and imposing a long-run restriction and the log of the ratio kurtosis over frequency of price changes (left panel), the log of frequency of price changes (center panel), the log of kurtosis of price changes (right panel). The top panel reports results for PPI products whereas the bottom panel reports results for CPI products.

E Kurtosis Measurement with Unobserved Heterogeneity

The measure of Kurtosis is particularly sensitive to unobserved heterogeneity. Measured kurtosis is in particular known to suffer from an upward bias when a sample is composed of two (or more) sub-populations with different variances. To investigate the robustness of our results with respect to such a concern, we use an alternative measure of kurtosis derived along the lines of [Alvarez, Lippi, and Oskolkov \(2021\)](#). The assumption underlying this correction, is that within a given product category, there are several varieties (indexed by $i = 1, \dots, I$) that are pooled. For instance, one could have various brands of soda, in the case the brand of soda is not collected by the statistical office, or not disclosed to the researcher. At a given date t , the price change for all varieties is driven by a common factor $\Delta\tilde{p}_t$, but the variance differs across varieties, according to a scaling factor b_i .

$$\Delta p_{it} = b_i \Delta\tilde{p}_t \text{ for } i \in I \text{ and } t \in T(i)$$

where $T(i)$ is the set of adjustment instances for variety i . Under the assumption that $\Delta\tilde{p}_t$ is serially uncorrelated, and some other general assumptions, [Alvarez, Lippi, and Oskolkov \(2021\)](#) show that the following property then holds:

$$Kurt(\Delta\tilde{p}_t) = Kurt(\Delta p_{it}) \frac{E[(\Delta p_{it}^2)]^2}{E[(\Delta p_{it}^2)(\Delta p_{is}^2)]} \text{ for } t \neq s$$

or equivalently

$$Kurt(\Delta\tilde{p}_t) = \frac{Kurt(\Delta p_{it})}{1 + corr(\Delta p_{it}^2, \Delta p_{is}^2) CV(\Delta p_{it}^2) CV(\Delta p_{is}^2)} \text{ for } t \neq s$$

where $CV(\cdot)$ denotes the coefficient of variation and $corr(\cdot, \cdot)$ the correlation coefficient.

We use these equations to compute a measure of kurtosis robust to unobserved heterogeneity. In practice, we want to use information from several possible lags (the s 's as different from t), rather than picking up a single particular lag s .

To compute the covariance terms in the expression above we as use an estimator of $E = E[(\Delta p_{it}^2)(\Delta p_{is}^2)]$ the following expression:

$$E = (1/\#Terms) \sum_{t,s \in T(i), t \neq s} (\Delta p_{it})^2 (\Delta p_{is})^2 \quad (33)$$

In practice, we consider the first K lags of squared price changes. So, the numerator of the formula (33) above is computed as:

$$S = 2 * \left[\sum_{t=2}^T (\Delta p_t)^2 (\Delta p_{t-1})^2 + \sum_{t=3}^T (\Delta p_t)^2 (\Delta p_{t-2})^2 + \dots + \sum_{t=K+1}^T (\Delta p_t)^2 (\Delta p_{t-K})^2 \right] \quad (34)$$

Denotig by NN the number of terms in equation (34), then $\#Terms = 2 * NN$, where:

$$NN = (T - 1) + (T - 2) + \dots + (T - k) = T(T - 1)/2 - (T - K - 1) * (T - K)/2 \quad (35)$$

So when $K = T - 1$, $\#Terms = 2 * T(T - 1)/2 = T(T - 1)$ Then we recover

$$E = \frac{S}{T(T - 1)}$$

F Measurement Error

This appendix assesses the impact of (one form of) measurement errors on the micro moments of price adjustment and their ratio $Kurt/Freq$.

Assume measurement errors are of the following type: for a given store, measurement errors materialize at some points by extra spurious price changes, and these spurious price changes are small. Such patterns of error is plausible (as discussed in [Alvarez, Le Bihan, and Lippi \(2016\)](#)), both for CPI data because small coding error can stay undetected by the error checking procedures of the statistical institute, and for scanner data as the price is typically computed as the ratio of value purchased to quantity sold (and the numerator can vary reflecting e.g. coupons). These spurious price changes will increase both the measured Kurtosis, as well as the measured Frequency of price changes - with the size of the bias being a function of the fraction of spurious price changes. However, as is formally shown below, such measurement errors will leave ratio Kurtosis/Frequency unchanged. As a result, not only theory indicates that the ratio Kurtosis/Frequency is the relevant covariate, but it is also the case that this ratio should be more robust to measurement errors than each of the moments taken separately.

Formally, let $N_{\Delta p}$ be the number of “true” price changes per period (i.e. the frequency of price changes). Assume Δp , the price changes, have mean zero, variance $Var(\Delta p) = \sigma_{\Delta p}^2$ and Kurtosis $Kurt(\Delta p) = m_{4,\Delta p}/\sigma_{\Delta p}^4$, where $m_{4,\Delta p}$ is the fourth moment of variable Δp . Let N_e denote the number of spurious price changes per unit of time. Assume that spurious price changes, denoted e , have mean zero and variance $Var(e) = \sigma_e^2$, and kurtosis $Kurt(e) = m_{4,e}/\sigma_e^4$. Assume spurious and true price changes to be statistically independent. Then the observed (measured) frequency of price changes will be $\tilde{N} = N_{\Delta p} + N_e$. The distribution of the observed price changes, denoted $\tilde{\Delta p}$'s, will have mean zero and its Kurtosis will be

$$Kurt(\tilde{\Delta p}) = \frac{\theta Kurt(\Delta p)\sigma_{\Delta p}^4 + (1 - \theta)Kurt(e)\sigma_e^4}{\left(\theta\sigma_{\Delta p}^2 + (1 - \theta)\sigma_e^2\right)^2}$$

with $\theta \equiv \frac{N_{\Delta p}}{\tilde{N}}$ the fraction of “true” price changes. We consider the case of arbitrarily small measurement errors. From the above it results that $\lim_{\sigma_e^2 \rightarrow 0} Kurt(\tilde{\Delta p}) = \frac{Kurt(\Delta p)}{\theta}$. Then we have $\lim_{\sigma_e^2 \rightarrow 0} \frac{Kurt(\tilde{\Delta p})}{\tilde{N}} = \frac{Kurt(\Delta p)}{N_{\Delta p}}$. Thus, the ratio Kurtosis over Frequency is unaffected by these presence of small measurement error.

