

Macroeconomía y Costos de Ajuste

Cátedras 5 y 6

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III. Investment

- ① Basics
- ② Lumpy investment in partial equilibrium:
 - ① The cautionary effect of uncertainty
 - ② Increasing hazards and time-varying IRFs
 - ③ Uncertainty shocks and an application to 9/11
- ③ Lumpy investment in general equilibrium

3. Lumpy Investment in General Equilibrium

- Thomas (JPE, 2002), Khan and Thomas (JME, 2003; ECMA, 2008)
- Bachmann, Caballero and Engel (2008)

Thomas and Khan-Thomas papers

- If prices fixed: micro lumpiness has macro impact (as in Caballero and Engel, 1999)
- Yet in a DSGE model, with an endogenous interest rate, Khan-Thomas's calibration is such that lumpiness washes away and aggregate investment dynamics becomes indistinguishable from how it behaves in an economy with no adjustment costs
- Next slide: figure from Khan and Thomas (2003) with aggregate capital after a positive productivity shock followed by a negative shock:
 - two sample paths: standard RBC, lumpy adjustment
 - indistinguishable

Irrelevance of Lumpy Adjustment in General DSGE?

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A. Khan, J.K. Thomas / Journal of Monetary Economics 50 (2003) 331–360

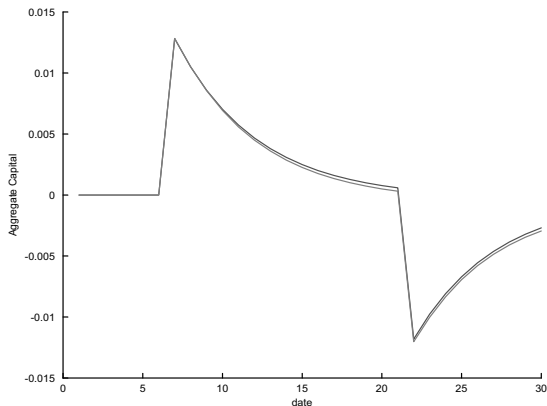


Fig. 8. The absence of nonlinearities in equilibrium.

Bachmann, Caballero, and Engel (2008)

- 1 Motivation
- 2 Model
- 3 Calibration
- 4 Aggregate Investment Dynamics
- 5 Conclusion

3.1. Motivation

The question

- Is lumpy capital adjustment relevant for aggregate investment dynamics?

The answer

- **Yes** ... in **partial** equilibrium: Caballero–Engel–Haltiwanger (1995), Caballero and Engel (1999), Cooper–Haltiwanger–Power (1999).
- **No** ... in **general** equilibrium: Thomas (2002), Khan–Thomas (2003, 2006)
- **Yes** ... in **general** equilibrium: This paper

What does it mean to be relevant?

Three possibilities (CE-1999)

- A. Skewness and kurtosis of aggregate investment series
- B. Better out-of-sample forecasts
- C. Time-varying impulse response function

How should we evaluate these “relevance” criteria

- Enough statistical power?
- Relevant for macroeconomics?

A. Skewness and kurtosis

Statistical power

- Caballero–Engel (1999): 20 2-digit manufacturing sectors ✓
- General equilibrium models (Thomas, Khan–Thomas, this paper): work with one aggregate investment series: ✗
 - less skewness and kurtosis at higher levels of aggregation
 - less data

Should macroeconomists care?

