

# Title: Productivity, Job Creation and Entrepreneurship

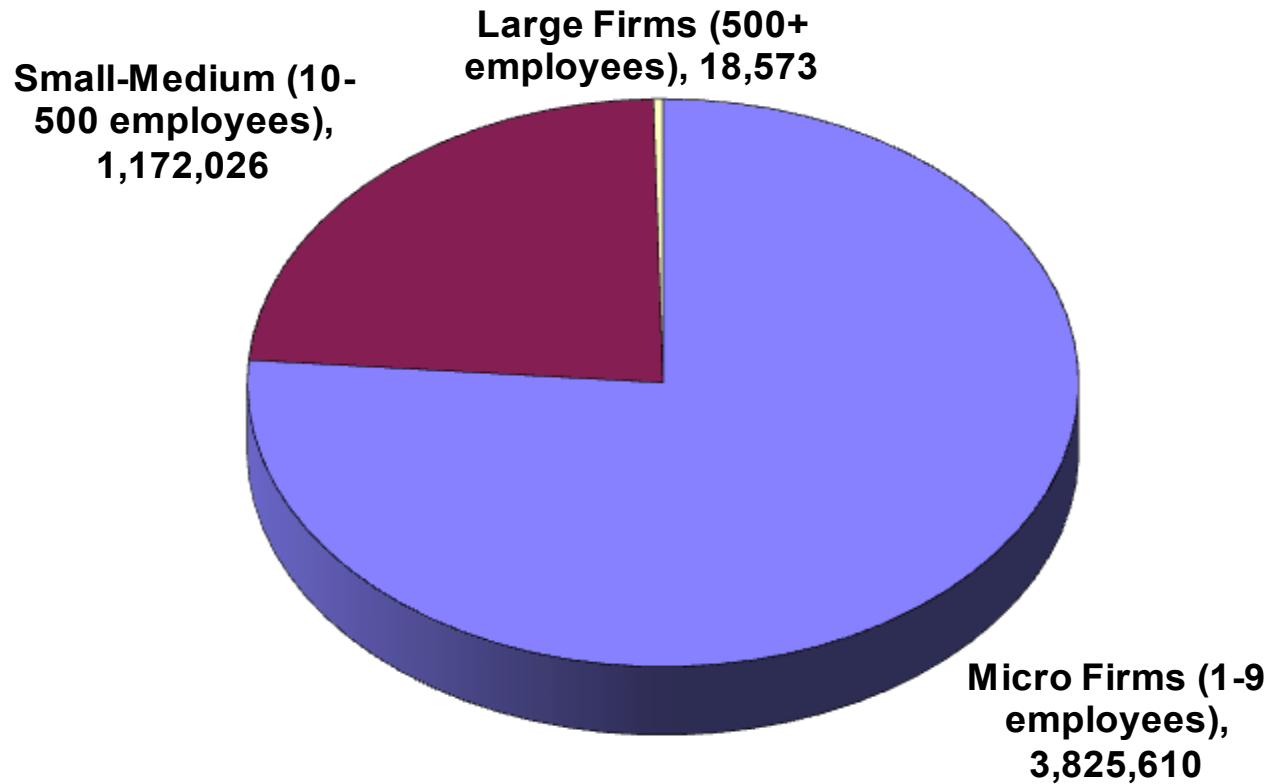
Lecture Notes for NBER Entrepreneurship Bootcamp

July 2013

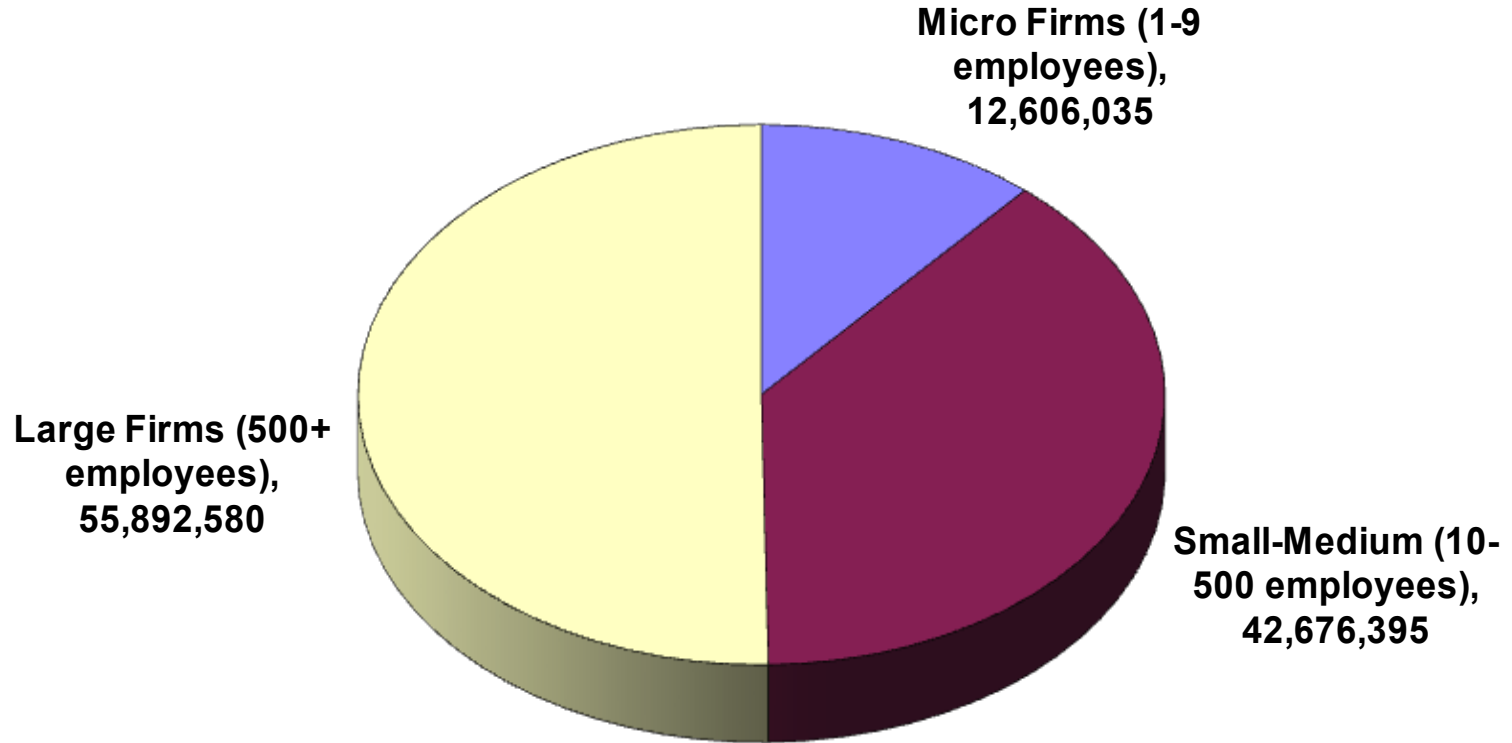
By John Haltiwanger

University of Maryland and NBER

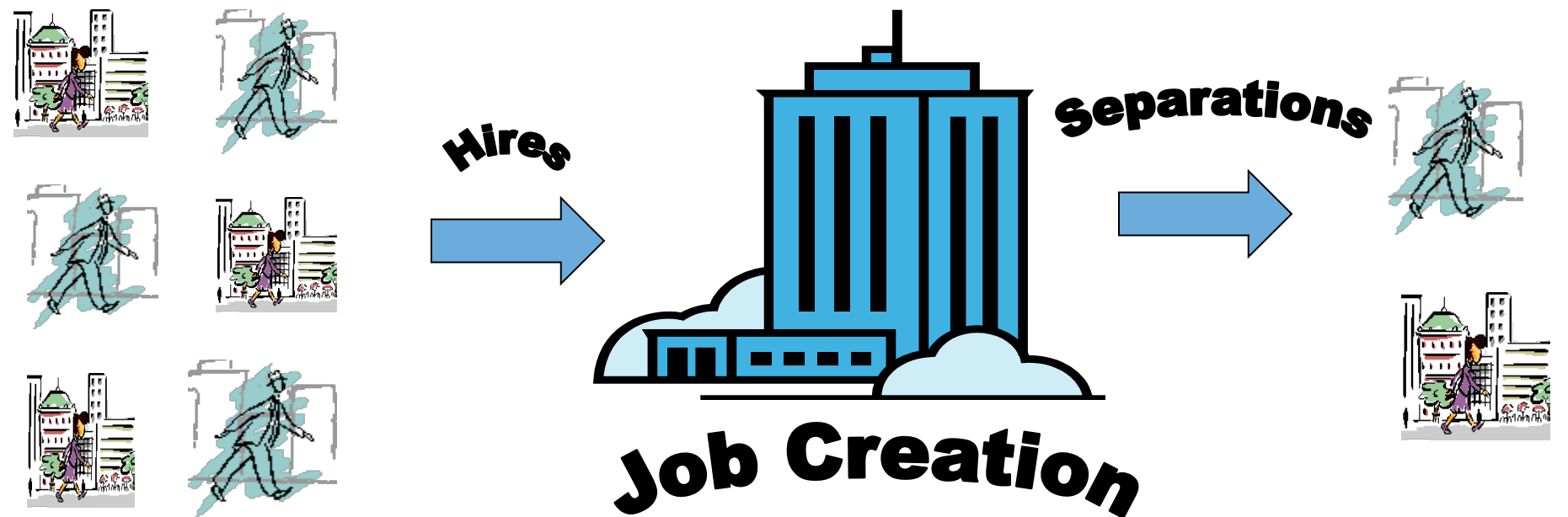
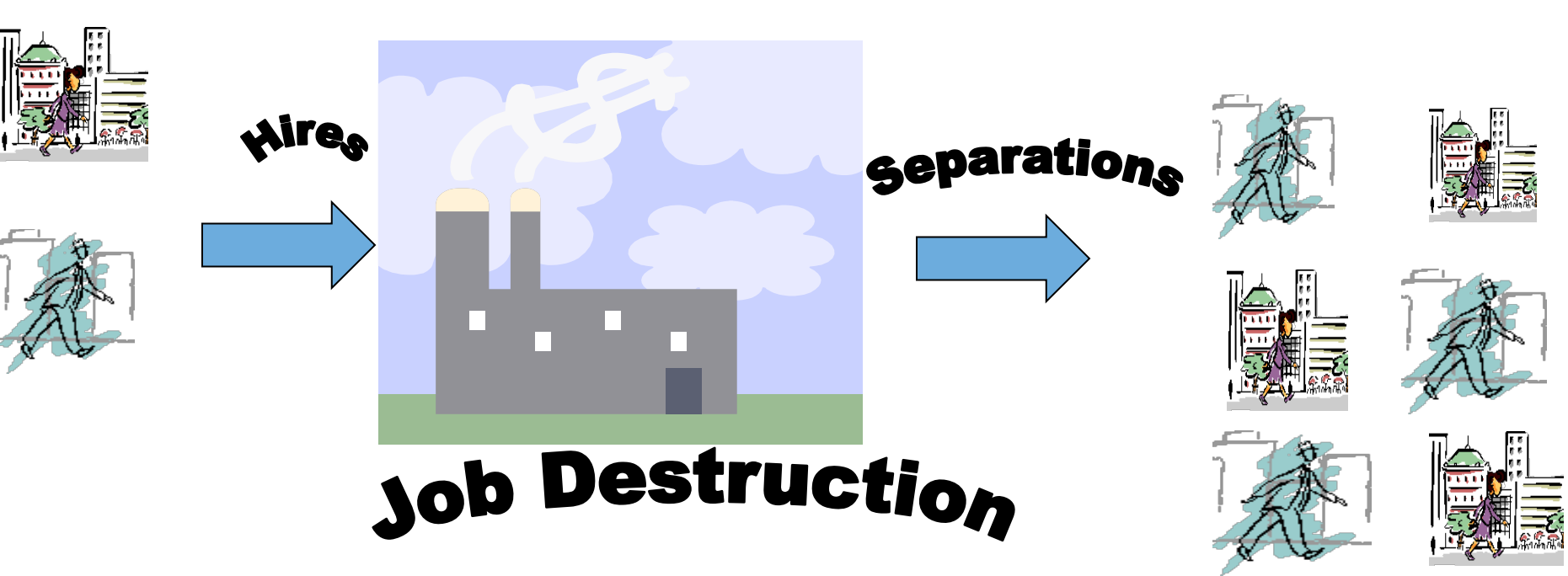
## Share of Firms by Firm Size, 2010



