

# Fewer but Better:

## Sudden Stops, Firm Entry, and Financial Selection

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- ▶ **Sudden stop:** Abrupt reduction in net capital flows into an economy.

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**Question:**

- ▶ **What is the contribution to productivity of the forgone entrants?**
  
  
  
  
  
  
  
  
  
  
- ▶ **Sudden stop:** Abrupt reduction in net capital flows into an economy.

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- ▶ Is the **mass-composition margin** quantitatively important for the macroeconomic consequences of a sudden stop?

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## 3. Quantitative analysis:

- ▶ Medium-run: amplification and persistence.
- ▶ Long-run: no heterogeneity doubles productivity loss, 30% larger consumption equivalent welfare cost.

## Persistent Effects of Crises:

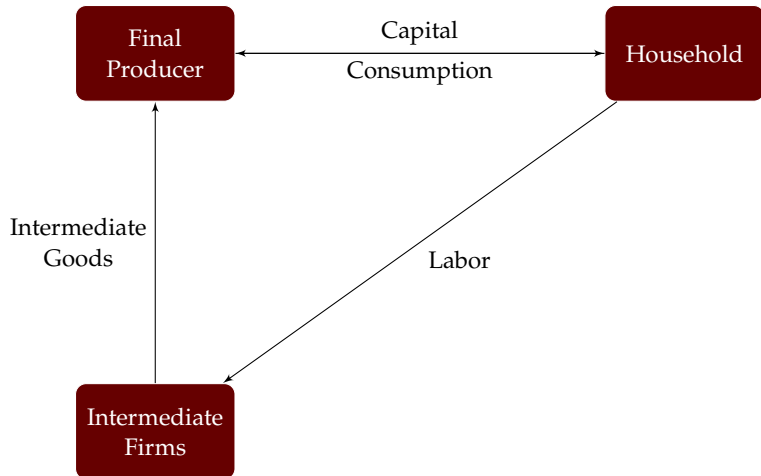
- ▶ **Motivation:** Cerra and Saxena (2008), Reinhart and Rogoff (2014).
- ▶ **Linking the short and long run:** Comin and Gertler (2006), Queralto (2014), Gornemann (2014), Guerron-Quintana and Jinnai (2014).

## Firm Heterogeneity and Entry:

- ▶ **Short run:** Bilbiie, Ghironi and Melitz (2012), Clementi, Khan, Palazzo and Thomas (2014).
- ▶ **Long run:** Akcigit and Kerr (2013).