- Skewness and kurtosis do not matter per se

B. Out-of-sample forecasts

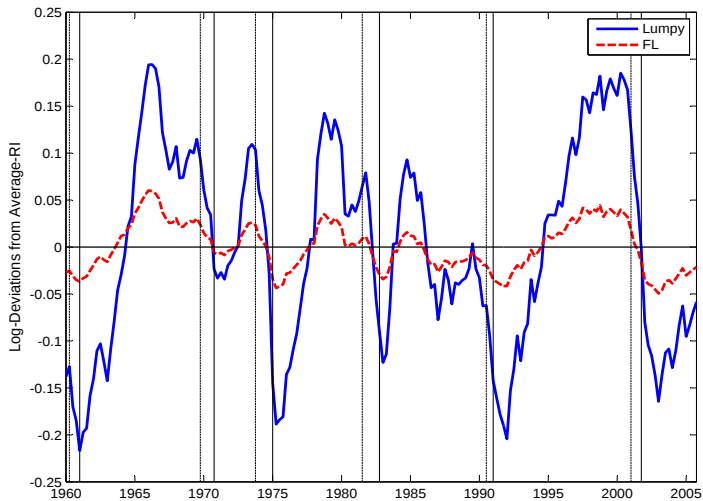
Statistical power

- Caballero–Engel (1999): 20 2-digit manufacturing sectors ✓
 - Equipment: 6.6% avge. reduction in forecast error
 - Structures: 31.3% avge. reduction in forecast error
- General equilibrium models (Thomas, Khan–Thomas, this paper): work with one aggregate investment series ✕

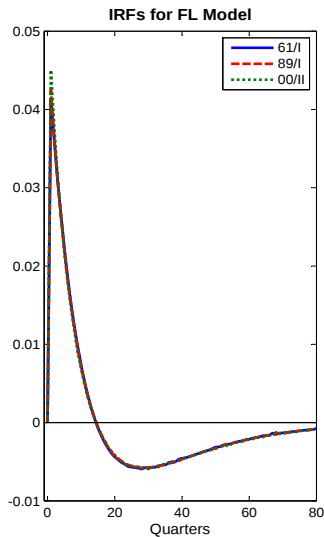
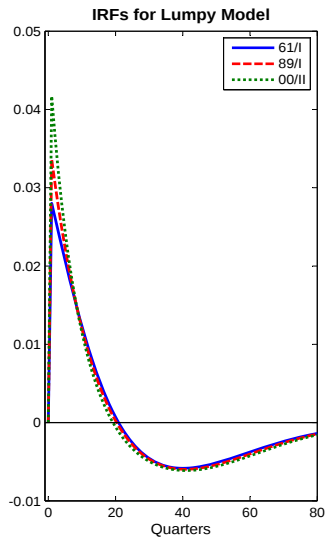
Should macroeconomists care?

- Yes? Maybe?

C. Time-varying IRF_0 implied by the **model**



C. Time-varying IRF implied by the **model**



Should macroeconomists care?

- Yes!

Statistical power: A Time-Series Approach

- Latest version of the paper manages to detect the type of time-varying IRF predicted by lumpy-adjustment models

Time-varying IRF: A Time Series Approach

- Let: $x_t \equiv I_t/K_{t-1}$
- Consider a GARCH model:

$$x_t = \phi x_{t-1} + \sigma_t e_t, \quad (1a)$$

$$\sigma_t = h(x_{t-1}, x_{t-2}, \dots) \quad (1b)$$

where $h(x_{t-1}, x_{t-2}, \dots)$ summarizes “recent investment” and the e_t are i.i.d. (0,1)

- Will have:

$$\text{IRF}_0 = \frac{\partial x_t}{\partial \varepsilon_t} = \sigma_t = h(x_{t-1}, x_{t-2}, \dots).$$

Time-Series Evidence on time-varying IRFs

- Use quarterly U.S. data, 1960.I–2005.IV, BLS.
- Use OLS to estimate

$$x_t = \phi x_{t-1} + e_t$$

- Two possibilities to estimate h :

- 1 Estimate via OLS:

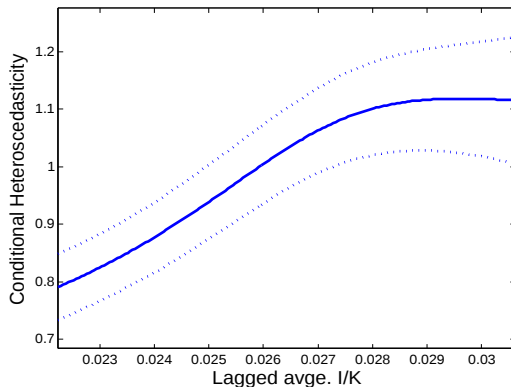
$$|\hat{e}_t| = \hat{\alpha}_0 + \hat{\alpha}_1 \bar{x}_{t-1}^k + \text{error},$$

- 2 Use a Gaussian kernel smoother and cross-validation to estimate $h(x_{t-1})$ based on $|\hat{v}_t|$ s

Time-Series Evidence on time-varying IRFs

	k						
	1	2	3	4	5	6	7
ϕ :	0.9599 (0.0299)	0.9599 (0.0300)	0.9599 (0.0298)	0.9599 (0.0302)	0.9599 (0.0300)	0.9599 (0.0295)	0.9599 (0.0296)
$10^3 \times \alpha_0$:	0.4828 (0.0351)	0.4849 (0.0353)	0.4823 (0.0355)	0.4822 (0.0363)	0.4805 (0.0386)	0.4820 (0.0390)	0.4837 (0.0397)
α_1 :	0.0209 (0.0172)	0.0239 (0.0173)	0.0282 (0.0172)	0.0300 (0.0174)	0.0330 (0.0179)	0.0336 (0.0179)	0.0331 (0.0182)
p -value ($\alpha_1 > 0$):	0.0507	0.0369	0.0222	0.0179	0.0170	0.0189	0.0210
No. obs. 1st regr.:	183	183	183	183	183	183	183
No. obs. 2nd regr.:	183	182	181	180	179	178	177

Time-Series Evidence on time-varying IRFs



- Q.: How do we determine whether lumpy investment is relevant for aggregate investment dynamics?
- A.: See whether lumpy adjustment affects the impulse response function, in particular, it it leads to a time-varying IRF
- ARCH-evidence suggests that this is the case: implied ratio over our sample of maximum and minimum IRF_0 is approximately 1.5
- Objective: build a DSGE model that delivers a time-varying IRF for aggregate investment similar to that obtained via time-series methods from the actual data

3.2. Model

- We follow closely Khan and Thomas (2005)
- Departures:
 - 1 sector specific productivity shocks
 - 2 maintenance investment

Production Units

- No entry or exit
- Unit's production function:

$$y_t = z_{t \in S, t \in I, t} k_t^\theta n_t^\nu.$$

with log-AR(1) shocks

- $\theta + \nu < 1$
- I.i.d. cost of adjusting capital, ξ , drawn from a $U[0, \bar{\xi}]$, measured in units of labor

Households

- A continuum of identical households with access to a complete set of state-contingent claims
- Felicity function:

$$U(C, N^h) = \begin{cases} \frac{C^{1-\sigma_c}}{1-\sigma_c} - AN^h, & \text{if } \sigma_c \neq 0 \\ \log C - AN^h, & \text{otherwise.} \end{cases}$$

- The intertemporal price:

$$p(\mathbf{z}, \mu) \equiv U_C(C, N^h) = C(\mathbf{z}, \mu)^{-\sigma_c}.$$

- The intratemporal price:

$$\omega(\mathbf{z}, \mu) = -\frac{U_N(C, N^h)}{p(\mathbf{z}, \mu)} = \frac{A}{p(\mathbf{z}, \mu)}.$$