## Share of Employment by Firm Size, 2010



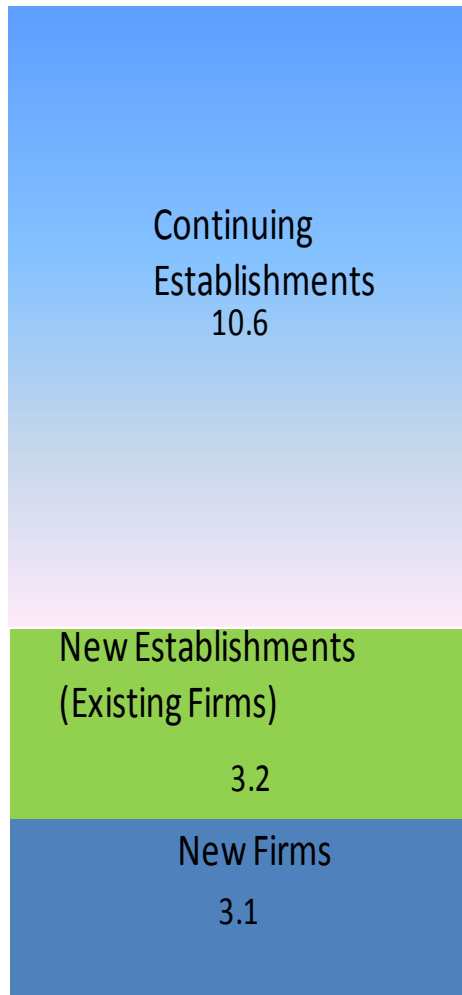




# Job Creation and Destruction, U.S. Private Sector, Annual Rates (Percent of Employment), 1980-2009

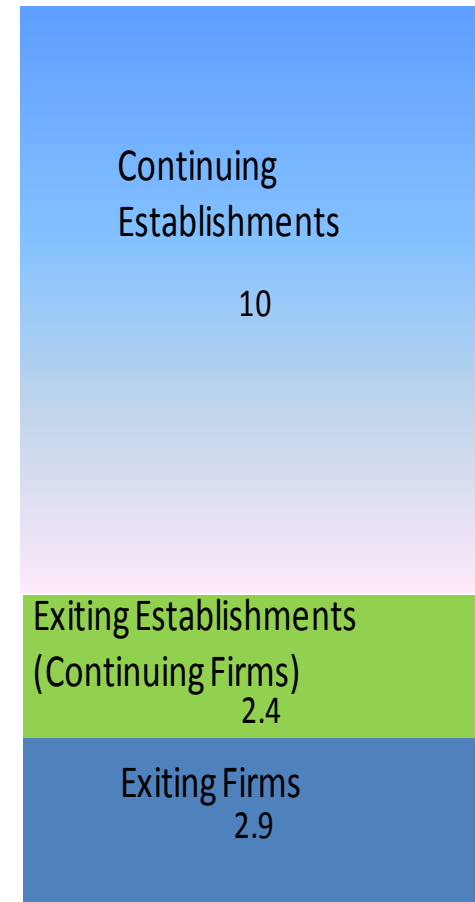
Source: BDS

Total=16.9



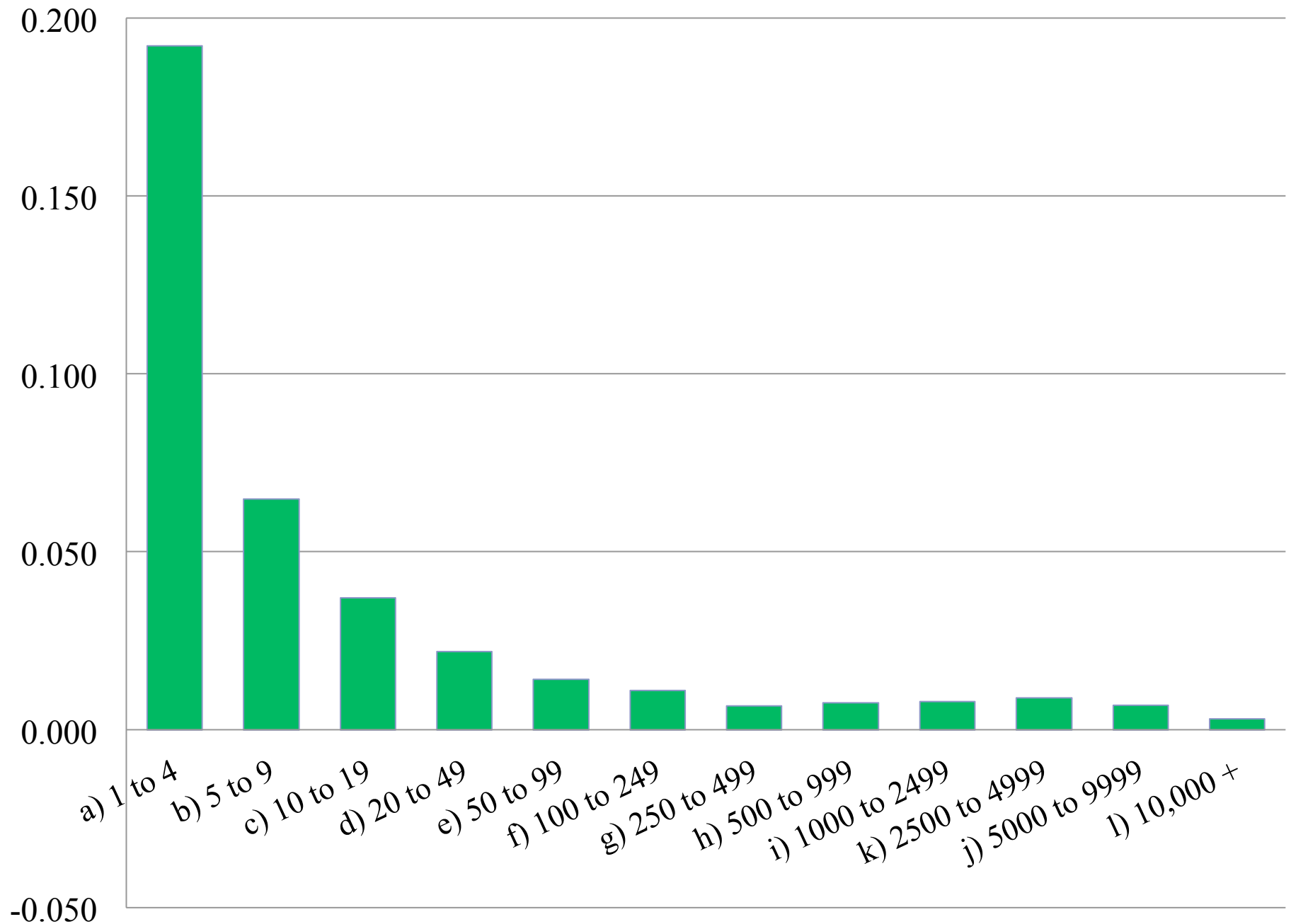
Job Creation

Total=15.3

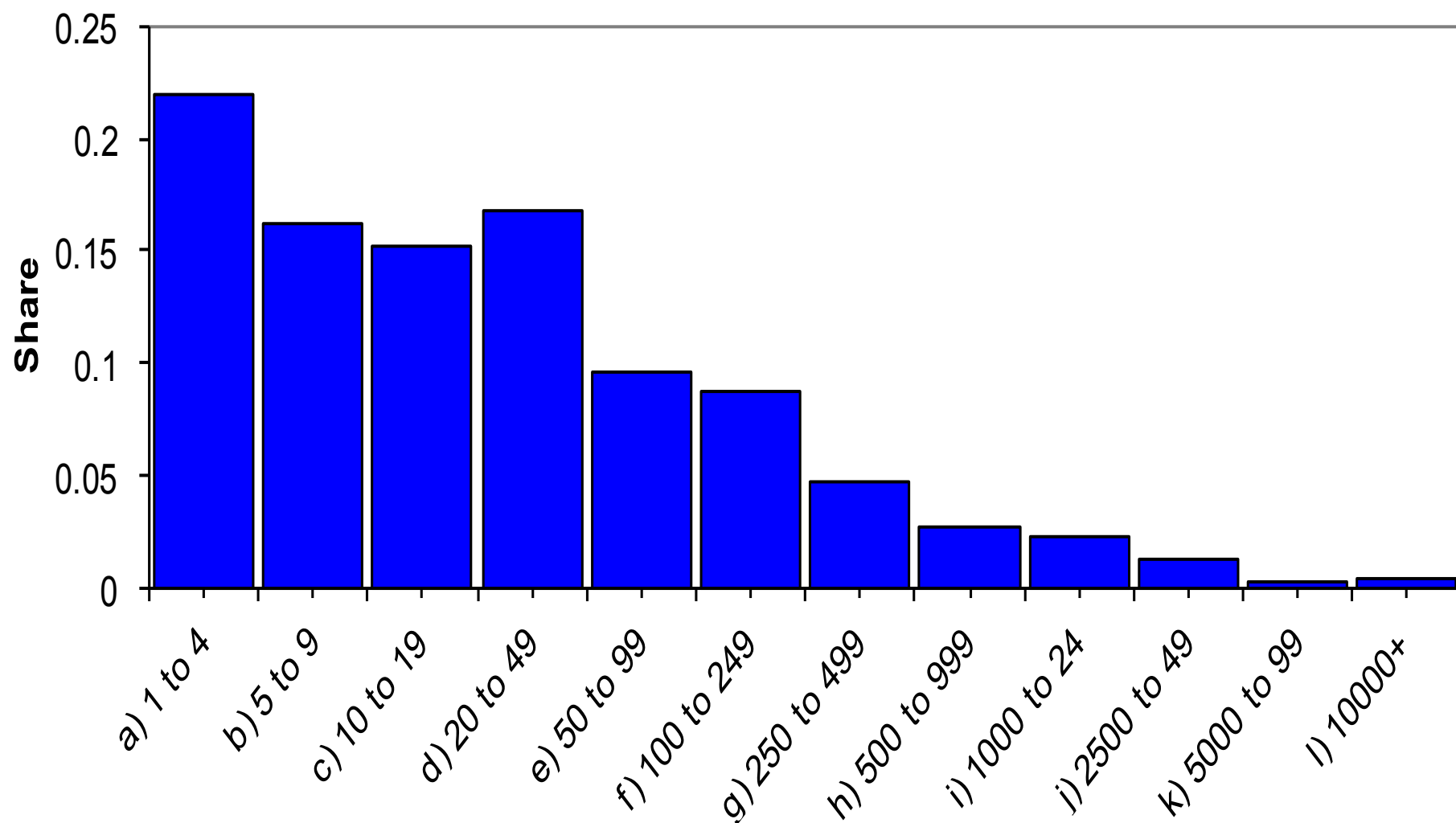


Job Destruction

# Net Employment Growth by Base Year Firm Size

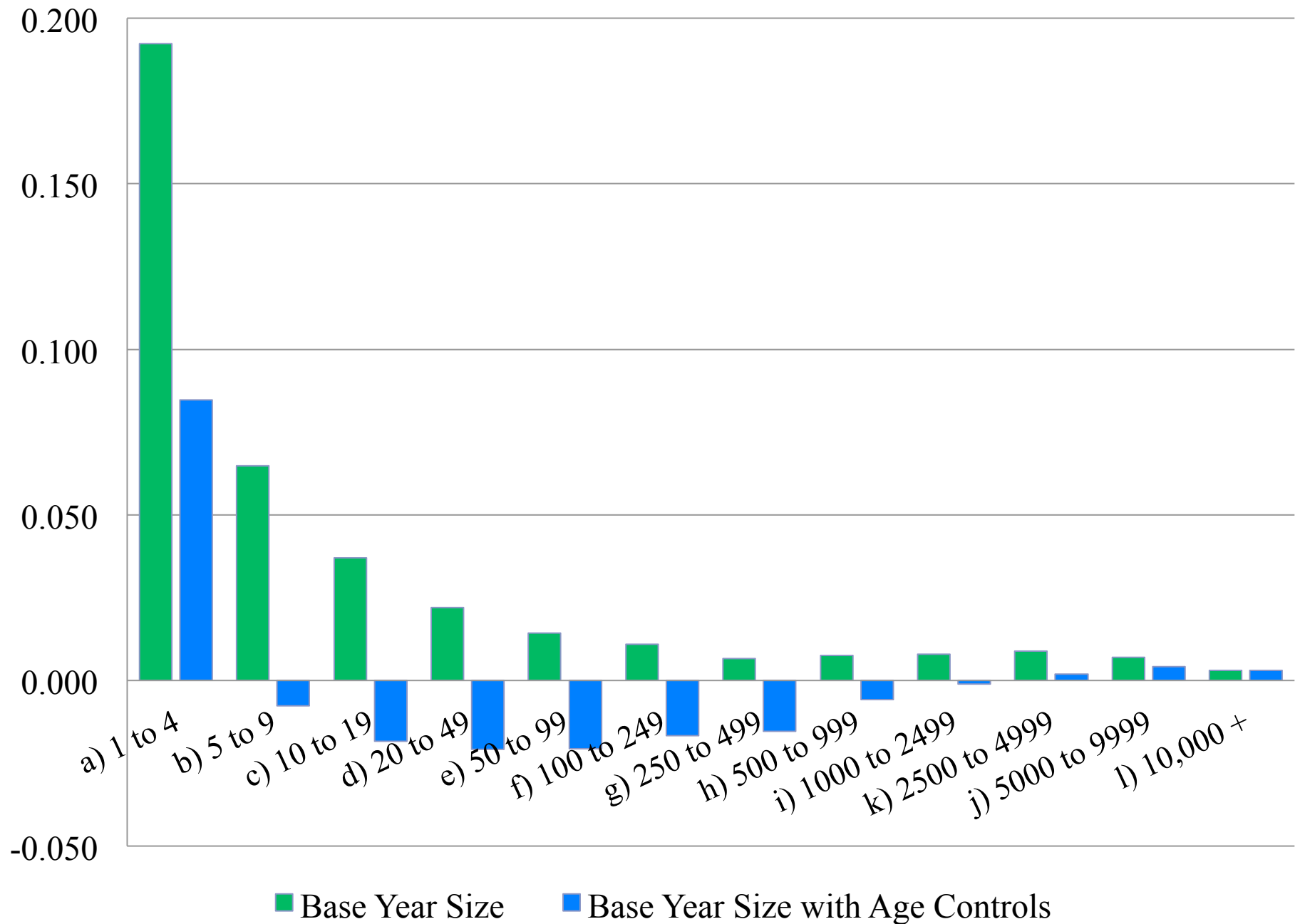


# Share of Employment in Startups by Firm Size Class

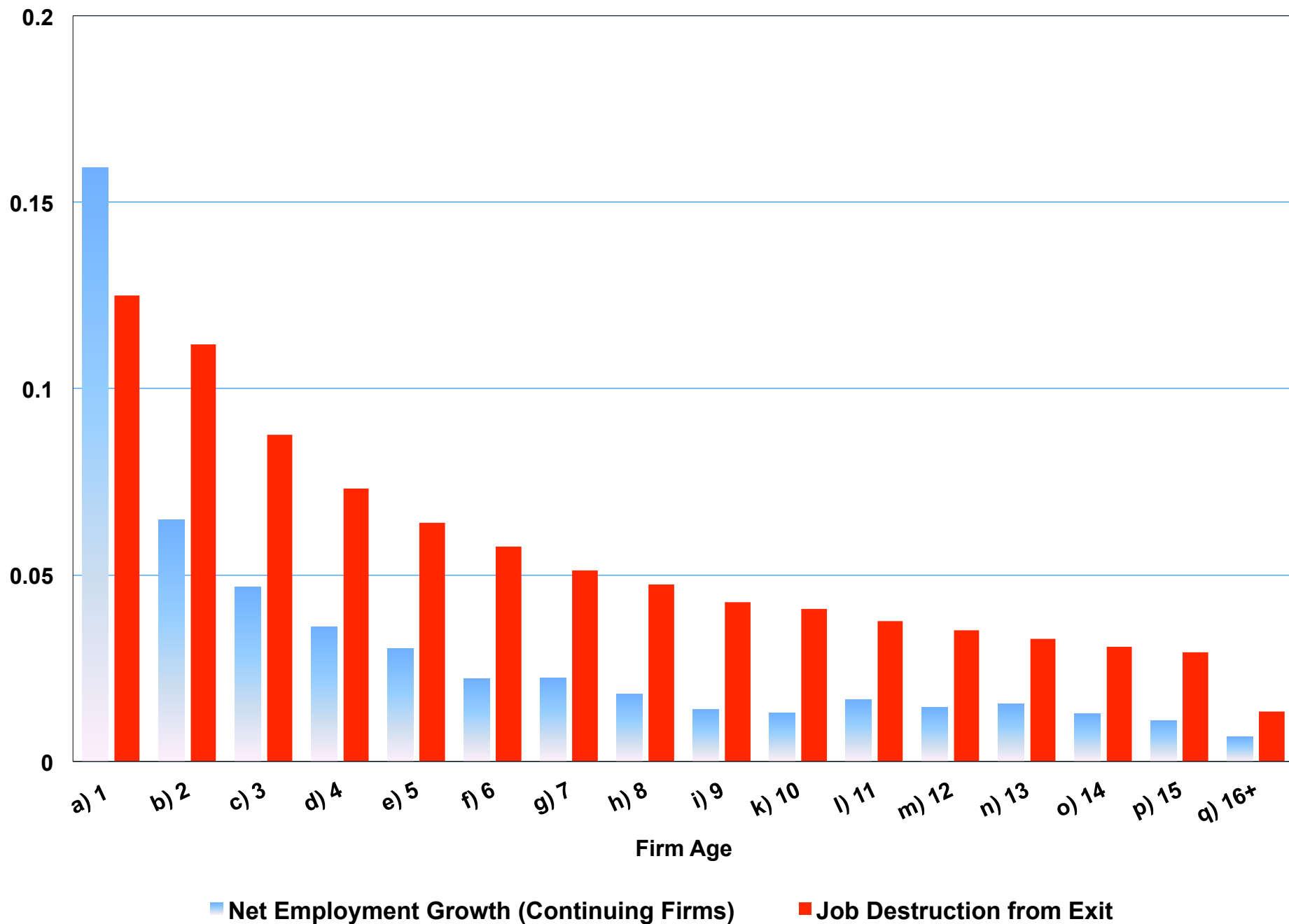




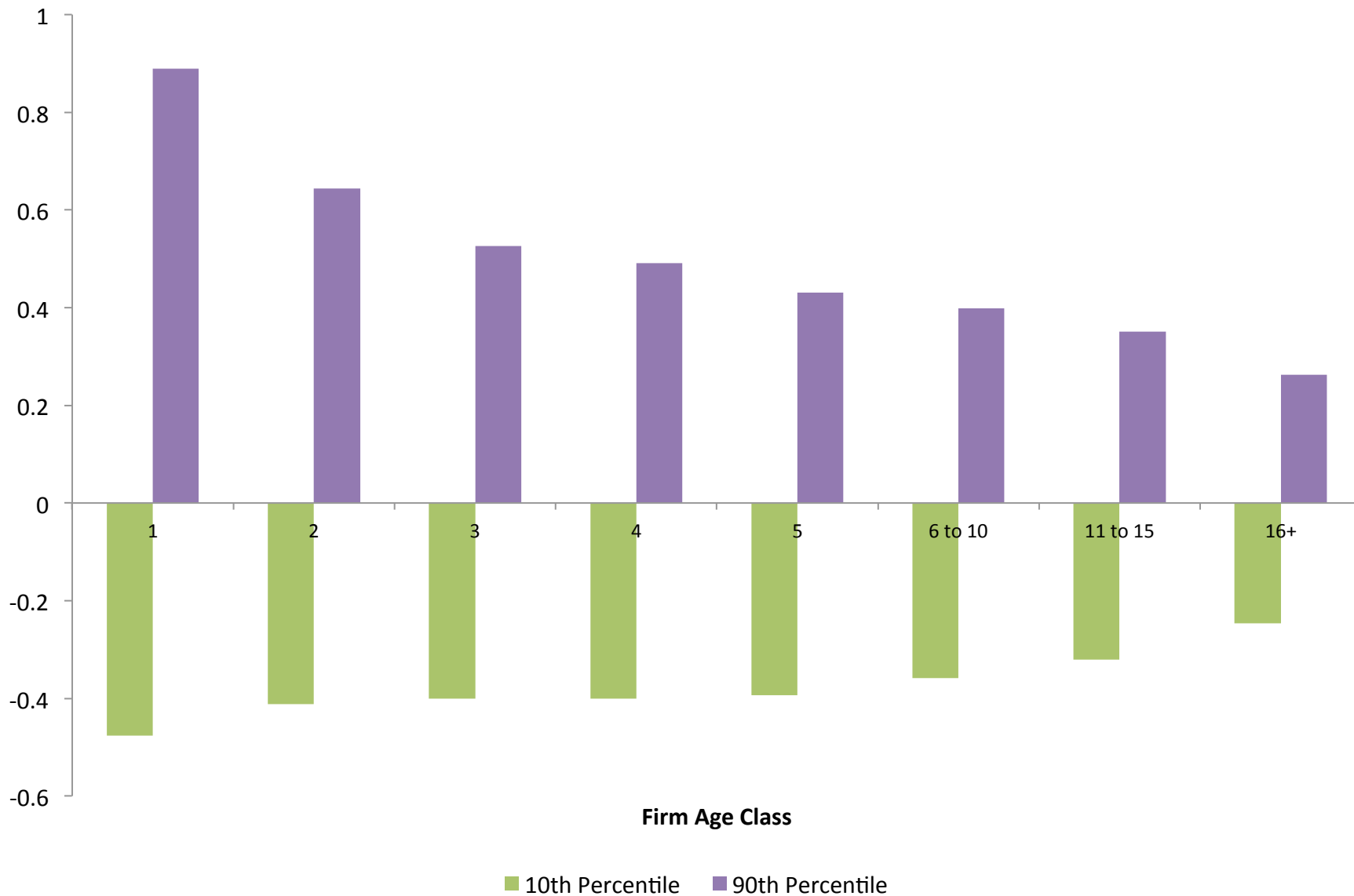
# Net Employment Growth by Base Year Firm Size



## Up or Out Dynamics of Young U.S. Firms

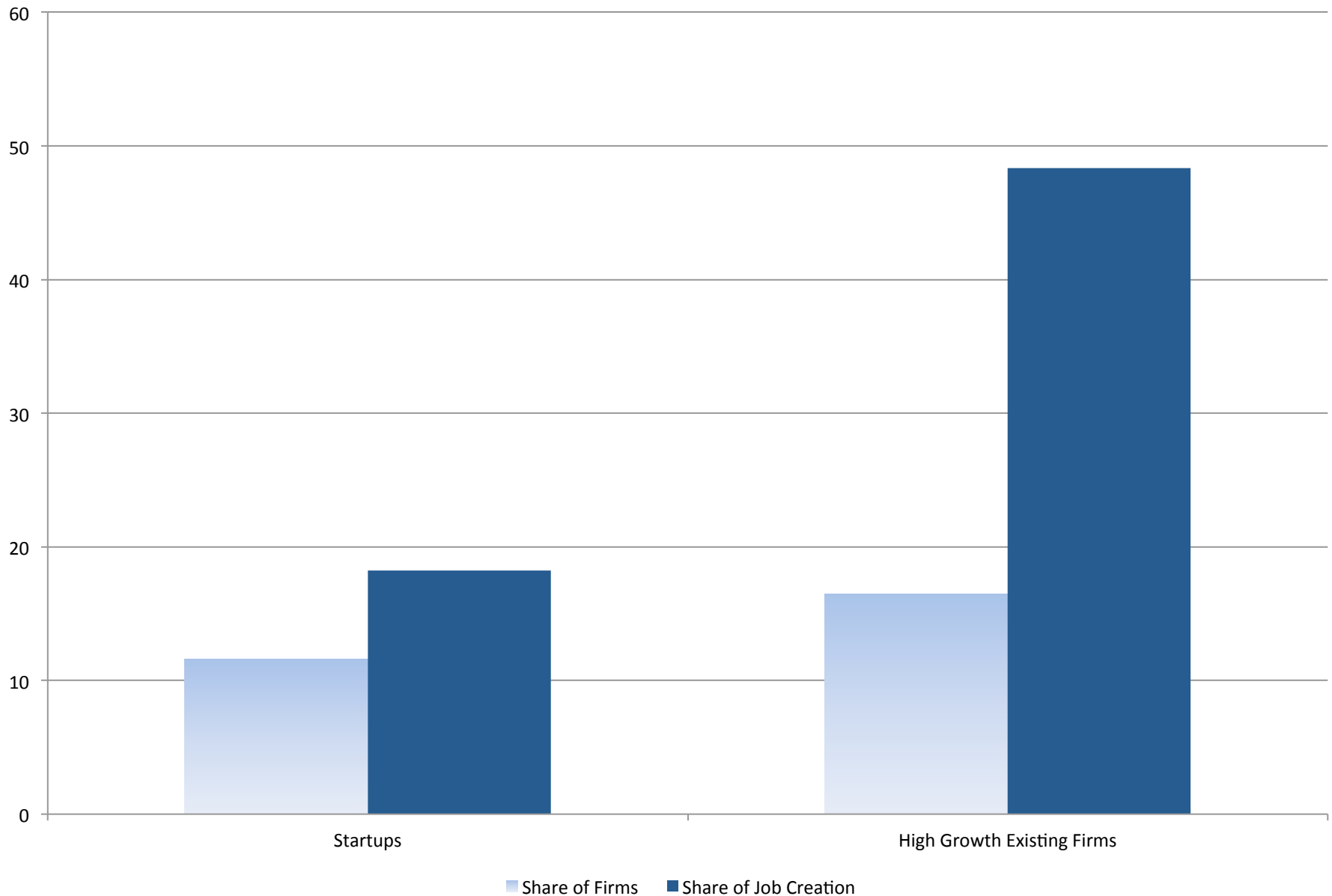


### 90th and 10th Percentiles of Net Employment Growth Rates for Surviving U.S. Private Sector Firms by Firm Age



Source: Firm-level data used by Haltiwanger, Jarmin and Miranda (2011)

## Startups and High Growth (Annual Growth>25 percent) Existing Firms Disproportionately Create Jobs, U.S. Private Sector

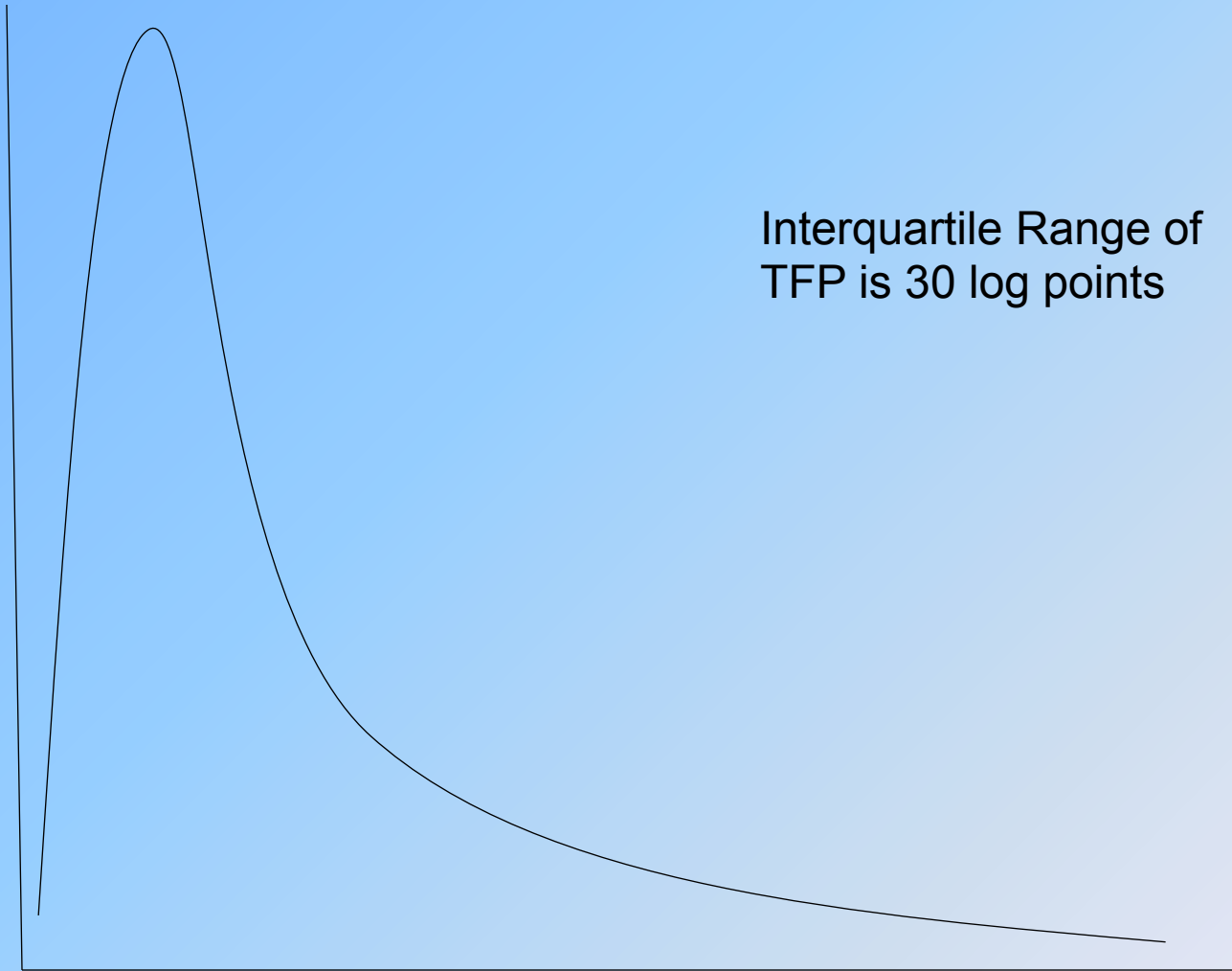


Source: Tabulations from Firm-Level Data Used in Haltiwanger, Jarmin and Miranda (

# What accounts for cross sectional and dynamic patterns?

- Very skewed size distribution
- Constant state of churning
  - Wave of entering firms contributes substantially to job creation each year
  - Most exit
  - Conditional on survival, young businesses grow quickly
  - Even amongst large, mature businesses high pace of churning of jobs and businesses

