

Chaos Before Order: Productivity Patterns in U.S. Manufacturing

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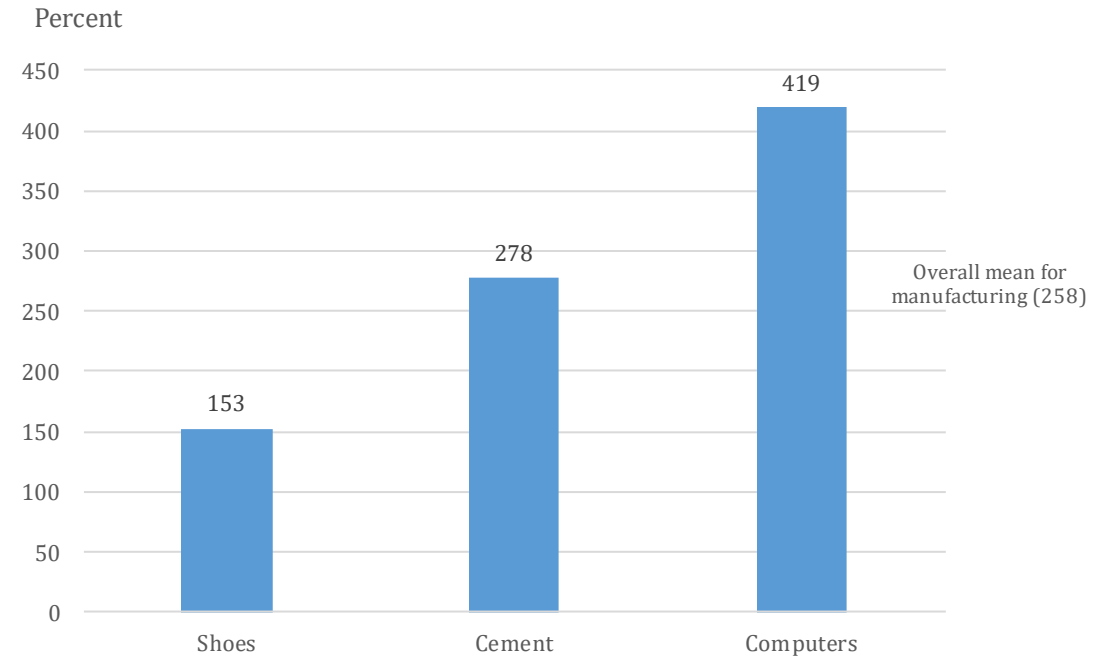
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Dispersion Statistics on Productivity (DiSP)

- BLS-Census joint experimental data product
 - 86 4-digit NAICS manufacturing industries
 - Two productivity measures: Output per hour (LP) and total factor productivity (TFP)
 - Three dispersion statistics: Interquartile range (IQR), interdecile range (90-10), standard deviation
 - Activity-weighted and non-activity-weighted estimates
 - 1997–2016 (Released September 2020)
- Key Findings
 - Enormous within-industry dispersion in productivity
 - Large differences between industries as well
 - Size, state, and age don't explain the dispersion

Productivity Differences Between More and Less Productive Establishments Within an Industry: 2016

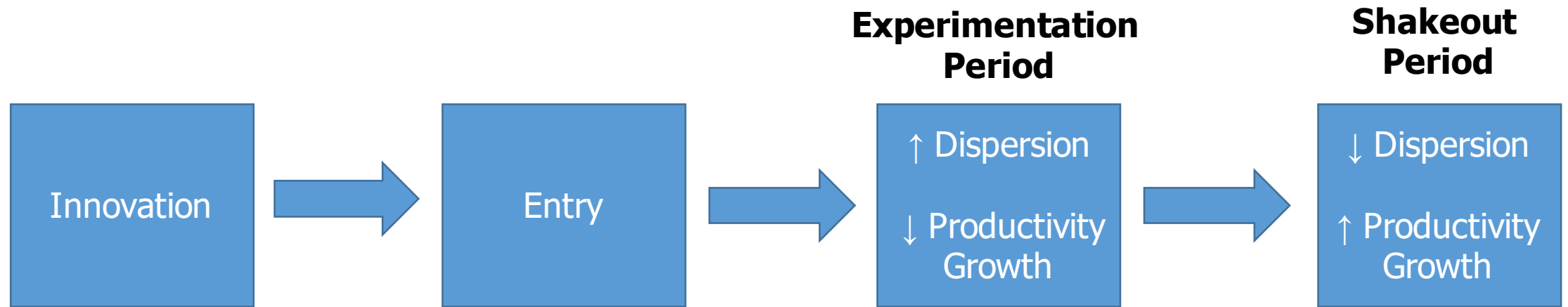
(Differences shown are real revenue per hour)



What accounts for the enormous variation in productivity across establishments within detailed industries?

