

# Price Selection in the Microdata

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# Sticky-prices in a GHF framework: overview of paper

- ▶ Empirically analyze price setting behavior through a GHF framework
- ▶ Discuss implications for shock propagation: how important is selection?
- ▶ More specifically: measure firms' desired adjustment  $\hat{x}_{i,t}$  and :
  1. use micro data to estimates a GHF :  $\Lambda(\hat{x}_{i,t})$  , and other moments
  2. identify a time series for aggregate "monetary" shocks:  $\epsilon_t$
  3. use OLS to study effect of  $\hat{x}_{i,t}$  ,  $\epsilon_t$  ,  $\underbrace{\hat{x}_{i,t} \cdot \epsilon_t}_{\text{"selection"}}$  on prob. of adjustment

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- ▶ Main results (emphasized in the paper):
  - ▶ Price-setting behavior shows strong elements of state dependence
  - ▶ The GHF linear in  $|\hat{x}|$ , as in Eichenbaum, Jaimovich, Rebelo (AER 2011)
  - ▶ Find no role for interaction term:  $\hat{x}_{i,t} \cdot \epsilon_t$  (selection is overrated!)

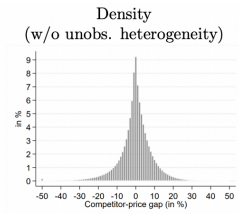
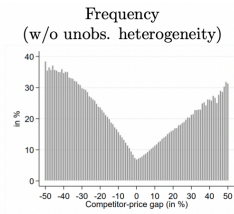
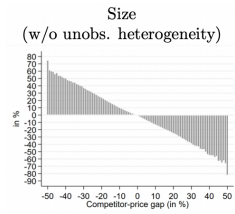
# Short summary of GHF models (Caballero-Engel)

- ▶ Setup for models with fixed cost of adjustment:
  - ▶ Firm  $i$  controls gap:  $x_i \equiv (p_i - mc_i) - \mu^*$  where  $\mu^*$  is the ideal markup
  - ▶ Uncontrolled state  $x_i$  follows diffusion:  $dx_i = \sigma dW_i$
  - ▶ Optimal policy gives GHF:  $\Lambda(x_i)$  if  $x_i \in (\underline{x}, \bar{x})$ , adjust otherwise
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  - ▶ Upon adjustment state is reset to  $x^* = 0$  (“closing the gap”)
- ▶ Aggregation for many firms: Given  $\{\Lambda(\cdot), \underline{x}, \bar{x}, \sigma\}$  we have
  - ▶ cross-section distribution of gaps:  $f(x)$  KFE:  $\Lambda(x)f(x) = \frac{\sigma^2}{2} f''(x)$
  - ▶ Frequency of price changes:  $N$ ,  $N = 2 \int_0^{\bar{x}} \Lambda(x)f(x)dx - \sigma^2 f'(\bar{x})$
  - ▶ cross-section distribution of price changes:  $q(\Delta x)$ ,  $q(-\Delta x) = \frac{\Lambda(x)f(x)}{N}$

# Interesting Results: New important facts (fig. 2)

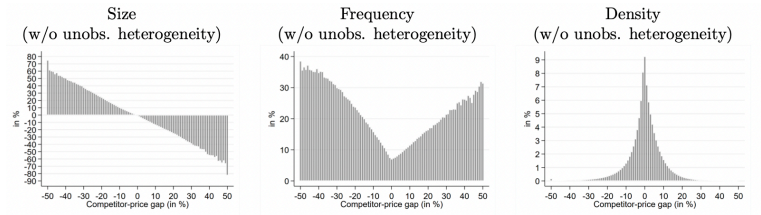


Price changes:  $\Delta x$

GHF  $\Lambda(x)$

density:  $f(x)$

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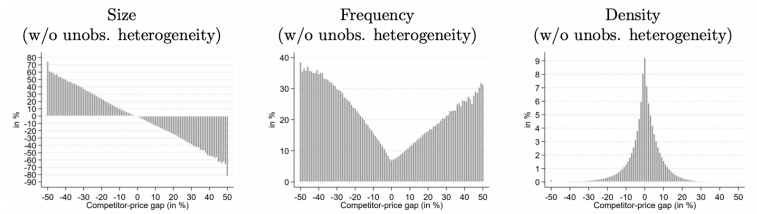


Price changes:  $\Delta x$   
closing the gap!