G OLS regressions - Additional results and robustness

Table B: Testing Model's Predictions

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
	<i>PANEL A: PRODUCER PRICES</i>					
<i>Constrained model</i>						
P-val $\beta = 1/6$	0.003	0.053	0.125	0.025	0.681	0.351
P-val $\alpha = -T$	0.111	0.702	0.042	0.763	0.006	0.000
Ratio α/β	-307.6	-360.9	-69.58	-72.82	-178.5	-216.0
<i>Unconstrained model</i>						
P-val $\beta_f = -\beta_k$	0.566	0.457	0.648	0.643	0.819	0.857
P-val $\beta_f = -\frac{\bar{K}}{6\bar{F}}$	0.130	0.325	0.111	0.047	0.577	0.460
P-val $\beta_k = \frac{\bar{K}}{6\bar{F}}$	0.679	0.915	0.098	0.044	0.477	0.389
P-val $\gamma = -T + \frac{\bar{K}}{6\bar{F}}$	0.743	0.687	0.161	0.170	0.100	0.007
<i>PANEL B: CONSUMER PRICES</i>						
<i>Constrained model</i>						
P-val $\beta = 1/6$	0.000	0.000	0.030	0.802	0.000	0.003
P-val $\alpha = -T$	0.000	0.009	0.001	0.630	0.003	0.000
Ratio α/β	685.5	11,160	-184.4	-205.0	-696.0	-759.4
<i>Unconstrained model</i>						
P-val $\beta_f = -\beta_k$	0.877	0.492	0.001	0.000	0.049	0.039
P-val $\beta_f = -\frac{\bar{K}}{6\bar{F}}$	0.281	0.860	0.003	0.000	0.032	0.006
P-val $\beta_k = \frac{\bar{K}}{6\bar{F}}$	0.124	0.377	0.208	0.591	0.710	0.664
P-val $\gamma = -T + \frac{\bar{K}}{6\bar{F}}$	0.352	0.433	0.000	0.000	0.340	0.202

Note: we report p-values of Wald tests performed on the parameters of our baseline OLS regressions presented in [Table 2](#) and [Table 2](#). These tests correspond to model's predictions presented in [equation \(10\)](#) and [equation \(11\)](#). We perform four different tests: (i) in the constrained version of the model we test whether β (parameter associated with the ratio $Kurt/Freq$ is equal to $-\delta/6$ (where δ is the MP shock here normalised to 1%); (ii) we test whether the constant of the constrained model (α) is equal to $-T$ and in the unconstrained model, whether γ is equal to $-T + \frac{\bar{K}}{6\bar{F}}$ and (iii) in the unconstrained model, we test whether the parameter associated with frequency (β_f) is equal to minus the parameter associated with kurtosis ($-\beta_k$); (iv) in the unconstrained version, we also perform tests on the parameter associated with frequency and kurtosis, they are predicted to be equal to $\frac{\bar{K}}{6\bar{F}}$ where \bar{K} and \bar{F} are sample averages of kurtosis and frequency. We also report the ratio of the estimated coefficients in OLS regressions.

Table C: Regression Results: Alternative Specifications - Log

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: Producer prices - Constrained model</i>						
Log(Ratio)	3.948** (1.649)	5.301*** (1.723)	18.01*** (5.377)	28.84*** (8.383)	10.07*** (3.398)	12.39*** (4.541)
Constant	-31.63*** (6.440)	-49.67*** (6.681)	-72.29*** (20.67)	-117.7*** (32.13)	-61.54*** (12.97)	-85.53*** (17.35)
R^2	0.103	0.175	0.255	0.282	0.259	0.223
<i>PANEL B: Producer prices - Unconstrained model</i>						
Log(Freq)	-3.748** (1.833)	-5.155*** (1.873)	-18.23*** (5.589)	-29.39*** (8.676)	-10.56*** (3.582)	-13.11*** (4.845)
Log(Kurt)	4.834** (2.291)	5.948** (2.414)	17.05** (7.033)	26.39** (10.72)	7.895* (4.061)	9.189* (5.274)
Constant	-15.37*** (4.982)	-26.66*** (4.836)	12.74 (12.09)	20.39 (17.77)	-10.47 (7.124)	-21.55** (9.852)
R^2	0.104	0.176	0.255	0.282	0.262	0.226
Observations	118	118	118	118	118	118
<i>PANEL C: Consumer prices - Constrained model</i>						
Log(Ratio)	1.354 (2.289)	3.509 (2.877)	15.97** (7.563)	27.98** (12.05)	8.643** (3.580)	11.63** (4.539)
Constant	-18.72* (10.50)	-41.94*** (13.07)	-72.36** (33.78)	-131.7** (53.77)	-65.36*** (15.72)	-95.81*** (19.89)
R^2	0.003	0.017	0.092	0.111	0.082	0.086
<i>PANEL D: Consumer prices - Unconstrained model</i>						
Log(Freq)	1.460 (2.715)	-0.824 (3.442)	-17.54* (9.270)	-32.62** (14.71)	-8.005* (4.559)	-11.23* (5.792)
Log(Kurt)	11.07*** (3.020)	12.77*** (3.338)	10.57** (4.800)	11.94 (7.711)	10.84*** (3.289)	12.99*** (4.426)
Constant	-32.83*** (5.901)	-45.19*** (7.093)	12.52 (16.42)	30.70 (26.61)	-30.17*** (10.52)	-45.12*** (13.88)
R^2	0.051	0.050	0.095	0.121	0.084	0.087
Observations	223	223	223	223	223	223

Note: this table reports results of OLS regressions (equation (10)) where the endogenous variable is the product-specific $CIR_T^{P_j}$ (expressed in %) and the RHS variables include the log of the product-specific ratio $Kurt/freq$ in the constrained model and the log of product specific frequency and the log of product-specific kurtosis. Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table D: Regression Results - Placebo Unconstrained Specification