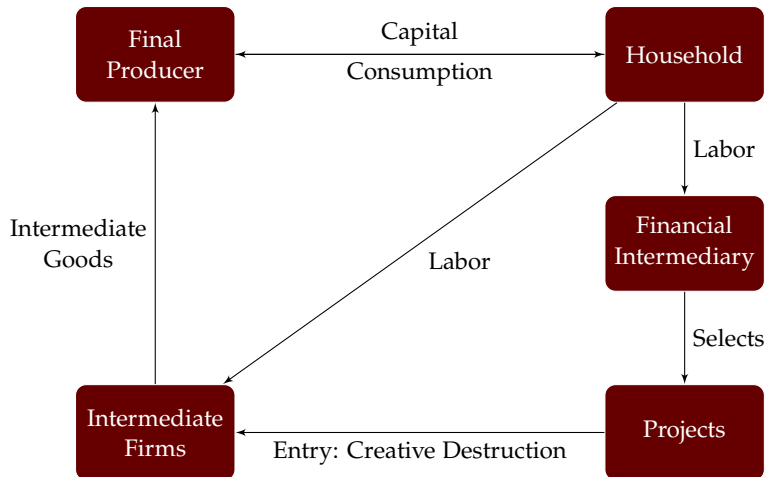
# Model

# Model Overview: Standard Components

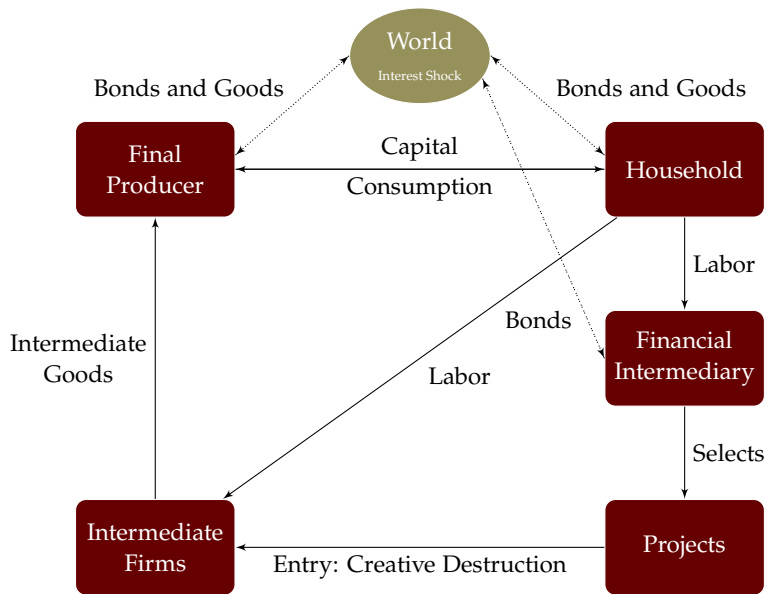




# Model Overview: Endogenous Technological Change



# Model Overview: Small Open Economy



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Every leader of type  $d \in \{H, L\}$  earns the same profits ( $\sigma^H > \sigma^L$ ).

# Selection Technology

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- ▶ **Implied proportion of  $H$ -type in the entrant cohort:**

$$\tilde{\mu}^{H_{\bar{z}^*}}(s^t) = \underbrace{\frac{1}{\nu + 1}}_{\text{Proportion if random selection}} \times \underbrace{\left[ 1 - \rho + \rho \frac{1 - (\bar{z}^*(s^t))^{\nu+1}}{1 - \bar{z}^*(s^t)} \right]}_{\geq 1 \text{ for } \nu > 0}$$

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$$\max_{\bar{z}(s^t)} \left\{ \underbrace{\lambda(1 - \bar{z}(s^t))}_{\text{Cohort's mass}} \times \left[ \underbrace{\tilde{\mu}^H(\bar{z}(s^t))V^H(s^t) + (1 - \tilde{\mu}^H(\bar{z}(s^t)))V^L(s^t)}_{\text{Cohort's expected value}} \right] - \underbrace{(1 - \bar{z}(s^t))R(s^t)W(s^t)\kappa}_{\text{Enaction cost}} \right\}$$

$\lambda$ : Entry probability.

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# Empirical Results

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### 3. **Exogeneous sudden stop.**

- ▶ August 1998: Russia defaulted on domestic debt and declared a moratorium on foreign creditors.
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Were firms born during the sudden stop *fewer, but better?*

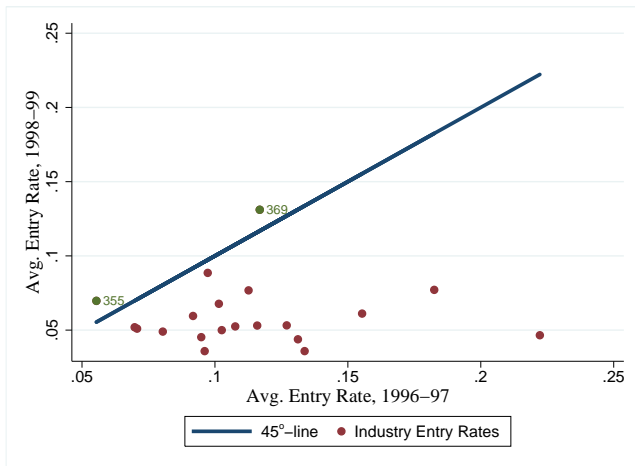


Figure: Mass (quantity)

- **Mass:** Firm entry drops on average by 40% during the crisis.

## Summary:

1. The average firm born during crisis is *9pp* more profitable.
2. The average firm born during crisis has higher labor productivity.
3. The average firm born during crisis accumulates capital faster.
4. The average firm born during crisis does not face higher exit risk.
5. Even during tranquil times, larger cohorts at the industry level are associated with lower average profitability.

