

Chapter 22: Heterogenous Firms

Graduate Macroeconomics Slides

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Outline

- Introduction
- A Simple Model of Heterogeneous Firms
- Firm Heterogeneity in the Data
- Reallocation and Misallocation
- Firm Heterogeneity in General Equilibrium
- Alternative Market Arrangements
- Business Cycles and Heterogeneous Firms
- Endogenous Productivity
- Conclusion



Motivation

Classical assumption: Aggregate production function $Y = F(K, L)$ and homogeneous firms.

Reality: Firms vary in size, age, productivity, growth, etc.

- Large vs. small firms
- Young vs. old
- Productive vs. unproductive

Question: How does firm heterogeneity affect macroeconomic outcomes?



Motivation (continued)

Policy relevance:

- Entry barriers: Should we encourage or discourage new firms?
- Growth policies: How to support fast-growing, more productive firms?
- Dominance of mega-firms: Effects on competition, labor share, and business cycles.
- Decline in business dynamism: Slowing entry and reallocation across firms.

Aggregate production function alone cannot capture these nuances.



Key Topics of the chapter

1. A Simple Model of Heterogeneous Firms
2. Empirical Evidence: U.S. Firm-Level Data
3. Reallocation and Misallocation
4. A General Equilibrium Model with Firm Heterogeneity
5. Alternative Market Arrangements (Monopolistic competition, Oligopoly)
6. Business Cycle Implications
7. Endogenous Productivity (Klette-Kortum model)



Setup: Basic Structure

- A continuum of firms $i \in [0, 1]$.
- Production function of firm i :

$$y_i = a_i F(x_i)^\gamma, \quad \gamma \in (0, 1).$$

- a_i is firm-specific productivity (heterogeneous).
- $F(x_i)$ is constant returns to scale, but $F(x_i)^\gamma$ has decreasing returns overall.



Optimal Input Allocation

Firm's cost-minimization (two steps):

1. Minimize cost subject to $F(x_i) = 1$.

$$\min_{x_i} \{p \cdot x_i : F(x_i) = 1\} \implies x_i^* \quad \text{and} \quad c \equiv p \cdot x_i^*$$

2. Scale of production:

$$\max_{m_i} \{a_i m_i^\gamma - c m_i\}.$$

Key result: Equilibrium aggregate production can be written as

$$Y = AF(X)^\gamma,$$

where A depends on the distribution of $\{a_i\}$.



Role of Dispersion in a_i

- If $\ln(a_i)$ is lognormally distributed with variance σ^2 , then

$$A = \exp\left(\nu + \frac{\gamma}{1-\gamma} \frac{\sigma^2}{2}\right).$$

- Higher σ^2 (greater dispersion) \implies higher A .
- **Intuition:** Productive firms (in the right tail of a_i) get allocated more inputs, boosting aggregate productivity beyond the mean.



Firm Size Distribution (U.S.)

- Over 5 million firms in the U.S.
- Most firms are very small (1–4 employees).
- Large firms (over 1,000 employees) still number in thousands.

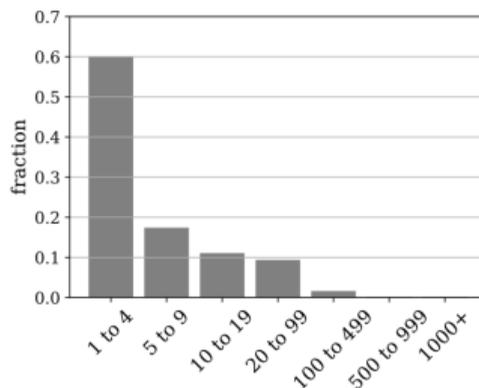


Figure 22.1: Distribution of firm size in 2022.

Source: Business Dynamics Statistics.

Figure: Hypothetical representation of the 2019 BDS firm size distribution. (Replace with actual figure if desired.)



Employment Share by Firm Size

- Despite many small firms, large firms account for a substantial share of total employment.
- Firms with 1,000+ employees cover roughly half of U.S. employment.

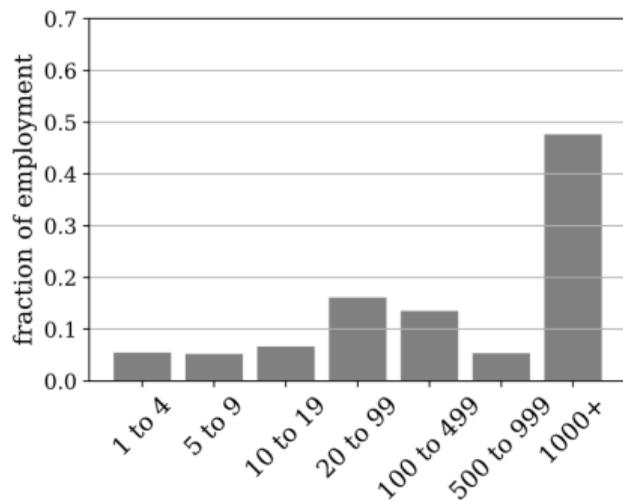


Figure 22.2: Employment share of each size category in 2022.

