Motivation

Data

Gap

Selection

Credit shock

Selection

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Price Selection in Micro Data

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February 2022

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Motivation Se	lection
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Motivation

- Rigidity of the price level influences
 - Real effects of monetary policy
 - Amplification through 'demand' channels

Gap

- Prices change infrequently (Bils and Klenow, 2004)
- In standard price-setting models (Calvo, 1983)
 - Low frequency implies rigid price level
- In models microfounded by fixed (menu) costs of adjustment (Caplin and Spulber, 1987; Golosov and Lucas, 2007)
 - ▶ Price level stays flexible even if a small fraction adjusts, because
 - Large price changes are selected



Selection of large price changes

- Why are large price changes selected?
- > Menu costs: optimal to concentrate on the products with the largest price misalignment
- When an aggregate shock hits
 - Adjusted prices are the most misaligned,
 - They change by a lot,
 - Raise the flexibility of the price level.



- ▶ Revisit the Golosov and Lucas (2007)-critique to price-rigidity
- By establishing new facts using microdata
 - Generate proxies for price misalignment (price gap)
 - Identify aggregate shocks
 - Measure selection as the impact of the gap-shock (micro-macro) interaction on price-change probability
 - Are prices with large gaps changed with higher probability than those with small gaps, conditional on a shock?



- State dependence: price-change probability and size increases with gap
- ► No selection: gap immaterial with respect to aggregate shock
- ▶ Uniform shift between price increases versus price decreases (gross extensive margin)
- Provides guidance for model choice and policy implications
- Consistent with mildly state-dependent models with linear and flat hazard function and sizable monetary non-neutrality

Robustness

Selection: Theory (Caballero and Engel, 2007)

Price adjustment frictions: lumpy price adjustment

- Price gap $x_{it} = p_{it} p_{it}^*$
 - *p_{it}* (log) price of product *i*: adjusts occasionally
 - *p*^{*}_{it} (log) optimal price: influenced continuously by both product-level and aggregate factors
 - Dispersed distribution



Robustness

Discussion

Conclusion

References

Selection: Theory (Caballero and Engel, 2007)

• Focus: shape of the adjustment hazard $(\Lambda(x_{it}))$

- Menu cost (S,s) model
 - Step function



Robustness

Selection: Theory (Caballero and Engel, 2007)

Price changes are large in normal times (not selection)

- Menu cost (S,s) model
 - Price changes are the product of
 - Probability of adjustment and gap density
 - Size of adjustment: $-x_{it}$



Credit shock

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References

Selection: Theory (Caballero and Engel, 2007)

Selection: new adjusters after a shock

- Menu cost (S,s) model
 - New adjusters after a shock are large
- Calvo (1983) model
 - Flat hazard
 - No new adjusters: no selection



Discussion

Selection: Theory (Caballero and Engel, 2007)

Selection: reduces real effects of a monetary shock (Golosov and Lucas, 2007)

- Menu cost (S,s) model
 - New adjusters after a shock are large
- Calvo (1983) model
 - Flat hazard
 - ▶ No new adjusters: no selection



Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion	Refer
Data									
Data									

- IRi supermarket scanner data (pprox 15% of CPI)
 - Very granular: 170 000 products
 - ▶ Wide coverage: 50 markets across the US, over 3000 stores
 - 12 years of weekly data (2001-2012)
- Suitable dataset
 - Granularity: high-quality information about close substitutes
 - Long time series: can identify aggregate fluctuations
- Baseline data Data cleaning Expenditure weights
 - Reference prices: filter out temporary discounts Sales filtering
 - Time-aggregation: monthly mode



Posted, reference and sales-price indices





IRi supermarket index

- Similar business-cycle fluctuations as CPI food-at-home
- Trend inflation lower than CPI food-at-home
 - Main reason: new products
 - Higher-quality higher-price than existing products
 - CPI takes this into account we only use surviving products



- A relevant component of the gap is observable
 - Distance from the average price of close competitors,
 - Controlling for store fixed effects (regional variation, amenities)
 - Stores wants to avoid price misalignments; higher: low demand; lower: low markup
- Competitiors' reference-price gap

$$x_{pst} = p_{pst}^f - \bar{p}_{pt}^f - \hat{\alpha}_s,$$

where p_{pst}^{f} is the sales-filtered reference price and $\hat{\alpha}_{s}$ is the store-FE in $p_{pst}^{f} - \bar{p}_{pt}^{f} = \alpha_{s}$.

- Control for unobserved heterogeneity
 - Deduct estimated product-store FE
 - Raise all estimates with the average product-store FE

Robustness

Discussion

Conclusion

References

Competitors' price gap, density

Data

Density:

- Sizable dispersion, fat tails
- Despite sales-filtering and store-FE



Robustness

Conclusion

Discussion

References

Competitors' price gap, size

Size

Almost (inverse) one-on-one btw

gap and size, on average

Relevant component of the gap



Robustness

Discussion Conclusion

References

Competitors' price gap, frequency

- Adjustment hazard in the data: (comp. Gagnon and López-Salido, 2014; Eichenbaum et al., 2011)
 - Increases with distance from 0
 - Mildly asymmetric, positive at 0
 - Approximately (piecewise) linear



Impulse response to a credit shock

- Sizable, exogenous tightening of credit conditions
- Identified with timing restrictions (Gilchrist and Zakrajšek, 2012)
 - Increase in the excess bond premium (default-free corporate spread)
 - No contemporaneous effect on activity, prices and interest rate



Run a series of OLS regressions h (Jordà, 2005)

$$x_{t+h} - x_t = \alpha_h + ebp_t + \Gamma_h \Psi(L) X_t + u_{t,h},$$

- x: variable of interest, e.g. (log) price level
- ebp_t: credit shock
- $\Gamma_h \Psi(L) X_t$: set of controls: contemporaneous cpi, ip, 1y and 1-12m lags of cpi, ip, 1y, ebp
- Monthly aggregates, seasonally adjusted
- ▶ 95% confidence bands