Production Units: Bellman Equation

- Unit's problem:

$$V^1(\epsilon_S, \epsilon_I, k, \xi; \mathbf{z}, \mu) = \max_n \{ \mathbf{CF} + \max(V_i, \max_{k'} [-\mathbf{AC} + V_a]) \},$$

where

$$\begin{aligned} \mathbf{CF} &= [\mathbf{z} \epsilon_S \epsilon_I k^\theta n^\nu - \omega(\mathbf{z}, \mu) n - i^M] p(\mathbf{z}, \mu), \\ V_i &= \beta E[V^0(\epsilon'_S, \epsilon'_I, \psi(1 - \delta)k/\gamma; \mathbf{z}', \mu')], \\ \mathbf{AC} &= \xi \omega(\mathbf{z}, \mu) p(\mathbf{z}, \mu), \\ V_a &= -ip(\mathbf{z}, \mu) + \beta E[V^0(\epsilon'_S, \epsilon'_I, k'; \mathbf{z}', \mu')], \\ \mu &= \text{distribution of } (\epsilon_S, \epsilon_I, k). \end{aligned}$$

Recursive Equilibrium

A **recursive competitive equilibrium** is a set of functions

$$\omega, p, V^1, N, K', C, N^h, \Gamma$$

such that

- ① *Production unit optimality*: Taking ω , p and Γ as given, demand N and K'
- ② *Household optimality*: Taking ω and p as given, the household optimally chooses consumption C and labor N^h
- ③ *Commodity market clearing*
- ④ *Labor market clearing*
- ⑤ *Model consistent dynamics*: $\mu' = \Gamma(z, \mu)$.

Equilibrium Computation

- μ : infinite dimensional
- We follow Krusell and Smith:
 - approximate μ by its first moment over capital
 - approximate $\mu' = \Gamma(z, \mu)$ by a log-linear rule
- To simplify computations: $\rho_S = \rho_I$, the unit then only cares about $\epsilon \equiv \epsilon_S \epsilon_I$.

Mandated Investment

- We have:

$$k' = \begin{cases} k^*(\epsilon; \mathbf{z}, \bar{k}), & \text{if } \xi \leq \xi^T(\epsilon, k; \mathbf{z}, \bar{k}), \\ (1 - \delta + \chi\delta)k, & \text{otherwise.} \end{cases}$$

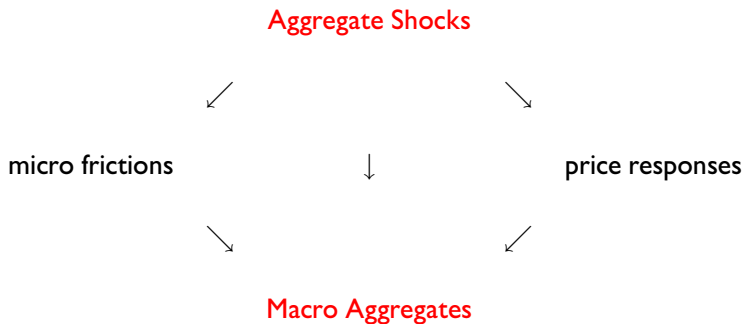
- We define **mandated investment** for a unit with current state $(\epsilon, \mathbf{z}, \bar{k})$ and current capital k as:

$$x(\epsilon; \mathbf{z}, \bar{k}) \equiv \log k^*(\epsilon; \mathbf{z}, \bar{k}) - \log[1 - \delta + \chi\delta]k.$$

3.3. Calibration

- A central element of this paper is its calibration strategy
- It differs in fundamental way from calibrations in the Thomas and Khan-Thomas papers
- This difference is relevant for calibration of macro models in general

Sources of Smoothing in Macroeconomics



Sources of Smoothing: Lumpy Investment Models

① Micro frictions $\equiv$ PE smoothing:

- it isn't only the size of adjustment costs
- **aggregation** is central
- Caplin and Spulber (1987) as an extreme example

② Price responses $\equiv$ GE smoothing:

- quasi labor supply
- supply of funds

Calibration in Khan-Thomas (2008)

- Khan and Thomas: choose parameters to various statistics of the distribution of establishment level adjustments in the LRD, for example, the fraction of establishments that adjust by more than 20%
- We argue that using plant level moments is not robust:
 - how many micro units in the **model** correspond to one **observed** micro unit?