Source: Business Dynamics Statistics.



Business Dynamism: Job Creation and Destruction

- Large annual turnover rates (job creation and job destruction).
- **Cyclical**: During recessions, job destruction rises, job creation falls.
- **Declining trend** in both since the 1980s: the so-called “decline in business dynamism”.

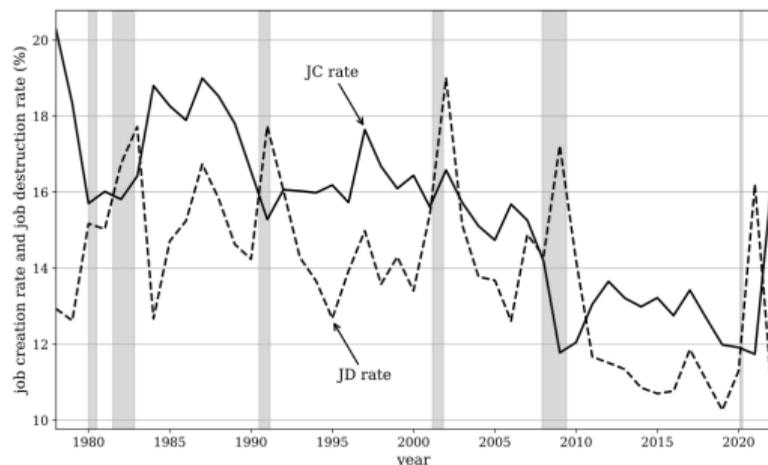


Figure 22.5: Annual job creation and destruction rates (establishment level).

Source: Business Dynamics Statistics.



Entry and Exit Rates

- Similar cyclical pattern and declining trend in establishment entry/exit rates.
- \implies Less reallocation activity across businesses.

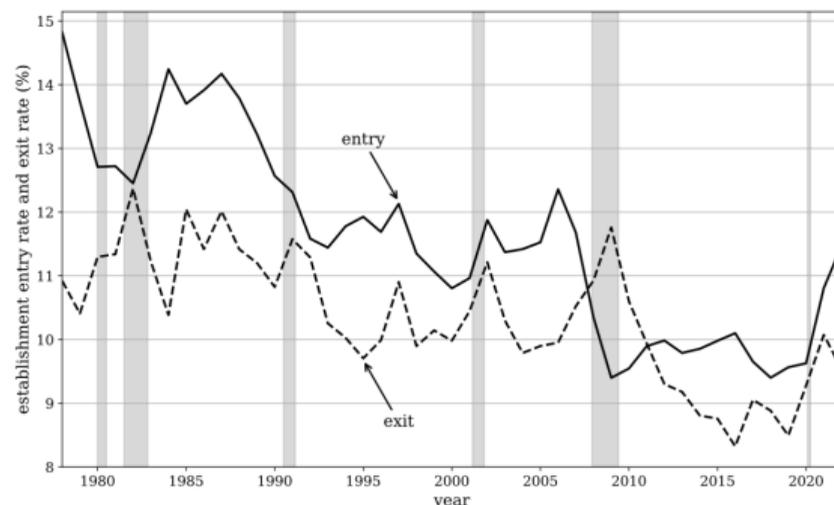


Figure 22.6: Annual establishment entry and exit rates.

Source: Business Dynamics Statistics.

Figure: Establishment entry and exit rates. (BDS)



Rise of Mega-Firms & Market Power

- Large firms (10,000+ employees) are growing in share (now $\sim 30\%$ of total employment).
- Empirical evidence of increasing average markups since the 1980s (De Loecker et al., 2020).

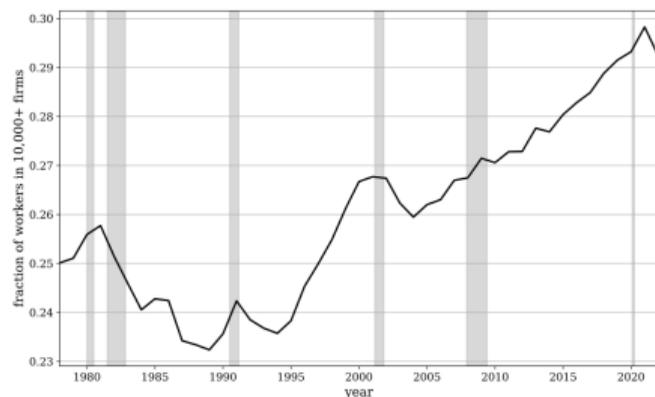


Figure 22.7: Fraction of employees working in firms with 10,000+ employee.