Discussion

References

Credit shock, 2001-2012



Credit shock

Selection

Robustness

Conclusion

Discussion

References

Response of the supermarket-price index

Supermarket-price level



- Gradual response, not unlike core CPI
- Peak effect not before 24 months

Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion	References
<u> </u>									
Selectio	on								

- ▶ With a product-level proxy and an aggregate shock: we can now assess selection.
- Do the new adjusters after a shock have large gaps?
- Approach: Selection is an interaction between
 - Aggregate shock and
 - Product-level proxy.
- Framework: Linear probability model of price adjustment
 - Does the interaction term influences adjustment probability?



Linear probability model

$$I_{pst,t+h}^{\pm} = \beta_{xih}^{\pm} x_{pst-1} \hat{ebp}_t + \beta_{xh}^{\pm} x_{pst-1} + \beta_{ih}^{\pm} ebp_t + \gamma_h^{\pm} T_{pst-1} + \Gamma_h^{\pm} \Phi(L) X_t + \alpha_{psh}^{\pm} + \alpha_{mh}^{\pm} + \varepsilon_{psth}^{\pm},$$

- $I_{pst,t+h}^{\pm}$ indicator of price increase (resp. decrease) of product p in store s between t and t+h
- ► x_{pst-1}: price gap (to control for its regular effect)
- ebp_t is the aggregate shock (to control for its average effect)
- x_{pst-1}ebp_t gap-shock interaction (selection: focus of analysis)



Linear probability model, cont.

$$\begin{split} I_{pst,t+h}^{\pm} &= \beta_{xih}^{\pm} x_{pst-1} \hat{ebp}_t + \beta_{xh}^{\pm} x_{pst-1} + \beta_{ih}^{\pm} ebp_t + \\ &\gamma_h^{\pm} T_{pst-1} + \Gamma_h^{\pm} \Phi(L) X_t + \alpha_{psh}^{\pm} + \alpha_{mh}^{\pm} + \varepsilon_{psth}^{\pm}, \end{split}$$

- ► *T_{pst}* (log) age of price (to control for time dependence)
- $\Gamma_h^{\pm} \Phi(L) X_t$ aggregate controls
- $\alpha \pm_{psh}$ product-store FE (to control for unexplained cross-sectional heterogeneity)
- α_{mh}^{\pm} are calendar-month FE (to control for seasonality)
- Standard errors are clustered across categories and time

Discussion

Results, competitors' price gap, credit shock, h=24m

Gap

	(1) Price increase $\left(I_{pst,t+24}^{+}\right)$	(2) Price decrease $\left(I_{\textit{pst},t+24}^{-} ight)$
$Gap(x_{pst-1})$	-1.75***	1.55***
Shock (ebp_t)	-0.03***	0.03***
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.00	0.01
Age (T_{pst-1})	0.02***	0.00**
Product × store FE	1	1
Calendar-month FE	\checkmark	1
Time FE	×	×
N	16.1 <i>M</i>	16.1 <i>M</i>
within R^2	18.5%	17.3%

Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion	References
lune milion	at a ma								
Implica	tions								

- State dependence: Gap raises frequency Spec.
 - ▶ Probability of price increase 26 pp. lower btw 1st and 3rd quartile (decrease 23 pp higher)
- Adjustment on the (gross) extensive margin: aggregate shock shifts the probability of price increases vs price decreases
 - Probability of price increase 1pp lower after a 1sd credit tightening (30 bps)
 - Probability of price decrease 1pp higher after a similar tightening



Implications, cont.

No selection: Specification

- No evidence of significant interaction
- Conditional on the shock, not adjusting the prices with larger gap

Time dependence

Older prices are changed with higher probability



Robustness to non-linearity, alternative gap, shock, data

Conclusion

Conceptual framework (Caballero and Engel, 2007)

- Lumpy price adjustment: identify channels of adjustment
- Caballero and Engel (2007): two channels
 - Intensive margin: only channel in time-dependent
 - Extensive margin: new channel in state dependent
- Our contribution: generalize Caballero and Engel (2007)
 - Separate extensive margin into two channels
 - Gross extensive margin: shift between price increases vs decreases
 - ► Selection: large gaps adjust with higher probability, conditional on shock

Discussion

Decomposing inflation: An accounting identity

$$\pi_t = \pi_t^+ + \pi_t^- = \int_{x < 0} -x\Lambda(x)f_t(x)dx + \int_{x \ge 0} -x\Lambda(x)f_t(x)dx$$

- π^- : inflation from positive gaps
- Density: $f_t(x)$
- ► Hazard: Λ(x)
- Desired change = gap: -x





Evidence for state-dependence

Decomposition

$$\pi_t^- = \int_{x \ge 0} -x\Lambda(x)f(x)dx = -\bar{x}^-\bar{\Lambda}^- + \underbrace{\operatorname{Cov}\left(-x,\Lambda(x)|x \ge 0\right)}_{\text{state-dependence}},$$

• 'State-dependence': increasing hazard (Λ): higher gaps change w/ higher probability

- We brought evidence
- Inconsistent with time-dependent (constant hazard) models (Calvo, 1983)



• Conditional on a permanent shock m; x ex ante gap



- Intensive margin: those that adjust, adjust by less
- Gross extensive margin: more decreases, less increases
- Selection: new decreases after the shock are far from their optimum



Our evidence broadly consistent with mildly state-dependent models (Dotsey et al., 1999; Woodford, 2009) with (close to) linear and flat hazard