- Frictions and distortions that inhibit productivity-enhancing reallocation
 - Costs of adjusting factors of production (Decker et al. 2020)
 - Barriers to entry
 - Regulations that prevent the equalization of marginal products across plants
 - Establishment-specific markups (De Loecker et al. 2020)
- Differences in management skills (Bloom et al. 2019)
- Differences in technologies (Zolas et al. 2020)
- Innovation dynamics (Foster et al. 2021)
 - Building on Gort and Klepper (1982)

Innovation Hypothesis



Data

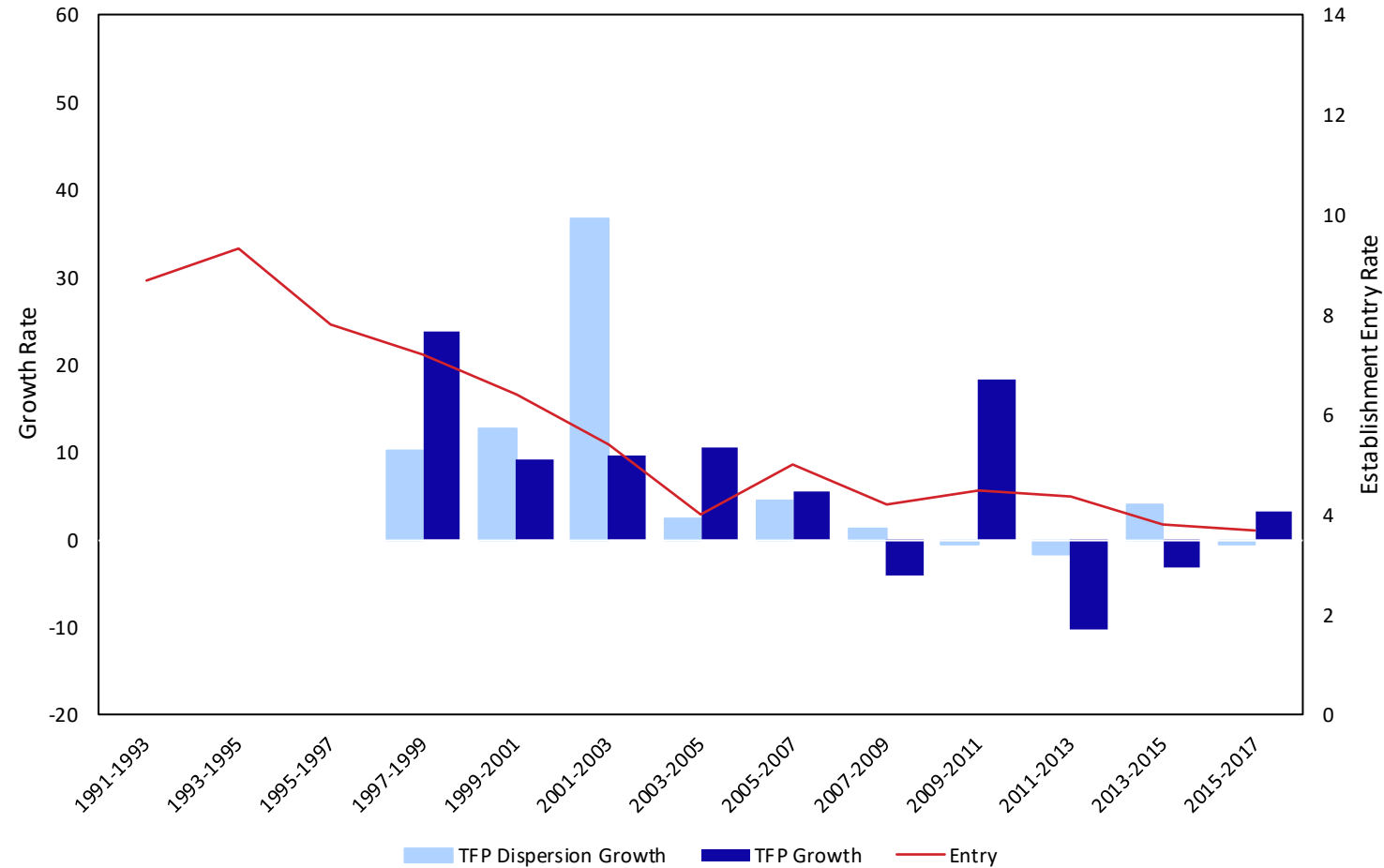
Sources:

- BLS Aggregate Industry Productivity Statistics
 - Labor productivity (LP)
 - Total factor productivity (TFP)
- Dispersion Statistics on Productivity (DiSP)
 - Activity-weighted dispersion measures (IQR, 75–50, 50–25)
- Business Dynamics Statistics (BDS)
 - Establishment and firm age distributions allow us to calculate entry rates

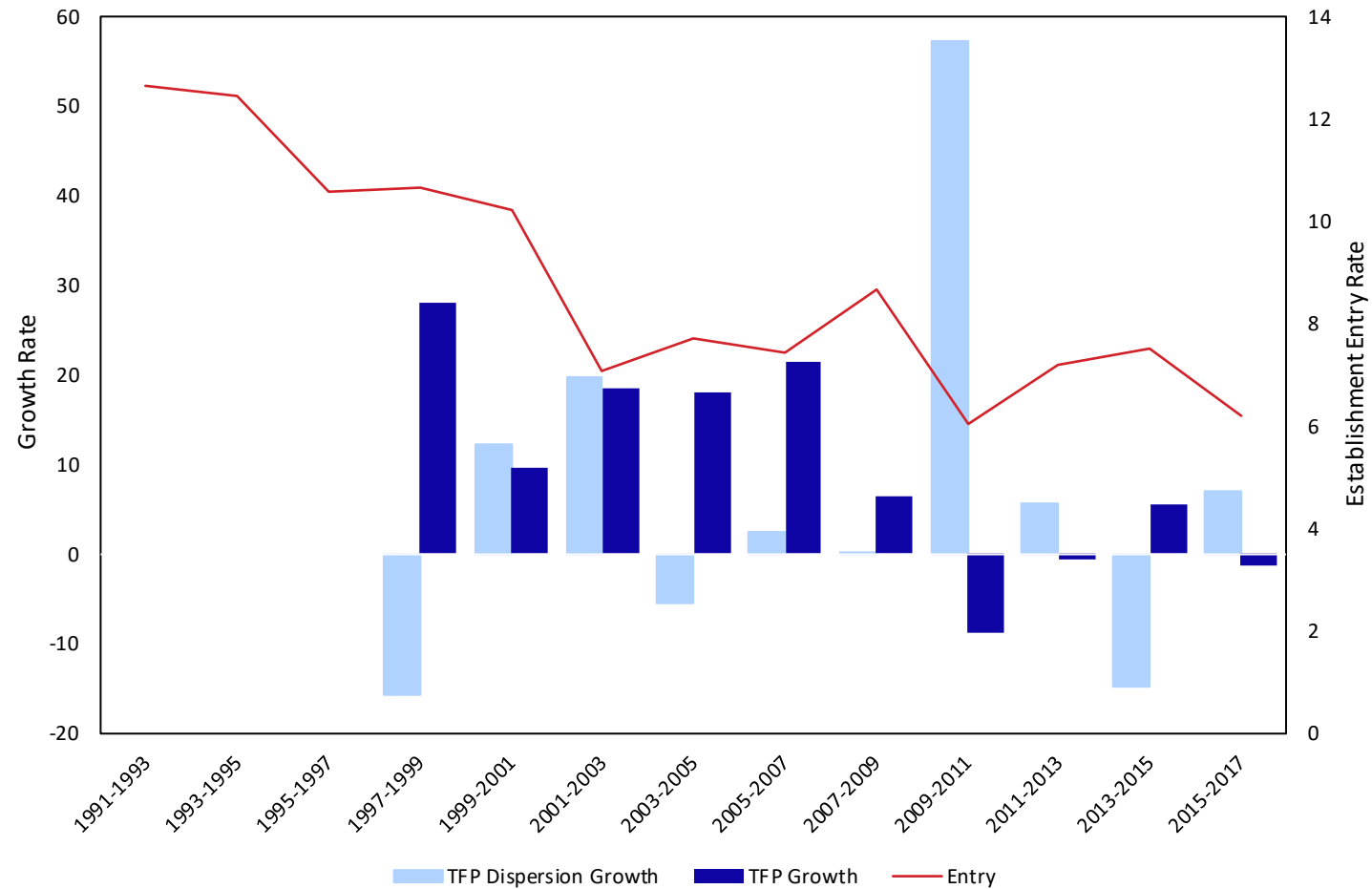
Sample:

- 86 4-digit NAICS manufacturing industries
- Construct the average annual growth rate for two-year periods for productivity dispersion and growth (to abstract from business cycle dynamics)
- Construct average entry rates for two-year periods

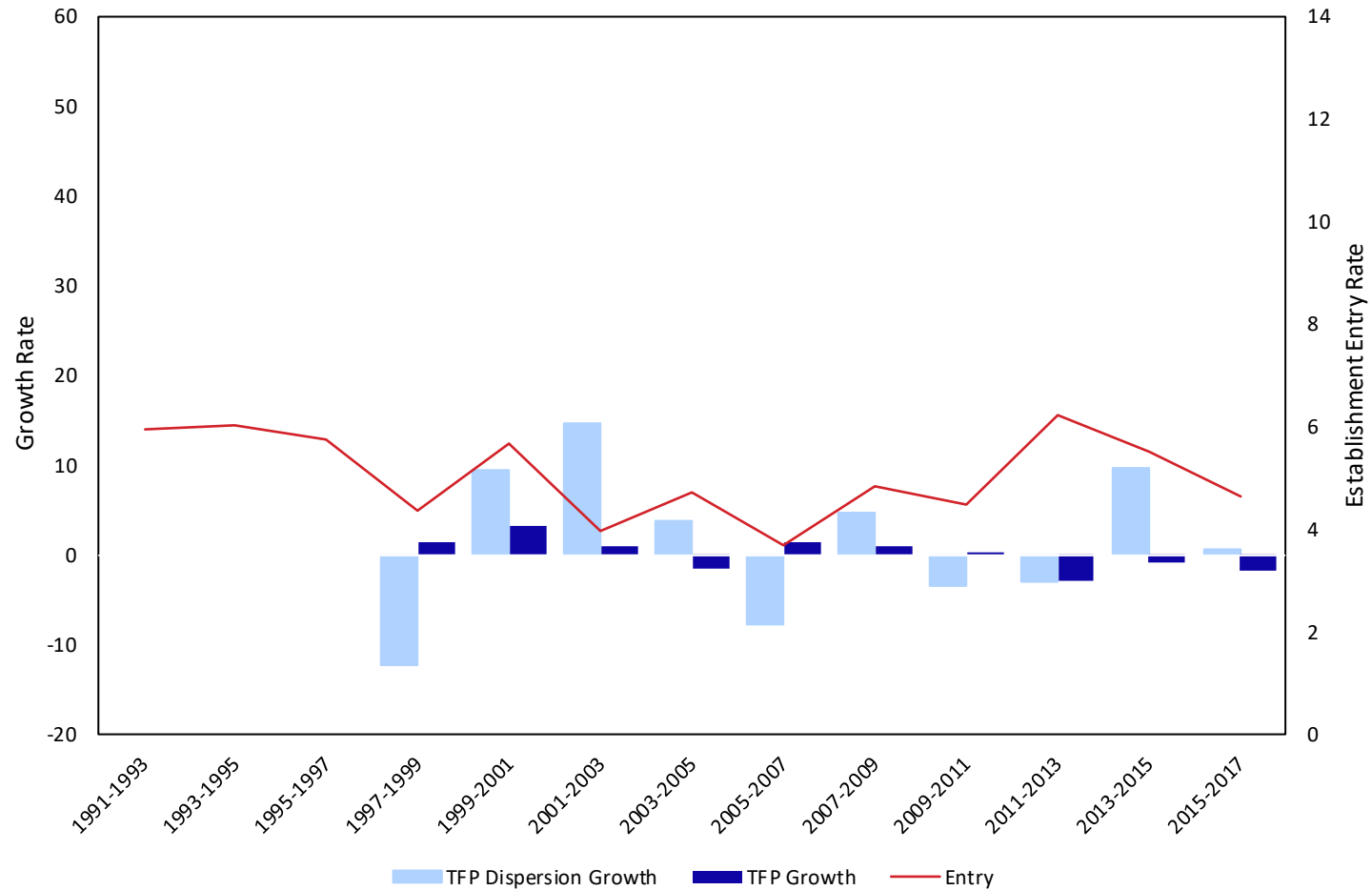
Semiconductor and Other Electronic Component Manufacturing



Computer and Peripheral Equipment Manufacturing



Grain and Oilseed Manufacturing



Model

$$Y_{i,t} = \alpha + \lambda_t + \lambda_i + \sum_{k=1}^3 [\beta_k Entry_{i,t-k} + \delta_k Entry_{i,t-k} \times Tech_i] + \varepsilon_{i,t}$$