Source: Business Dynamics Statistics.

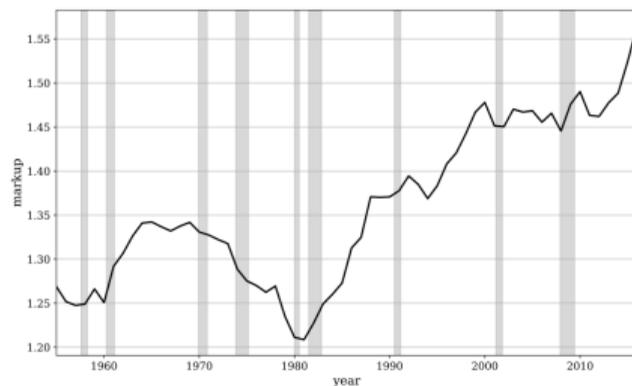


Figure 22.8: Markup of the U.S. public firms.

Source: De Loecker et al. (2020).

Figure: Fraction of employees at 10,000+ firms (BDS data) and average markups.



Reallocation & Productivity

(Foster et al., 2001): Decomposition

$$\Delta \bar{A}_t = \sum_{i \in C} s_{i,t-1} \Delta a_{it} + \sum_{i \in C} (a_{i,t-1} - \bar{A}_{t-1}) \Delta s_{it} + \dots$$

- Reallocation effects (shifts in market shares) account for **55%** of manufacturing productivity growth in U.S. (1977–1987).
- Highlights the importance of how resources move across firms.



Misallocation

- If distortions (e.g., taxes/subsidies) vary by firm, resource allocation is suboptimal.
- Example: (Restuccia and Rogerson, 2008), (Hsieh and Klenow, 2009).
- Leads to lower aggregate A (“misallocation”).

Simple illustration

$$\max_{m_i} \sum_i (1 - \tau_i) a_i m_i^\gamma - c m_i.$$

Distortions τ_i shrink aggregate output if they systematically reduce m_i for the most productive a_i .



(Hopenhayn and Rogerson, 1993) Framework

Key elements:

- Continuum of firms, each with idiosyncratic productivity a_t .
- Firing tax $\tau \max(0, l_{t-1} - l_t)$.
- Fixed operation cost c_f , free entry with cost c_e .
- Block-recursive structure: wage w determined from free-entry + labor supply.



Stationary Equilibrium

Firm problem:

$$W(a, l_{t-1}) = \max_{l_t} [a l_t^\gamma - w l_t - c_f - \tau \max(0, l_{t-1} - l_t)] + \beta \mathbb{E}[W(a', l_t)].$$

- Free entry: $W^e = c_e + \kappa$ (policy cost).
- In equilibrium, w solves labor market clearing with $L_d(w) = L_s(w)$.



Quantitative Effects of Firing Taxes

- Calibrate $\tau = 0$ baseline to match U.S. data (exit rate, size distribution).
- Increase τ from 0 to 0.1, 0.2 (equivalent to months of wages).
- **Outcome:** Lower employment, lower labor productivity, reduced job creation and destruction.

	$\tau = 0$	$\tau = 0.1$	$\tau = 0.2$
w	1.00	0.977	0.957
Total output	100	97.7	95.7
Total employment	100	98.3	97.4
Labor productivity	100	99.4	98.3
JC rate(%)	28	25	21



Effect of Entry Barriers

- Introduce extra entry cost κ .
- Raises firm value \implies lowers wage \implies more low-productivity firms survive \implies lowers aggregate productivity.

	$\kappa = 0$	$\kappa = 0.5$	$\kappa = 5.0$
w	1.00	0.986	0.879
Total output	100	98.6	87.9
Total emp.	100	99.5	96.2
Labor productivity	100	99.1	91.4
JC rate(%)	28	28	28



Monopolistic Competition

- Differentiated intermediate goods, final good has CES aggregator.
- With elasticity of substitution $\sigma > 1$, each producer faces demand:

$$p_i = \left(\frac{y_i}{Y} \right)^{-1/\sigma} .$$

- Constant markup: $\frac{\sigma}{\sigma-1}$ in basic CES case.
- Heterogeneity \implies different a_i but same markup with standard CES assumption.



Oligopoly with Endogenous Markups

Motivation: (De Loecker et al., 2020) find rising markups \implies Markups may vary endogenously with competition.

Atkeson & Burstein (2008) approach:

- Within-sector competition among J brands, across-sector elasticity σ .
- If J large, markup is close to $\sigma/(\sigma - 1)$; if $J = 1$ (monopoly), markup can be larger.
- Firm-level sales share s_{ij} determines elasticity $\varepsilon(s_{ij}) \implies$ markup changes with share.