- Gross extensive margin: aggregate shock shifts increase/decrease frequencies
- No selection: insignificant interaction





Empirical decomposition

• We use empirical density and hazard



Relative contributions of different channels

Intensive	Gross extensive	Selection
margin	margin	effect
73.4%	26.5%	0.2%

Margins of adjustment, cont. • Calvo (1983) • Ss

Gap

Data

Motivation

Selection

	Data	Time-	(S,s) & Convex	Linear
		dependent	hazard	hazard
Intensive margin	1	1	1	1
Gross extensive margin	1	×	1	1
Selection	×	×	\checkmark	×

Selection

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Credit shock



- Use granular supermarket and PPI data to measure selection
- We have found that
 - 1. State dependence: adjustment probability increases with gap
 - 2. No selection: Conditional adjustment independent of price gap
 - 3. Adjustment through the intensive and gross extensive margin
- Implications
 - Inconsistent with standard time-dependent (Calvo, 1983) or state-dependent (Golosov and Lucas, 2007) models
 - Consistent with mildly state-dependent models with linear and flat hazard
 - Implies sizable monetary non-neutrality


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- Selection is a robust prediction of menu cost models
- Classic papers (Caplin and Spulber, 1987; Golosov and Lucas, 2007)
- More recent iterations:
 - ▶ Karadi and Reiff (2019): even if idiosyncratic shocks have fat tails (Midrigan, 2011)
 - > ?: even with multiproduct firms (Alvarez and Lippi, 2014)
- Selection weakens with information frictions (Woodford, 2009; Costain and Nakov, 2011), which also microfound 'random menu cost' models (Dotsey et al., 1999; ?; ?)
- Us: Empirical question



- Minimal structure (vs. suff. statistic ?)
 - Implicit hazard-function approach (Caballero and Engel, 2007)
 - Estimate density and hazard function by matching moments
 - Sizable selection (Berger and Vavra, 2018; Petrella, Santoro and Simonsen, 2019)
 - Weak selection (??)
 - ▶ Us: explicit hazard function (Gagnon, López-Salido and Vincent, 2012)
- Construct informative moments that reveals selection
 - > ?: preset-price-relative vs. inflation
 - Dedola et al. (2019): selection bias in Danish PPI
 - Us: shock-gap interaction on frequency

IRi: data cleaning

Posted prices:

$$P_{psw} = \frac{TR_{psw}}{Q_{psw}},$$

► *TR* is the total revenue

- Q is the quantity sold for each product
- p in store s in week w
- Cleaning
 - Round to the nearest penny (8.7%)
 - Private label products: new products at relabeling
 - Drop products that are not available the whole year

Data

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IRi: sales-filtering

- Sales: high-frequency noise (Anderson et al., 2017)
- Modal-price filter of ?
- Reference prices P^f_{psw} on weekly data
 - 13-week two-sided modal price
 - Iterative updating to align the change of P_{psw}^{f} with P_{psw}
 - Reference price changes less than a third of posted price changes
- Results are robust to using posted prices
- Monthly prices P_{pst}: mode of weekly prices



Fixed-weight index (as CPI). Annual weights $t \in y$

$$\omega_{psy} = \frac{TR_{psy}}{\sum_{p} \sum_{s} TR_{psy}}$$

• Posted and reference-price inflation (i = p, f)

$$\pi_{t}^{i} = \sum_{s} \sum_{p} \omega_{pst} \left(p_{pst}^{i} - p_{pst-1}^{i} \right)$$

Sales-price inflation

$$\pi_t^s = \pi_t^p - \pi_t^f$$

Seasonal adjustment using monthly dummies



- Focus: aggregate shock price-gap interaction term
- Price increases I_{pst}^+ : expected sign is positive
 - Driven by products with negative gap $(x_{pst-1} \leq 0)$
 - Credit tightening $(\hat{ebp}_t \ge 0)$: less price increases
 - Credit easing $(\hat{ebp}_t < 0)$: more price increases
- Price decreases I⁻_{pst}: expected sign is positive
 - Driven by products with positive gap $(x_{pst-1} \ge 0)$
 - Credit tightening $(\hat{ebp}_t \ge 0)$: more price decreases
 - Credit easing $(\hat{bp}_t < 0)$: less price decreases



Specification, cont.

- Additional interest
- Impact of the price gap β_{xh} : expected sign: negative for I_{pst}^+ (positive for I_{pst}^-)
 - More negative gap: more price increases
 - (More positive gap: more price decreases)
- ▶ Impact of aggregate shock β_{ih} : expected sign: negative for I_{pst}^+ (positive for I_{pst}^-)
 - Credit tightening $(\hat{ebp}_t > 0)$ less increases, more decreases
 - Credit easing $(\hat{ebp}_t < 0)$ more increases, less decreases



Match hazard function with elasticity of substitution of 7



Calibration misses left tail



Alternative calibration, cont.

Higher estimated information friction parameter

θ	0	2.562	∞
	(S,s)	uniform	calvo
Frequency $(\overline{\Lambda})$	8.5	13.6	27.1
eta	42.1	18.8	27.1
$eta/\overline{\Lambda}$	4.95	1.38	1

Still high monetary non-neutrality



Specification, cont.