# Quantitative Exploration

Parameter	Symbol	Value	Main identification
Patience parameter	$\beta$	0.9975	$\beta = (1 + a)^{\gamma} / \bar{R}$
Success probability	$\lambda$	5.36%	Entry rate
Enaction cost	$\kappa$	6.65%	Entry cost
Labor disutility level	$\Theta_l$	1.73	Working time
Screening accuracy	$\rho$	69.7%	Fast exit
Scarcity	$\nu$	4.51	Growth
Capital adjustment cost	$\phi$	20	Investment volatility

Target	Model	Data	Expression
Entry rate	2.71%	2.71%	$\lambda (1 - \bar{z})$
Entry Cost	12.1%	12.1%	$\kappa(w/y)$
Working time	33.0%	33.0%	$L$
Fast exit	15.0%	15.0%	$(1 - \rho)\bar{z}$
Growth	0.62%	0.62%	$a = \left( (1 + \sigma^H)^{\mu^H} (1 + \sigma^L)^{1 - \mu^H} \right)^{\lambda(1 - \bar{z})} - 1$

- ▶ Ext. Calibrated Parameters: In accordance with SOE-RBC literature.

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# Validating the Model

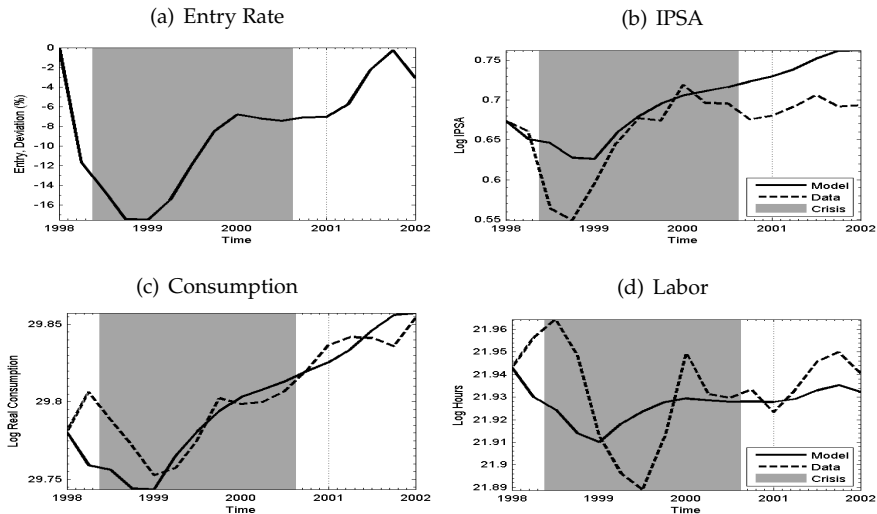


Figure: Non Targeted Macro Series

Two additional models for comparison with baseline model (*Base*):

## 1. Exogenous Growth Model (*Exo*):

- ▶ No entry: productivity grows at a constant and exogenous rate.
- ▶ No long-run cost: return to the original path after a shock.

## 2. No Heterogeneity Model (*NoHet*):

- ▶ Only one step size: homogeneous entry, no selection.
- ▶ Long-run cost: back to a parallel but lower path.