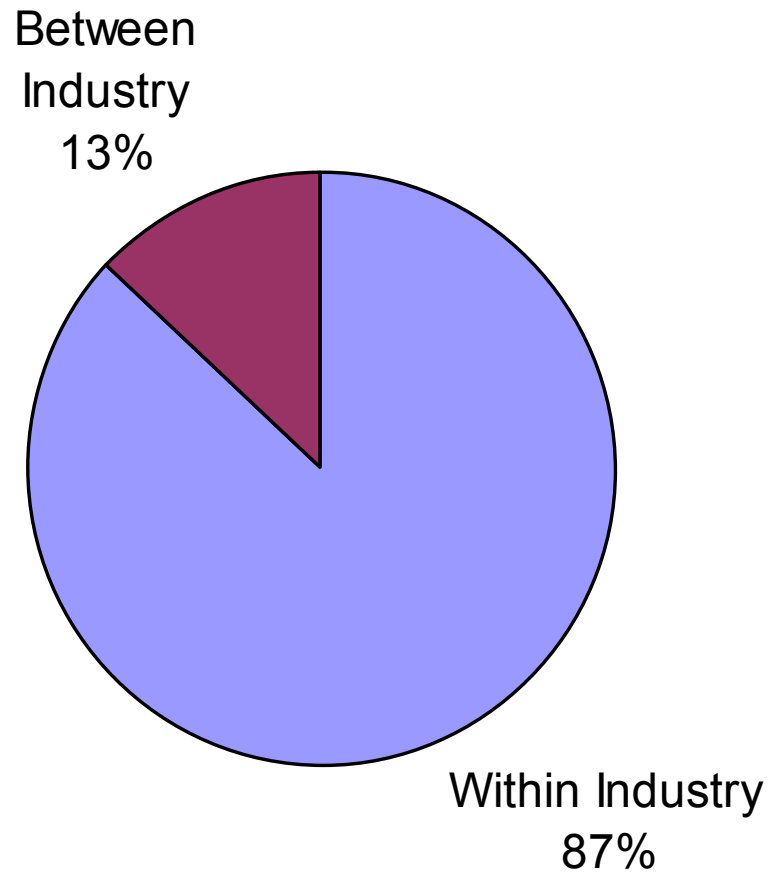
## Productivity Distribution Within Narrowly Defined Industries



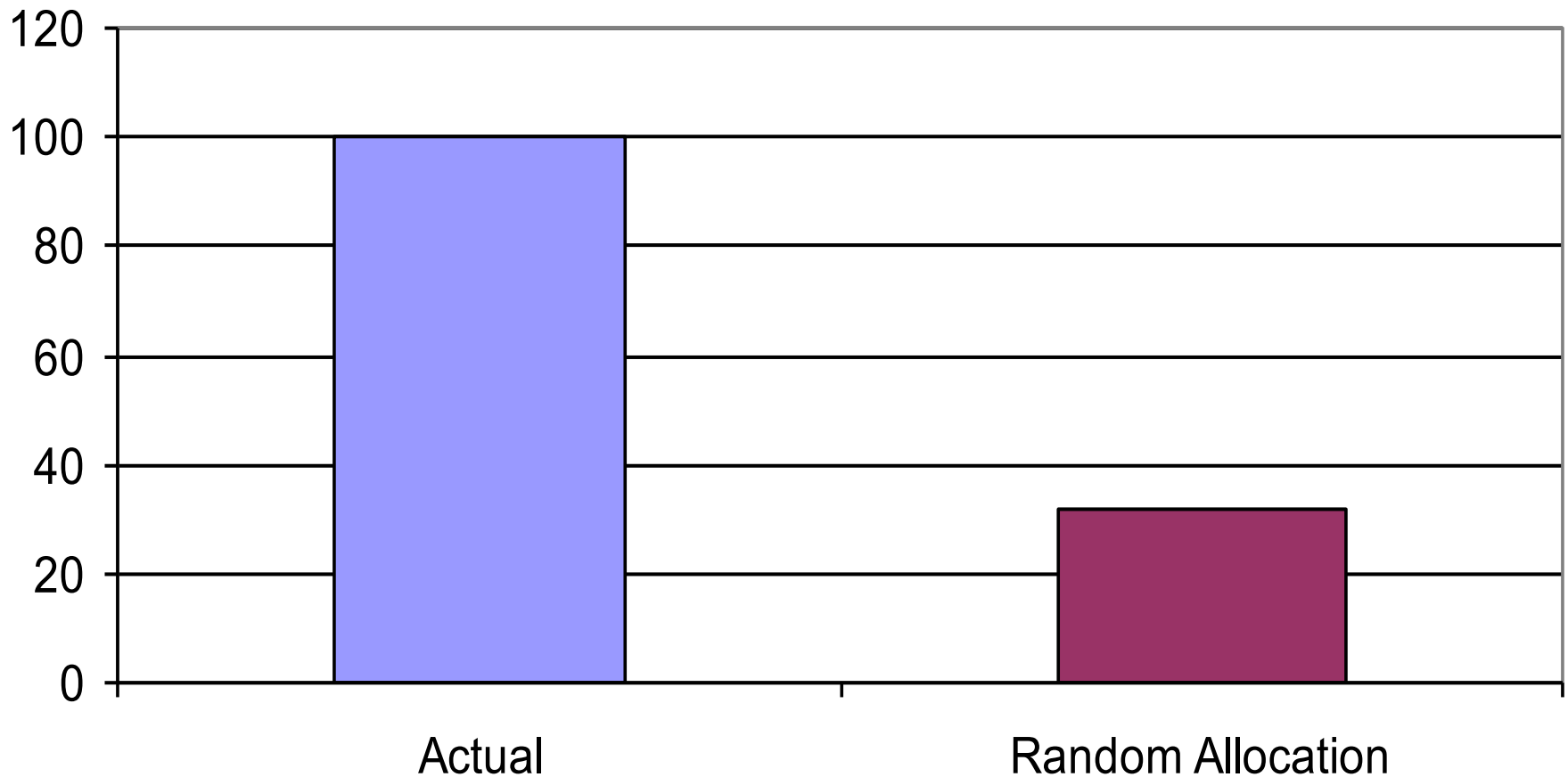
Interquartile Range of  
TFP is 30 log points

Productivity of Businesses

# Share of Reallocation Between and Within Detailed Industries

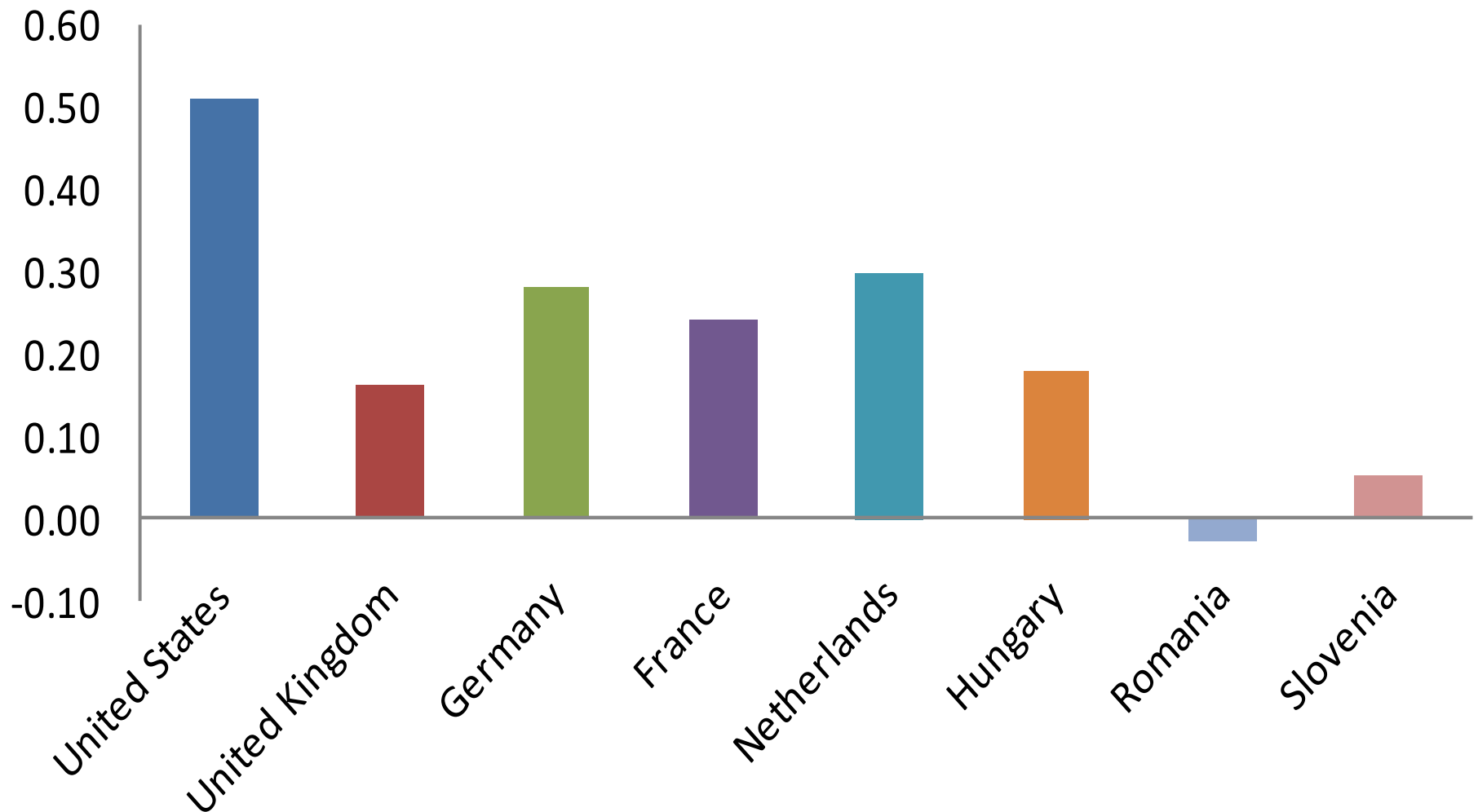


# U.S. Labor Productivity: Comparison Between Actual and Random Allocation of Size of Businesses

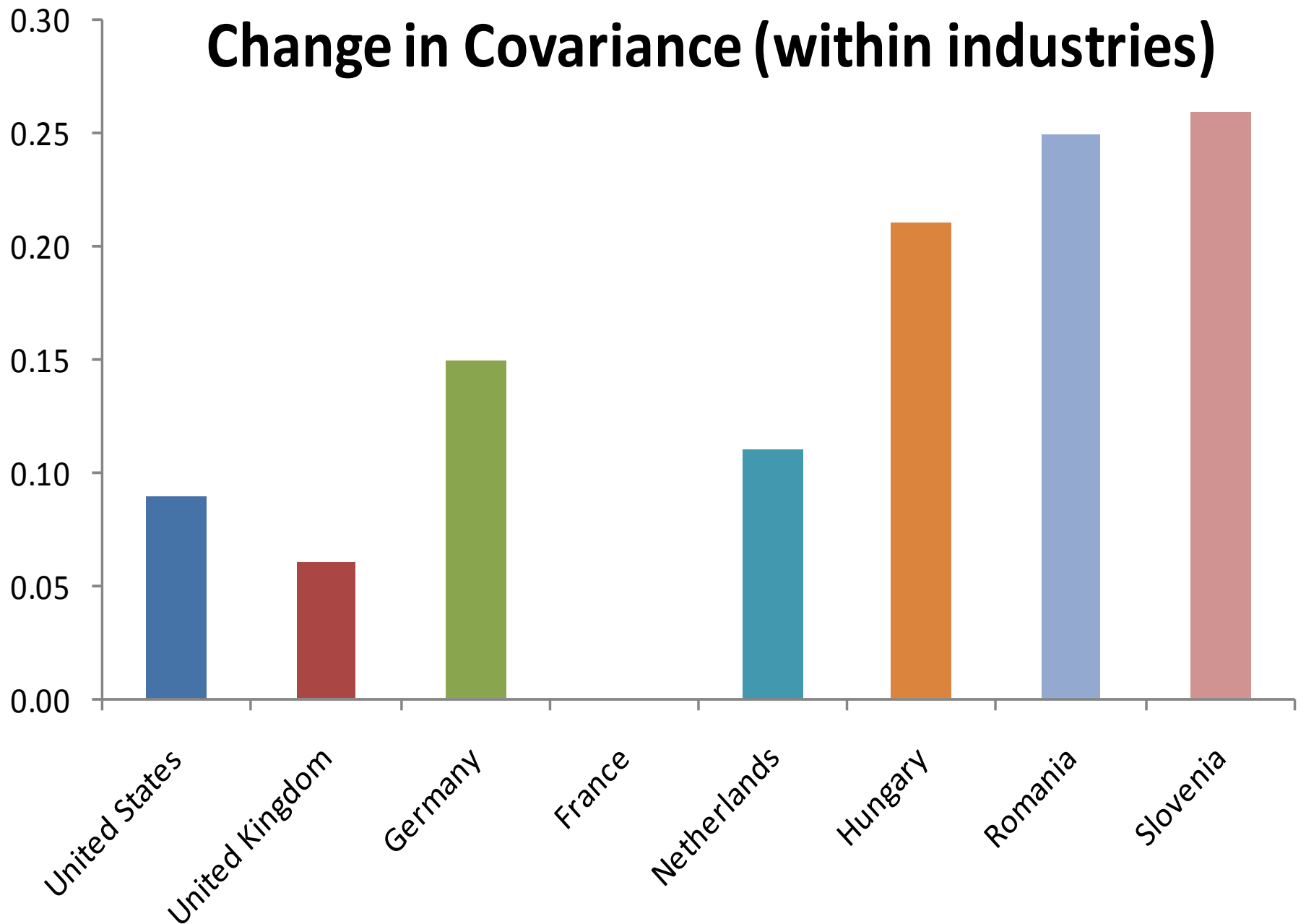




# Covariance Between Size and Productivity (within industries)

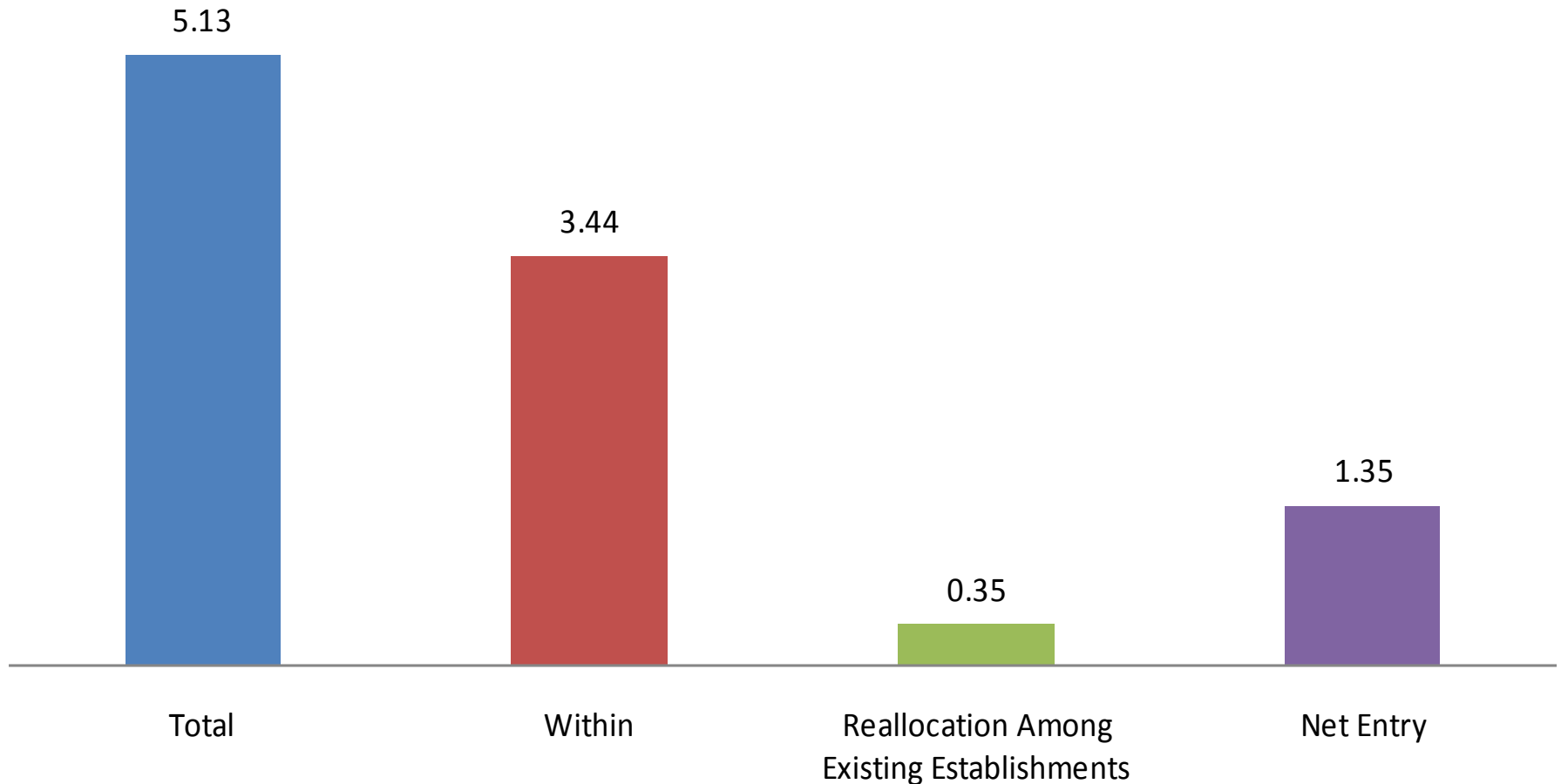


# Change in Covariance (within industries)

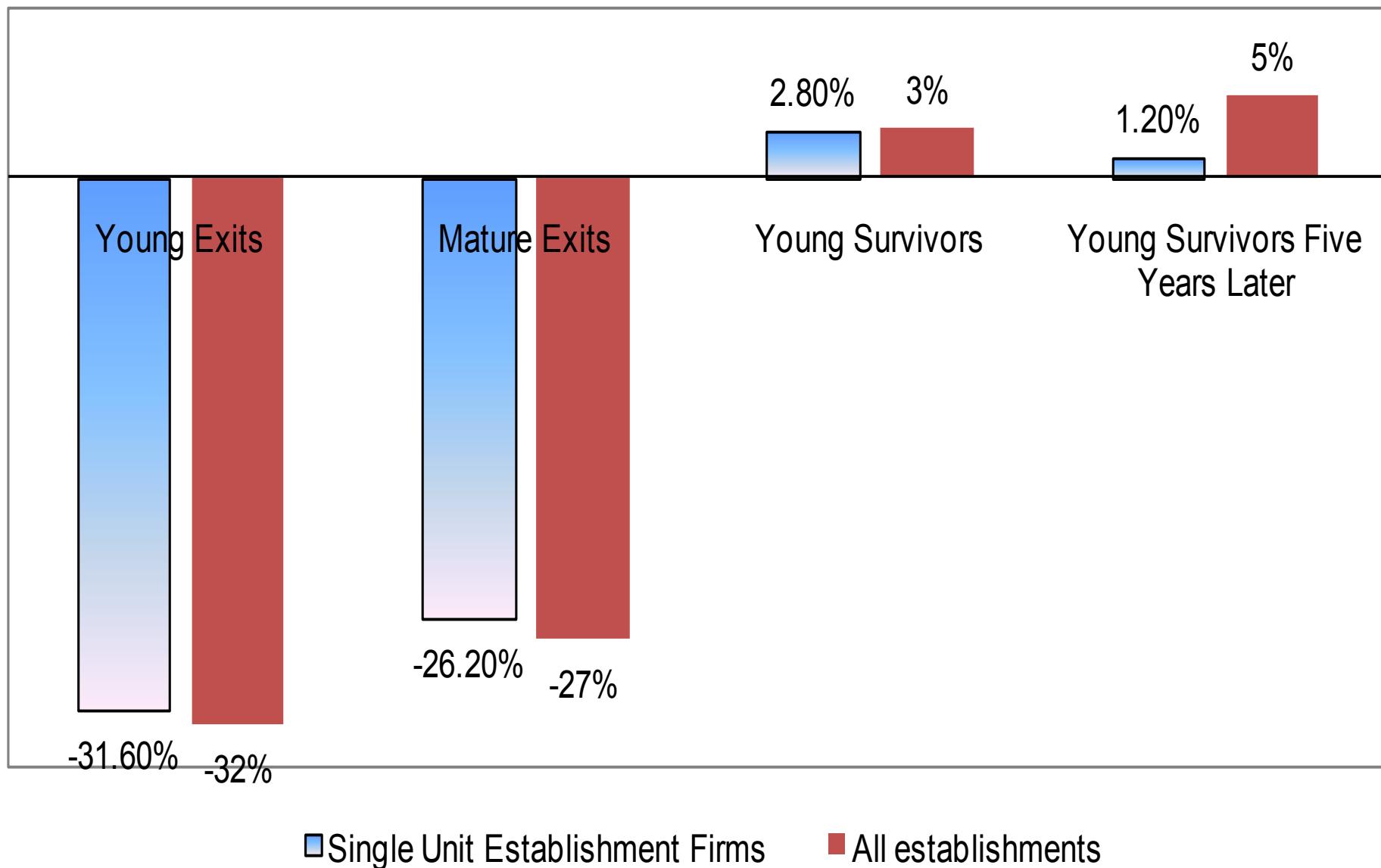


# Components of Total Factor Productivity Growth over Five-Year Horizons, 1977-1997, Selected Manufacturing Industries

■ Total   ■ Within   ■ Reallocation Among Existing Establishments   ■ Net Entry



# Productivity of Young Businesses Relative to Mature Surviving Incumbents, U.S. Retail Trade



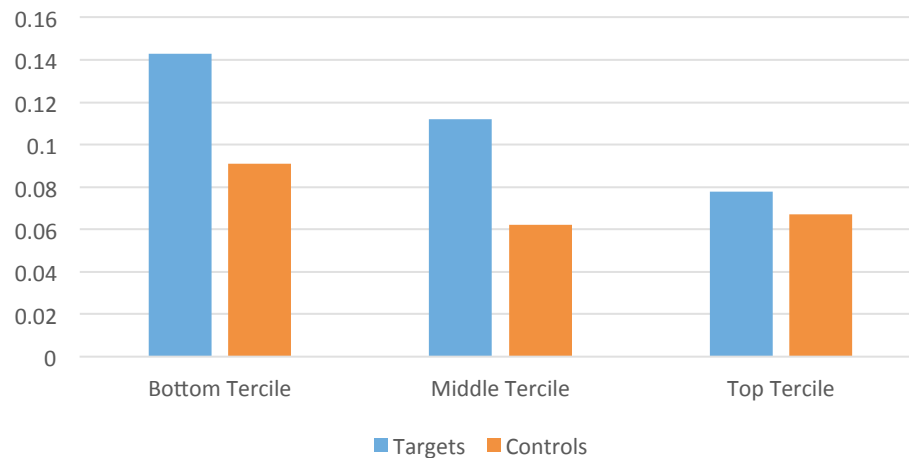
# Ownership Change, Management, Financing...