- 2 additional specifications for robustness
- Time-fixed effects (drop the direct impact of shock)
- Separate coefficients for positive and negative gaps

References

Results, competitors' price gap, credit shock, h=24m

Gap

	(1)	(2)	(3)	(4)	(5)	(6)
	Price inc	crease $\left(I_{pst,t+24}^{+}\right)$)	Price dec	rease $\left(I_{pst,t+24}^{-}\right)$)
$Gap(x_{pst-1})$	-1.75***	-1.75***		1.55***	1.55***	
Shock (ebp_t)	-0.03***			0.03***		
Selection $(x_{pst-1}\hat{ebp}_t)$	-0.00	-0.00		0.01	0.01	
Age (T_{pst-1})	0.02***	0.02***		0.00**	0.01***	
Pos. gap (x_{pst-1}^+)						
Neg. gap (x_{pst-1}^{-})						
Pos. sel. $(x_{pst-1}^+ e\hat{b}p)$						
Neg. sel. $(x_{pst-1}^{-}\hat{ebp})$						
Product × store FE	1	1		1	1	
Calendar-month FE	1	x		1	x	
Time FE	×	1		×	1	
N	16.1 <i>M</i>	16.1 <i>M</i>		16.1 <i>M</i>	16.1 <i>M</i>	
within R ²	18.5%	16.6%		17.3%	16.4%	

Discussion

References

Results, competitors' price gap, credit shock, h=24m

	(1)	(2)	(3)	(4)	(5)	(6)	
	Price	increase $\left(I_{pst,t+1}^{+}\right)$	24)	Price decrease $\left(I_{pst,t+24}^{-}\right)$			
$Gap(x_{pst-1})$	-1.75***	-1.75***		1.55***	1.55***		
Shock (ebp _t)	-0.03***		-0.04***	0.03***		0.03***	
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.00	-0.00		0.01	0.01		
Age (T_{pst-1})	0.02***	0.02***	0.02***	0.00**	0.01***	0.01***	
Pos. gap (x_{pst-1}^+)			-2.26***			2.29***	
Neg. gap (x_{pst-1}^{-})			-1.44^{***}			1.10***	
Pos. sel. $(x_{pst-1}^+ e\hat{b}p)$			0.04			-0.04	
Neg. sel. $(x_{pst-1}^{-} \hat{ebp})$			-0.03			0.04	
Product × store FE	1	1	1	1	1	1	
Calendar-month FE	1	×	1	1	×	1	
Time FE	×	1	×	x	1	×	
N	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	
within R^2	18.5%	16.6%	18.9%	17.3%	16.4%	18.2%	

Robustness

Discussion Conclusion

References

Gap group-dummies, within product-store, 24m

- Hazard close to linear and quite symmetric
 - Heterogeneity is controlled for (item, time FEs)
 - Predicted frequency in 24 months





Price decreases

Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion	References

Average moments

Annualize	ed inflation	Frequency			
Posted	Reference	Posted	Reference		
1.84 %	1.75%	36.2%	10.8%		
Reference	e frequency	Reference size			
Increase	Decrease	Increase	Decrease		
6.6%	4.2%	12.5%	-15.1%		

Selection: Theory (Caballero and Engel, 2007)

► Selection: reduces real effects of a monetary shock (Golosov and Lucas, 2007)

- ► Menu cost (S,s) model
 - New adjusters after a shock are large
- Calvo (1983) model
 - Flat hazard
 - ▶ No new adjusters: no selection





Micro-data: how do standard moments adjust to aggregate shocks Average moments

► Frequency:

$$\xi_{t,t+h}^{\pm} = \sum_{i} \overline{\omega}_{it,t+h} I_{it,t+h}^{\pm},$$

Size

$$\psi_{t,t+h}^{\pm} = \frac{\sum_{i} \overline{\omega}_{it,t+h} I_{it,t+h}^{\pm} (p_{it+h} - p_{it-1})}{\xi_{t,t+h}^{\pm}}.$$

Decomposition

$$p_{t+h} - p_{t-1} = \pi_{t,t+h} = \xi^+_{t,t+h} \psi^+_{t,t+h} + \xi^-_{t,t+h} \psi^-_{t,t+h},$$





Price changes



Average size declines

Broadly in line with both time-dependent (Calvo, 1983) and state-dependent (Golosov and Lucas, 2007) models

Data Gap

Credit shock

Selection

Robustness

Conclusion

Discussion

References

Less increases, more decreases



Discussion

Price setting with information frictions (Woodford, 2009)

- Starting point: a standard menu-cost model (Golosov and Lucas, 2007)
 - Monopolistic competition with differentiated goods
 - ► Idiosyncratic cost shocks $A_t(i) = A_{t-1}(i) + \varepsilon_t, \varepsilon \sim N(0, \sigma_A^2)$
 - Price gap $(x_t(i) = p_t(i) p^*(i))$ determines profit
 - Fixed (menu) cost of a price review κ
- Timing of price review: rational inattention
 - Costly signal f(x) about the state (cost \uparrow w/ informativeness: $\theta I = -\theta E \left[\log (f) (x) \right] \right)$
 - ▶ Result #1: optimal policy described by a hazard function (adjustment (signal) probability as a function of current gap A(x))
 - Result #2: Functional form of hazard function is well defined, depends on θ (θ = ∞: constant hazard, calvo; θ = 0: step function, (S,s)).



- Use density and hazard estimated using the competitors'-reset prices
 - ▶ Valid measure if stores set prices to $p_t^*(i)$, when they change it,
 - Calibrate (i) review cost (κ), (ii) standard deviation of idiosyncratic shocks (σ_A), information cost (θ) to minimize expected deviation from the
 - Hazard function (weighted w/ data density), frequency of price changes, size of price changes.