New Calibration Strategy: Overview

- Match volatility of sectoral (3-digit) I/K instead of moments at the establishment level
- Mainly (assume only) partial equilibrium effects at this level
- Also match volatility of aggregate (entire economy) I/K
- And to capture time-varying IRF we match the **heteroscedasticity range**: $\pm \log(\sigma_{\max}/\sigma_{\min})$ with $+$ iff the max lies to the right of the min

Calibration: Two Approaches

Most parameters: standard values and/or values obtained from TFP data (when possible)

Two approaches:

1 EIS=1 approach:

- set $EIS = 1$
- attractive if you take the EIS in these model literally
- choose χ , $\bar{\xi}$ and σ_A to match the heteroscedasticity range and the volatility of sectoral and aggregate I/K

2 EIS estimated:

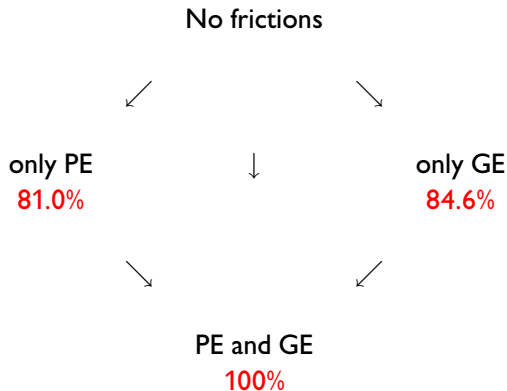
- for those that believe the EIS is a shortcut for many things in these models
- RBC model shortcomings for investment:
 - flatter supply of funds (who finances US investment?)
 - flatter labor supply?
- fix σ_A at values standard value

- Reasonable values for adjustment costs
- Very high EIS for the second approach
- Time-varying IRF in both cases (and very similar)
- Unless stated otherwise, in what follows we concentrate on the $EIS = 1$ case

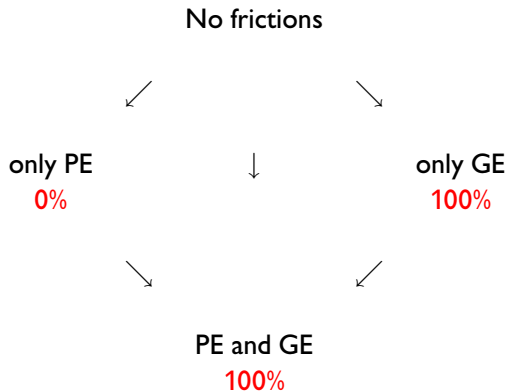
Results: Economic Magnitude of Adjustment Costs

Model	Tot. adj. costs/ Aggr. Output (1)	Tot. adj. costs/ Aggr. Investment (2)	Adj. costs/ Unit Output (3)	Adj. co Unit Wa (4)
Lumpy quarterly:	0.35%	2.41%	9.53%	14.8
Lumpy annual:	0.41%	2.84%	3.60%	5.62

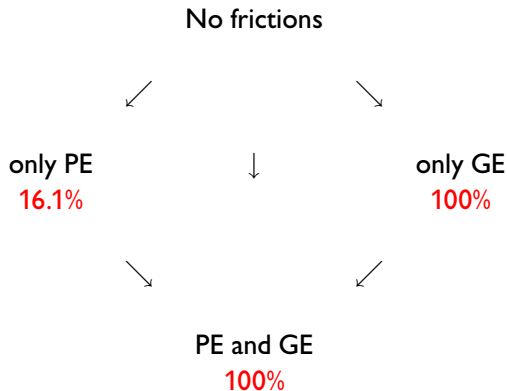
Smoothing and $\sigma(I/K)$: This Paper in a Nutshell



Smoothing and $\sigma(I/K)$: RBC



Smoothing and $\sigma(I/K)$: Khan and Thomas (2005)



Why the Difference?