- Many factors underlie the ongoing restructuring and reallocation of businesses
- For allocative efficiency, financial markets need to be facilitating the reallocation of resources to the most productive businesses
- Ownership/management practices and changes are part of these dynamics.
- Example: Private Equity

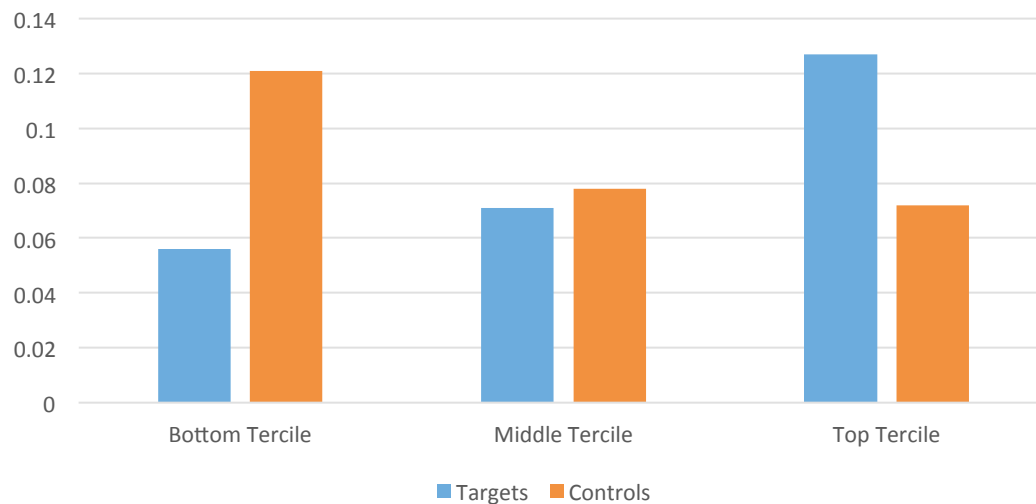
# Impact of Private Equity on Net and Job Reallocation



Exit Probability of Targets and Controls by  
Terciles of Within Industry Productivity  
Distribution



Entry Probability of Targets and Controls by Terciles of  
Within Industry Productivity Distribution



## Two Year Productivity Growth Impact From Private Equity

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Total Productivity Growth Differential	2.09
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Excluding Acquisition/Divestiture	1.96
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Share of Total from:

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Continuing Establishments	0.20
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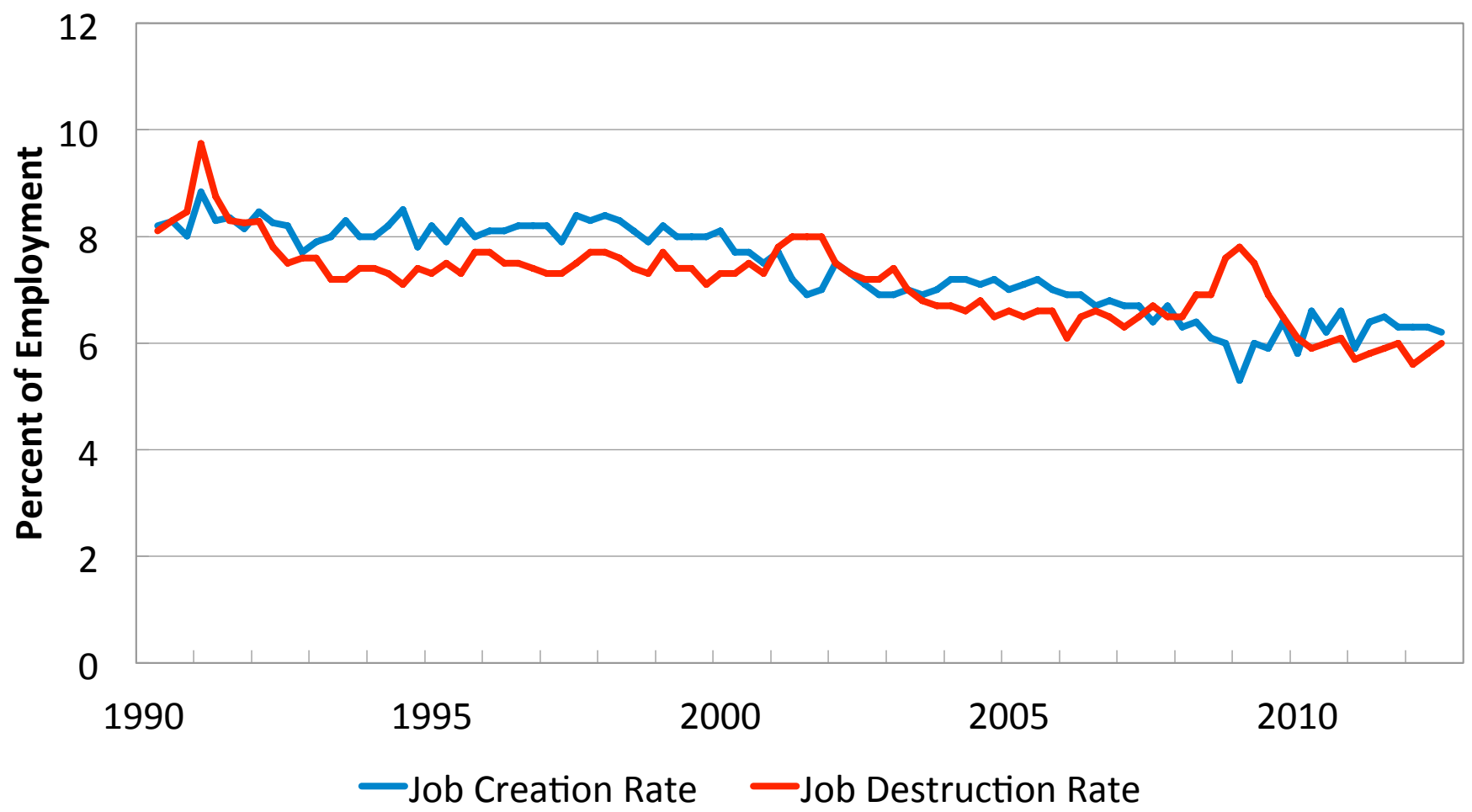
Net Entry	0.74
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Net Acquisition	0.06
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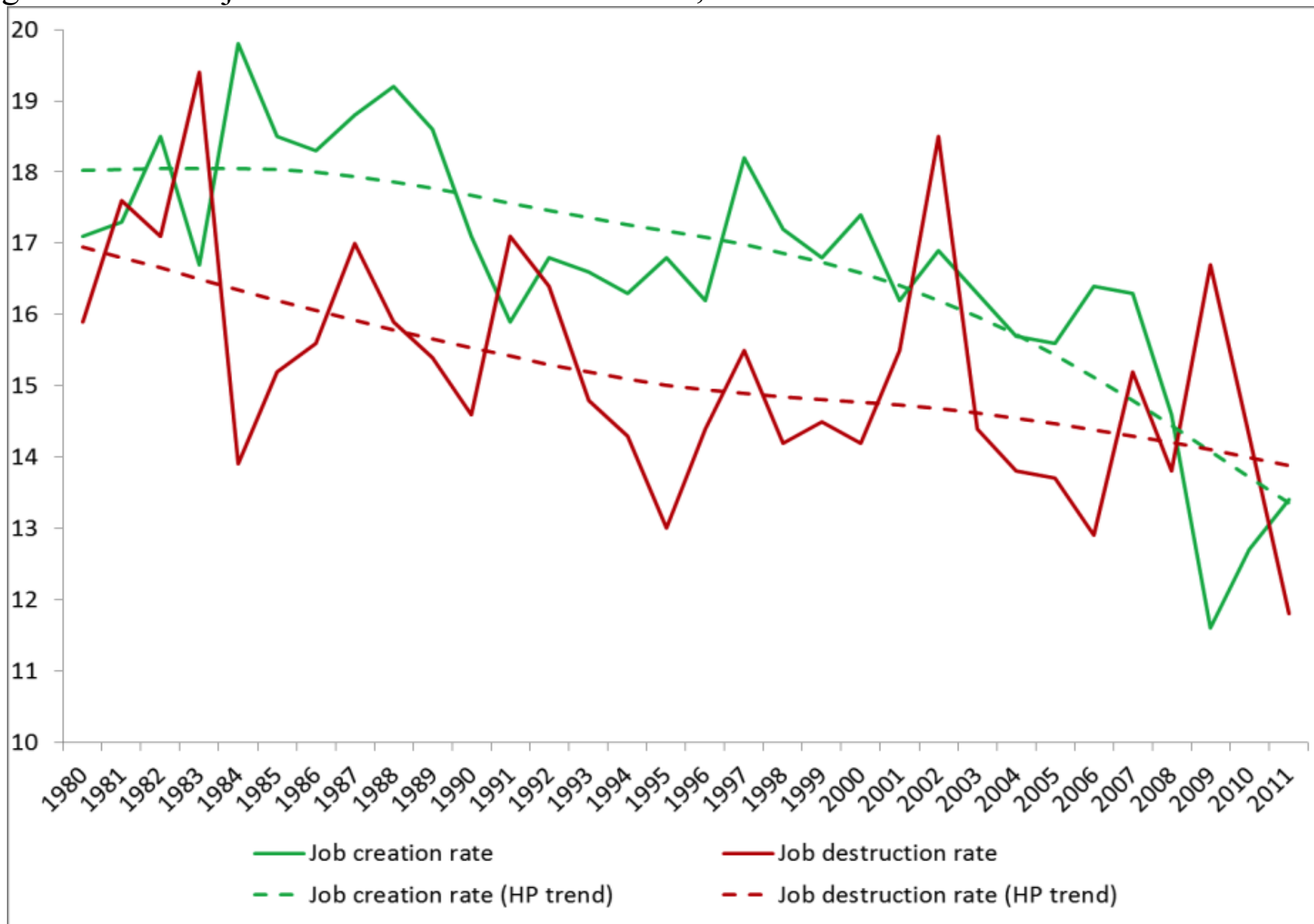
Some Disturbing Trends?

# Declining Pace of Creation and Destruction in BED



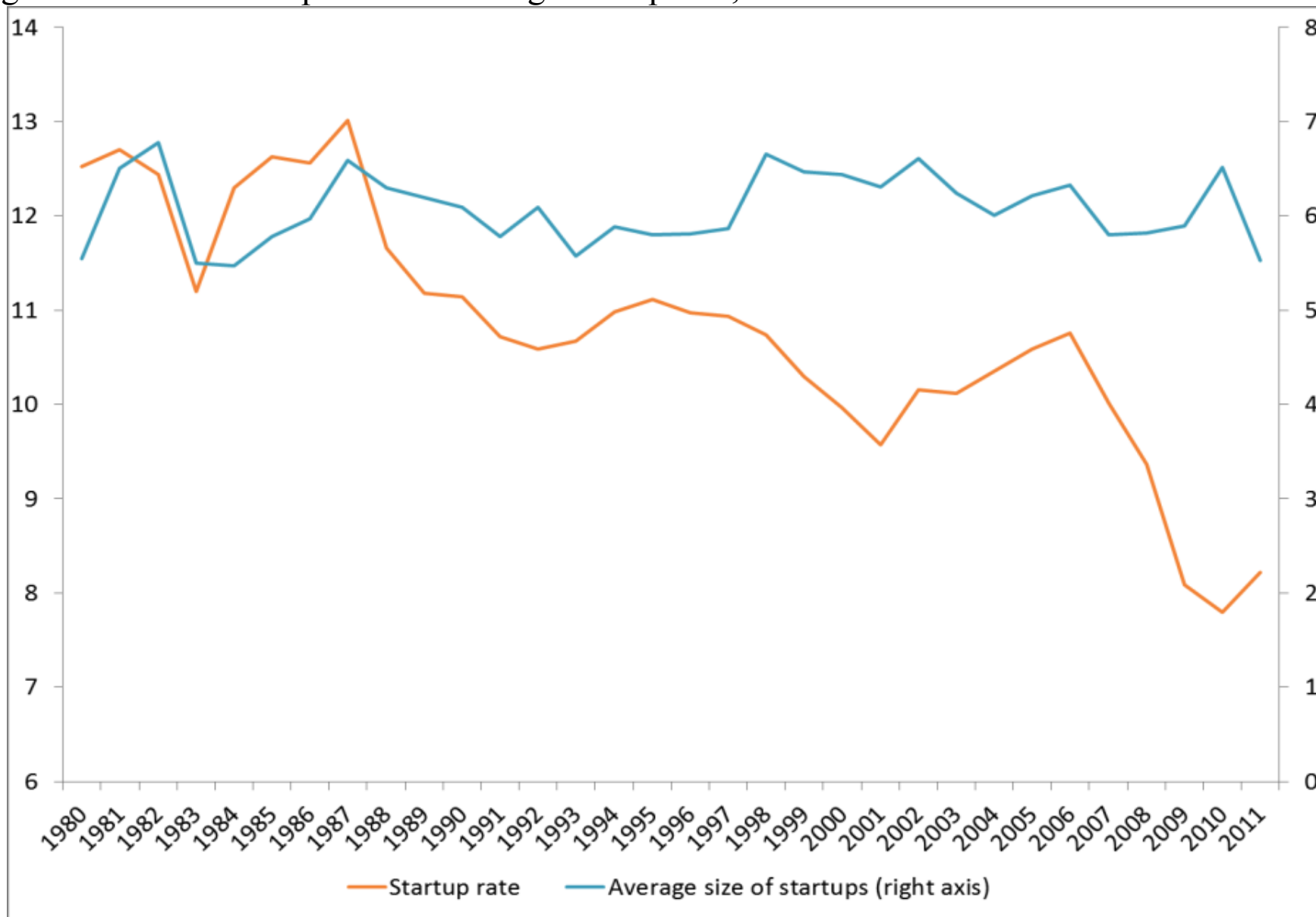
Source: BLS BED DATA

Figure 1: Annual job creation and destruction rates, 1980-2011



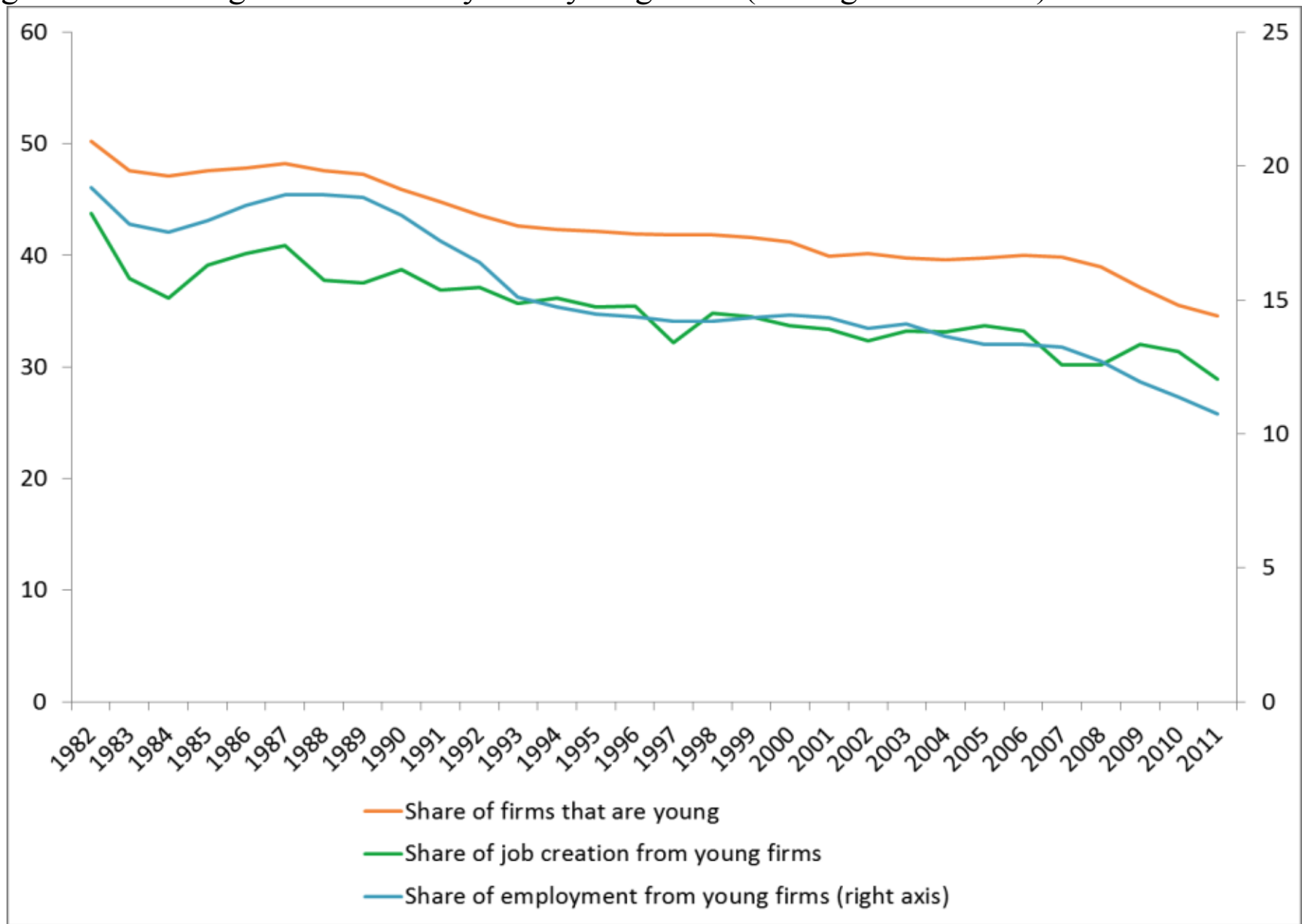
Notes: Author calculations from the U.S. Census Bureau's Business Dynamics Statistics. Filter is Hodrick-Prescott with multiplier 400. Vertical axis does not begin at zero.

Figure 2: Annual startup rate and average startup size, 1980-2011



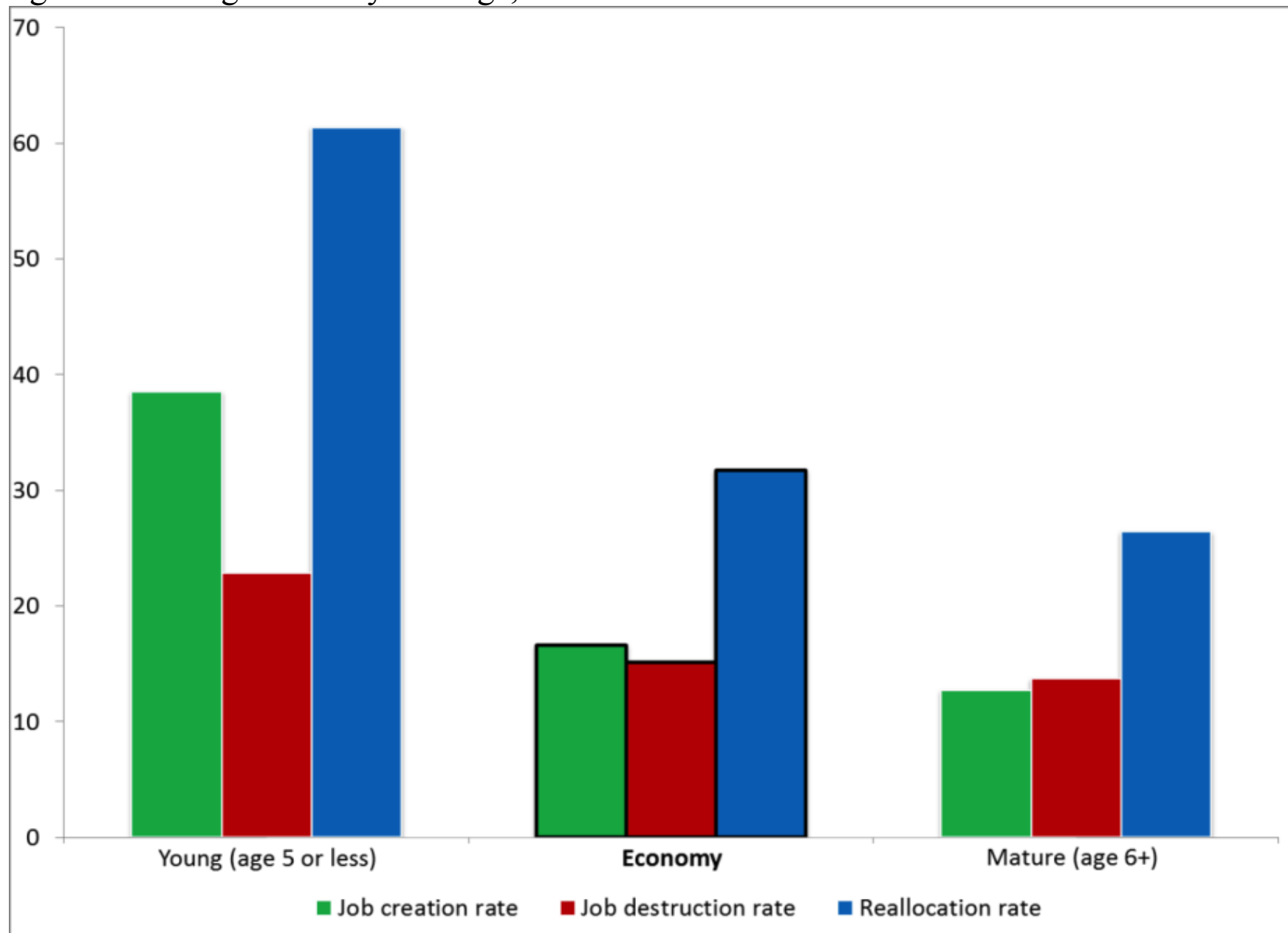
Notes: Author calculations from the U.S. Census Bureau's Business Dynamics Statistics. Left vertical axis does not begin at zero.