Evidence for state-dependence

Decomposition

$$\pi_t^- = \int_{x \ge 0} -x \Lambda(x) f(x) dx = \bar{x}^- \bar{\Lambda}^- + \underbrace{\int_{x \ge 0} -x \left(\Lambda(x) - \bar{\Lambda}^-\right) f_t(x)}_{\text{state-dependence}},$$

- 'State-dependence': increasing hazard (Λ): higher gaps change w/ higher probability
- We brought evidence
- Inconsistent with time-dependent (constant hazard) models (Calvo, 1983)



▶ Conditional on a permanent shock *m*; *x* ex-shock gap



Intensive margin: those that adjust, adjust by less

Gross extensive margin: more decreases, less increases

Selection: new decreases after the shock are far from their optimum

Discussion

Time-dependent model (Calvo, 1983)





Discussion

References

Selection in an sS model (Golosov and Lucas, 2007)





Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion	References

Nonlinearity II: Probit

	(1)	(2)	(3)
	Multinon	nial probit	Ordered probit
	Incr. $\left(I_{pst,t+24}^{+}\right)$	Decr. $\left(I_{pst,t+24}^{-}\right)$	Change $(I_{pst,t+24})$
$Gap(x_{pst-1})$	-3.15***	3.37***	-4.24***
Shock (ebp_t)	-0.11^{***}	0.05***	-0.10^{***}
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.05	-0.21**	0.04
Age (T_{pst-1})	0.01*	-0.03***	0.02***
Freq. incr. (ξ^+_{psM})	5.17***	2.91***	1.79***
Freq. decr. (ξ^{psM})	3.02***	5.84***	-1.33***
Product × store FE	×	×	×
Calendar-month FE	1	1	1
Time FE	×	×	×
Ν	16.1 <i>M</i>	16.1 <i>M</i>	14.3 <i>M</i>

Credit shock

Selection R

Robustness

Conclusion

Discussion

References

Heterogeneity across product categories

- Heterogeneous demand elasticities might bias our baseline
- Separate estimates across product categories: price increases

Gap

Gap

Shock

Selection

	Effect Size		Effect Size			Effect Size with 95% CI
category	with 95% CI	category	with 95% CI	category		
all	-1.75 [-1.88, -1.62]	al —	-0.03 [-0.05, -0.02]	all		0 [-0.09, 0.09]
beer	-2.99 [-3.03, -2.95]	beer	0.01 [-0.00, 0.03]	beer		5 [-0.09, 0.19]
blades	-1.94 [-1.99, -1.88]	blades	0.00 [-0.01, 0.02]	blades		6 [-0.15, 0.04]
carbbev	-1.55 [-1.57, -1.52]	carbbev	-0.02 [-0.05, -0.00]	carbbev		1 [-0.07, 0.09]
coldcer	-1.82 [-1.86, -1.78]	coldcer	-0.05 [-0.08, -0.02]	coldcer		1 [-0.15, 0.13]
deod	-1.66 [-1.72, -1.61]	deod	0.01 [-0.01, 0.02]	deod		4 [-0.03, 0.10]
factiss	-1.72 [-1.79, -1.64]	factiss I	-0.05 [-0.09, -0.01]	factiss	-0.0	2 [-0.13, 0.09]
fzdinent	-1.58 [-1.63, -1.53]	fzdinent -	-0.04 [-0.05, -0.02]	fzdinent	-0.0	2[-0.17, 0.14]
fzpizza	-1.51 [-1.54, -1.48]	fzpizza 🔤	-0.05 [-0.09, -0.01]	fzpizza	-0.0	9 [-0.22, 0.04]
hhclean	-2.25 [-2.30, -2.21]	hhclean	-0.01 [-0.03, 0.01]	hhclean		4 [-0.05, 0.12]
hotdog	-1.50 [-1.55, -1.46]	hotdog	0.00 [-0.03, 0.03]	hotdog		3 [-0.11, 0.18]
laundet	-2.08 [-2.12, -2.04]	laundet -	-0.01 [-0.02, 0.01]	laundet	0.0	8 [-0.07, 0.24]
mayo	-1.68 [-1.77, -1.59]	mayo	-0.03 [-0.05, -0.00]	mayo		11 [-0.15, 0.17]
mustketc	-1.77 [-1.82, -1.72]	mustketc	0.03 [-0.04, -0.01]	mustketc		0 [-0.10, 0.10]
paptowl	-1.89 [-1.96, -1.82]	paptowl	-0.04 [-0.06, -0.01]	paptowl	0.0	2 [-0.20, 0.24]
peanbutr	-2.13 [-2.26, -2.00]	peanbutr	-0.06 [-0.09, -0.03]	peanbutr	-0.0	0 [-0.26, 0.26]
razors -	-3.21 [-3.35, -3.08]	razors	-0.03 [-0.07, 0.01]	razors	-0.2	2 [-0.54, 0.10]
saltsnck	-1.48 [-1.52, -1.43]	saltsnck -	-0.01 [-0.03, 0.00]	saltsnck		3 [-0.05, 0.11]
shamp	-2.07 [-2.10, -2.04]	shamp	-0.00 [-0.02, 0.01]	shamp		0 [-0.08, 0.09]
soup	-1.70 [-1.74, -1.66]	soup —	-0.05 [-0.06, -0.03]	soup	-0.0	3 [-0.17, 0.11]
spagsauc	-1.85 [-1.88, -1.82]	spagsauc —	-0.03 [-0.05, -0.01]	spagsauc	-0.0	2 [-0.14, 0.10]
sugarsub	-2.32 [-2.37, -2.26]	sugarsub	0.00 [-0.03, 0.03]	sugarsub	-0.0	2 [-0.16, 0.13]
toothbr	-1.55 [-1.59, -1.51]	toothbr -	0.02 [-0.03, -0.01]	toothbr		0 [-0.07, 0.06]
toothpa	-1.84 [-1.88, -1.80]	toothpa	-0.02 [-0.04, -0.01]	toothpa	-0.0	1 [-0.09, 0.06]
yogurt	-2.07 [-2.11, -2.04]	yogurt	-0.03 [-0.05, -0.01]	yogurt	0.0	6 [-0.10, 0.21]
	-1.92 [-2.08, -1.75]		-0.02 [-0.03, -0.01]		o.c 🖕	0 [-0.02, 0.02]

Gap

Selection

Robustness

Discussion

Selection

Conclusion

References

Heterogeneity across product categories, cont.