Model	$\log(\sigma_{\max}/\sigma_{\min})$
<i>Data</i>	<i>0.3971</i>
This paper:	0.4008
Frictionless:	0.0767
Khan-Thomas (2008):	0.0998

Why the Difference?

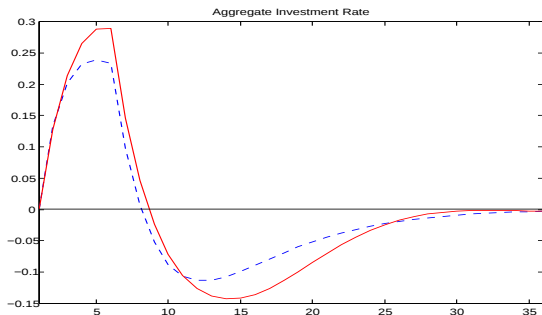
- When you calibrate a model you **implicitly** decide on the relative importance of PE and GE smoothing
- There are many combinations of PE and GE smoothing that achieve the same degree of aggregate smoothing
- We use 3-digit sectoral data to calibrate the relative importance of PE and GE smoothing: at this level of aggregation we mainly have PE-smoothing
- Also: the importance of matching aggregate IRF volatility

3.4. Aggregate Investment Dynamics

An Illustrative Exercise

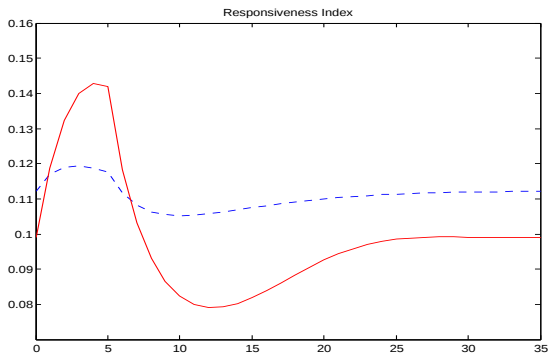
- Particular sample path of aggregate shocks:
 - 5 consecutive innovations of $2\sigma_A$
 - followed by no innovations
- Economy starts at steady-state
- Compare:
 - frictionless economy
 - our lumpy economy ($\chi = 0.50$)

Evolution of Aggregate Investment Rates



- Eventually the lumpy economy responds by more to further positive shocks
- Lumpy economy's larger boom comes with a more protracted slump

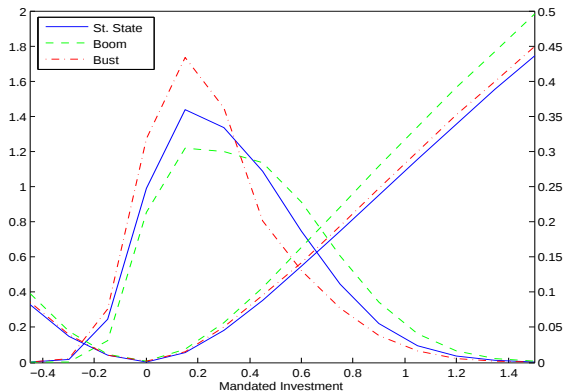
Evolution of the Responsiveness Index



- First element in IRF
- Conditional on history of shocks (summarized by μ)

- A decline in the strength of PE-smoothing explains the rise in the index during the boom phase
 - the responsiveness index fluctuates much less in the frictionless economy
 - frictionless economy only has GE-smoothing
 - hence: contribution of GE smoothing to fluctuations in responsiveness index of lumpy economy is small
- As the boom proceeds, the economy comes “closer” to the Caplin-Spulber scenario