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: PRODUCER PRICES</i>						
<i>Freq/\bar{F}</i>	-2.865* (1.454)	-3.337** (1.491)	-11.50** (4.922)	-17.79** (7.743)	-5.488* (3.258)	-6.347 (4.406)
<i>Kurt/\bar{K}</i>	3.026 (3.048)	4.066 (2.796)	10.13* (5.749)	15.95* (8.154)	5.823* (3.177)	7.153 (4.673)
Mean	-0.254 (0.792)	-0.186 (0.851)	-2.353 (2.242)	-3.699 (3.486)	-0.855 (1.285)	-0.927 (1.699)
Skewness	1.798 (3.359)	0.793 (2.989)	-2.645 (7.065)	-6.643 (9.668)	-4.708* (2.589)	-7.159* (3.855)
Standard dev.	-0.916 (1.297)	-0.625 (1.306)	-2.533 (4.158)	-3.331 (6.425)	0.245 (2.460)	0.837 (3.229)
Constant	-13.33* (7.379)	-28.68*** (7.539)	4.485 (22.87)	1.130 (35.69)	-27.87* (14.08)	-47.18** (18.59)
Observations	118	118	118	118	118	118
R^2	0.118	0.164	0.246	0.264	0.228	0.195
<i>PANEL B: CONSUMER PRICES</i>						
<i>Freq/\bar{F}</i>	-6.170* (3.170)	-10.12*** (3.640)	-37.39*** (9.143)	-62.37*** (13.96)	-17.81*** (3.969)	-23.07*** (4.877)
<i>Kurt/\bar{K}</i>	-4.732* (2.733)	-2.640 (3.216)	6.454 (6.959)	16.08 (11.74)	7.629 (4.688)	11.55* (6.434)
Mean	0.0898 (0.783)	-0.366 (0.822)	-3.328** (1.451)	-6.085*** (2.258)	-1.505** (0.718)	-2.092** (0.956)
Skewness	5.111 (4.335)	7.162 (4.393)	9.659* (5.724)	14.77 (9.326)	8.489** (3.586)	10.98** (4.974)
Standard dev.	-2.972*** (1.078)	-2.767** (1.248)	0.0440 (2.524)	2.281 (3.908)	-0.00409 (1.211)	0.554 (1.571)
Constant	21.50* (11.82)	8.500 (13.60)	30.21 (29.17)	23.08 (45.66)	-15.71 (14.13)	-35.47* (18.48)
Observations	223	223	223	223	223	223
R^2	0.108	0.165	0.509	0.572	0.383	0.380

Note: this table reports results of OLS regressions ([equation \(11\)](#)) where the endogenous variable is the product-specific $CIR_T^{P_j}$ (expressed in %) and the RHS variables include the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$, but also three other moments of the product-specific price change distribution: the average price change $Mean$, the skewness of price changes $Skewness$, and the standard deviation of price changes $StandardDev.$. Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table E: Regression Results: Outliers - Constrained - Producer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: CIR</i>						
Kurt/Freq	0.0451* (0.0230)	0.0713** (0.0275)	0.200*** (0.0632)	0.331*** (0.0916)	0.131*** (0.0341)	0.172*** (0.0452)
Constant	-19.18*** (1.490)	-33.53*** (1.680)	-14.56*** (3.956)	-26.00*** (5.648)	-30.61*** (2.039)	-48.11*** (2.711)
R^2	0.042	0.087	0.111	0.135	0.144	0.138
<i>PANEL B: Ratio</i>						
Kurt/Freq	0.0762* (0.0392)	0.108*** (0.0352)	0.320*** (0.0824)	0.517*** (0.116)	0.181*** (0.0444)	0.226*** (0.0668)
Constant	-20.39*** (2.388)	-34.85*** (2.146)	-19.84*** (5.050)	-33.92*** (7.031)	-32.18*** (2.580)	-49.54*** (3.861)
R^2	0.039	0.086	0.121	0.149	0.146	0.116
<i>PANEL C: Kurtosis</i>						
Kurt/Freq	0.0933** (0.0404)	0.134*** (0.0419)	0.439*** (0.123)	0.713*** (0.189)	0.261*** (0.0714)	0.328*** (0.0942)
Constant	-21.63*** (2.452)	-36.67*** (2.479)	-27.37*** (7.198)	-46.28*** (10.94)	-37.23*** (4.073)	-55.95*** (5.406)
R^2	0.057	0.112	0.151	0.172	0.176	0.158
<i>PANEL D: Frequency</i>						
Kurt/Freq	0.0491 (0.0308)	0.0866** (0.0369)	0.321*** (0.113)	0.542*** (0.179)	0.215*** (0.0679)	0.281*** (0.0865)
Constant	-19.54*** (2.049)	-34.53*** (2.281)	-22.12*** (6.795)	-38.86*** (10.55)	-35.46*** (3.839)	-54.31*** (4.886)
R^2	0.023	0.064	0.107	0.128	0.153	0.155
Observations	112	112	112	112	112	112

Note: this table reports results of OLS regressions (equation (10)) for PPI products relating the product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$. For each of the 4 regressions, we remove products with "extreme" values of CIR (Panel A); ratio $Kurt/Freq$ (Panel B), kurtosis (Panel C), frequency of price changes (Panel D). "Extreme values" are defined as products below the 2.5th percentile or above the 97.5th percentile of the distribution of each statistic (10 products for CPI and 6 for PPI). Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table F: Regression Results: Outliers - Constrained - Consumer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: CIR</i>						
Kurt/Freq	-0.0320*** (0.0108)	-0.0302** (0.0117)	0.00321 (0.0139)	0.0366* (0.0221)	0.0212 (0.0160)	0.0384* (0.0220)
Constant	-9.067*** (1.654)	-22.80*** (1.616)	-1.922 (1.878)	-12.32*** (2.903)	-29.25*** (1.631)	-48.51*** (2.361)
R^2	0.036	0.029	0.000	0.012	0.009	0.015
<i>PANEL B: Ratio</i>						
Kurt/Freq	-0.0353* (0.0184)	-0.0244 (0.0214)	0.0317 (0.0312)	0.0876* (0.0481)	0.0247 (0.0248)	0.0419 (0.0330)
Constant	-9.372*** (2.811)	-24.45*** (3.135)	-6.259 (4.480)	-19.24*** (6.824)	-30.66*** (2.668)	-50.04*** (3.514)
R^2	0.015	0.006	0.005	0.017	0.006	0.010
<i>PANEL C: Kurtosis</i>						
Kurt/Freq	-0.0170 (0.0187)	-0.00270 (0.0226)	0.0786* (0.0475)	0.157** (0.0757)	0.0429 (0.0276)	0.0630* (0.0359)
Constant	-11.68*** (2.988)	-27.53*** (3.505)	-14.62** (7.360)	-32.45*** (11.67)	-34.42*** (3.681)	-54.70*** (4.734)
R^2	0.003	0.000	0.014	0.022	0.013	0.016
<i>PANEL D: Frequency</i>						
Kurt/Freq	-0.0326** (0.0153)	-0.0230 (0.0172)	0.00332 (0.0193)	0.0371 (0.0285)	0.0270 (0.0172)	0.0450* (0.0230)
Constant	-8.829*** (2.553)	-23.48*** (2.774)	-1.463 (2.925)	-11.16*** (4.180)	-29.32*** (2.046)	-48.33*** (2.747)
R^2	0.016	0.007	0.000	0.007	0.012	0.017
Observations	213	213	213	213	213	213