Gap

Separate estimates across product categories: price decreases

category	Effect Size with 95% CI	category	Effect Size with 95% CI	category		Effect Size with 95% CI
ili — 🖬 —	1.55 [1.42, 1.68]	all —	0.03 [0.01, 0.04]	all		0.01 [-0.10, 0.11
beer	2.27 [2.22, 2.32]	beer -	-0.01 [-0.02, 0.00]	beer		-0.06 [-0.20, 0.08
olades 🗾	1.52 [1.48, 1.56]	blades -	0.00 [-0.01, 0.01]	blades		0.05 [-0.06, 0.15
oldcer I	1.60 [1.56, 1.64]	coldcer	0.03 [0.01, 0.05]	coldcer		0.01 [-0.13, 0.16
deod I	1.52 [1.48, 1.55]	deod 📲	0.00 [-0.01, 0.01]	deod		0.01 [-0.07, 0.10
	1.71 [1.67, 1.76]	fzdinent -	0.03 [0.02, 0.05]	fzdinent		0.00 [-0.14, 0.14
zpizza	1.53 [1.49, 1.56]	fzpizza	0.03 [-0.00, 0.05]	fzpizza		- 0.02 [-0.14, 0.19
hclean	1.65 [1.61, 1.68]	hhclean -	0.01 [0.00, 0.02]	hhclean	-	-0.03 [-0.11, 0.06
notdog	1.11 [1.08, 1.14]	hotdog -1	+0.00 [+0.01, 0.01]	hotdog	-	-0.02 [-0.10, 0.07
aundet	1.86 [1.83, 1.89]	laundet -	- 0.02 [0.01, 0.03]	laundet		-0.04 [-0.15, 0.06
nargbutr 🚹	1.32 [1.25, 1.38]	margbutr	0.06 [0.04, 0.08]	margbutr		0.01 [-0.13, 0.1
nayo 🚹	1.30 [1.24, 1.37]	mayo 🚽	- 0.02 [0.01, 0.03]	mayo	-	0.01 [-0.10, 0.1
nilk	1.91 [1.82, 1.99]	milk	0.05 [0.01, 0.09]	milk		0.09 [-0.06, 0.23
mustketc	1.29 [1.26, 1.32]	mustketc	0.02 [0.01, 0.03]	mustketc		-0.04 [-0.11, 0.03
beanbutr	1.84 [1.72, 1.96]	peanbutr	0.05 [0.03, 0.07]	peanbutr -		-0.03 [-0.28, 0.2
azors		razors	-0.00 [-0.02, 0.02]	razors		-0.08 [-0.34, 0.1
saltsnck	1.14 [1.10, 1.17]	saltsnck -	0.01 [0.01, 0.02]	saltsnck	-	-0.01 [-0.10, 0.0
shamp 💽	1.86 [1.83, 1.89]	shamp -	0.02 [0.01, 0.03]	shamp	-	-0.04 [-0.13, 0.0
ioup 💽	1.42 [1.38, 1.46]	soup -	0.02 [0.01, 0.03]	soup		- 0.05 [-0.09, 0.1
spagsauc 💽	1.71 [1.66, 1.76]	spagsauc —	- 0.02 [0.01, 0.04]	spagsauc		- 0.04 [-0.11, 0.2
sugarsub	1.80 [1.76, 1.85]	sugarsub –	0.01 [-0.01, 0.02]	sugarsub		-0.01 [-0.14, 0.1
oothbr 🗾	1.21 [1.17, 1.24]	toothbr -	0.03 [0.02, 0.05]	toothbr		0.02 [-0.04, 0.04
oothpa 🗾	1.65 [1.61, 1.68]	toothpa -	0.01 [0.00, 0.02]	toothpa		0.00 [-0.10, 0.1
rogurt 💶	2.02 [1.98, 2.06]	yogurt		yogurt		-0.04 [-0.17, 0.10
•	1.64 [1.50, 1.77]	•	0.02 [0.01, 0.02]			-0.00 [-0.02, 0.02

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Shock

▲ Back

Reset-price gap

- Alternative price-gap proxy
- Reference price reset gap $x_{pst} = p_{pst}^f p_{pst}^{f*}$
- Reset-price (p_{pst}^{f*}) is as in Bils et al. (2012)

$$p_{pst}^{f*} = \left\{ egin{array}{cc} p_{pst}^f & ext{if } I_{pst} = 1 \ p_{pst-1}^{f*} + \pi_{ct}^{f*} & ext{otherwise,} \end{array}
ight.$$

where π_{ct}^{f*} is category-level reset-price inflation:

$$\pi_{ct}^{f*} = \sum_{p \in c} \frac{\omega_{pst} I_{pst} \left(p_{pst}^{f*} - p_{pst-1}^{f*} \right)}{\sum_{p \in c} \omega_{pst} I_{pst}}$$





Discussion

References

Results, reset-price gap, credit shock, h=24m

	(1)	(2)	(3)	(4)	(5)	(6)
	Price i	ncreases $(I_{pst,t+}^+)$	24)	Price d	lecreases $(I^{pst,t+})$	-24)
$Gap(x_{pst-1})$	-0.45***	-0.48***		0.34***	0.37***	
Shock (\hat{ebp}_t)	-0.04***		-0.04***	0.03***		0.03***
Selection $(x_{pst-1} \hat{b} p_t)$	-0.14	-0.13		0.12	0.14	
Age (T_{pst-1})	0.01***	0.01***	0.01***	0.01***	0.02***	0.01***
Positive gap (x_{pst-1}^+)			-0.39***			0.33***
Negative gap (x_{pst-1}^{-})			-0.49***			0.35***
Pos. sel. $(x_{pst-1}^+ \hat{ebp}_t)$			0.11			-0.03
Neg. sel. $(x_{pst-1}^{-} \hat{ebp}_t)$			-0.27**			0.21*
Ν	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>
within R^2	2.6%	0.3%	2.6%	1.3%	0.3%	1.3%
Data