**Same common parameters and same long run growth.**

# IRF: Mass and Composition

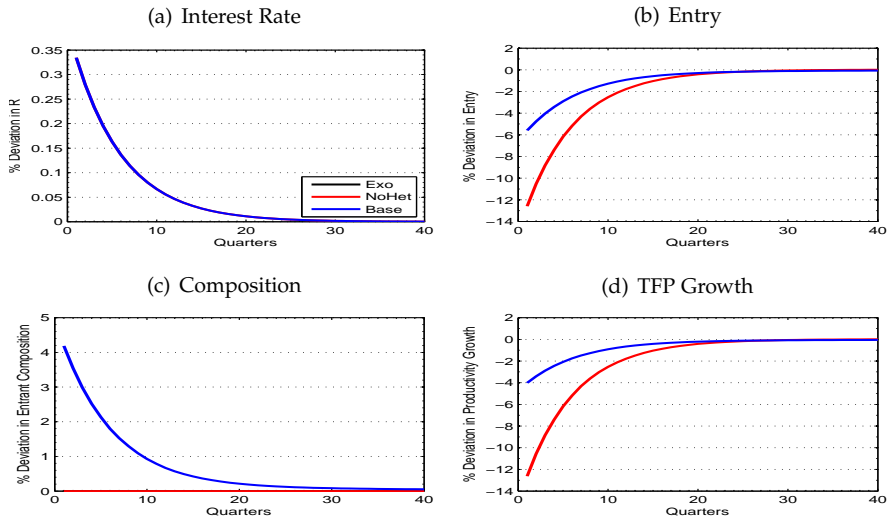


Figure: 33 basis point increase in interest rate.

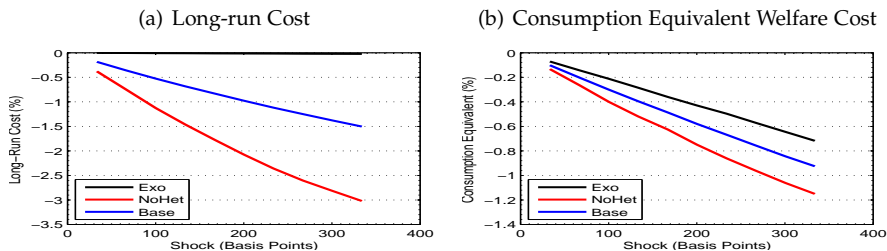


Figure: The Impact of Selection.

1. *Exo*: No long-run cost by construction.
2. *Base* versus *Exo*: Long-run cost is 1/3 of welfare cost.
3. *NoHet* versus *Base*: 2 times higher long-run cost. 30% higher welfare cost.

# Financial Development

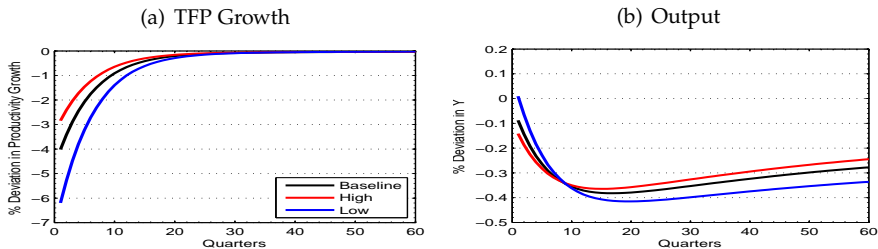


Figure: Financial Development.

1. i) **Baseline**:  $\rho = 70\%$  ii) **High**:  $\rho = 91\%$  iii) **Low**:  $\rho = 49\%$ .
2. Lower  $\rho$ : more medium-run amplification and persistence, higher long-run cost and lower short-run impact.

1. Tractable framework for studying heterogeneity and financial selection in a dynamic stochastic small open economy model.
2. Firm level evidence of novel *mass-composition* trade-off: Cohorts born during the Chilean sudden stop were *fewer, but better*.
3. Heterogeneity and selection are quantitatively important:
  - 3.1 No heterogeneity doubles the long-run cost.
  - 3.2 No heterogeneity increases the welfare cost by 30%.
4. Financial development introduces a trade-off between short-run impact, and long-run cost.

# APPENDIX



Intermediate inputs and capital are combined to produce the final good:

$$\ln Y(s^t) = \alpha \int_0^1 \ln X_j^D(s^t) dj + (1 - \alpha) \ln K^D(s^{t-1})$$

Working capital constraint on intermediate goods.

$$\max_{K(s^{t-1}), \{X_j^D(s^t)\}_{j \in [0,1]}} \left\{ Y(s^t) - \underbrace{\left( 1 + \eta(R(s^t) - 1) \right)}_{\text{Cost wedge}} \int_0^1 X_{j,t}^D p_j(s^t) dj - K^D(s^{t-1}) r(s^t) \right\}$$

Demand for variety  $j$ :

$$X_j^D(s^t) = \frac{\alpha Y(s^t)}{p_j(s^t) (1 + \eta(R(s^t) - 1))} \equiv \frac{\Gamma(s^t)}{p_j(s^t)}$$

- ▶  $\eta$ : Fraction of intermediate expenditure to be held as working capital.
- ▶ Final good is the numeraire.