$Y_{i,t}$ – within-industry/period productivity dispersion growth or industry/period productivity growth (TFP)

Entry – establishment or firm entry rates

Tech – indicator for high-tech industry (16 out of 86 4-digit NAICS industries based on share of STEM jobs)

λ_t – period effects

λ_i – industry effects

TFP Dispersion Growth, TFP Growth, and Establishment Entry Rates

	Dispersion	Productivity
Lag 1 Entry	0.01 (0.44)	0.01 (0.07)
Lag 2 Entry	-0.35 (0.42)	-0.20** (0.09)
Lag 3 Entry	0.33 (0.40)	-0.05 (0.09)
Lag 1 Entry x Tech	2.90* (1.49)	-0.67** (0.30)
Lag 2 Entry x Tech	-4.24** (1.63)	0.78*** (0.46)
Lag 3 Entry x Tech	0.85 (1.94)	0.83** (0.57)
Joint Hypothesis Tests:		
Lag 1 Entry + Lag 1 Entry x Tech	2.91** (1.41)	-0.66** (0.31)
Lag 2 Entry + Lag 2 Entry x Tech	-4.59*** (1.60)	0.58 (0.45)
Lag 3 Entry + Lag 3 Entry x Tech	1.18 (1.90)	0.79 (0.58)

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

TFP Dispersion Growth and Establishment Entry Rates

	75–50 range	50-25 range
Lag 1 Entry	-0.39 (0.53)	0.05 (0.75)
Lag 2 Entry	0.11 (0.56)	-1.27 (0.79)
Lag 3 Entry	-0.94* (0.49)	0.97* (0.56)
Lag 1 Entry x Tech	3.34 (2.22)	3.04 (1.86)
Lag 2 Entry x Tech	-7.20*** (2.72)	-1.54 (1.50)
Lag 3 Entry x Tech	4.51*** (1.69)	-1.02 (1.66)
Joint Hypothesis Tests:		
Lag 1 Entry + Lag 1 Entry x Tech	2.95 (2.20)	3.54** (1.76)
Lag 2 Entry + Lag 2 Entry x Tech	-7.09*** (2.69)	-2.80** (1.36)
Lag 3 Entry + Lag 3 Entry x Tech	3.57** (1.65)	-0.05 (1.61)

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

TFP Dispersion Growth, TFP Growth, and Firm Entry Rates

	Dispersion	Productivity
Lag 1 Entry	-0.23 (0.35)	-0.04 (0.07)
Lag 2 Entry	-0.01 (0.28)	-0.11* (0.06)
Lag 3 Entry	0.14 (0.32)	-0.07 (0.09)
Lag 1 Entry x Tech	2.00 (1.75)	-0.19 (0.32)
Lag 2 Entry x Tech	-4.96** (1.75)	1.08** (0.42)
Lag 3 Entry x Tech	2.08 (1.58)	1.76*** (0.43)
Joint Hypothesis Tests:		
Lag 1 Entry + Lag 1 Entry x Tech	1.27 (1.80)	-0.13 (0.27)
Lag 2 Entry + Lag 2 Entry x Tech	-4.95*** (1.82)	0.23 (0.36)
Lag 3 Entry + Lag 3 Entry x Tech	2.48 (1.54)	0.90* (0.48)

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

Conclusions

- Innovation (proxied by entry rates) can initially lead to a rise in productivity dispersion and a decline in aggregate productivity.
- After a shakeout period, dispersion will eventually fall and aggregate productivity will rise.
- Dynamic relationship between entry, dispersion, and growth
 - High-tech industries
 - Stronger using establishment entry rates than firm entry rates
 - Stronger for TFP than LP
- Recent (pre-pandemic) lower entry rates suggest slower productivity growth to come.
 - Preliminary evidence of surge in entry during pandemic (Dinlersoz et al. 2021; Haltiwanger 2021)



Shaded areas indicate U.S. recessions.

Source: U.S. Census Bureau

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Data Availability

- Dispersion Statistics on Productivity (DiSP)

<https://www.bls.gov/lpc/productivity-dispersion.htm>

<https://www.census.gov/disp>

- Industry Productivity Statistics

[https://www.bls.gov/lpc/tables by sector and industry.htm](https://www.bls.gov/lpc/tables_by_sector_and_industry.htm)

- Business Dynamics Statistics (BDS)

<https://www.census.gov/programs-surveys/bds.html>

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