Figure 3: Declining share of activity from young firms (firm age five or less)



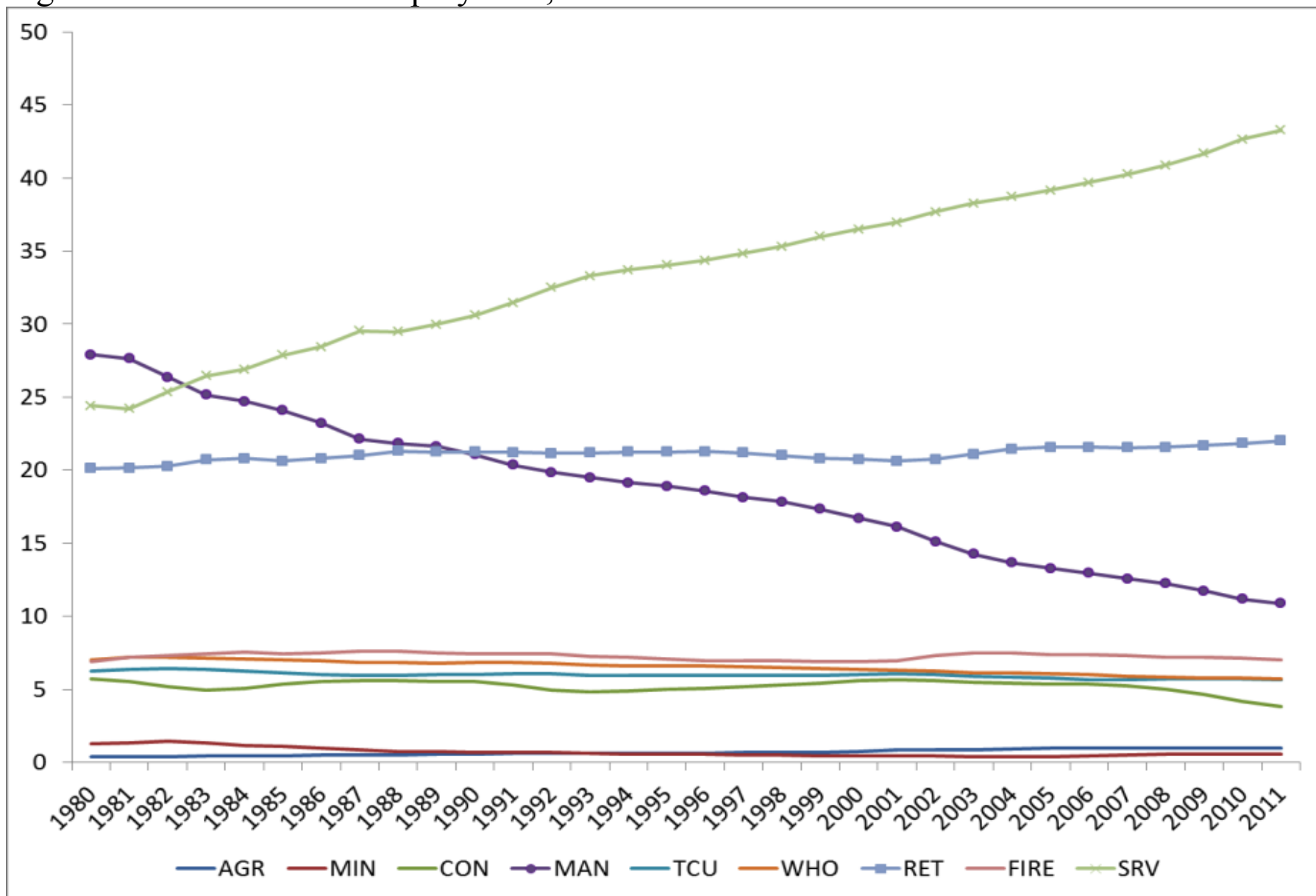
Notes: Author calculations from the U.S. Census Bureau's Business Dynamics Statistics. Employment shares in each period based on the average of employment in period t-1 and t (the denominator of the DHS growth rate).

Figure 4: Average flows by firm age, 1982-2011



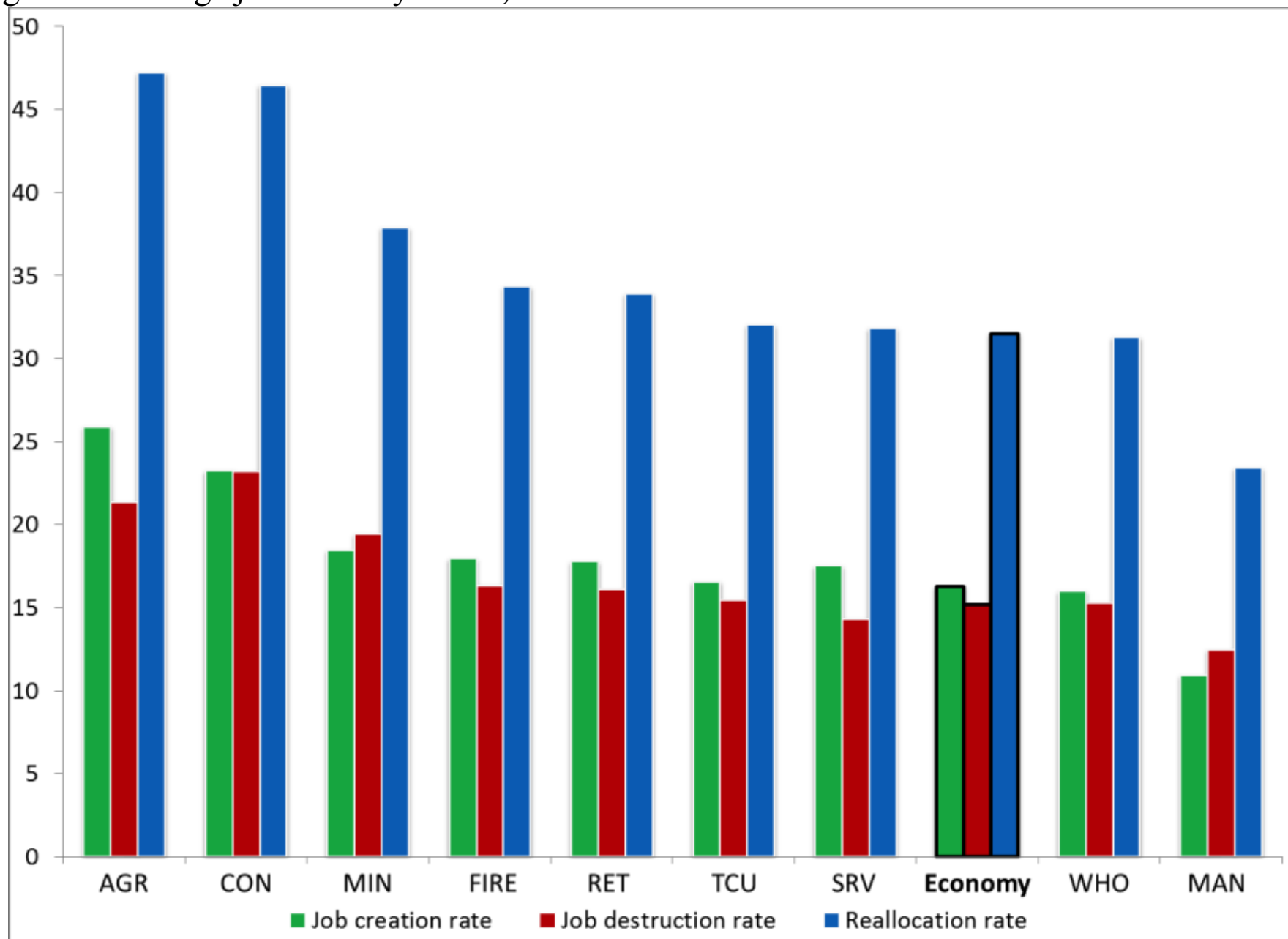
Notes: Author calculations from the U.S. Census Bureau's Business Dynamics Statistics.

Figure 5: Sector share of employment, 1980-2011



Notes: Author calculations from the U.S. Census Bureau's Business Dynamics Statistics. Broad sectors are on SIC basis. AGR= Agricultural Services, MIN=Mining, CON=Construction, MAN=Manufacturing, TCU=Transportation, Communication and Utilities, WHO=Wholesale Trade, RET=Retail Trade, FIRE = Finance, Insurance and Real Estate, and SRV=Services.

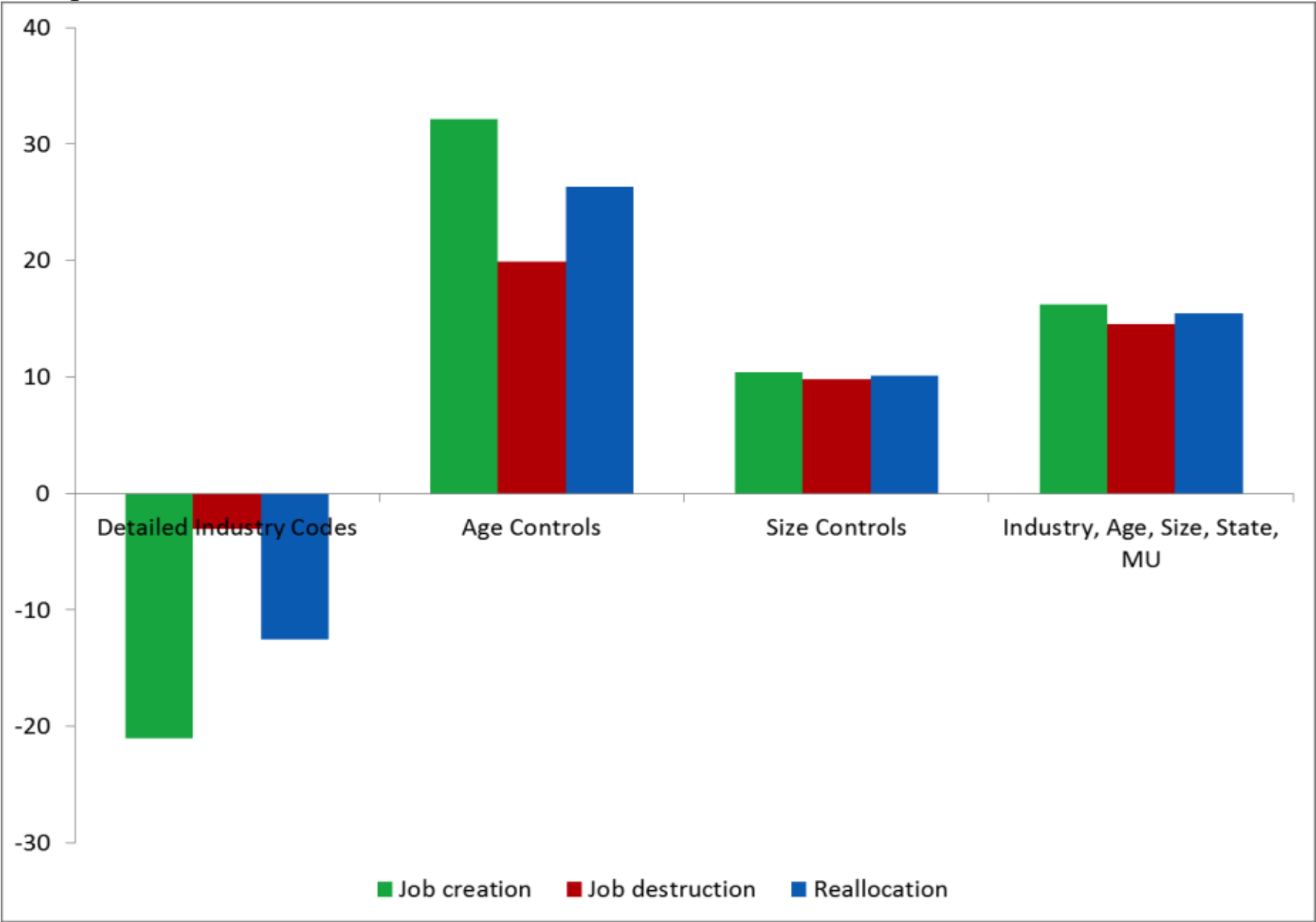
Figure 6: Average job flows by sector, 1980-2011



Notes: Author calculations from the U.S. Census Bureau's Business Dynamics Statistics. Sector definitions from SIC. See notes from Figure 5 for details on sectoral abbreviations.



Figure 9: Percent of decline in job flows from 1987/89 to 2004/06 (averages) accounted for by composition effects



Notes: Author calculations from the U.S. Census Bureau's Longitudinal Business Database. See text for details of the decomposition used to generate these calculations.

# NUTS AND BOLTS of PRODUCTIVITY MEASUREMENT

## Measurement of Plant-level Productivity

$$tfp_i = y_i - \alpha_l l_i - \alpha_k k_i - \alpha_m m_i - \alpha_e e_t$$

All variables in logs, difficult measurement Issues on outputs and inputs and factor elasticities

Typical to assume Cobb-Douglass or to have Divisia index approach approximation

# Measurement issues

- Factor inputs:
  - Labor quality
  - Capital stock (book value vs. perpetual inventory)
- Factor elasticities:
  - Cost shares, estimated elasticities using OLS, IV, proxy methods
  - All typically estimate factor elasticities at the industry level
    - Time invariant with estimated approach typically given Cobb-Douglass assumptions
  - Estimates vary in literature but measures of TFP highly correlated across these methods. Other issues (below) appear to matter more.
- Plant-level heterogeneity in output and input prices
- Plant-level heterogeneity in factor elasticities

# More Basic Measures of Productivity Are Often Used

- Labor productivity Measures at the Establishment (or Firm level)

- Real Value Added Per Worker

$$RLP_{et} = (VA_{et} / TE_{et}) = (Y_{et} - M_{et}) / TE_{et}$$

Where  $Y_{et}$  = Real Gross Output

$M_{et}$  = Real Materials (including energy)

$Te_{et}$  = Total Employment

Use detailed industry output and material price deflators

Often best available measure is real gross output per worker – comparable within industries

## Example of proxy method

$$y_{jt} = \beta_0 + \beta_k k_{jt} + \beta_a a_{jt} + \beta_l l_{jt} + \omega_{jt} + \eta_{jt} \quad (24)$$

$$i_{jt} = i(k_{jt}, a_{jt}, \omega_{jt}, \Delta_t) = i_t(k_{jt}, a_{jt}, \omega_{jt}). \quad (27)$$

$$\omega_{jt} = h_t(k_{jt}, a_{jt}, i_{jt}). \quad (28)$$

$$y_{jt} = \beta_0 + \beta_k k_{jt} + \beta_a a_{jt} + \beta_l l_{jt} + h_t(k_{jt}, a_{jt}, i_{jt}) + \eta_{jt}. \quad (29)$$

$$y_{jt} - \beta_l l_{jt} = \beta_0 + \beta_k k_{jt} + \beta_a a_{jt} + \omega_{jt} + \eta_{jt}. \quad (33)$$

$$\begin{aligned} y_{jt} - \beta_l l_{jt} \\ = \beta_0 + \beta_k k_{jt} + \beta_a a_{jt} + g(\omega_{jt-1}) + \xi_{jt} + \eta_{jt} \end{aligned} \quad (34a)$$

$$\begin{aligned} &= \beta_0 + \beta_k k_{jt} + \beta_a a_{jt} + g(\phi_{jt-1} - \beta_0 - \beta_k k_{jt-1} - \beta_a a_{jt-1}) + \xi_{jt} + \eta_{jt} \\ &= \beta_k k_{jt} + \beta_a a_{jt} + \tilde{g}(\phi_{jt-1} - \beta_k k_{jt-1} - \beta_a a_{jt-1}) + \xi_{jt} + \eta_{jt}, \end{aligned} \quad (34b)$$

Depends critically on the invertibility amongst other assumptions

$$Y = AK^\alpha L^{1-\alpha}$$

$$Y = \frac{D^{\varepsilon-1}}{P^\varepsilon}$$

$$\frac{rK}{wL} = \frac{\alpha}{1-\alpha}$$

$$APK = \frac{PY}{K} = \frac{\varepsilon}{\varepsilon-1} \frac{r}{\alpha}$$

$$APL = \frac{PY}{L} = \frac{\varepsilon}{\varepsilon-1} \frac{w}{1-\alpha}$$

$$TFPR = PA = \frac{\varepsilon}{\varepsilon-1} \left( \frac{w}{1-\alpha} \right)^{1-\alpha} \left( \frac{r}{\alpha} \right)^\alpha$$

$$TFPQ = A$$

Cobb-Douglas Technology, CRS  
Isoelastic Demand, No Frictions,  
Price takers in factor markets

No dispersion in factor cost  
share ratio, Revenue average  
product of capital, revenue average  
product of labor, TFPR

Even though there is dispersion  
In TFPQ

# Why is there so much dispersion in productivity across businesses in narrowly defined sectors?

- Background facts:
  - Interquartile range of log of Revenue TFP (TFPR) is 0.29
  - Interquartile range of log of Revenue Labor Productivity (RLP) is 0.65
  - Dispersion in TFPQ, TFPR, and output price within narrow product classes (7-digit) in U.S. (Source: FHS (2008)):
    - Std. Dev of  $\log(\text{TFPQ})$  is: 0.26
    - Std. Dev of  $\log(\text{TFPR})$  is: 0.22
    - Std. Dev of  $\log(\text{RLP})$  is: 0.65
    - Std. Dev of  $\log(P)$  is: 0.18
    - Std. Dev of  $\log(Q)$  is: 1.05
    - $\text{Corr}(\log(\text{TFPQ}), \log(P))$  is: -0.54
    - $\text{Corr}(\log(\text{TFPQ}), \log(Q))$  is: 0.28
    - $\text{Corr}(\log(\text{TFPQ}), \log(\text{TFPR}))$  is: 0.75
    - $\text{Corr}(\log(\text{TFPQ}), \log(\text{RLP}))$  is: 0.56



# Frictions + Distortions

- Costs of Entry (and exit)
  - Including costs of entering new markets
  - Hopenhayn (1992), Melitz (2003), Melitz and Ottaviano (2005)
- Learning (initial conditions and after changing products/processes)
  - Jovanovic (1982) and Ericson and Pakes (1998)
  - Experimentation
- Adjustment costs for factors of production (capital, labor, intangible capital)
  - Convex vs. Nonconvex
- Economies of scope and control
- Product Differentiation:
  - Horizontal (e.g., spatial) vs. Vertical
- Output and input price dispersion and determination
- Imperfections in product, labor, capital, credit markets
- Distortions to all of the above + market institutions
  - Idiosyncratic distortions as in Banerjee and Duflo (2003), Restuccia and Rogerson (2008), Hsieh and Klenow (2009), Bartelsman, Haltiwanger and Scarpetta (2013)

# What frictions matter the most?

- Many studies showing evidence of entry costs, labor adjustment costs, capital adjustment costs, trade costs, product differentiation, and so on.
- Many open questions and issues:
  - Not practical to include all frictions in all models – but caution about identification since we are all using same data
  - How do frictions vary across advanced vs. emerging vs. transition?
- Important to distinguish between those frictions that yield some plants persistently higher productivity than others as opposed to adjustment dynamics