Value of a type  $d$  product line:

$$V^d(s^t) = \underbrace{(1 - \tau)\Pi^d(s^t)}_{\text{After-tax Profits}} + E \left[ \underbrace{m(s^t, s_{t+1})}_{\text{Stochastic Discount}} \underbrace{(1 - \lambda M(s^t, s_{t+1}))}_{\text{Survival Probability}} V^d(s^t, s_{t+1}) | s^t \right]$$

$\tau$ : Corporate tax.

$M(s^t, s_{t+1})$ : Mass of projects enacted.

$\lambda$ : Entry probability.

$m(s^t, s_{t+1})$ : Stochastic discount factor of the household.

$$\max_{\{B(s^t), C(s^t), L(s^t), I(s^t)\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \sum_{s^t} \beta^t \pi(s^t) \frac{1}{1-\gamma} \left( C(s^t) - \Theta_l A(s^t) (L(s^t))^{\nu} \right)^{1-\gamma}$$

subject to:

$$C(s^t) \leq W(s^t)L(s^t) + r(s^t)K(s^{t-1}) + B(s^{t-1})R(s^{t-1}) + T(s^t) - I(s^t) - B(s^t) - \psi(\bullet)$$

where

$$I(s^t) = K(s^t) - (1-\delta)K(s^{t-1}) + \Phi(\bullet)$$

As in **Neumeyer and Perri (2005)**:

- ▶ Preferences: Greenwood, Hercowitz, and Huffman (1988).
- ▶ Bond holding costs:

$$\Psi(B(s^t), Y(s^t)) = \frac{\psi}{2} Y(s^t) \left( \frac{B(s^t)}{Y(s^t)} - \bar{b} \right)^2$$

- ▶ Capital adjustment costs:

$$\Phi(K(s^{t-1}), K(s^t)) = \frac{\phi}{2} K(s^{t-1}) \left[ \frac{K(s^t)}{K(s^{t-1})} - (1+\bar{g}) \right]^2$$

1. Representative household as in Neumeyer and Perri (2005)
2. The interest rate  $R(s^t)$ :

$$\ln \left( \frac{R(s^t)}{\bar{R}} \right) = \rho_R \ln \left( \frac{R(s^{t-1})}{\bar{R}} \right) + \sigma_s \epsilon(s^t)$$

3. Net exports:

$$NX(s^t) = Y(s^t) - C(s^t) - I(s^t) - \Psi(\bullet)$$

4. Debt position of the country:

$$D(s^t) = B(s^{t-1}) \underbrace{- \eta H(s^t) - (1 - \bar{z}(s^t)) \kappa W(s^t)}_{\text{Working Capital and Project Enaction}}$$

5. Composition of the intermediate good producers:

$$\mu(s^t) = \mu(s^{t-1}) + \lambda(1 - \bar{z}(s^t)) \left( \tilde{\mu}(\bar{z}(s^t)) - \mu(s^{t-1}) \right)$$

Let a lower case variable, e.g.  $e(s^t) = \frac{E(s^t)}{A(s^t)}$ , denote normalized variables where

$$\ln(A(s^t)) \equiv \int_0^1 \ln q_j(s^t) dj.$$

This transformation renders the model stationary.

## Equilibrium

A competitive equilibrium for this small open economy, given initial conditions:

1. Households optimally choose  $\{c(s^t), b(s^t), k(s^t), L(s^t)\}$ .
2. Final good producers optimally choose  $\left\{ \left\{ x_j^D(s^t) \right\}_{j \in [0,1]}, k^D(s^{t-1}) \right\}$ .
3. Intermediate good producers optimally choose  $\left\{ x_j(s^t), p_j(s^t), L_j(s^t) \right\}_{j \in [0,1]}$ .
4. Financial intermediary optimally chooses  $\{\bar{z}(s^t)\}$ ;
5. Government budget is balanced every period.
6. Labor, asset, capital, final and intermediate good markets clear.
7.  $\{q_j(s^t), v_j(s^t)\}_{j \in [0,1]}$  and  $\{\mu(s^t), \tilde{\mu}(s^t)\}$  evolve according to their law of motion.