Note: this table reports results of OLS regressions ([equation \(10\)](#)) for CPI products relating the product-specific $CIR_T^{P_i}$ (expressed in %) to the ratio $Kurt/freq$. For each of the 4 regressions, we remove products with "extreme" values of CIR (Panel A), ratio $Kurt/freq$ (Panel B), kurtosis (Panel C), frequency of price changes (Panel D). "Extreme values" are defined as products below the 2.5th percentile or above the 97.5th percentile of the distribution of each statistic (10 products for CPI and 6 for PPI). Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table G: Regression Results: Outliers - Unconstrained - Producer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
	<i>PANEL A: CIR</i>					
<i>Freq/\bar{F}</i>	-1.226 (0.868)	-2.415** (1.169)	-5.927*** (2.229)	-9.038*** (2.717)	-3.461*** (1.136)	-4.898*** (1.560)
<i>Kurt/\bar{K}</i>	1.615 (1.168)	2.547* (1.431)	6.249** (2.680)	9.878** (3.938)	4.014** (1.946)	4.302* (2.419)
Constant	-17.64*** (1.253)	-30.59*** (1.480)	-6.404* (3.261)	-12.74** (5.072)	-25.66*** (2.502)	-40.28*** (3.124)
R^2	0.047	0.156	0.140	0.146	0.129	0.134
<i>PANEL B: Ratio</i>						
<i>Freq/\bar{F}</i>	-1.984 (1.547)	-2.240* (1.164)	-7.069*** (2.564)	-10.75*** (3.161)	-2.874 (1.772)	-3.162 (2.990)
<i>Kurt/\bar{K}</i>	4.770* (2.431)	5.526** (2.116)	12.71** (5.138)	18.81*** (7.113)	4.710* (2.702)	4.962 (4.070)
Constant	-20.04*** (2.445)	-33.65*** (2.274)	-12.35** (6.041)	-20.77** (8.624)	-26.57*** (2.887)	-42.02*** (3.999)
R^2	0.062	0.088	0.118	0.127	0.071	0.043
<i>PANEL C: Kurtosis</i>						
<i>Freq/\bar{F}</i>	-2.514** (1.264)	-3.146** (1.294)	-11.23*** (4.168)	-17.74*** (6.550)	-5.924** (2.761)	-7.117* (3.754)
<i>Kurt/\bar{K}</i>	6.798** (2.865)	8.448*** (2.899)	23.34*** (8.291)	36.12*** (12.52)	11.90** (4.575)	14.06** (5.991)
Constant	-21.80*** (3.010)	-36.09*** (2.882)	-20.32** (8.236)	-33.63*** (12.06)	-31.89*** (3.969)	-48.71*** (5.237)
R^2	0.125	0.186	0.259	0.277	0.235	0.192
<i>PANEL D: Frequency</i>						
<i>Freq/\bar{F}</i>	-1.929 (1.512)	-3.319* (1.881)	-14.06** (6.139)	-23.62** (9.670)	-9.641*** (3.484)	-12.52*** (4.261)
<i>Kurt/\bar{K}</i>	3.094* (1.744)	4.156** (1.841)	11.69** (5.030)	18.55** (7.579)	6.122** (2.860)	7.471** (3.703)
Constant	-18.73*** (2.184)	-31.88*** (2.140)	-7.005 (6.097)	-12.26 (9.007)	-23.44*** (3.227)	-38.15*** (4.255)
R^2	0.058	0.132	0.251	0.295	0.362	0.358
Observations	112	112	112	112	112	112

Note: This table reports results of OLS regressions ([equation \(11\)](#)) relating the product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. For each of the 4 regressions, we remove products with "extreme" values of CIR (Panel A); ratio $Kurt/Freq$ (Panel B), kurtosis (Panel C), frequency of price changes (Panel D). "Extreme values" are defined as products below the 2.5th percentile or above the 97.5th percentile of the distribution of each statistic. Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table H: Regression Results: Outliers - Unconstrained - Consumer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
	<i>PANEL A: CIR</i>					
<i>Freq/\bar{F}</i>	4.572** (2.110)	4.832** (2.450)	-3.749 (2.512)	-10.71*** (3.456)	-5.233* (2.703)	-8.578** (3.670)
<i>Kurto/\bar{K}</i>	3.575* (1.872)	4.757** (2.031)	5.826*** (2.022)	8.653*** (2.610)	6.617*** (1.700)	8.495*** (2.170)
Constant	-19.77*** (2.289)	-34.72*** (2.855)	-4.218 (3.138)	-8.168* (4.730)	-29.19*** (3.755)	-45.70*** (5.075)
R^2	0.067	0.082	0.063	0.100	0.091	0.097
<i>PANEL B: Ratio</i>						
<i>Freq/\bar{F}</i>	0.207 (4.469)	-2.562 (4.697)	-20.52** (9.468)	-37.25*** (13.28)	-10.28*** (3.494)	-14.15*** (4.012)
<i>Kurto/\bar{K}</i>	8.072*** (2.779)	10.01*** (3.040)	9.452*** (3.312)	12.32** (5.105)	10.67*** (2.700)	13.37*** (3.586)
Constant	-20.49*** (4.440)	-33.98*** (4.756)	6.168 (8.615)	10.64 (12.53)	-29.47*** (4.981)	-46.43*** (6.418)
R^2	0.031	0.047	0.249	0.327	0.153	0.152
<i>PANEL C: Kurtosis</i>						
<i>Freq/\bar{F}</i>	-5.606** (2.758)	-9.607*** (3.162)	-38.46*** (8.108)	-64.60*** (12.46)	-17.86*** (3.666)	-23.21*** (4.536)
<i>Kurto/\bar{K}</i>	12.83*** (3.347)	15.14*** (3.757)	14.20** (5.638)	17.43* (9.197)	13.23*** (3.855)	16.00*** (5.120)
Constant	-19.94*** (4.172)	-32.85*** (4.782)	16.23* (8.592)	27.69** (13.67)	-25.88*** (5.801)	-41.83*** (7.637)
R^2	0.108	0.181	0.554	0.607	0.388	0.376
<i>PANEL D: Frequency</i>						
<i>Freq/\bar{F}</i>	6.723** (2.839)	5.051 (3.094)	-1.024 (4.345)	-7.915 (6.509)	-4.369* (2.587)	-7.549** (3.423)
<i>Kurto/\bar{K}</i>	4.587* (2.507)	5.874** (2.731)	5.296** (2.313)	7.061** (3.045)	6.992*** (1.937)	8.912*** (2.428)
Constant	-22.50*** (2.988)	-36.09*** (3.240)	-5.618 (3.679)	-7.837 (5.820)	-30.05*** (3.737)	-46.52*** (5.023)
R^2	0.055	0.042	0.019	0.035	0.067	0.073
Observations	213	213	213	213	213	213