# Lots of margins for distortions...

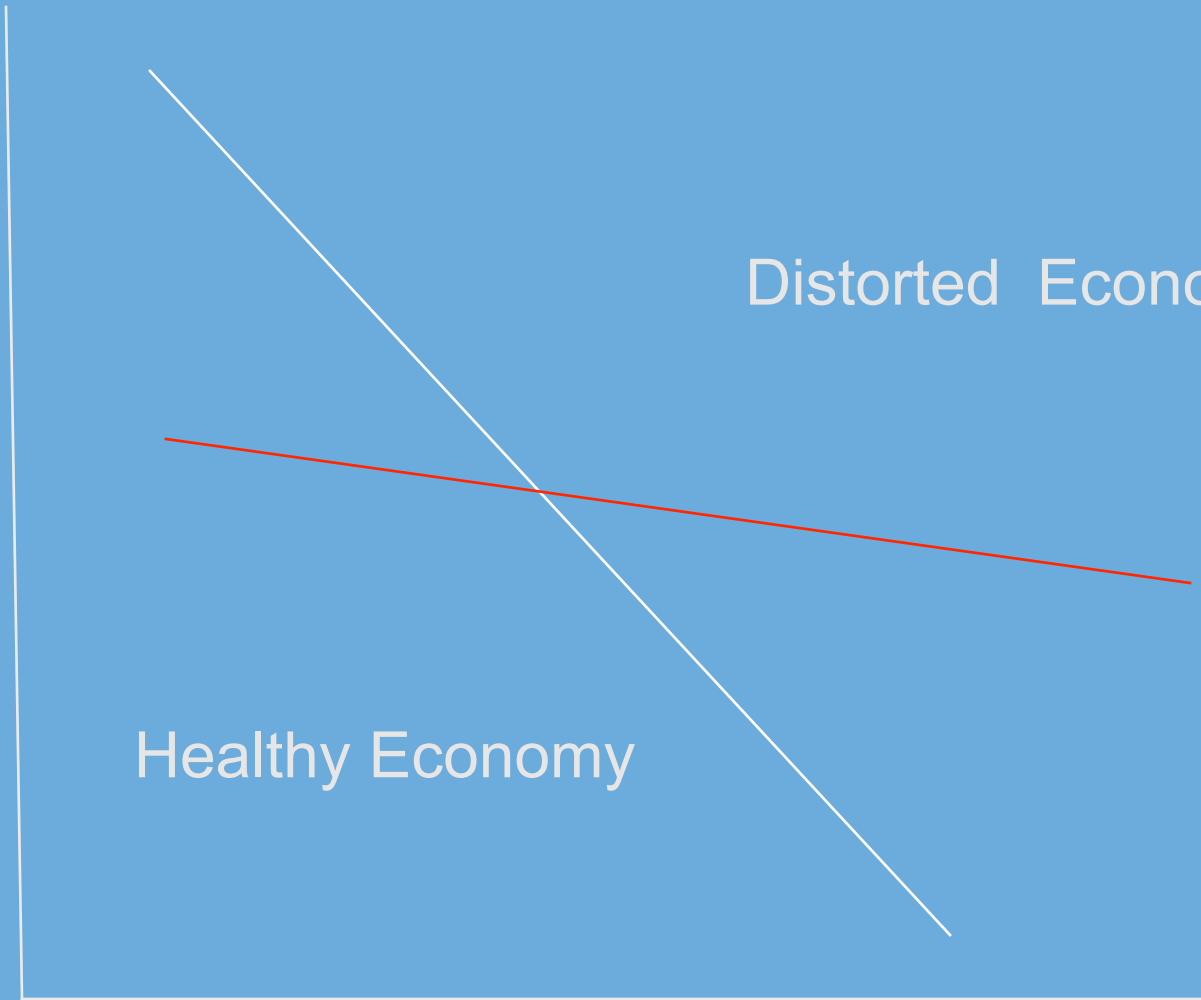
- Cross sectional misallocation
- Dynamic distortions:
  - Startups
  - Post-entry up or out dynamic
  - Creative Destruction
- Secular vs. Cyclical Distortions

Prob  
Of  
Exit  
(firm)

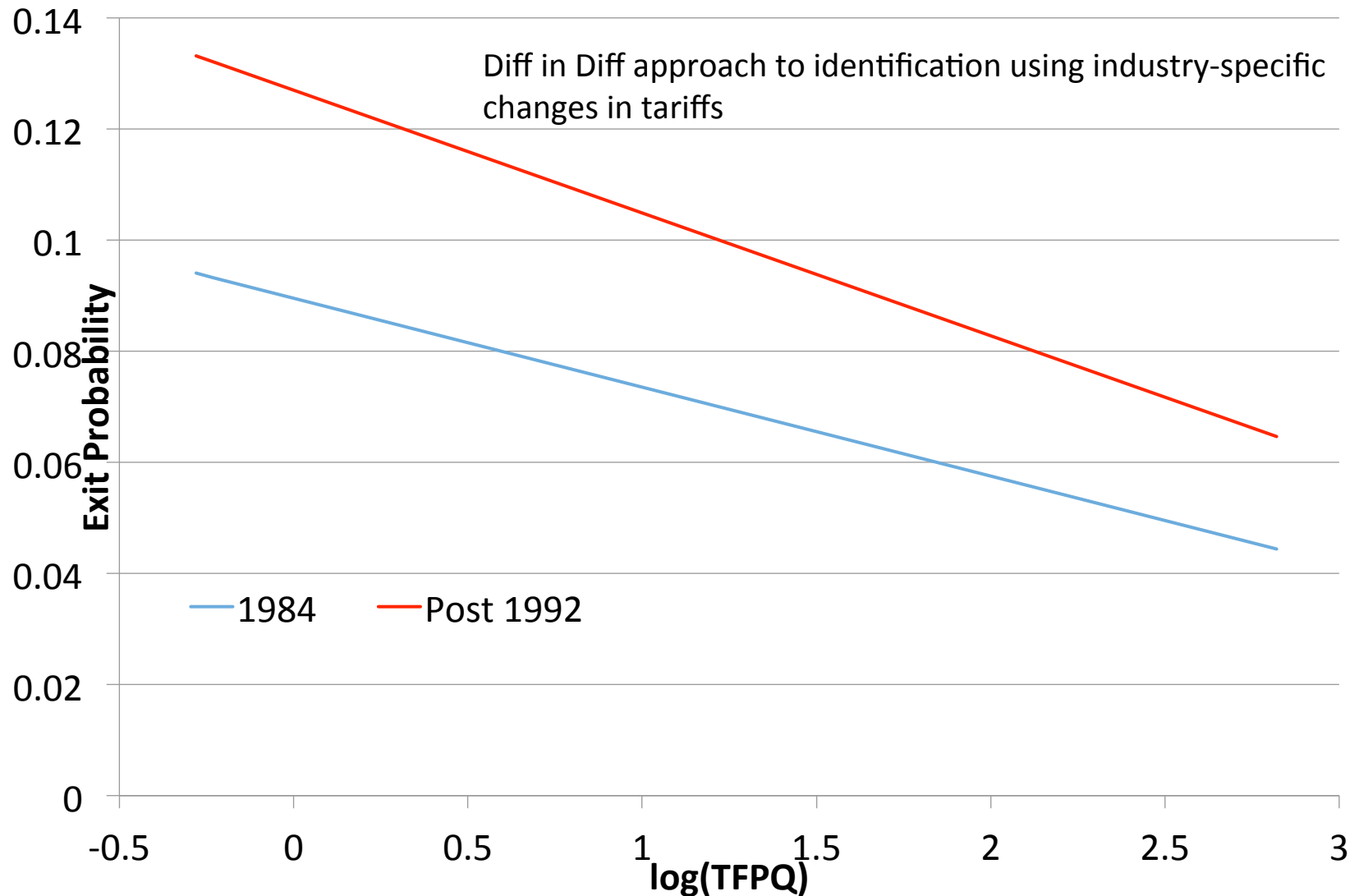
Distorted Economy

Healthy Economy

Firm Productivity



## Impact of Trade Reform on Plant Exit Hazard in Colombia



Source: Eslava, Haltiwanger, Kugler and Kugler (2012)

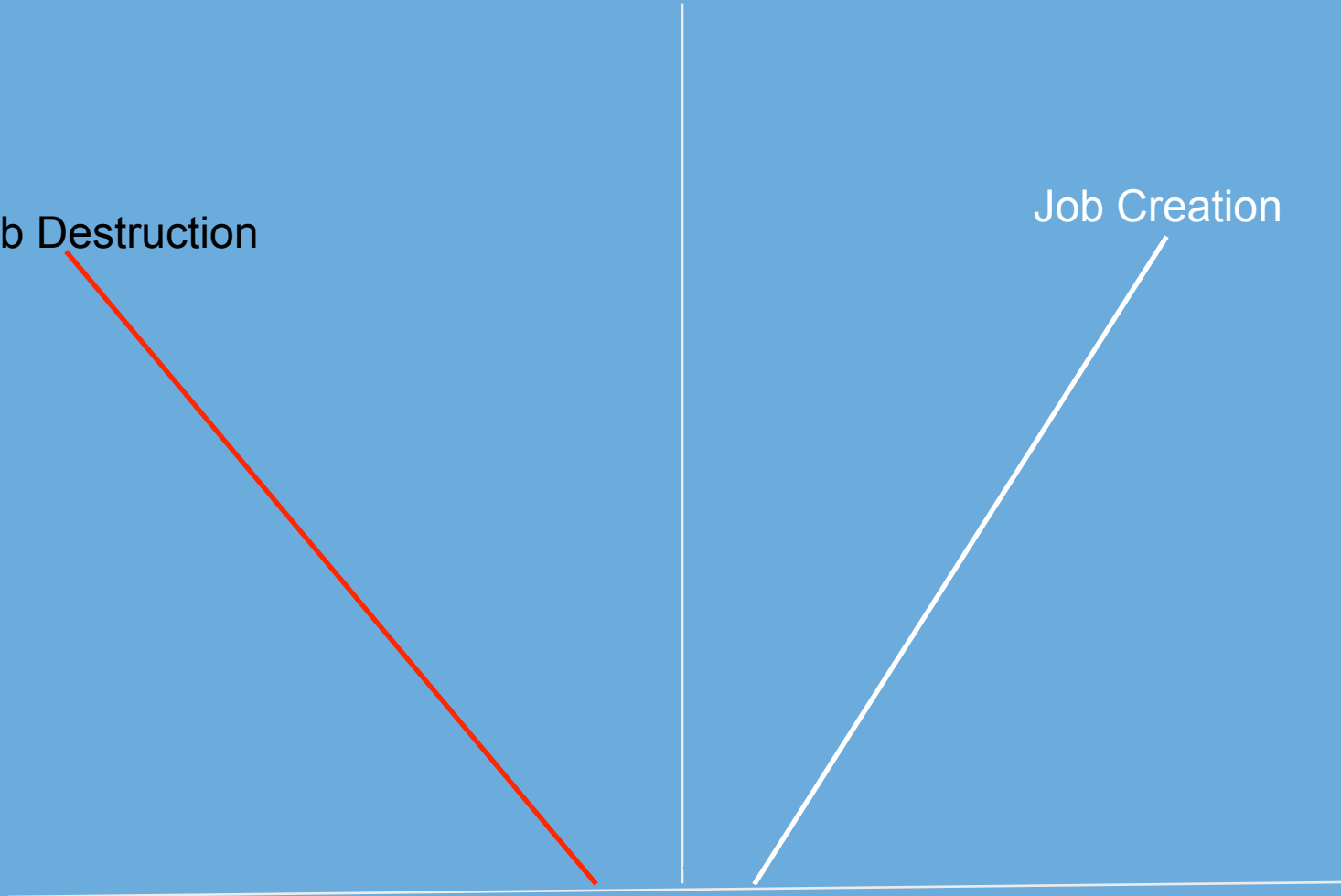
# Firm Employment Changes

Job Destruction

Job Creation

Range of Inaction

Firm Productivity Shock  
(Profitability)



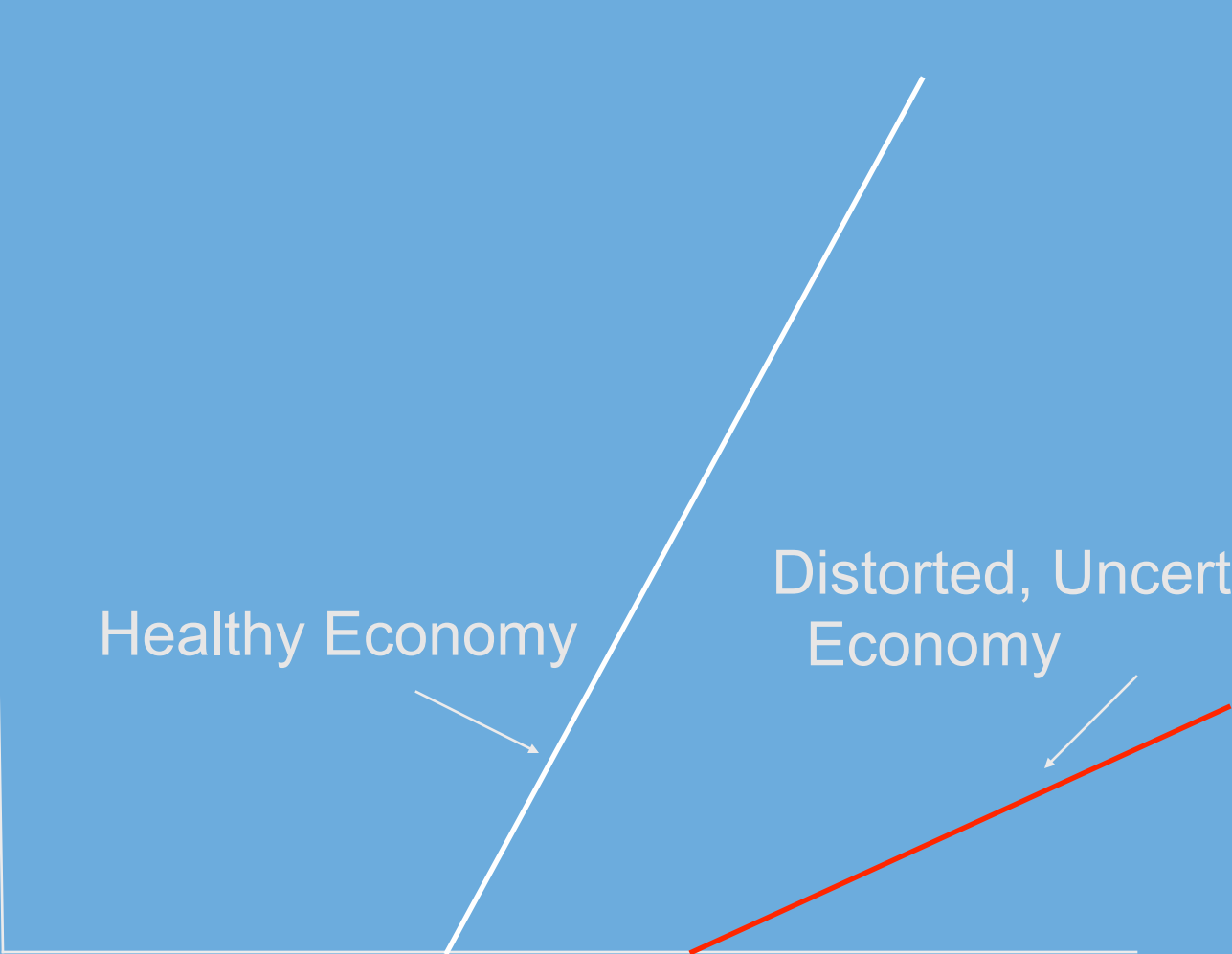
Job  
Creation

Healthy Economy

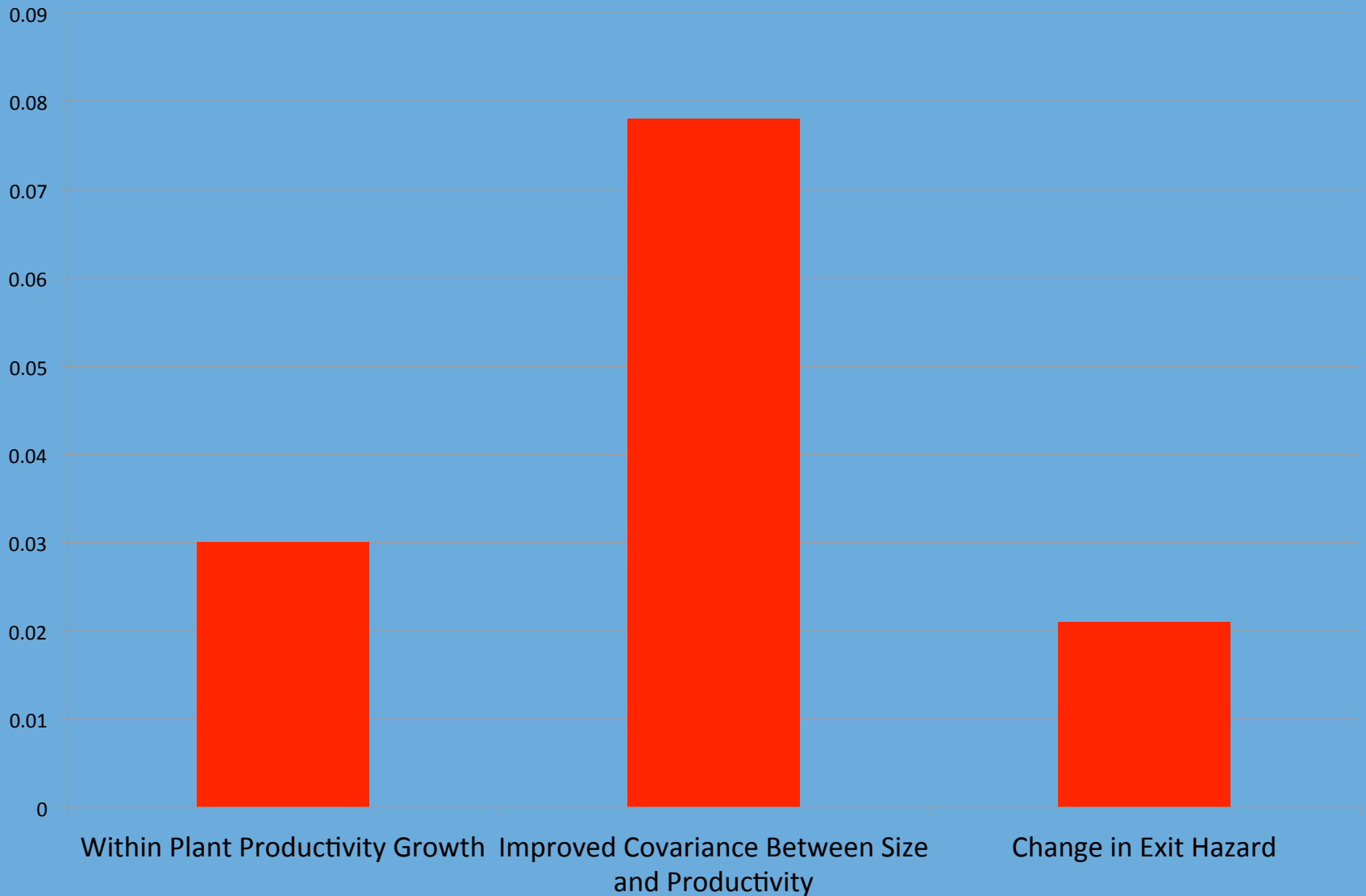
Distorted, Uncertain  
Economy

Range of inaction  
(increases with uncertainty and distortions)

Firm Productivity Shock



## Impact of Trade Reform on TFP(Q) in Colombia



Source: Eslava, Haltiwanger, Kugler and Kugler (2012)



# Taking Stock

- High pace of churning of businesses within narrowly defined industries
- Startups and young businesses play an important role in these dynamics
- Up or out dynamics
- These dynamics connected to productivity (and demand) dynamics at the micro level
- Identifying the frictions and how they vary across industry, time, and country ongoing activity
- But what about before entry?

How Do We measure the  
CONTRIBUTION OF  
REALLOCATION?