- ▶ **Superstar:** one standard deviation above average  $P_t = \frac{\text{Revenue}_t - \text{Cost}_t}{\text{Revenue}_t}$ .

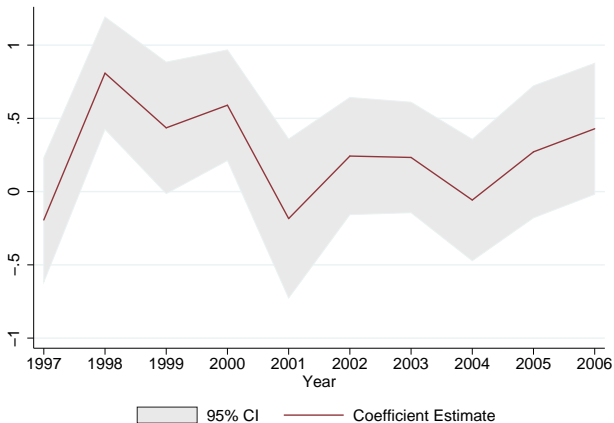
$$\Pr(\text{Superstar} = 1 | \text{age} = 1) = \frac{e^{x'_i \beta}}{1 + e^{x'_i \beta}} \quad \text{where} \quad x'_i \beta = \alpha + \alpha_j + \alpha_r + \beta \ln(L_{i,0}) + \gamma_{\text{cohort}} + u_{i,t}$$

- **Superstar:** one standard deviation above average  $P_t = \frac{Revenue_t - Cost_t}{Revenue_t}$ .

$$Pr(\text{Superstar} = 1 | \text{age} = 1) = \frac{e^{x'_i \beta}}{1 + e^{x'_i \beta}} \quad \text{where} \quad x'_i \beta = \alpha + \alpha_j + \alpha_r + \beta \ln(L_{i,0}) + \gamma_{\text{cohort}} + u_{i,t}$$

	(1)	(2)	(3)	(4)	(5)
	Superstar at age 1			Superstar at age 0	Superstar at age 2
Crisis Dummy	0.540*** (0.110)			0.295*** (0.0970)	0.312** (0.135)
In Crisis		0.697*** (0.134)			
After Crisis		0.240* (0.126)			
entry <sub>j,0</sub>			-1.575** (0.803)		
ln(L <sub>i,0</sub> )	0.222*** (0.0527)	0.216*** (0.0526)	0.209*** (0.0521)	0.146*** (0.0436)	0.153** (0.0605)
Observations	3197	3197	3197	4220	2618

Standard errors in parentheses, bootstrapped (250), \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$



**Figure:** Logit Estimation by Cohort



## Main Equation:

$$\text{Profitability}_{i,t} = \alpha + \beta X_{i,t} + \gamma Z_i + \bar{\gamma} \text{Born in Crisis} + \mu_i + \epsilon_{i,t}$$

- ▶  $X_{i,t}$ : Firm level variables (e.g: size) and macro variables (e.g: unemployment).
- ▶  $Z_i$ : Initial conditions (e.g: size at entry) and industry and region controls.
- ▶ *Born in Crisis* is 1 if the firm was born in 1998 – 2000. Main focus is to estimate  $\bar{\gamma}$ .

## Estimation by Hausman and Taylor (1981).

- ▶  $\text{profitability} = \frac{\text{Revenue} - \text{Cost}}{\text{Revenue}}$

# Fewer but Better: 9 p.p. more profitable

	(1)	(2)	(3)	(4)	(5)	(6)
	$P_{i,t}$	$P_{i,t}$	$P_{i,t}$	$P_{i,t}$	$\log \frac{Y_{i,t}}{L_{i,t}}$	$\frac{K_{i,t} - K_{i,t-1}}{K_{i,t}}$
Crisis dummy	0.0877** (0.0423)			0.0814*** (0.0313)	0.325** (0.136)	0.0527** (0.0233)
In Crisis		0.0861** (0.0397)				
After Crisis		0.00952 (0.0241)				
avg. Entry $_{j,t_0}$			-0.682** (0.337)			
Relative effect at means	-31.2%	-31.3%	—	-28.4%	-32.5%	-29.2%
Sargan-Hansen (p)	0.4545	0.2333	0.1230	0.0476	0.0395	0.7702
Observations	16834	16834	16834	16371	15583	16388