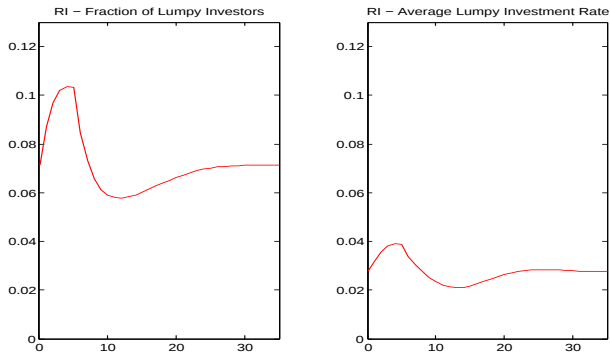
PE-Smoothing and the Boom-Bust Episode



Evolution of the Cross-section and Hazards

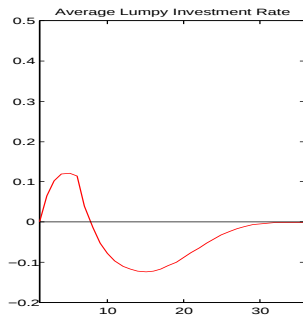
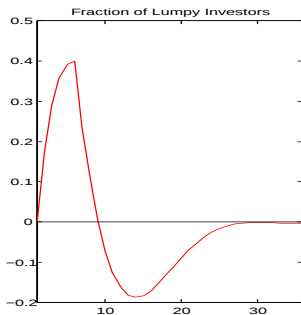
- Have the **increasing hazard** property
- During the boom:
 - the fraction of units with mandated investment close to zero decreases
 - the fraction of units with mandated investment above 40% increases
 - the fraction of units with negative mandated investment decreases
- During the boom the x-section moves into regions where the probability of adjusting is higher
 - this effect is not present in a frictionless (or Calvo) model

Decomposing the Responsiveness Index



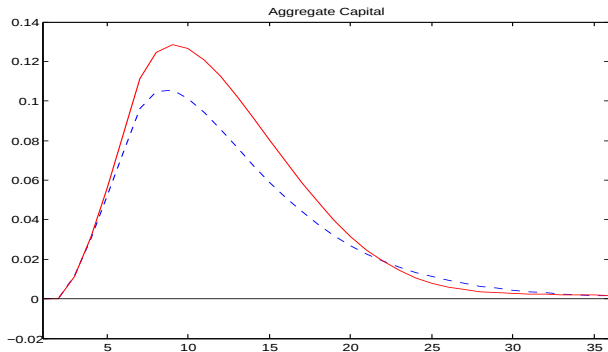
Fluctuations in responsiveness index driven mainly by variations in the **fraction** of units adjusting (**extensive** margin)

Decomposing Aggregate I/K



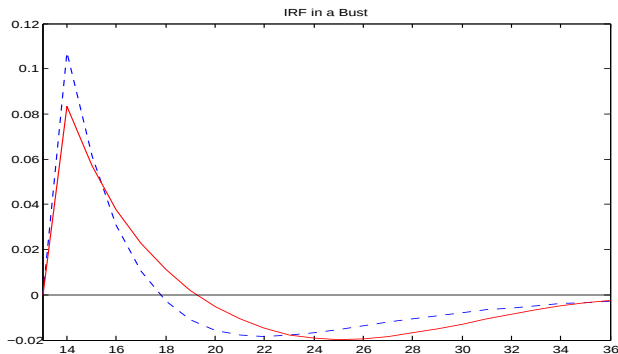
Doms and Dunne (1998): it's the **fraction** of units undergoing major investment episodes

Understanding the Bust



- More capital accumulation in the lumpy economy
- Large fraction in region where units are unresponsive to shocks (see previous figure)

Impulse Responses at the Trough



Lumpy economy less responsive to a positive stimulus

3.5. Conclusion

- The data suggest time-varying IRFs
- Lumpy adjustment DSGE models with mainly GE-smoothing forces cannot deliver history dependent IRFs
- Lumpy adjustment DSGE models where both PE and GE-smoothing are relevant deliver time-varying IRFs
- Bonus: reducing the burden of GE-smoothing leads to a substantial improvement in moment-matching of Y , C and N