# Size/productivity relationship within industries

$$\begin{aligned}\Omega_t &= \sum_i s_{it} \omega_{it} \\ &= (1/N_t) \sum_i \omega_{it} + \sum_i (s_{it} - (1/N_t) \sum_i s_{it}) (\omega_{it} - (1/N_t) \sum_i \omega_{it})\end{aligned}$$

Olley and Pakes (1996) decomposition

$$\begin{aligned}\Delta \Omega_t &= \sum_i s_{it} \omega_{it} - \sum_i s_{it-1} \omega_{it-1} \\ &= \sum_{i \in C} \overline{s_{it}} \Delta \omega_{it} + \sum_{i \in C} \Delta s_{it} (\overline{\omega_{it}} - \overline{\Omega_t}) + \sum_{i \in N} s_{it} (\omega_{it} - \overline{\Omega_t}) - \sum_{i \in X} s_{it-1} (\omega_{it-1} - \overline{\Omega_t}) \\ &= \textit{within} + \textit{reallocation} + \textit{entry} - \textit{exit}\end{aligned}$$

Modified Baily, Hulten and Campbell (1992) and Griliches and Regev (1995) decomposition

# Comments on Decomposition in Literature

- Some questions about how to interpret industry-level index defined in this manner
  - Typical check (e.g., BHC and FHK) to see how this index performs relative to standard aggregate **industry** measures
    - Common result – magnitudes very similar and correlations high in most studies
    - Cautions:
      - These measures very sensitive to measurement error since depend on measuring within industry productivity (log) level dispersion accurately
      - Not appropriate for decompositions that exploit between industry variation (measurement and index problems)
  - Standard decomposition summarizes changes in activity weighted micro distribution
  - Decompositions more closely tied to aggregate welfare and productivity have been developed (Petrin and Levinsohn (2008), Basu and Fernald (2002))
  - Alternatively, these decompositions can be used as moments to match in a calibration or indirect inference approach (see, e.g., Bartelsman, Haltiwanger and Scarpetta (2009))

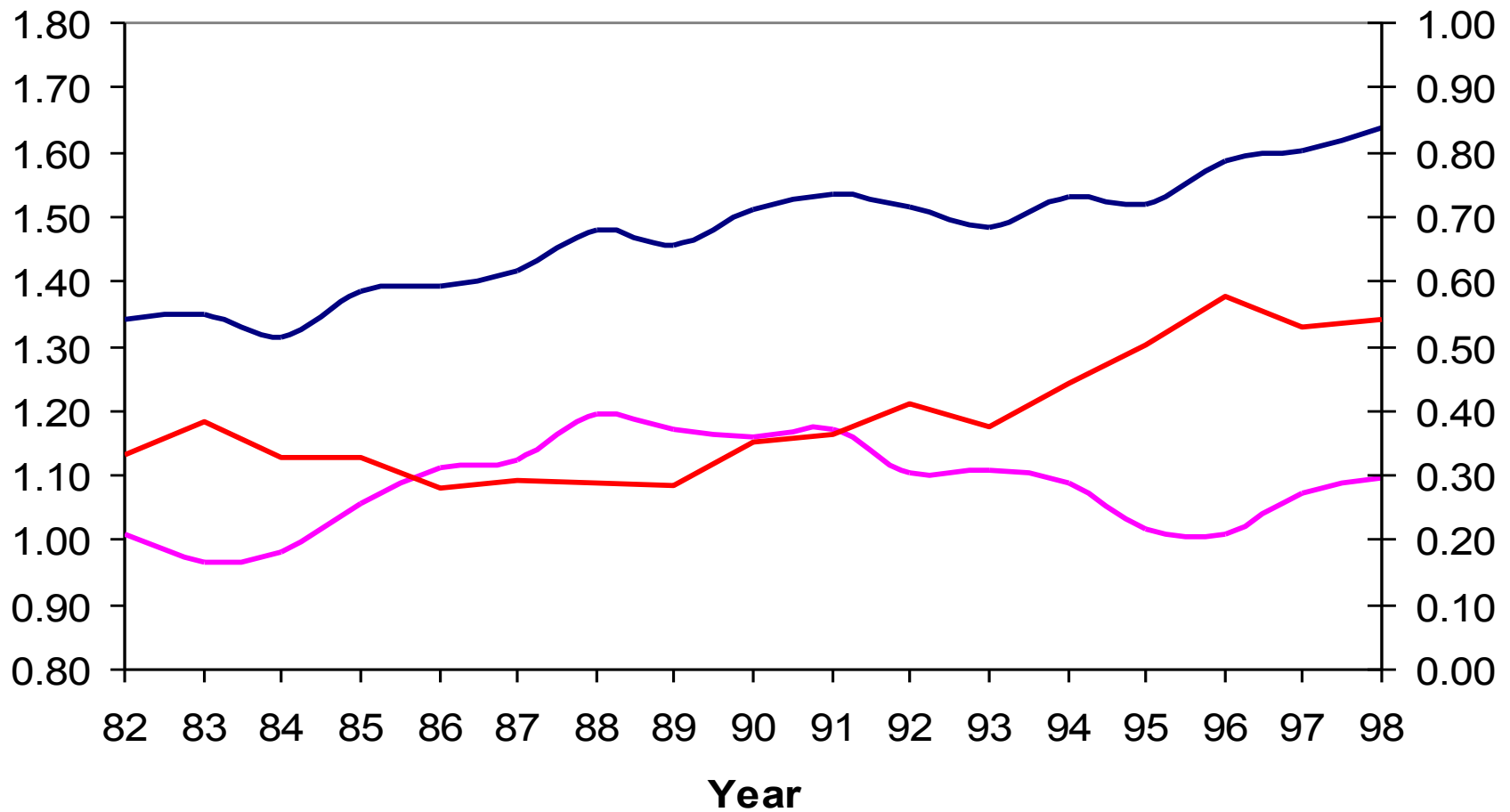
# Olley and Pakes (1996) results for Telecommunications equipment

TABLE XI  
DECOMPOSITION OF PRODUCTIVITY<sup>a</sup>  
(EQUATION (16))

Year	$p_t$	$\bar{p}_t$	$\Sigma_t \Delta s_{it} \Delta p_{it}$	$\rho(p_t, k_t)$
1974	1.00	0.90	0.01	-0.07
1975	0.72	0.66	0.06	-0.11
1976	0.77	0.69	0.07	-0.12
1977	0.75	0.72	0.03	-0.09
1978	0.92	0.80	0.12	-0.05
1979	0.95	0.84	0.12	-0.05
1980	1.12	0.84	0.28	-0.02
1981	1.11	0.76	0.35	0.02
1982	1.08	0.77	0.31	-0.01
1983	0.84	0.76	0.08	-0.07
1984	0.90	0.83	0.07	-0.09
1985	0.99	0.72	0.26	0.02
1986	0.92	0.72	0.20	0.03
1987	0.97	0.66	0.32	0.10

<sup>a</sup>See text for details.

## Olley-Pakes Decomposition for Colombian Manufacturing

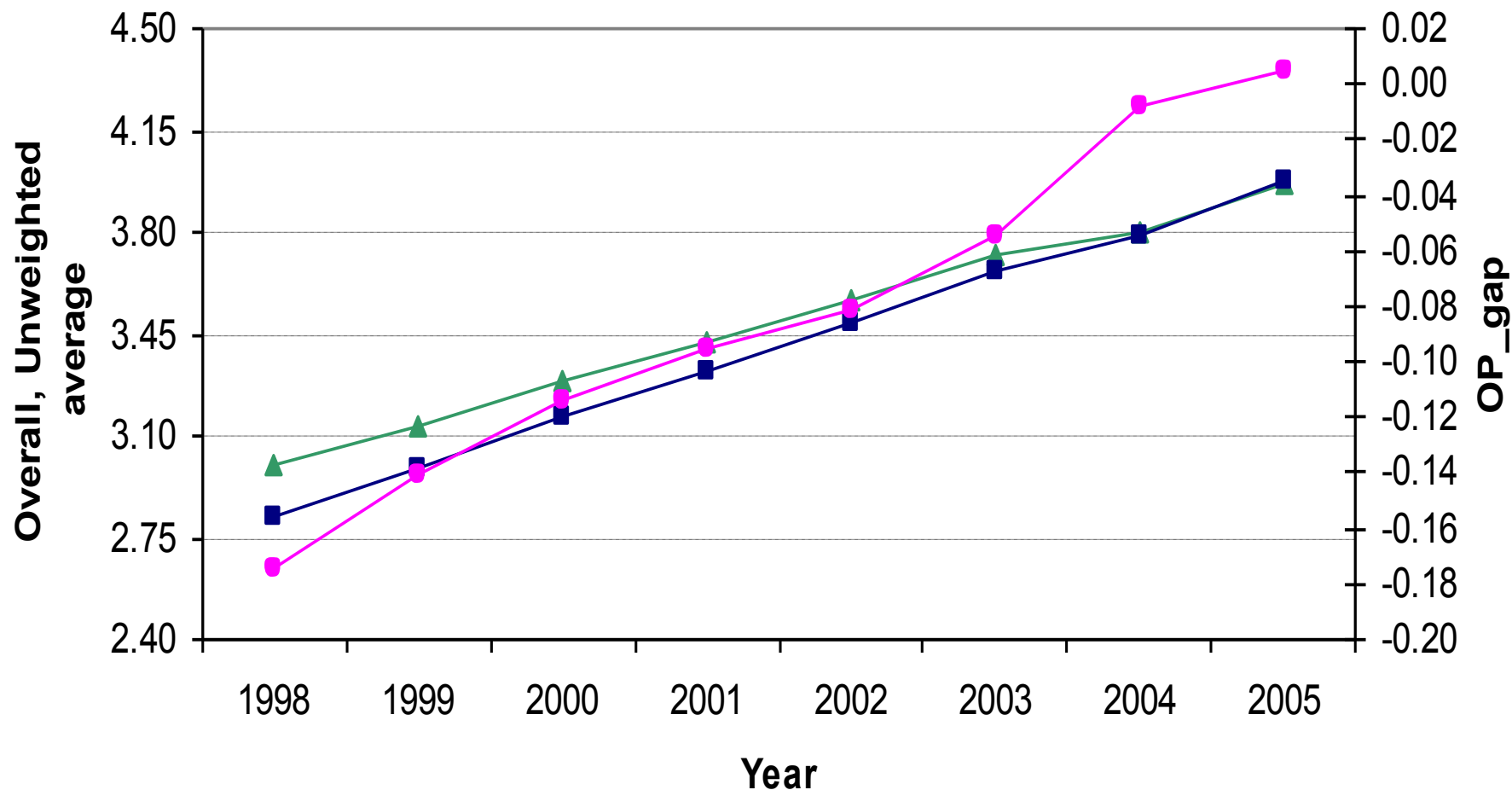


— Aggregate (Weighted) — Simple Average — Cross-term

Source: Eslava et al. (2005)

# Olley Pakes Decomposition of Labor Productivity (Average Industry)

China



—▲— Unweighted Average —■— Overall —●— OP\_gap

# Pre-Entry History of entrepreneurs

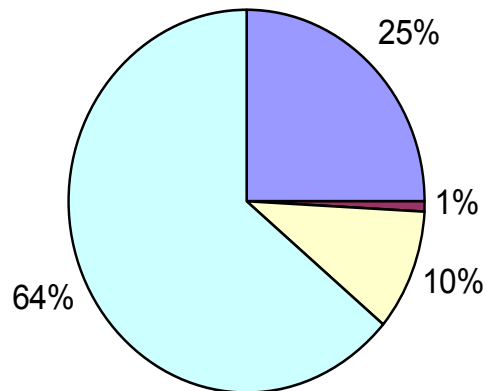


# “Before” Entry....

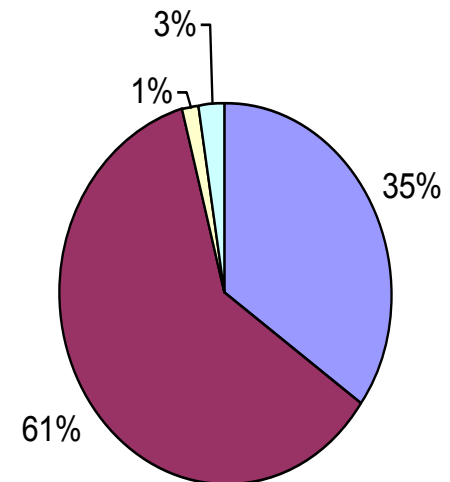
- Entrepreneurial dynamics starts at micro business level
- Entrepreneurs start with an idea – often while employed elsewhere
- New longitudinal databases at U.S. Census Bureau tracking this process
  - ILBD: Nonemployers (e.g., sole props without employees) + Employers
  - LEHD/SED: Tracking transitions from W&S jobs to self-employed jobs

Micro Businesses constitute a large share of businesses and a small share of revenue...

**Distribution of Businesses by Business Type, 2000**

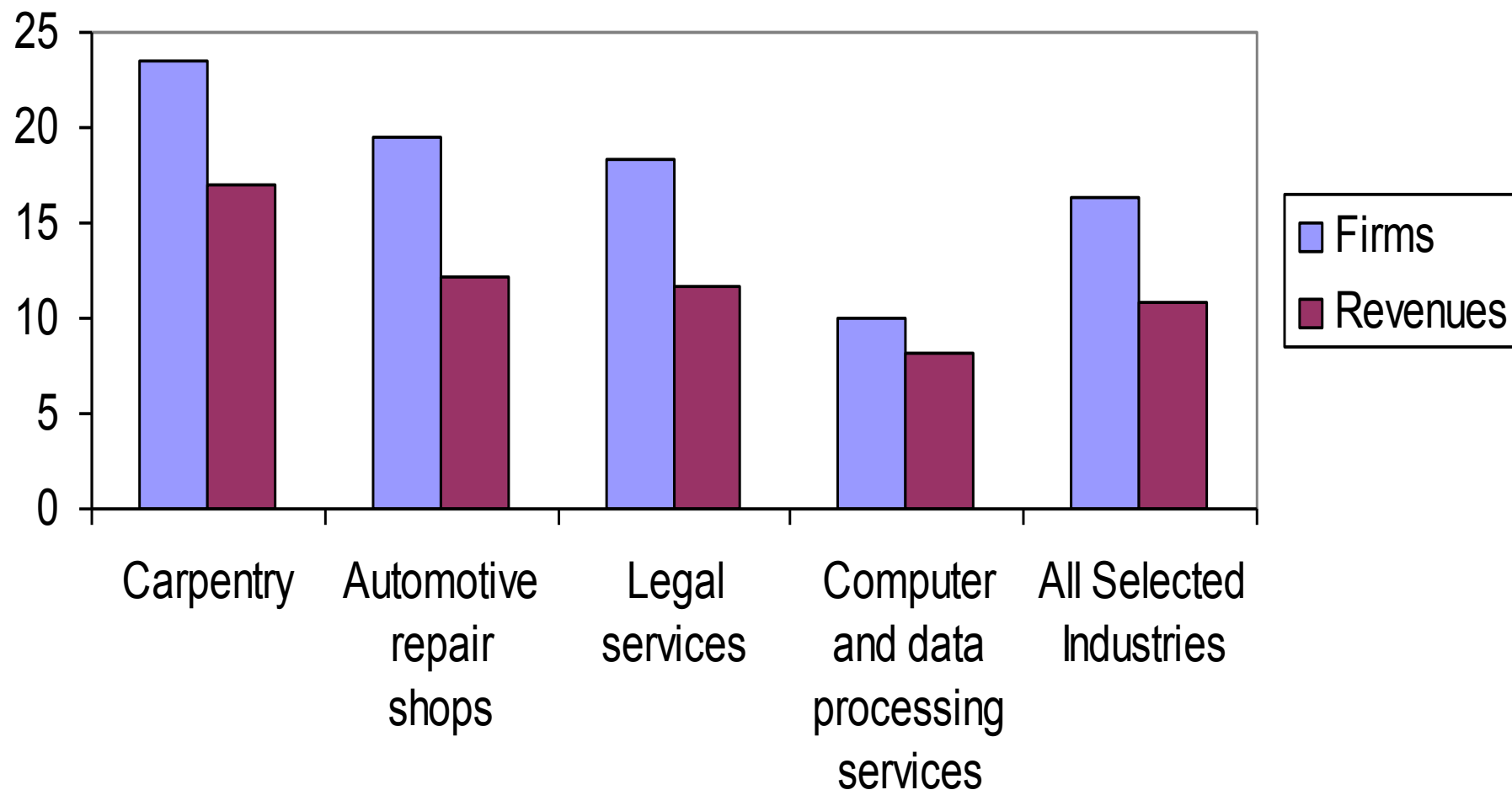


**Distribution of Revenue By Business Type, 2000**



Source: Davis et. al. (2008)

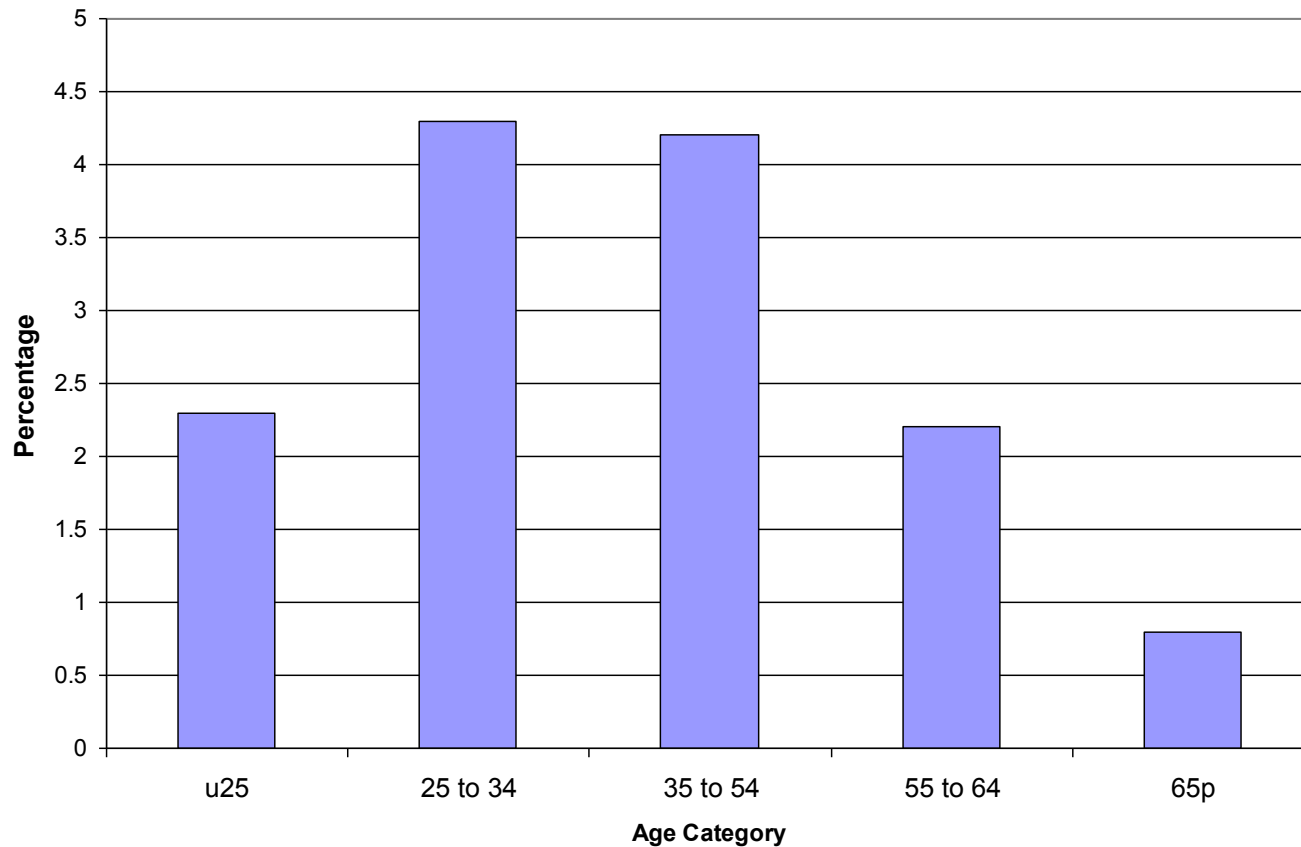
# Shares of New Employer Businesses in 1997 with Pre-History as Nonemployer Businesses



Source: Davis et al. (2008)

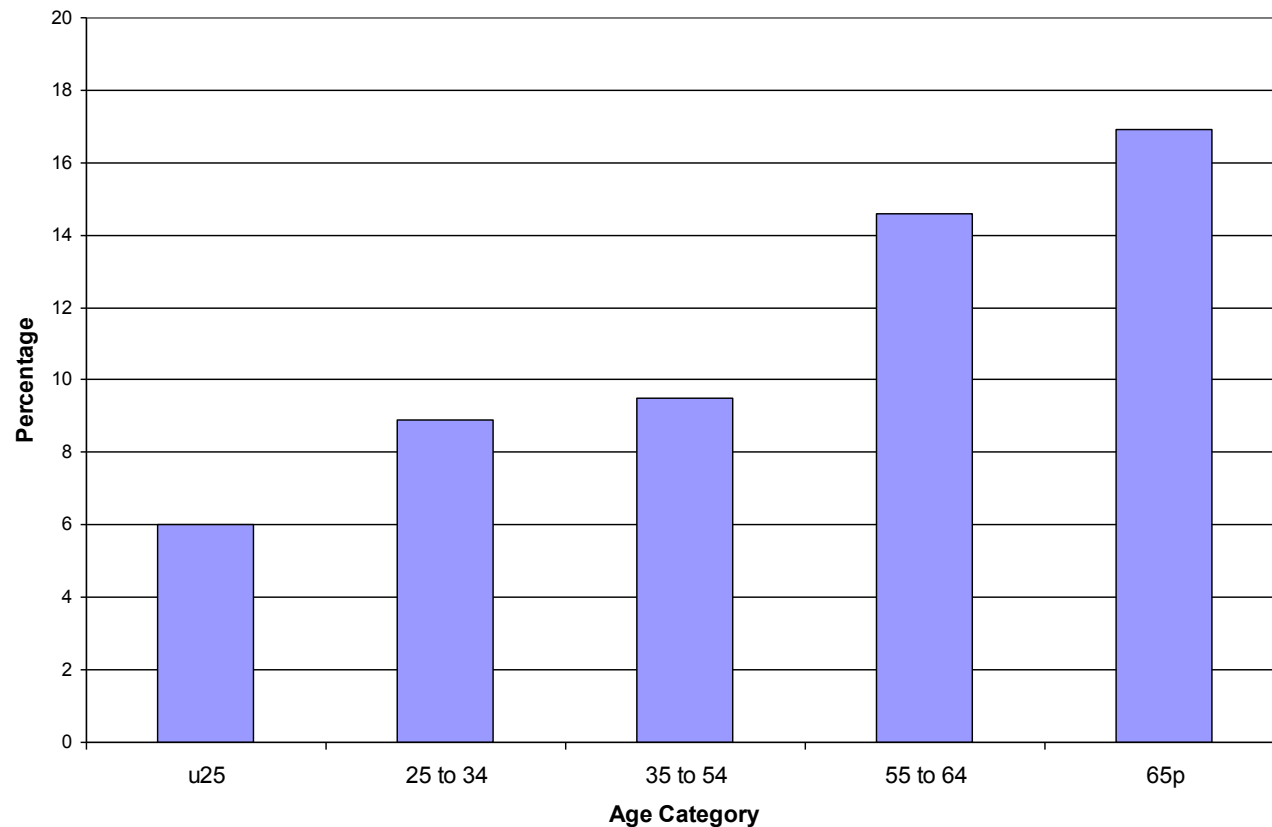
# Propensity to Diversify in Labor Market Varies in Important Ways Across Worker Life Cycle