Note: This table reports results of OLS regressions ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurto/\bar{K}$. For each of the 4 regressions, we remove products with "extreme" values of CIR (Panel A); ratio $Kurto/Freq$ (Panel B), kurtosis (Panel C), frequency of price changes (Panel D). "Extreme values" are defined as products below the 2.5th percentile or above the 97.5th percentile of the distribution of each statistic. Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table I: Regression Results: Kurtosis Measurement - Heterogeneity

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: Producer Prices - Constrained model</i>						
Kurt/Freq	0.0837** (0.0345)	0.115*** (0.0384)	0.348*** (0.112)	0.557*** (0.175)	0.190*** (0.0662)	0.234*** (0.0862)
Constant	-20.49*** (1.807)	-34.79*** (1.917)	-20.30*** (5.699)	-34.47*** (8.772)	-32.33*** (3.278)	-49.60*** (4.293)
R^2	0.045	0.080	0.093	0.103	0.090	0.078
<i>PANEL B: Producer Prices - Unconstrained model</i>						
Freq/ \bar{F}	-2.408* (1.280)	-3.066** (1.327)	-11.16** (4.341)	-17.71** (6.821)	-6.048** (2.822)	-7.328* (3.803)
Kurt/ \bar{K}	4.458** (2.027)	4.376** (1.746)	6.163*** (2.263)	7.230** (3.135)	-0.641 (2.236)	-2.262 (3.571)
Constant	-19.71*** (2.272)	-32.22*** (1.985)	-3.540 (3.317)	-5.158 (4.669)	-19.21*** (2.556)	-32.08*** (3.995)
R^2	0.146	0.185	0.220	0.231	0.192	0.162
Observations	118	118	118	118	118	118
<i>PANEL C: Consumer Prices - Constrained model</i>						
Kurt/Freq	-0.0152 (0.0148)	-0.00274 (0.0175)	0.0675* (0.0400)	0.136** (0.0666)	0.0446* (0.0228)	0.0646** (0.0303)
Constant	-12.21*** (2.264)	-27.40*** (2.634)	-11.26** (5.598)	-25.81*** (8.930)	-32.81*** (2.792)	-52.32*** (3.616)
R^2	0.003	0.000	0.010	0.015	0.013	0.016
<i>PANEL D: Consumer Prices - Unconstrained model</i>						
Freq/ \bar{F}	-4.959* (2.820)	-8.601** (3.344)	-34.89*** (8.873)	-58.69*** (13.73)	-16.44*** (3.911)	-21.39*** (4.832)
Kurt/ \bar{K}	3.637* (2.084)	4.424** (2.225)	3.393 (2.672)	4.015 (4.153)	5.771*** (1.827)	7.303*** (2.337)
Constant	-11.85*** (3.538)	-23.40*** (4.079)	24.53*** (8.622)	37.49*** (13.35)	-19.31*** (4.801)	-34.12*** (6.128)
R^2	0.061	0.126	0.476	0.528	0.342	0.335
Observations	223	223	223	223	223	223

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. The measure of kurtosis takes into account for possible product heterogeneity following the methodology presented in [Appendix E](#) and using $S = 5$. Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table J: Regression Results: Kurtosis Measurement - Producer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
	<i>PANEL A: Outlier threshold - small price changes - Constrained model</i>					
Kurt/Freq	0.0554* (0.0283)	0.0806** (0.0323)	0.271*** (0.0933)	0.441*** (0.146)	0.158*** (0.0561)	0.199*** (0.0728)
Constant	-19.78*** (1.801)	-33.99*** (1.878)	-18.89*** (5.404)	-32.51*** (8.261)	-31.95*** (3.075)	-49.28*** (4.048)
R^2	0.031	0.062	0.088	0.101	0.098	0.088
<i>PANEL B: Outlier threshold - small price changes - Unconstrained model</i>						
Freq/ \bar{F}	-2.417* (1.298)	-3.037** (1.327)	-10.90** (4.266)	-17.23** (6.675)	-5.819** (2.767)	-7.011* (3.743)
Kurt/ \bar{K}	2.490 (1.760)	3.403* (1.954)	10.32** (5.048)	16.51** (7.618)	5.349* (2.852)	6.565* (3.670)
Constant	-17.73*** (2.321)	-31.28*** (2.331)	-7.961 (6.351)	-14.92 (9.352)	-25.43*** (3.310)	-41.23*** (4.402)
R^2	0.096	0.147	0.227	0.245	0.209	0.173
<i>PANEL C: Outlier threshold - large price changes - Constrained model</i>						
Kurt/Freq	0.0273 (0.0193)	0.0415* (0.0245)	0.148** (0.0701)	0.244** (0.112)	0.0881* (0.0453)	0.112* (0.0587)
Constant	-19.49*** (1.906)	-33.69*** (2.116)	-18.46*** (6.073)	-31.96*** (9.448)	-31.81*** (3.652)	-49.18*** (4.786)
R^2	0.023	0.050	0.081	0.094	0.093	0.085
<i>PANEL D: Outlier threshold - large price changes - Unconstrained model</i>						
Freq/ \bar{F}	-2.521* (1.306)	-3.181** (1.345)	-11.34*** (4.308)	-17.94*** (6.752)	-6.053** (2.806)	-7.298* (3.794)
Kurt/ \bar{K}	1.062 (1.267)	1.653 (1.513)	6.948 (4.257)	11.50* (6.687)	4.095 (2.812)	5.210 (3.647)
Constant	-16.20*** (1.623)	-29.38*** (1.675)	-4.141 (4.776)	-9.199 (7.203)	-23.94*** (2.774)	-39.59*** (3.682)
R^2	0.089	0.137	0.223	0.243	0.210	0.176
Observations	118	118	118	118	118	118