Selection

Robustness

Discussion

Conclusion

References

PPI microdata

Coverage

- 1981-2012 monthly data
- Representative of the US economy
- No sales filtering











PPI: gaps

Size: clear negative relationship with the gaps

- ► Frequency:
 - Increases with competitors' gap eventually
 - Initially decreases with higher gap

Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion

Credit shock

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References

Gap

Selection

References

Results, competitors' price gap, credit shock, h=24m, PPI

	(1) Increases (I_p^+)	(2)	(3) Decreases (Ip	(4)
	increases (1p	st,t+24)	Decreases (1p	st,t+24)
$Gap(x_{pst-1})$	-0.23***	-0.23^{***}	0.22***	0.22***
Shock (ebp_t)	-0.023***		0.021***	
Selection $(x_{pst-1} \hat{ebp}_t)$	0.00	-0.00	-0.00	-0.00
Age (T_{pst-1})	0.035***	0.035***	0.01***	0.01***
Product × store FE	1	1	1	1
Calendar-month FE	1	×	1	x
Time FE	×	1	×	1
N	9.7 <i>M</i>	9.7 <i>M</i>	9.7 <i>M</i>	9.7 <i>M</i>
Within R ²	4.4%	3.5%	4.3%	3.7%



- Results are robust using longer and wider-coverage data
- ► Gap: significant unconditional impact on frequency
- Aggregate shock: shifts the probability of adjustment
- ► No selection:
 - No evidence of interaction:
 - Conditional on the shock, not adjusting prices with larger gap



Impulse responses to monetary policy shocks

- High-frequency identification of monetary policy shocks (Gertler and Karadi, 2015; Nakamura and Steinsson, 2018)
 - Intra-day financial market surprises around press statements
 - Control for information shocks using the co-movement of interest rates and stock prices (Jarociński and Karadi, 2020)
- Calculate relevant price-setting moments
- Estimate impulse responses using local projections (Jordà, 2005)

High-frequency identification of monetary policy shocks

- Central bank announcements generate unexpected variation in interest rates: can be used to assess monetary non-neutrality.
- Surprises
 - ▶ Measure change in interest rates in a 30-minute window around policy announcements
 - Only central bank announcements systematically impacts surprises
- FOMC press statements (8 times a year)

Robustness

Discussion

References

Conclusion

High-frequency surprises





Motivation	Selection	Data	Gap	Credit shock	Selection	Robustness	Discussion	Conclusion	References

Interest rate

- Preferred interest rate: 3-months federal funds futures rate
 - Closely controlled by the FOMC
 - Incorporates next FOMC meeting: with near-term forward guidance
 - Does not affected by 'timing' surprises
 - It stays active after ZLB is reached

bustness

Discussion

Conclusion

References

Controlling for central bank information shocks

- Issue: announcements can reveal information
 - not just about policy,
 - but also about the central bank's economic outlook.
- Use responses in stock markets (Jarociński and Karadi, 2020) Scatter
 - ▶ Negative co-movement in interest rates and stock prices: monetary policy shocks
 - Positive co-movement: central bank information shocks
- ▶ 'Poor man's sign restriction': use events when the co-movement was negative



Local projections

Run a series of OLS regressions h (Jordà, 2005)

$$x_{t+h} - x_t = \alpha_h + \beta_h \Delta i_t + \Gamma_h \Psi(L) X_t + u_{t,h},$$

- x: variable of interest, e.g. (log) price level
- Δi_t : high-frequency monetary policy shock
- $\Gamma_h \Psi(L) X_t$: set of controls: various lags of cpi, ip, dely

 Motivation
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 Impulse responses of key macroeconomic variables to a monetary policy

tightening



Motivation Selection Gap Credit shock Selection Robustness Conclusion Impulse responses of key macroeconomic variables to a monetary policy tightening

Data



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Discussion



Price changes



Aggregate frequency drops

► Size declines



Gap

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Less increases more decreases



Selection

Robustness

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References

Results, competitors' price gap, MP shock, h=12m

Gap

	(1)	(2)	(3)	(4)	(5)	(6)	
	Price i	Price increases $\left(I_{pst,t+12}^{+}\right)$			Price decreases $(I_{pst,t+1}^{-})$		
$Gap(x_{pst-1})$	-1.71^{***}	-1.71^{***}		1.36***	1.36***		
Shock (Δi_t)	-0.03*		-0.03	0.01*		0.01*	
Selection $(x_{pst-1}\Delta i_t)$	-0.07	-0.07		0.07	0.07		
Age (T_{pst-1})	0.03***	0.03***	0.03***	0.01***	0.01***	0.01***	
Positive gap (x_{pst-1}^+)			-1.92***			1.93***	
Negative gap (x_{pst-1}^{-})			-1.58^{***}			1.01***	
Pos. selection $(x_{pst-1}^+ \Delta i_t)$			-0.05			0.05	
Neg. selection $(x_{pst-1}^{-}\Delta i_t)$			-0.08			0.08	
Product × store FE	1	1	1	1	1	1	
Calendar-month FE	1	×	1	1	×	1	
Time FE	×	1	×	×	1	×	
N	23.7M	23.7M	23.7M	23.7M	23.7 <i>M</i>	23.7 <i>M</i>	
Within R ²	16.4%	14.7%	16.5%	13.3%	12.7%	13.8%	