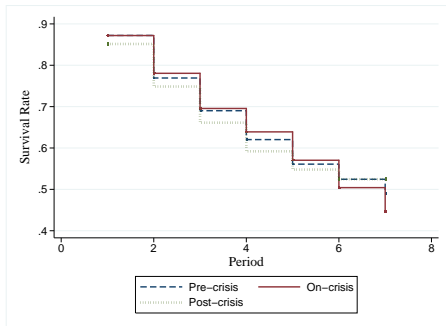
Standard errors in parentheses (bootstrapped (250), clustered by firm)

Controls: Macro controls (unemp. manuf. prod. labor cost, PPI), Elec. cons., labor, capital, age, initial HHI, initial workers, industry, and geography.

Regressions (4) and (5) use initial capital to control for entry size instead of workers.

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$

$$h_{mn}(t, X_i) = h_{0mn}(t) \exp[\beta_1 \log(\text{elec}_{it}) + \beta_2 \log(\text{worker}_{it}) + \beta_3 \log(\text{worker}_{i0}) + \beta_4 \log(\text{elec}_{i0}) + \beta_5 \log(\text{prft}_{jt}) + \gamma \cdot \text{industry}]^1$$

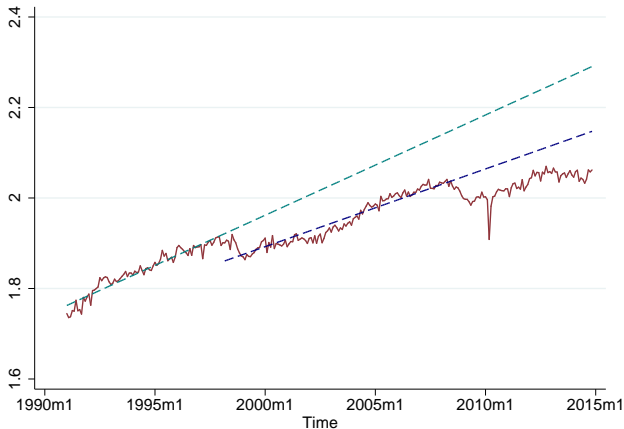


1. Proportional hazard model is not rejected.
2. **Firms born in crisis do not die more: *ex ante* Selection!**

<sup>1</sup>Stratified by region ( $m$ ) and period ( $n$ )

Parameter	Symbol	Value	Source
Capital share	$1 - \alpha$	0.32	Mendoza (1991)
Elasticity of Substitution ( $1/\gamma$ )	$\gamma$	2	Mendoza (1991)
Frisch Elasticity ( $1/(1 - \chi)$ )	$\chi$	1.455	Mendoza (1991)
Working Capital	$\eta$	1	Neumeyer and Perri (2005)
Debt adjustment cost	$\psi$	0.0001	Low
Depreciation rate	$\delta$	1.94%	Bergoeing et al (2002)
Corporate tax rate	$\tau$	0.17	Data
Long-run interest rate	$\bar{R}$	1.015	Chilean Central Bank Data
Persistence of interest rate	$\rho_r$	0.836	Chilean Central Bank Data
Dispersion of interest rate shock	$\sigma_r$	0.33%	Chilean Central Bank Data
Long-run debt to GDP ratio	$\bar{b}$	$4 * (-0.44)$	Chilean Central Bank Data
Low profitability ( $\sigma^L / (1 + \sigma^L)$ )	$\sigma^L$	14.5%	ENIA
High profitability ( $\sigma^H / (1 + \sigma^H)$ )	$\sigma^H$	55.5%	ENIA

- ▶ In accordance with SOE-RBC literature.



**Figure:** Monthly Manufacturing Production (log)