Percent of 1992 Wage and Salary Earners moving to Partial Self-Employment by 1997:  
By Age Category



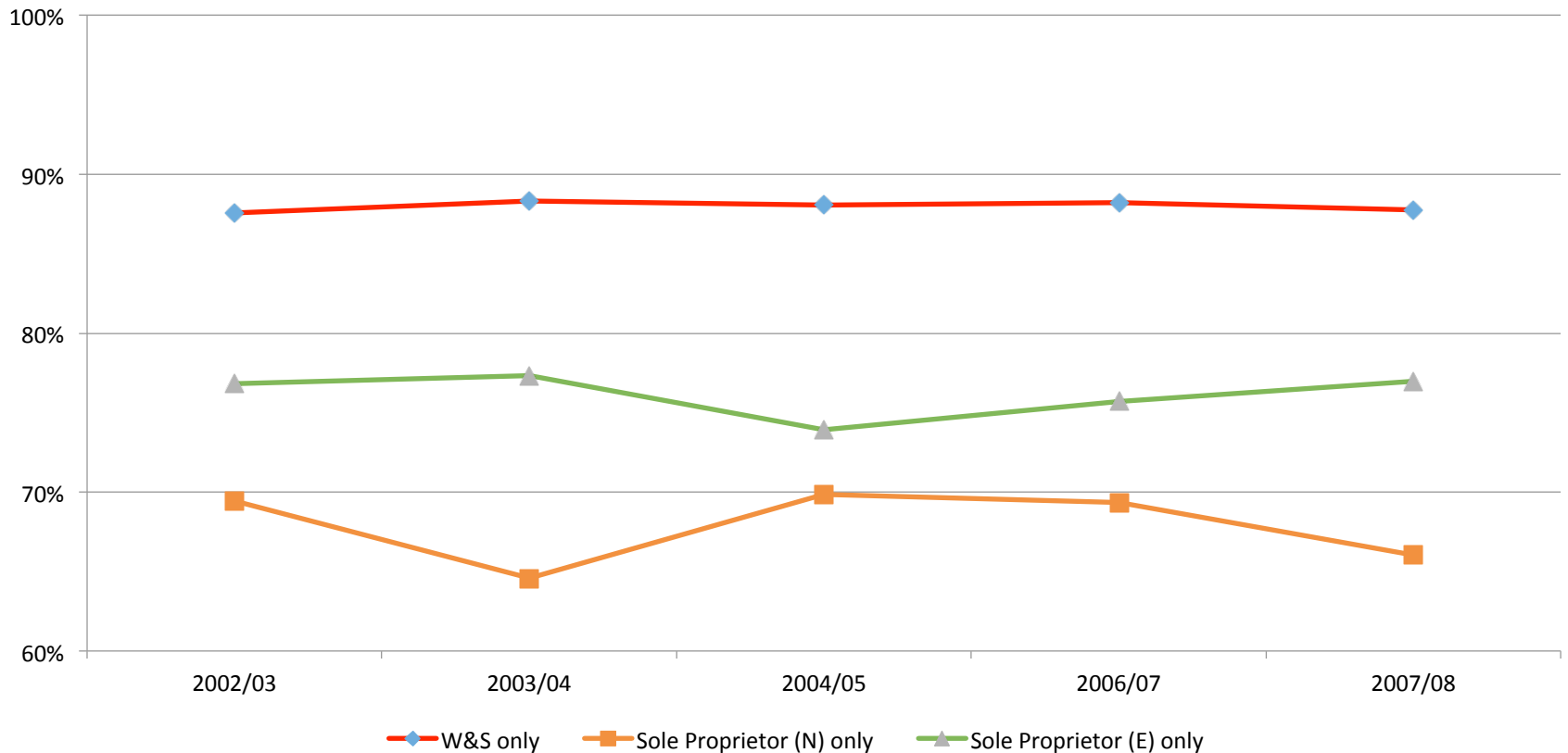
# Propensity to Diversify in Labor Market Varies in Important Ways Across Worker Life Cycle

Percent of 1992 Partially Self-Employed moving to Full Self-Employment by 1997: By Age Category



# Small Businesses With and Without Paid Employees Differ in Fundamental Ways

## Job Stability -- Likelihood of Staying in Same Labor Market State



# Data

- Tracking U.S. Business Dynamics
  - The Longitudinal Business Database
    - 1975-2005 (08) – long time series permits analysis by firm age
    - Private Non Farm Economy
    - Establishment level with Firm identifiers
    - High quality establishment links to identify entry/exit
      - Need both firm and establishment level data to get dynamics right
    - Firm Size: constructed by aggregating employment up to firm
    - Firm Age: constructed from age of oldest establishment at time of firm birth
    - Other: Payroll, Industry, Location (Lat/Lon possible)
    - Can be integrated with data from Economic Censuses and Annual Surveys as well as external data (COMPUSTAT, Venture Capital, Private Equity)

# Micro Productivity Data in U.S.

- Manufacturing:

- Annual Survey of Manufactures and Census of Manufactures
  - Nominal revenue and expenditures
  - Can construct measures of real outputs and inputs
  - Five year panel rotation so longitudinal analysis possible (but requires careful treatment of data)
  - Selected products have physical quantities

- Retail Trade

- Census of Retail Trade
  - Nominal revenue so a gross output per store measure feasible



# New data on micro businesses

- ILBD:
  - Tracks all nonemployer and employer businesses including transitions
- LEHD:
  - Tracks all employer-employee matches in U.S.
  - Can be integrated with ILBD
  - Enables tracking of transitions between W&S, an owner of nonemployer business and owner of employer business

# Availability of data

- Public domain tabulations available at:

[http://www.ces.census.gov/index.php/bds/bds\\_home](http://www.ces.census.gov/index.php/bds/bds_home)

- Census NSF/RDC access at:

<http://www.ces.census.gov/index.php/ces/researchguidelines>

- Sensitive data:

- Must work in enclave (NBER, NYCRDC, Washington, D.C., Chicago Fed, Duke, UCLA, UC-Berkeley, Univ. of Michigan, Cornell, Stanford , Univ. of Minn., Atlanta, ...)
- Predominant purpose must benefit U.S. Census

# Extra Slides on Firm Dynamics model

# *Standard Heterogeneous-Producer Industry Models*

The Workhorse:

- Producers  $i$  differ in a profitability component  $\omega_i$ , usually taken to represent costs/productivity
- Profits depend on  $\omega_i$  and industry state  $S$ :  $\pi_i = \pi_i(\omega_i, S)$     $\omega_i \sim G(\omega)$
- There is some critical  $\omega^*$  such that producers with  $\omega_i < \omega^*$  have NPVs below outside option and therefore exit the industry
- Industry state  $S$  typically depends on endogenously determined distribution of  $\omega_i$  among producers (add'l free entry assumption)
- Examples: Jovanovic (1982), Hopenhayn (1992), Melitz (2003), Asplund and Nocke (2007)

## *Closely Related Issue – Size Distribution of Activity*

- $\pi_i = \pi_i(\omega_i, S)$  has curvature either from decreasing returns (e.g., Lucas (1978)) or product differentiation (e.g., Melitz (2003))
- Curvature pins down the size distribution of activity and permits studying the evolution of the size distribution of activity
- In healthy market economies, most productive plants are the largest – allocative efficiency
- Active literature attempting to explain cross country differences in productivity (e.g., Hsieh and Klenow (2009)) using distortions on this margin

## *Model: Melitz/Ottaviano (2005) and FHS (2008)*

Industry is comprised of a continuum of producers of measure  $N$ .

Each produces a single variety (indexed by  $i$ ) of industry product.

Representative consumer's utility function

$$\begin{aligned} U &= y + \int_{i \in I} (\alpha + \delta_i) q_i di - \frac{1}{2} \eta \left( \int_{i \in I} q_i di \right)^2 - \frac{1}{2} \gamma \int_{i \in I} q_i^2 di \\ &= y + \alpha \int_{i \in I} q_i di - \frac{1}{2} \left( \eta + \frac{\gamma}{N} \right) \left( \int_{i \in I} q_i di \right)^2 + \int_{i \in I} \delta_i q_i di - \frac{1}{2} \gamma \int_{i \in I} (q_i - \bar{q})^2 di \end{aligned}$$

$\alpha > 0$ ,  $\eta > 0$ , and  $\gamma \geq 0$ .

$y$  = numeraire good

$\delta_i$  = variety-specific, mean-zero taste shifter

$q_i$  = quantity of good  $i$  consumed

$$\bar{q} = \frac{1}{N} \int_{i \in I} q_i di$$

The implied demand curve:

$$q_i = \frac{1}{\eta N + \gamma} \alpha - \frac{1}{\gamma} \frac{\eta N}{\eta N + \gamma} \bar{\delta} + \frac{1}{\gamma} \frac{\eta N}{\eta N + \gamma} \bar{p} + \frac{1}{\gamma} \delta_i - \frac{1}{\gamma} p_i$$

## Model: Supply

Production Function:  $q_i = \omega_i x_i$

Producers face (potentially idiosyncratic) factor price  $w_i$

$\Rightarrow$  marginal cost =  $w_i/\omega_i$

Profits:

$$\pi_i = \left( \frac{1}{\eta N + \gamma} \alpha - \frac{1}{\gamma} \frac{\eta N}{\eta N + \gamma} \bar{\delta} + \frac{1}{\gamma} \frac{\eta N}{\eta N + \gamma} \bar{p} + \frac{1}{\gamma} \delta_i - \frac{1}{\gamma} p_i \right) \left( p_i - \frac{w_i}{\omega_i} \right)$$

Profit-maximizing price (constant marginal cost  $c_i$ ):

$$p_i = \frac{1}{2} \frac{\gamma}{\eta N + \gamma} \alpha - \frac{1}{2} \frac{\eta N}{\eta N + \gamma} \bar{\delta} + \frac{1}{2} \frac{\eta N}{\eta N + \gamma} \bar{p} + \frac{1}{2} \delta_i + \frac{1}{2} \frac{w_i}{\omega_i}$$

Deviation from industry-average price:

$$p_i - \bar{p} = \frac{1}{2} (\delta_i - \bar{\delta}) + \frac{1}{2} \left( \frac{w_i}{\omega_i} - \overline{\left( \frac{w}{\omega} \right)} \right)$$

Maximized profits:

$$\pi_i = \frac{1}{4\gamma} \left( \frac{\gamma}{\eta N + \gamma} \alpha - \frac{\eta N}{\eta N + \gamma} \bar{\delta} + \frac{\eta N}{\eta N + \gamma} \bar{p} + \delta_i - \frac{w_i}{\omega_i} \right)^2$$

## Model: Equilibrium

Equilibrium Condition 1: The marginal producer in the industry makes zero profits

Define “profitability index”  $\phi_i \equiv \delta_i - \frac{w_i}{\omega_i}$ . Then marginal producer has index equal to:

$$\phi^* = -\frac{\gamma}{\eta N + \gamma} \alpha + \frac{\eta N}{\eta N + \gamma} \bar{\delta} - \frac{\eta N}{\eta N + \gamma} \bar{p}$$

Profits can be rewritten in terms of this marginal profitability level

$$\pi_i = \frac{1}{4\gamma} (\phi_i - \phi^*)^2$$

Profits increase in demand ( $\delta_i$ ) and efficiency ( $\omega_i$ ), decrease in factor price ( $w_i$ )

Equilibrium Condition 2: Potential entrants decide whether to pay sunk entry cost  $s$  to learn  $\delta_i, \omega_i, w_i$ . Expected value of entry is 0.

$$V^e = \int_0^{w_u} \int_{\omega_l}^{\omega_u} \int_{\phi^* + \frac{w}{\omega}}^{\delta_e} \frac{1}{4\gamma} (\phi_i - \phi^*)^2 f(\delta, \omega, w) d\delta d\omega dw - s = 0$$



Selection effect:

- Only high-profitability producers operate in equilibrium
- Low types exit

Sunk costs, market power and dispersion:

- Sunk costs make entry costly
- Curvature yields equilibrium size distribution

Many models of selection also include fixed costs of operating each period

## Model: Empirical Implications

Output-based productivity:

$$TFPQ_i = \frac{q_i}{x_i} = \frac{\omega_i x_i}{x_i} = \omega_i$$

Revenue-based productivity (literature standard):

$$TFPR_i = \frac{p_i q_i}{x_i} = p_i \omega_i = \frac{1}{2} \frac{\gamma \alpha}{\eta N + \gamma} \omega_i + \frac{1}{2} \frac{\eta N}{\eta N + \gamma} (\bar{p} - \bar{\delta}) \omega_i + \frac{1}{2} \delta_i \omega_i + \frac{1}{2} w_i$$

---

Plant price deviation from industry deflator depends on both demand (enters positively into profits) and costs (enter negatively):

$$p_i - \bar{p} = \frac{1}{2} (\delta_i - \bar{\delta}) + \frac{1}{2} \left( \frac{w_i}{\omega_i} - \overline{\left( \frac{w}{\omega} \right)} \right)$$

---

Comparative Statics:

- $\frac{d\phi^*}{d\gamma} < 0$ : Lower substitutability (higher  $\gamma$ ) lowers  $\phi^*$
- $\frac{d\phi^*}{ds} < 0$ : Higher sunk entry cost lowers  $\phi^*$

# Start with Foster, Haltiwanger and Syverson (2008)

- Source data: Census of Manufactures
  - High quality coverage
  - Limited number of products with physical quantity data

Correlations								
Variables	Trad'l. Output	Revenue Output	Physical Output	Price	Trad'l. TFP	Revenue TFP	Physical TFP	Capital
Traditional Output	1.00							
Revenue Output	0.99	1.00						
Physical Output	0.98	0.99	1.00					
Price	-0.03	-0.03	-0.19	1.00				
Traditional TFP	0.19	0.18	0.15	0.13	1.00			
Revenue TFP	0.17	0.21	0.18	0.16	0.86	1.00		
Physical TFP	0.17	0.20	0.28	-0.54	0.64	0.75	1.00	
Capital	0.86	0.85	0.84	-0.04	0.00	-0.00	0.03	1.00
Standard Deviations								
Standard Deviations	1.03	1.03	1.05	0.18	0.21	0.22	0.26	1.14

# Measuring Plant-Level Demand

Estimate product demand curves; plant-specific residual is idio. demand

$$\ln q_{it} = \alpha_o + \alpha_1 \ln p_{it} + \alpha_2 \ln(INCOME_{mt}) + \sum_t \alpha_t YEAR_t + \eta_{it}$$

$q_{it}$ —physical output of plant  $i$  in year  $t$

$p_{it}$ —plant unit price

$INCOME_{mt}$ —average income in the plant's local market  $m$

$YEAR_t$ —year dummy

$\eta_{it}$ —plant-year disturbance term

Plant demand:

$$\hat{\delta}_{it} = \hat{\eta}_{it} + \hat{\alpha}_2 \ln(INCOME_{mt}) = \ln q_{it} - \hat{\alpha}_o - \hat{\alpha}_1 \ln p_{it} - \sum_t \hat{\alpha}_t YEAR_t$$

I.e., residual is plant quantity sold that can't be accounted for by unit price or local income differences

- Use TFPQ<sub>it</sub> to instrument for prices (captures production costs)

Product	IV Estimation		OLS Estimation	
	Price Coefficient		Income Coefficient	Price Coefficient
	$(\alpha_1)$		$(\alpha_2)$	$(\alpha_1)$
Boxes	-3.02		-0.03	-2.19
	<i>0.17</i>	[0.61]	<i>0.02</i>	<i>0.12</i>
Bread	-3.09		0.12	-0.89
	<i>0.42</i>	[0.33]	<i>0.05</i>	<i>0.15</i>
Carbon Black	-0.52		-0.21	-0.57
	<i>0.38</i>	[0.50]	<i>0.11</i>	<i>0.21</i>
Coffee	-3.63		0.22	-1.03
	<i>0.98</i>	[0.41]	<i>0.14</i>	<i>0.32</i>
Concrete	-5.93		0.13	-0.83
	<i>0.36</i>	[0.10]	<i>0.01</i>	<i>0.09</i>
Hardwood Flooring	-1.67		-0.20	-0.87
	<i>0.48</i>	[0.61]	<i>0.18</i>	<i>0.47</i>
Gasoline	-1.42		0.23	-0.16
	<i>2.72</i>	[0.20]	<i>0.07</i>	<i>0.80</i>
Block Ice	-2.05		0.00	-0.63
	<i>0.46</i>	[0.32]	<i>0.11</i>	<i>0.20</i>
Processed Ice	-1.48		0.18	-0.70
	<i>0.27</i>	[0.37]	<i>0.03</i>	<i>0.13</i>
Plywood	-1.21		-0.23	-1.19
	<i>0.14</i>	[0.89]	<i>0.10</i>	<i>0.13</i>
Sugar	-2.52		0.76	-1.04
	<i>1.01</i>	[0.15]	<i>0.13</i>	<i>0.55</i>

Dependent Variable	Five-Year Horizon		Implied One-Year Persistence Rates	
	Unweighted Regression	Weighted Regression	Unweighted Regression	Weighted Regression
Traditional TFP	0.249 <i>0.017</i>	0.316 <i>0.042</i>	0.757	0.794
Revenue TFP	0.277 <i>0.021</i>	0.316 <i>0.042</i>	0.774	0.794
Physical TFP	0.312 <i>0.019</i>	0.358 <i>0.049</i>	0.792	0.814
Price	0.365 <i>0.025</i>	0.384 <i>0.066</i>	0.817	0.826
Demand Shock	0.619 <i>0.013</i>	0.843 <i>0.021</i>	0.909	0.966

Variable	Plant Age Dummies			
	Exit	Entry	Young	Medium
Unweighted Regressions				
Traditional TFP	-0.0211	0.0044	0.0074	0.0061
	<i>0.0042</i>	<i>0.0044</i>	<i>0.0048</i>	<i>0.0048</i>
Revenue TFP	-0.0220	0.0133	0.0075	0.0028
	<i>0.0044</i>	<i>0.0047</i>	<i>0.0051</i>	<i>0.0053</i>
Physical TFP	-0.0186	0.0128	0.0046	-0.0039
	<i>0.0050</i>	<i>0.0053</i>	<i>0.0058</i>	<i>0.0062</i>
Price	-0.0034	0.0005	0.0029	0.0067
	<i>0.0031</i>	<i>0.0034</i>	<i>0.0038</i>	<i>0.0042</i>
Demand Shock	-0.3466	-0.5557	-0.3985	-0.3183
	<i>0.0227</i>	<i>0.0264</i>	<i>0.0263</i>	<i>0.0267</i>



# Determinants of Market Selection

Specification:	[1]	[2]	[3]	[4]	[5]	[6]	[7]
Traditional TFP	-0.073 <i>0.015</i>						
Revenue TFP		-0.063 <i>0.014</i>					
Physical TFP			-0.040 <i>0.012</i>			-0.062 <i>0.014</i>	-0.034 <i>0.012</i>
Prices				-0.021 <i>0.018</i>		-0.069 <i>0.021</i>	
Demand Shock					-0.047 <i>0.003</i>		-0.047 <i>0.003</i>

Note: Much greater dispersion in demand shocks than physical TFP

# Establishment-level Productivity Empirical Patterns

- Dispersion (large), persistence (high) evolution (consistent with learning and selection)
- Selection
  - Lower productivity plants exit
  - Other determinants of productivity matter
  - Open questions: Impact of distortions on selection?
    - Models like Melitz (2003) and Restuccia and Rogerson (2007) imply reduced distortions will improve selection
    - Eslava et. al. (2009) find evidence that trade liberalization improves market selection
- These patterns both support basic models and can be used to test and estimate models
- One other approach has to been to explore the covariance between size and productivity within industries.
  - Basic prediction of virtually all models is positive correlation between size and profitability/productivity

# Within Industry Dynamic Decomposition Applied to FHS (2008) data

		Components of Decomposition (GR)				
		Within	Between	Entry	Exit	Net Entry
Traditional	2.30	1.40	0.18	0.44	0.27	0.72
Revenue	5.13	4.03	0.16	0.55	0.39	0.94
Physical	5.13	3.82	-0.05	1.04	0.32	1.36

Extra Slides

## Growth Identities: Establishment

$$g_{it} = (E_{it} - E_{it-1}) / X_{it}$$

where

$$X_{it} = .5 * (E_{it} + E_{it-1})$$

Then

$$JC_{it} = \max(g_{it}, 0)$$

$$JD_{it} = \max(-g_{it}, 0)$$

From Entry/Exit

$$JC_{it} = \max(g_{it}, 0) * I\{g_{it} = 2\}$$

$$JD_{it} = \max(-g_{it}, 0) * I\{-g_{it} = 2\}$$

# Growth Identities: Aggregate Measures (any level)

$$JC_t = \sum_i (X_{it} / X_t) \max\{g_{it}, 0\} \quad JD_t = \sum_i (X_{it} / X_t) \max\{-g_{it}, 0\}$$

$$JC\_Entry_t = \sum_i (X_{it} / X_t) I\{g_{it} = 2\} \max(g_{it}, 0)$$

$$JD\_Exit_t = \sum_i (X_{it} / X_t) I\{g_{it} = -2\} \max(-g_{it}, 0)$$

$$g_t = JC_t - JD_t$$

$$JC_t = JC\_Cont_t + JC\_Entry_t$$

$$JD_t = JD\_Cont_t + JD\_Exit_t$$