

Aggregate Dynamics in Lumpy Economies

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Overview

- How do **aggregate shocks propagate** in economies with ...
 - Heterogeneous agents
 - Lumpiness = Inaction + Large adjustment
- We develop a **sufficient statistic** approach:
 1. Propagation = $f(\text{ergodic moments})$
 2. Ergodic moments = $g(\text{microdata on adjustments})$

Overview

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 - Heterogeneous agents
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- We develop a **sufficient statistic** approach:
 1. Propagation = $f(\text{ergodic moments})$
 - (1) $\text{Var}[\text{capital/productivity}]$
 - (2) $\text{Cov}[\text{capital/productivity, time elapsed since last adjustment}]$
 2. Ergodic moments = $g(\text{microdata on adjustments})$
 - Plant-level investment data recovers $\text{Var}[\cdot]$ and $\text{Cov}[\cdot]$
- **Investment dynamics** following an aggregate productivity shock

Simplified Environment in Investment

- Continuum of firms, steady state
- Firm's state: capital gap (unobservable)

$$x_t \equiv \log(\text{capital}_t/\text{tfp}_t) - \underbrace{\mathbb{E}[\log(\text{capital}/\text{tfp})]}_{\text{cross-sectional mean}}$$

- Between adjustments, capital gap follows:

$$dx_t = -\nu dt + \sigma dW_t$$

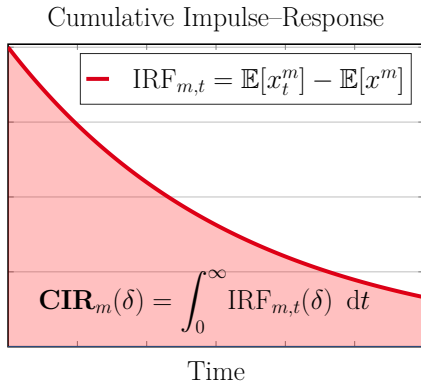
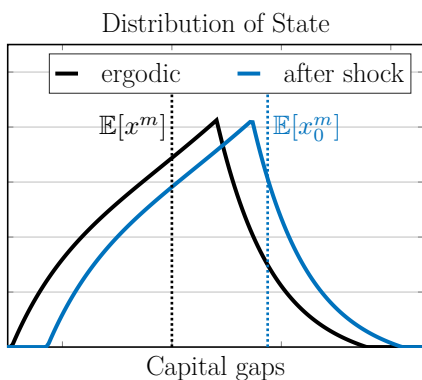
- Firm policy: reset capital gap to \hat{x} at dates $\{T_k\}_k$ if
 - Capital gap is too large (Ss policy)
 - Or/and at random *i.i.d.* dates (Calvo, sticky info)

- Observable actions in panel data: $(\Delta x, \tau, a)$

- Adjustment size (investment): $\Delta x_k = \hat{x} - x_{T_k^-}$
- Duration of completed spells: $\tau_k = T_k - T_{k-1}$
- Duration of uncompleted spells (age): $a_{tk} = t - T_{k-1}$

Propagation of Aggregate Shocks

- Unanticipated decrease in productivity of size δ (small) for all firms



- Cumulative Impulse-Response:** area under IRF
 - Real output ($m = 1$), capital misallocation ($m = 2$)
 - Measure of impact & persistence in one scalar

Propagation = f (ergodic moments)

Propagation = $f(\text{ergodic moments})$

- **Input:** Stochastic process between adjustments + firm policy
 - Assume firms use steady-state policy
- **Output:** CIR as linear combination of two ergodic moments

$$\frac{\text{CIR}_m(\delta)}{\delta} = \frac{\mathbb{E}[x^{m+1}] + \nu \text{Cov}[x^m, a]}{\sigma^2} + o(\delta)$$

- Ergodic moments encode agents' **responsiveness** to shocks
 - ⇒ Information about speed of convergence

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- Ergodic moments encode agents' **responsiveness** to shocks
 - \Rightarrow Information about speed of convergence
- Real output response ($m = 1$):

$$\frac{\text{CIR}_1(\delta)}{\delta} = \frac{\text{Var}[x] + \nu \text{Cov}[x, a]}{\sigma^2} + o(\delta)$$