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) for PPI products relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. In Panels A and B, we have modified the thresholds defining very small price changes for the calculation of kurtosis: we have removed all price changes below 0.5% in absolute values (instead 0.1% in our baseline). In Panels C and D, we have modified thresholds defining very large price changes for the calculation of kurtosis (25% for instead of 15% in the baseline). Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table K: Regression Results: Kurtosis Measurement - Consumer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
	<i>PANEL A: Outlier threshold - small price changes - Constrained model</i>					
Kurt/Freq	-0.0144 (0.0166)	0.00129 (0.0199)	0.0811* (0.0423)	0.161** (0.0673)	0.0542** (0.0240)	0.0781** (0.0312)
Constant	-11.86*** (2.758)	-27.59*** (3.226)	-13.77** (6.820)	-30.79*** (10.80)	-34.59*** (3.369)	-54.89*** (4.337)
R^2	0.003	0.000	0.016	0.025	0.022	0.027
<i>PANEL B: Outlier threshold - small price changes - Unconstrained model</i>						
Freq/ \bar{F}	-4.956* (2.845)	-8.610** (3.372)	-35.32*** (8.879)	-59.42*** (13.73)	-16.46*** (3.953)	-21.42*** (4.886)
Kurt/ \bar{K}	4.087* (2.381)	5.031* (2.557)	1.675 (2.554)	0.994 (4.235)	5.693*** (1.950)	7.245*** (2.490)
Constant	-12.22*** (3.881)	-23.90*** (4.449)	26.86*** (8.872)	41.50*** (13.78)	-19.16*** (5.045)	-34.00*** (6.433)
R^2	0.065	0.132	0.482	0.534	0.345	0.337
<i>PANEL C: Outlier threshold - large price changes - Constrained model</i>						
Kurt/Freq	-0.00911 (0.0128)	0.00346 (0.0154)	0.0639* (0.0329)	0.126** (0.0525)	0.0441** (0.0183)	0.0632*** (0.0238)
Constant	-12.17*** (2.744)	-27.96*** (3.206)	-14.00** (6.789)	-31.01*** (10.77)	-34.83*** (3.340)	-55.16*** (4.299)
R^2	0.002	0.000	0.017	0.025	0.024	0.029
<i>PANEL D: Outlier threshold - large price changes - Unconstrained model</i>						
Freq/ \bar{F}	-4.922* (2.783)	-8.546** (3.301)	-34.81*** (8.831)	-58.57*** (13.68)	-16.39*** (3.867)	-21.34*** (4.780)
Kurt/ \bar{K}	4.246* (2.540)	5.404* (2.854)	5.016* (2.713)	6.692* (3.727)	6.289*** (2.198)	7.988*** (2.696)
Constant	-12.50*** (3.682)	-24.44*** (4.268)	22.83*** (8.103)	34.69*** (12.43)	-19.86*** (4.708)	-34.85*** (5.972)
R^2	0.069	0.137	0.479	0.530	0.351	0.343
Observations	223	223	223	223	223	223

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) for CPI products relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. In Panels A and B, we have modified the thresholds defining very small price changes for the calculation of kurtosis: we have removed all price changes below 0.5% in absolute values (instead 0.1% in our baseline). In Panels C and D, we have modified thresholds defining very large price changes for the calculation of kurtosis (35% for instead of 25% in the baseline). Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table L: Reverse OLS Regression Results

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: PRODUCER PRICES</i>						
$CIR_T^{P_j}$	0.612*** (0.220)	0.845*** (0.186)	0.358*** (0.0672)	0.252*** (0.0477)	0.681*** (0.130)	0.488*** (0.103)
Constant	54.38*** (5.117)	69.69*** (7.050)	46.64*** (2.665)	47.52*** (2.698)	61.22*** (4.765)	63.92*** (5.781)
Observations	118	118	118	118	118	118
R^2	0.041	0.082	0.117	0.135	0.131	0.118
<i>PANEL B: CONSUMER PRICES</i>						
$CIR_T^{P_j}$	-0.227 (0.272)	-0.0248 (0.208)	0.192*** (0.0489)	0.153*** (0.0335)	0.394** (0.157)	0.332*** (0.125)
Constant	87.22*** (5.890)	89.53*** (7.526)	91.55*** (5.477)	92.85*** (5.544)	102.0*** (7.622)	106.2*** (8.634)
Observations	223	223	223	223	223	223
R^2	0.004	0.000	0.014	0.023	0.019	0.024

Note: this table reports results of OLS regressions where the endogenous variable is the product-specific ratio $\frac{Kurt}{Freq}$ and the RHS variable is the $CIR_T^{P_j}$ (expressed in %). Each observation corresponds to a disaggregate CPI or PPI product. For CPI, the level of disaggregation is 5 digit-level of the ECOICOP classification (ie. '01.1.1.1') whereas for PPI, the product level is the 4-digit level of the NACE rev2 classification of sectors. PPI covers the manufacturing sectors whereas CPI covers about 60% of the whole French CPI (main products excluded are rents, cars, utilities like electricity). Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table M: Regression Results: Role of sales - Consumer Prices

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
	<i>PANEL A: Excluding food, clothing/footwear, furnishings - Constrained model</i>					
Kurt/Freq	0.0467** (0.0229)	0.0692** (0.0290)	0.159** (0.0688)	0.257** (0.111)	0.0993*** (0.0342)	0.129*** (0.0437)
Constant	-23.04*** (4.330)	-39.33*** (5.454)	-28.69** (12.67)	-48.78** (20.35)	-38.04*** (6.312)	-57.39*** (8.042)
R^2	0.034	0.048	0.053	0.054	0.078	0.079
<i>PANEL B: Excluding food, clothing/footwear, furnishings - Unconstrained model</i>						
$Freq/\bar{F}$	-8.726*** (1.218)	-12.88*** (1.514)	-41.62*** (5.832)	-67.93*** (9.581)	-19.92*** (2.792)	-25.41*** (3.633)
$Kurt/\bar{K}$	5.318** (2.683)	6.718** (3.062)	7.192** (3.581)	9.855* (5.421)	6.500*** (2.173)	8.122*** (2.724)
Constant	-14.39*** (3.494)	-25.38*** (3.996)	23.61*** (6.327)	38.21*** (10.11)	-13.44*** (3.537)	-25.64*** (4.559)
R^2	0.260	0.361	0.725	0.745	0.644	0.636
Observations	134	134	134	134	134	134
<i>PANEL C: % of sales prices below the median - Constrained model</i>						
Kurt/Freq	-0.000929 (0.0276)	0.0316 (0.0342)	0.157** (0.0783)	0.297** (0.124)	0.133*** (0.0374)	0.185*** (0.0473)
Constant	-12.32** (5.440)	-30.73*** (6.705)	-28.98* (14.92)	-58.63** (23.51)	-45.07*** (7.003)	-69.35*** (8.771)
R^2	0.000	0.009	0.046	0.064	0.130	0.153
<i>PANEL D: % of sales prices below the median - Unconstrained model</i>						
$Freq/\bar{F}$	-8.872*** (2.363)	-13.71*** (2.662)	-46.39*** (8.340)	-76.50*** (13.05)	-23.27*** (3.285)	-29.96*** (4.144)
$Kurt/\bar{K}$	0.410 (2.749)	2.409 (3.355)	1.933 (5.007)	4.958 (7.617)	7.194** (2.802)	10.26*** (3.579)
Constant	-3.958 (4.625)	-15.85*** (5.405)	33.29*** (9.066)	46.49*** (13.62)	-13.92*** (4.492)	-28.66*** (5.665)
R^2	0.166	0.273	0.645	0.693	0.676	0.690
Observations	111	111	111	111	111	111