Aggregate Shocks vs. Idiosyncratic Shocks

- Standard result: Idiosyncratic shocks tend to wash out in aggregate (Law of Large Numbers).
- $\text{Var}(Y) \approx \frac{\text{Var}(y_i)}{N}$ if $\{y_i\}$ are i.i.d. with finite variance.
- Not enough to generate big business cycles in standard large economy with many small firms.



“Granular” Effects: Large Firms

(Gabaix, 2011): If firm size distribution is *fat-tailed* (Pareto with shape near 1), large idiosyncratic shocks can affect aggregate outcomes significantly.

Key insights:

- Distribution with Pareto tail \implies top firms weigh heavily.
- Many real-world large firms (e.g., Amazon, Walmart) have huge sales \implies big Domar weights \implies big macro impact.



(Hulten, 1978) Theorem:

$$\frac{dY}{Y} = \sum_i D_i \frac{da_i}{a_i}, \quad D_i = \frac{p_i y_i}{\sum_j p_j c_j}.$$

- Domar weight D_i depends on *sales*, not just value-added.
- Downstream firms with large sales magnify shocks through the network.



(Klette and Kortum, 2004) Model

- Continuum of products, each produced by some firm.
- Innovation \implies new generation of product with higher quality $\lambda > 1$.
- Incumbents and entrants compete in R&D to take over product lines.
- **Firm size** = number of product lines it owns.



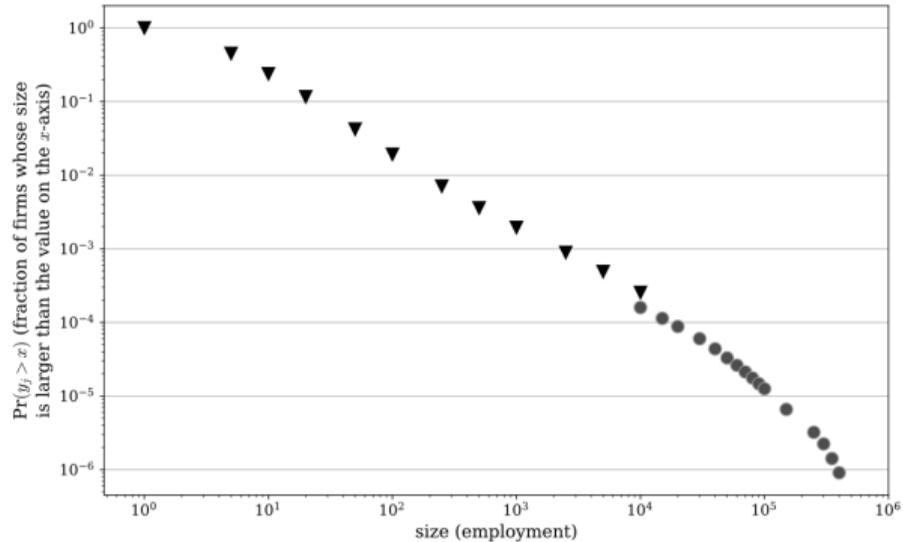


Figure 22.9: Distribution of firm size using log-log axes.

Source: Business Dynamics Statistics and Compustat. Reproduced from [Carvalho and Grassi \(2019\)](#).



Innovation & Growth

- **Limit pricing:** Leader charges $p = \lambda w$ to deter older-generation producers.
- Profit per product line: $\pi = (p - w) c$ with elasticity of substitution = 1.
- Incumbent invests in R&D to expand product lines; each line can be lost to competitor's innovation.
- Balanced-growth path: Growth rate g proportional to economy-wide innovation rate μ .

$$g = \mu \ln(\lambda).$$



Firm Dynamics

- Each product line evolves via *Bernoulli trials* of “innovate and keep” vs. “lose to rival”.
- In discrete time, we can track distribution of product lines across firms.
- Allows modeling of *entry/exit* and *endogenous* productivity process rather than exogenous a_j .



Limitations & Extensions

- Baseline model: Growth of big firms is *negative on average* (contrasts with empirical patterns).
- Tail index vs. growth suggests modifications (e.g., cumulative innovation, different R&D structure).
- Despite simplicity, the Klette-Kortum framework unifies:
 - Innovation-based growth
 - Entry/exit
 - Firm size and dynamics



Concluding Remarks

- Firm heterogeneity is critical to understanding macro outcomes, policy impacts, and productivity growth.
- Data show strong dispersion, reallocation, and evolving patterns (mega-firm rise, declining dynamism).
- Simple frameworks illustrate how dispersion affects aggregate productivity through allocation.
- General equilibrium models (e.g., [\(Hopenhayn and Rogerson, 1993\)](#)) show policy influences on hiring, firing, entry, exit.
- Endogenous growth perspectives [\(Klette and Kortum, 2004\)](#) connect innovation to firm turnover and macro growth.



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Thank you!

Questions or comments?

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