$\Rightarrow \nu = 0 : \text{CIR}_1(\delta)/\delta \approx \text{Var}[x]/\sigma^2$

$\Rightarrow \nu \neq 0 : \text{Cov}[x, a]$ corrects dispersion due to drift

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- **Result:** Ergodic moments and parameters (ν, σ^2)

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i. **Drift:** frequency \times size of investment

$$\nu = \frac{\mathbb{E}[\Delta x]}{\mathbb{E}[\tau]}$$

ii. **Volatility:** frequency \times dispersion of investment (adjusted for drift)

$$\sigma^2 = \frac{\mathbb{E}[\Delta x^2]}{\mathbb{E}[\tau]} - 2\nu\hat{x}$$

iii. **Reset capital gap:** $\hat{x} = \frac{\mathbb{E}[\Delta x]}{2} (1 - \mathbb{CV}^2[\tau]) + \frac{\text{Cov}[\tau, \Delta x]}{\mathbb{E}[\tau]}$

- Set $\hat{x} = 0$ for today

Ergodic moments = $g(\text{microdata})$

- Variance of capital gap

$$\mathbb{V}ar[x] = \frac{1}{3} \frac{\mathbb{E} [\Delta x^3]}{\mathbb{E} [\Delta x]}$$

- Covariance between capital gap and age

$$\mathbb{C}ov[x, a] = \frac{1}{2\nu} \left\{ \mathbb{V}ar[x] + \sigma^2 \mathbb{E}[a] - \frac{\mathbb{E}[\tau \Delta x^2]}{\mathbb{E}[\tau]} \right\}$$

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- Large investment rate \implies Large capital misallocation
- Dispersed investment rate \implies Indicative of Calvo adjustments

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$$\mathbb{C}ov[x, a] = \frac{1}{2\nu} \left\{ \underbrace{\mathbb{V}ar[x] + \sigma^2 \mathbb{E}[a]}_{>0} - \underbrace{\frac{\mathbb{E}[\tau \Delta x^2]}{\mathbb{E}[\tau]}}_{<0} \right\}$$

- If depreciation ν is large $\implies \mathbb{C}ov[x, a] < 0$
- If volatility σ/ν is large + irreversibility $\implies \mathbb{C}ov[x, a] > 0$

$$\text{Propagation} = f(g(\text{microdata}))$$

Main Contribution

- Alvarez, Le Bihan and Lippi (2016)'s environment:
 - Large set of price-setting models
 - Zero drift + Symmetric policies + $m = 1$
 - x = markup gap, Δx = price change

Output effect of a money shock

$$\underbrace{\frac{\text{CIR}_1(\delta)}{\delta}}_{\text{output dynamics}} = \underbrace{\frac{\mathbb{E}[\tau] \mathbb{K}ur[\Delta x]}{6}}_{\frac{\text{kurtosis of price changes}}{\text{freq. of price changes}}}$$

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Output effect of a money shock

$$\underbrace{\frac{\text{CIR}_1(\delta)}{\delta}}_{\text{output dynamics}} = \underbrace{\frac{\mathbb{V}[x]}{\sigma^2}}_{\text{markup dispersion}} = \underbrace{\frac{\mathbb{E}[\tau]\mathbb{K}ur[\Delta x]}{6}}_{\frac{\text{kurtosis of price changes}}{\text{freq. of price changes}}}$$

- By connecting to ergodic moments, we allow for:
 - Non-zero drift, asymmetries, $m \geq 1$, mean-preserving spreads...
- Sufficient statistics across fields in economics

Bringing the Theory to the Data

Investment fluctuations

- **Input:** Plant-level investment data from Chile (structures)
- **Output:** Parameters and ergodic moments

Parameters	Value	Interpretation
ν	0.11	6% depreciation + 5% (price + growth)
σ^2	0.08	In line w/Bachmann, Caballero, Engel (2013)
Ergodic moments		
$\mathbb{V}[x]$	0.23	Calvo (standard Ss model 0.002)
$\mathbb{Cov}[x, a]$	0.90	Fixed Costs + Irreversibility
Propagation ($\delta = -1\%$)		
CIR_1	4.3	Cumulative capital drop of 4.3%
Half-life	3	3 years

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Our sufficient statistics are powerful devices for **discriminating between models** and **retrieving primitives** in lumpy economies