MP shock: selection

- Robustly no evidence for selection
- Significant shift in adjustment probability in supermarket prices

Gap

Selection

Robustness

References

Robustness to dropping fixed effects

	(1)	(2)	(3)	(4)
	Increases ($\left(\stackrel{+}{_{pst,t+24}} \right)$	Decreases ($I^{-}_{pst,t+24}\Big)$
$Gap(x_{pst-1})$	-1.75***	-0.99***	1.55***	0.90***
Shock (ebp_t)	-0.03***	-0.04***	0.03***	0.03**
Selection $(x_{pst-1}\hat{ebp}_t)$	-0.00	-0.01	0.01	0.02
Age (T_{pst-1})	0.02***	-0.01^{**}	0.00**	-0.03***
Product × store FE	1	×	1	×
Calendar-month FE	1	1	1	1
Time FE	×	×	×	×
N	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>
Within R^2	18.5%	8.9%	17.3%	9.3%

Robustness to using posted prices

	(1) Increases (1	(2) (+	(3) Decreases ((4) /
	Reference	Posted	Reference	Posted
$Gap(x_{pst-1})$	-1.75***	-1.46***	1.55***	1.25***
Shock (ebp_t)	-0.03***	-0.04***	0.03***	0.03***
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.00	-0.01	0.01	0.02
Age (T_{pst-1})	0.02***	0.01***	0.00**	-0.01^{***}
Product × store FE	1	1	1	1
Calendar-month FE	1	1	1	1
Time FE	×	×	×	×
N	16.1 <i>M</i>	18.6 <i>M</i>	16.1 <i>M</i>	18.6 <i>M</i>
Within R^2	18.5%	17.6%	17.3%	14.8%

Robustness

Discussion

References

Robustness to excluding the Great Recession

Gap

	(1)	(2)	(3)	(4)
	Increases ($\left(I_{pst,t+24}^{+}\right)$	Decreases	$\left(I_{pst,t+24}^{-}\right)$
	2001-2012	2001-2007	2001-2012	2001-2007
$Gap\;(x_{pst-1})$	-1.75***	-1.74***	1.55***	1.50***
Shock (ebp_t)	-0.03***	-0.03***	0.03***	0.02***
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.00	0.06	0.01	-0.06
Age (T_{pst-1})	0.02***	0.02***	0.00**	0.01***
$Product \times store \; FE$	1	1	1	1
Calendar-month FE	1	1	1	1
Time FE	×	×	×	×
N	16.1 <i>M</i>	9.9 <i>M</i>	16.1 <i>M</i>	9.9 <i>M</i>
Within R^2	18.5%	17.7%	17.3%	16.5%

Competitors' price gap, cont.

Increase frequency

Decrease frequency

Discussion



Competitors' price gap, cont.

Increase size

Decrease size









Decrease frequency



Reset price gap, cont.



Decrease size

Discussion



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Gap

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References

Robustness to dropping fixed effects

	(1)	(2)	(3)	(4)
	Increases ($\left(\stackrel{+}{_{pst,t+24}} \right)$	Decreases ($I^{-}_{pst,t+24}\Big)$
$Gap(x_{pst-1})$	-1.75***	-0.99***	1.55***	0.90***
Shock (ebp_t)	-0.03***	-0.04***	0.03***	0.03**
Selection $(x_{pst-1}\hat{ebp}_t)$	-0.00	-0.01	0.01	0.02
Age (T_{pst-1})	0.02***	-0.01^{**}	0.00**	-0.03***
Product x store FE	1	×	1	×
Calendar-month FE	1	1	1	1
Time FE	×	×	×	×
N	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>	16.1 <i>M</i>
Within R^2	18.5%	8.9%	17.3%	9.3%

Robustness to using posted prices

	(1)	(2)	(3)	(4)
	Increases ($I_{pst,t+24}^+$	Decreases ($I_{pst,t+24}^{-}$
	Reference	Posted	Reference	Posted
$Gap\;(x_{pst-1})$	-1.75***	-1.46***	1.55***	1.25***
Shock (ebp_t)	-0.03***	-0.04***	0.03***	0.03***
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.00	-0.01	0.01	0.02
Age (T_{pst-1})	0.02***	0.01***	0.00**	-0.01^{***}
Product × store FE	1	1	1	1
Calendar-month FE	1	1	1	1
Time FE	X	×	×	×
N	16.1 <i>M</i>	18.6 <i>M</i>	16.1 <i>M</i>	18.6 <i>M</i>
Within R^2	18.5%	17.6%	17.3%	14.8%

Robustness

Discussion

References

Robustness to excluding the Great Recession

Gap

	(1)	(2)	(3)	(4)
	Increases ($\left(I_{pst,t+24}^{+}\right)$	Decreases	$\left(I_{pst,t+24}^{-}\right)$
	2001-2012	2001-2007	2001-2012	2001-2007
$Gap\;(x_{pst-1})$	-1.75***	-1.74***	1.55***	1.50***
Shock (ebp_t)	-0.03***	-0.03***	0.03***	0.02***
Selection $(x_{pst-1} \hat{ebp}_t)$	-0.00	0.06	0.01	-0.06
Age (T_{pst-1})	0.02***	0.02***	0.00**	0.01***
$Product \times store \; FE$	1	1	1	1
Calendar-month FE	1	1	1	1
Time FE	×	×	×	×
N	16.1 <i>M</i>	9.9 <i>M</i>	16.1 <i>M</i>	9.9 <i>M</i>
Within R^2	18.5%	17.7%	17.3%	16.5%

Conclusion

Competitors' price gap, cont.

Increase frequency

Decrease frequency





Competitors' price gap, cont.

Increase size

Decrease size







Decrease frequency



Reset price gap, cont.



Decrease size

Discussion



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