GHF  $\Lambda(x)$   
linear and symmetric!  
(no fairies in sight)

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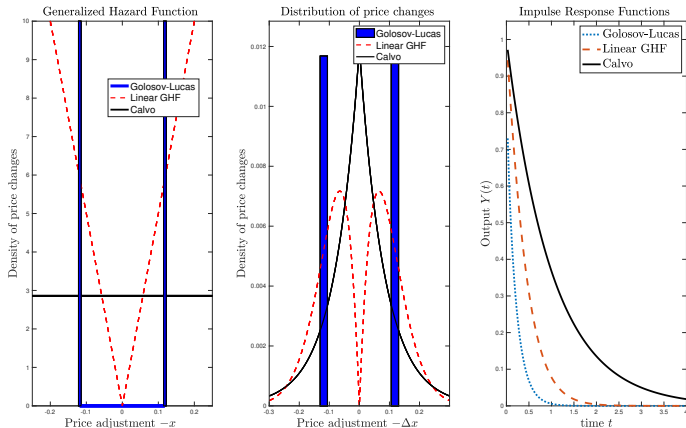
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**If price-setting follows GHF model, then shock propagation fully known**



# Shock propagation in GHF models: analytic results

three models with same frequency  $N$  and different aggregate flexibility



GHF encodes all you need to study shock propagation

# Authors' analysis of "selection" in the empirical model

- ▶ Price setting probability depends on "gap", "shock", and "selection"

## Linear probability model

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

- ▶  $I_{pst,t+h}^{\pm}$  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)

# Probability model: estimates

Table 3: Estimates, scanner data, competitor-price gap, credit shock

	(1)	(2)	(3)	(4)	(5)	(6)
	Price increase $(I_{pst,t+24}^+)$			Price decrease $(I_{pst,t+24}^-)$		
Gap $(x_{pst-1})$	-1.75*** (0.06)	-1.75*** (0.06)		1.55*** (0.06)	1.55*** (0.06)	
Shock $(ebp_t)$	-0.03*** (0.01)		-0.04*** (0.01)	0.03*** (0.01)		0.03*** (0.01)
Selection $(x_{pst-1}\hat{ebp}_t)$	-0.00 (0.04)	-0.00 (0.04)		0.01 (0.05)	0.01 (0.04)	
Age $(T_{pst-1})$	0.02*** (0.00)	0.02*** (0.00)	0.02*** (0.00)	0.00** (0.00)	0.01*** (0.00)	0.01*** (0.00)

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- ▶ monetary shock  $\epsilon_t$  affects firm's marginal cost  $\implies$  Prob P-change
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  - Paper measures gap  $\hat{x}_i = p_i - p$  where  $p$  is competitors avg. price
  - theory-based gap:  $x_i = \hat{x}_i + \alpha\epsilon_t$  and hazard:  $\Lambda(x_{i,t}) = \Lambda(\hat{x}_{i,t} + \alpha\epsilon_t)$

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example #1 :

$$\Lambda(\hat{x}_{i,t} + \alpha \epsilon_t) = \left| \hat{x}_{i,t} + \alpha \epsilon_t \right|$$

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example #2 :

$$\Lambda(\hat{x}_{i,t} + \alpha \epsilon_t) = \left( \hat{x}_{i,t} + \alpha \epsilon_t \right)^2$$

# Use theory as a LAB to test the metrics

example #1  $\Lambda(\hat{x}_{i,t-1} + \alpha \epsilon_t) = \left| \hat{x}_{i,t-1} + \alpha \epsilon_t \right|$  with  $\alpha = 1$