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) for CPI products relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. In Panels A and B, we have removed goods of three broad sectors where sales concentrate (COICOP01.1 Food, COICOP03 Clothing/Footwear, and COICOP05 Furnishing goods). In Panels C and D, we have removed products for which the share of sales and promotions represent more than 11% of all price changes (this threshold corresponds to the median of this ratio over all CPI products). Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table N: Regression Results: Alternative Specifications - Sector Fixed Effects

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: Producer prices - Constrained model</i>						
Kurt/Freq	0.0366 (0.0301)	0.0565* (0.0321)	0.187** (0.0746)	0.308*** (0.113)	0.119** (0.0483)	0.153** (0.0662)
Constant	-14.93*** (2.701)	-27.83*** (3.153)	-10.95 (6.680)	-21.18** (9.927)	-27.53*** (3.258)	-44.18*** (3.918)
R^2	0.371	0.440	0.521	0.549	0.527	0.476
<i>PANEL B: Producer prices- Unconstrained model</i>						
$Freq/\bar{F}$	-1.705 (1.310)	-1.951 (1.274)	-5.645 (3.587)	-8.545 (5.466)	-2.621 (2.336)	-2.957 (3.375)
$Kurt/\bar{K}$	2.562 (1.964)	2.722 (1.984)	10.22** (4.954)	15.54** (7.317)	3.630 (2.902)	3.823 (3.963)
Constant	-14.62*** (2.986)	-26.72*** (3.177)	-8.912 (6.631)	-17.16* (9.341)	-24.30*** (2.908)	-39.57*** (3.588)
R^2	0.396	0.462	0.544	0.567	0.525	0.467
Observations	118	118	118	118	118	118
<i>PANEL C: Consumer prices - Constrained model</i>						
Kurt/Freq	0.0246 (0.0187)	0.0422* (0.0224)	0.0869* (0.0505)	0.147* (0.0814)	0.0821*** (0.0259)	0.110*** (0.0337)
Constant	1.697 (2.922)	-14.68*** (2.760)	-0.568 (3.821)	-18.47*** (5.835)	-34.83*** (2.039)	-58.09*** (3.063)
R^2	0.530	0.544	0.334	0.338	0.486	0.491
<i>PANEL D: Consumer prices - Unconstrained model</i>						
$Freq/\bar{F}$	-11.03*** (1.623)	-14.88*** (1.883)	-44.40*** (6.174)	-70.72*** (9.915)	-19.96*** (2.697)	-24.87*** (3.495)
$Kurt/\bar{K}$	3.499 (2.458)	4.357* (2.456)	-2.829 (3.174)	-6.232 (5.412)	4.020** (1.988)	5.231* (2.688)
Constant	16.46*** (4.419)	5.828 (4.571)	73.59*** (10.25)	101.4*** (16.40)	-4.886 (4.972)	-20.75*** (6.628)
R^2	0.678	0.743	0.765	0.766	0.747	0.723
Observations	223	223	223	223	223	223

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. Regressions include sectoral fixed effects at the 2-digit level for both CPI and PPI products (38 sectors for CPI and 24 sectors for PPI). Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table O: Regression Results: 2-year German Bond - High-Frequency IV

Identification Long-run Restriction	High-Frequency IV Yes		High-Frequency IV No	
	24 months	36 months	24 months	36 months
	<i>PANEL A: Producer Prices - Constrained model</i>			
Kurt/Freq	0.186*** (0.0669)	0.244*** (0.0775)	0.293*** (0.0959)	0.463*** (0.154)
Constant	-20.34*** (4.393)	-34.78*** (4.993)	-25.38*** (5.932)	-45.14*** (9.271)
R^2	0.069	0.091	0.092	0.095
<i>PANEL B: Producer Prices - Unconstrained model</i>				
$Freq/\bar{F}$	-5.148* (2.627)	-6.623** (2.973)	-9.108** (3.684)	-14.76** (5.895)
$Kurt/\bar{K}$	8.553** (3.931)	10.98** (4.451)	7.716 (4.982)	9.263 (7.542)
Constant	-15.64*** (5.071)	-28.50*** (5.651)	-11.24* (6.250)	-19.45** (8.931)
R^2	0.104	0.131	0.144	0.149
Observations	118	118	118	118
<i>PANEL C: Consumer Prices - Constrained model</i>				
Kurt/Freq	-0.0265 (0.0171)	-0.00828 (0.0158)	0.0422* (0.0222)	0.133*** (0.0442)
Constant	-3.369 (2.802)	-16.95*** (2.516)	-9.440*** (3.347)	-29.42*** (6.229)
R^2	0.010	0.001	0.016	0.041
<i>PANEL D: Consumer Prices - Unconstrained model</i>				
$Freq/\bar{F}$	-3.898 (2.999)	-5.798* (3.015)	-15.37*** (2.927)	-29.36*** (4.435)
$Kurt/\bar{K}$	-1.131 (1.960)	1.167 (1.774)	-6.664*** (2.377)	-10.20** (4.764)
Constant	-0.732 (3.180)	-13.06*** (3.008)	16.40*** (4.000)	22.13*** (7.265)
R^2	0.032	0.091	0.324	0.304
Observations	223	223	223	223

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. CIR are here calculated using the 2-year German bond rate as a policy rate and the model is identified using a high frequency instrument method. Robust standard errors are reported in parentheses. *** $p < 0.01$, ** $p < 0.05$, * $p < 0.1$

Table P: Regression Results: FAVAR PPI only - Euribor vs German 2-year Bond

Identification	Euribor				High-Frequency IV			
	No		Yes		No		Yes	
Long-run Restriction	24 months	36 months	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: Constrained model</i>								
Kurt/Freq	0.348*** (0.117)	0.527*** (0.173)	0.0734** (0.0300)	0.0858*** (0.0279)	0.432*** (0.137)	0.571*** (0.185)	0.240*** (0.0766)	0.251*** (0.0752)
Constant	-29.50*** (7.065)	-47.13*** (10.30)	-21.41*** (1.976)	-34.12*** (1.761)	-38.37*** (8.034)	-56.29*** (10.60)	-30.89*** (4.777)	-43.82*** (4.614)
R ²	0.104	0.112	0.052	0.087	0.115	0.112	0.102	0.118
<i>PANEL B: Unconstrained model</i>								
Freq/ \bar{F}	-12.30*** (4.472)	-18.31*** (6.561)	-2.349** (1.132)	-2.306** (0.962)	-13.41*** (4.803)	-17.43*** (6.153)	-7.233** (3.016)	-7.120** (2.873)
Kurt/ \bar{K}	12.89** (6.110)	19.36** (8.872)	2.986* (1.794)	3.428** (1.588)	12.16* (6.747)	14.04 (8.807)	10.17** (4.267)	10.72** (4.112)
Constant	-14.94** (6.872)	-25.21** (9.685)	-18.85*** (2.379)	-31.51*** (2.065)	-18.28** (7.808)	-28.01*** (9.966)	-23.36*** (5.117)	-36.48*** (4.866)
R ²	0.218	0.227	0.095	0.117	0.181	0.167	0.169	0.178
Observations	118	118	118	118	118	118	118	118

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) relating product-specific $CIR_T^{P_i}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_i}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. CIR are here calculated from a FAVAR model estimated using only PPI product-level series (non-seasonally adjusted) using both the Euribor (with no instrument) and the 2-year German bond rate as a policy rate, in this latter model, the model is identified using a high frequency instrument method. Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table Q: Regression Results: FAVAR PPI only (including seasonal adjustment) - Euribor vs German 2-year Bond

Identification	Euribor				High-Frequency IV			
	No		Yes		No		Yes	
Long-run Restriction	24 months	36 months	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: Constrained model</i>								
Kurt/Freq	0.334*** (0.101)	0.526*** (0.155)	-0.00175 (0.0118)	0.00144 (0.00911)	0.402*** (0.122)	0.595*** (0.176)	-0.0379** (0.0150)	-0.0356*** (0.0129)
Constant	-24.45*** (6.149)	-39.20*** (9.375)	-19.19*** (0.837)	-30.97*** (0.646)	-29.46*** (7.485)	-43.98*** (10.74)	-20.93*** (0.931)	-31.77*** (0.793)
R ²	0.131	0.141	0.000	0.000	0.120	0.127	0.063	0.076
<i>PANEL B: Unconstrained model</i>								
Freq/ \bar{F}	-10.96*** (4.125)	-17.01*** (6.339)	-0.119 (0.564)	-0.0517 (0.417)	-12.80*** (4.757)	-18.61*** (6.788)	0.621 (0.626)	0.597 (0.530)
Kurt/ \bar{K}	14.43*** (5.376)	22.74*** (8.157)	-0.152 (0.765)	-0.0568 (0.591)	15.29** (6.355)	22.49** (9.050)	-1.493* (0.881)	-1.532** (0.741)
Constant	-13.36** (5.859)	-21.98** (8.704)	-18.99*** (1.055)	-30.80*** (0.816)	-14.42* (7.432)	-21.96** (10.47)	-21.70*** (1.058)	-32.39*** (0.887)
R ²	0.252	0.264	0.001	0.000	0.210	0.216	0.042	0.055
Observations	118	118	118	118	118	118	118	118

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) relating product-specific $CIR_T^{P_i}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_i}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. CIR are here calculated from a FAVAR model estimated using only PPI product-level series (seasonally adjusted) using both the Euribor (with no instrument) and the 2-year German bond rate as a policy rate, in this latter model, the model is identified using a high frequency instrument method. Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1

Table R: Regression Results: No drift - Sectoral Average Inflation < 5%

Identification Long-run Restriction	Cholesky Yes		Cholesky No		High-Frequency IV Yes	
	24 months	36 months	24 months	36 months	24 months	36 months
<i>PANEL A: Producer Prices - Constrained model</i>						
Kurt/Freq	0.0597* (0.0318)	0.0917** (0.0350)	0.309*** (0.103)	0.509*** (0.161)	0.190*** (0.0616)	0.242*** (0.0805)
Constant	-19.95*** (2.037)	-34.68*** (2.150)	-21.03*** (6.203)	-36.50*** (9.569)	-34.05*** (3.671)	-52.25*** (4.851)
R^2	0.042	0.085	0.127	0.142	0.129	0.118
<i>PANEL B: Producer Prices - Unconstrained model</i>						
$Freq/\bar{F}$	-2.615** (1.290)	-3.243** (1.324)	-11.65*** (4.279)	-18.37*** (6.712)	-6.092** (2.798)	-7.294* (3.787)
$Kurt/\bar{K}$	2.817* (1.604)	4.008** (1.808)	11.29** (4.838)	18.19** (7.521)	6.322** (3.095)	7.900* (4.056)
Constant	-17.54*** (1.637)	-31.44*** (1.790)	-7.149 (4.549)	-14.05** (7.011)	-26.00*** (2.985)	-42.29*** (4.029)
R^2	0.133	0.184	0.295	0.303	0.220	0.179
Observations	116	116	116	116	116	116
<i>PANEL C: Consumer Prices - Constrained model</i>						
Kurt/Freq	-0.0261 (0.0171)	-0.00991 (0.0207)	0.0730 (0.0443)	0.156** (0.0706)	0.0542** (0.0249)	0.0805** (0.0322)
Constant	-9.926*** (2.909)	-25.94*** (3.448)	-13.56* (7.372)	-32.02*** (11.66)	-35.30*** (3.613)	-56.26*** (4.624)
R^2	0.009	0.001	0.014	0.025	0.023	0.030
<i>PANEL D: Consumer Prices - Unconstrained model</i>						
$Freq/\bar{F}$	-4.783 (2.896)	-8.439** (3.398)	-34.78*** (8.877)	-58.65*** (13.67)	-16.45*** (3.852)	-21.44*** (4.740)
$Kurt/\bar{K}$	3.505 (2.169)	5.047** (2.524)	4.216 (3.022)	6.237 (4.729)	7.892*** (2.418)	10.43*** (3.105)
Constant	-11.07*** (3.655)	-23.47*** (4.291)	23.76*** (8.425)	34.83*** (12.97)	-21.73*** (4.977)	-37.79*** (6.361)
R^2	0.059	0.128	0.478	0.533	0.360	0.357
Observations	214	214	214	214	214	214

Note: This table reports results of OLS results of the constrained model ([equation \(10\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio $Kurt/freq$ and OLS results of the unconstrained model ([equation \(11\)](#)) relating product-specific $CIR_T^{P_j}$ (expressed in %) to the ratio of the product-level frequency over its average $Freq/\bar{F}$ and the ratio of the product-level kurtosis over its average $Kurt/\bar{K}$. We remove products for which the average annual inflation is above 5% (in absolute values) over the sample period. Robust standard errors are reported in parentheses. *** p<0.01, ** p<0.05, * p<0.1