Dependent variable: Prob of price decrease  $I_{t+h}^-$

h  
period

$$\left| \hat{x}_{i,t-1} + \alpha \epsilon_t \right|$$

t-stat

$$\hat{x}_{i,t-1}$$

t-stat

$$\epsilon_t$$


t-stat

$$\hat{x}_{i,t-1} \cdot \epsilon_t$$

t-stat

$$age_t$$

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12 aggr. shocks per year, 10 years data, 50 k products (  $\Rightarrow$  asymptotic stats) 




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h period	+1 daily
$\left  \hat{x}_{i,t-1} + \alpha \epsilon_t \right $ t-stat	0.055 130
$\hat{x}_{i,t-1}$ t-stat	-
$\epsilon_t$ t-stat	-
$\hat{x}_{i,t-1} \cdot \epsilon_t$ t-stat	-0.005 -0.1
$age_t$ t-stat	

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$\hat{x}_{i,t-1}$ t-stat	-	0.027 200
$\epsilon_t$ t-stat	-	0.026 14
$\hat{x}_{i,t-1} \cdot \epsilon_t$ t-stat	-0.005 -0.1	0.14 9
$age_t$ t-stat		0.0036 130


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Dependent variable: Prob of price decrease  $I_{t+h}^-$

h	+1	+1	+1
period	daily	daily	monthly
$ \hat{x}_{i,t-1} + \alpha \epsilon_t $	0.055		
t-stat	130		
$\hat{x}_{i,t-1}$	-	0.027	0.76
t-stat		200	400
$\epsilon_t$	-	0.026	0.84
t-stat		14	270
$\hat{x}_{i,t-1} \cdot \epsilon_t$	-0.005	0.14	4.3
t-stat	-0.1	9	140
$age_t$		0.0036	0.12
t-stat		130	430

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- ▶ Calvo or GL? neither really, and even a good GHF is still “not enough”  
Look for strategic complementarities?  $\Lambda(x, X)$

## Summing up

- ▶ The paper has some very interesting micro evidence
- ▶ Direct evidence on GHF and behavior upon adjustment (closing the gap)
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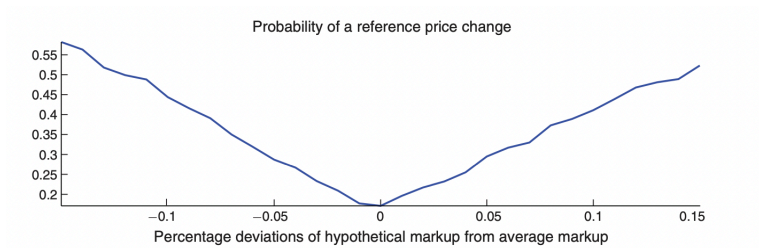
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- ▶ Direct evidence on GHF and behavior upon adjustment (closing the gap)
- ▶ Results are new and important for macro! (no Calvo behavior)
- ▶ The discussion of “selection” needs a tighter link to theory
  - Does “interaction” matter? depends on fct form of  $\Lambda$ , horizon  $h$ , sample size
- ▶ The data could be used to test strategic complementarities  $\Lambda(x, X)$

# Background material

# Eichenbaum, Jaimovich, Rebelo, AER 2011

Roughly linear hazard (in absolute value)



data from large US supermarket chain

# Pitfalls of the Heuristic approach

## State dependence (extending Caballero and Engel, 2007)

- ▶ Focus: shape of the adjustment hazard  $\Lambda(x)$ .
- ▶ Steep hazard: price changes are large unconditionally (state-dependence, not selection)

$$\pi^- = \int_{x \geq 0} -x \Lambda(x) f(x) dx = -\bar{x}^- \bar{\Lambda}^- + \underbrace{\text{Cov}(-x, \Lambda(x) | x \geq 0)}_{\text{state-dependence}}$$

