
Large Firms, Monopsony, and Concentration in the Labor Market

741 Macroeconomics
Topic 8

Masao Fukui

2025 Fall

Motivation

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... but no firm is really “large” in Hopenhayn-Rogerson – each firm is measure zero

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 - “Effective” number of firms: 3-9
 - Local labor market: 3-digit NAICS × commuting zone

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 - Local labor market: 3-digit NAICS × commuting zone
- Natural to expect that these firms exploit ***labor market power***
- Today: a model of oligopsony in the labor market

General Equilibrium Oligopsony Model

– Based on Berger-Mongey-Herkenhoff (2022)

Environment

- Static model
- Representative family
 - Continuum of labor markets $j \in [0,1]$
 - Labor market j has a fixed number of firms $i \in \{1,2,\dots,M_j\}$
 - Continuum of workers within a family, choosing where to work (i,j)
- Firms
 - Each firm produces final goods using $y_{ij} = z_{ij}^{1-\alpha} n_{ij}^{\alpha}$
- Markets
 - Local labor market: Cournot competition for labor

Representative Family

- Mass L of workers within the family
- Each worker $l \in [0, L]$ has efficiency unit of labor $\epsilon_{ij}(l)$ when working at (i, j)
- The family solves

$$\max_{C, \{\mathbb{I}_{ij}(l)\}} C$$

$$\text{s.t. } C = \int_0^1 \sum_{i=1}^{M_j} \int_0^L w_{ij} \epsilon_{ij}(l) \mathbb{I}_{ij}(l) dl dj + \Pi$$

- Assume the distribution of $\epsilon_{ij}(l)$ follow nested Fréchet (GEV)

$$\Pr \left(\{\epsilon_{ij}(l) \leq a_{ij}\}_{ij} \right) = \exp \left[-G \left(\{a_{ij}\}_{ij} \right) \right], \quad G(\{a_{ij}\}) = \int_0^1 \left(\sum_{i=1}^{M_j} a_{ij}^{-(\eta+1)} \right)^{\frac{\eta+1}{\theta+1}} dj$$

with $\eta > \theta$

Representation Result

- The family's problem can be equivalently represented as

$$\begin{aligned} & \max_{C, \{\ell_{ij}\}: \sum_{ij} \ell_{ij} = 1} C \\ \text{s.t. } & C = \int_0^1 \sum_{i=1}^{M_j} w_{ij} \ell_{ij} S_{ij}(\{\ell_{ij}\}) dj \times L + \Pi \\ & \int_0^1 \sum_j \ell_{ij} di = 1 \end{aligned}$$

where

$$S_{ij}(\{\ell_{ij}\}) = \left(\frac{\ell_{ij}}{\sum_i \ell_{ij}} \right)^{-1/(\eta+1)} \left(\sum_i \ell_{ij} \right)^{-1/(\theta+1)}$$

- ℓ_{ij} : share of workers working for firm i in market j
- S_{ij} : average efficiency of workers in (i, j) , and it captures selection:
more workers work in $(i, j) \Rightarrow$ average efficiency of workers worsens
- See Donald-Fukui-Miyauchi (2024) Appendix C for a proof

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$$\text{FOC: } w_{ij} \left[S_{ij} + \ell_{ij} \partial_{\ell_{ij}} S_{ij} \right] = \lambda$$

Nested CES Labor Supply System

Solutions: Given a vector of wages, $\{w_{ij}\}_{ij}$,

- The share of workers who choose to work in (i, j) is

$$\ell_{ij}(\{w_{ij}\}_{ij}) = \left(\frac{w_{ij}}{\mathbf{w}_j} \right)^{\eta+1} \left(\frac{\mathbf{w}_j}{\mathbf{W}} \right)^{\theta+1}$$

where $\mathbf{w}_j \equiv \left[\sum_i w_{ij}^{\eta+1} \right]^{1/(\eta+1)}$, $\mathbf{W} \equiv \left[\int_0^1 \mathbf{w}_j^{\theta+1} dj \right]^{1/(\theta+1)}$

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- The efficiency units of labor supply for (i, j) is

$$n_{ij}(\{w_{ij}\}_{ij}) \equiv \ell_{ij} S_{ij}(\{\ell_{ij}\}) L = \left(\frac{w_{ij}}{\mathbf{w}_j} \right)^{\eta} \left(\frac{\mathbf{w}_j}{\mathbf{W}} \right)^{\theta} L$$

Oligopsonistic Labor Market

- The inverse labor supply function is

$$w_{ij}(\{n_{ij}\}) = \left(\frac{n_{ij}}{\mathbf{n}_j} \right)^{\frac{1}{\eta}} \left(\frac{\mathbf{n}_j}{\mathbf{N}} \right)^{\frac{1}{\theta}} \quad (1)$$

$$\mathbf{n}_j \equiv \left[\sum_i n_{ij}^{\frac{\eta+1}{\eta}} \right]^{\frac{\eta}{\eta+1}}, \quad \mathbf{N} \equiv \left[\int_0^1 \mathbf{n}_j^{\frac{\theta+1}{\theta}} dj \right]^{\frac{\theta}{\theta+1}}$$

(2)

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- Firms engage in Cournot competition, taking competitor's hiring as given, $n_{-ij} = n_{-ij}^*$
- $$\max_{n_{ij}} z_{ij}^{1-\alpha} n_{ij}^{\alpha} - w_{ij}(n_{ij}, n_{-ij}^*) n_{ij} \quad (2)$$

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- General solution:

$$w_{ij} = \mu_{ij} \times \alpha z_{ij}^{1-\alpha} n_{ij}^{\alpha-1}, \quad \mu_{ij} \equiv \frac{\varepsilon_{ij}}{\varepsilon_{ij} + 1}, \quad \varepsilon_{ij} \equiv \frac{d \ln n_{ij}}{d \ln w_{ij}}$$

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wage markdown

MPL

Equilibrium Definition

A (Cournot) equilibrium consists of $\{w_{ij}(\{n_{ij}\}), n_{ij}\}$ such that

- $w_{ij}(\{n_{ij}\})$ is consistent with household's optimality (1)
- Taking $\{n_{-ij}\}$ as given, firm i solves (2)

Wage Markdown

- With our functional form assumption, the labor supply elasticity takes the form of

$$\varepsilon_{ij}(s_{ij}) = \left[\frac{1}{\eta}(1 - s_{ij}) + \frac{1}{\theta}s_{ij} \right]^{-1}, \quad \mu_{ij}(s_{ij}) = \frac{\varepsilon_{ij}(s_{ij})}{\varepsilon_{ij}(s_{ij}) + 1}$$

where

$$s_{ij} = \frac{w_{ij}n_{ij}}{\sum_k w_{kj}n_{kj}} \tag{3}$$

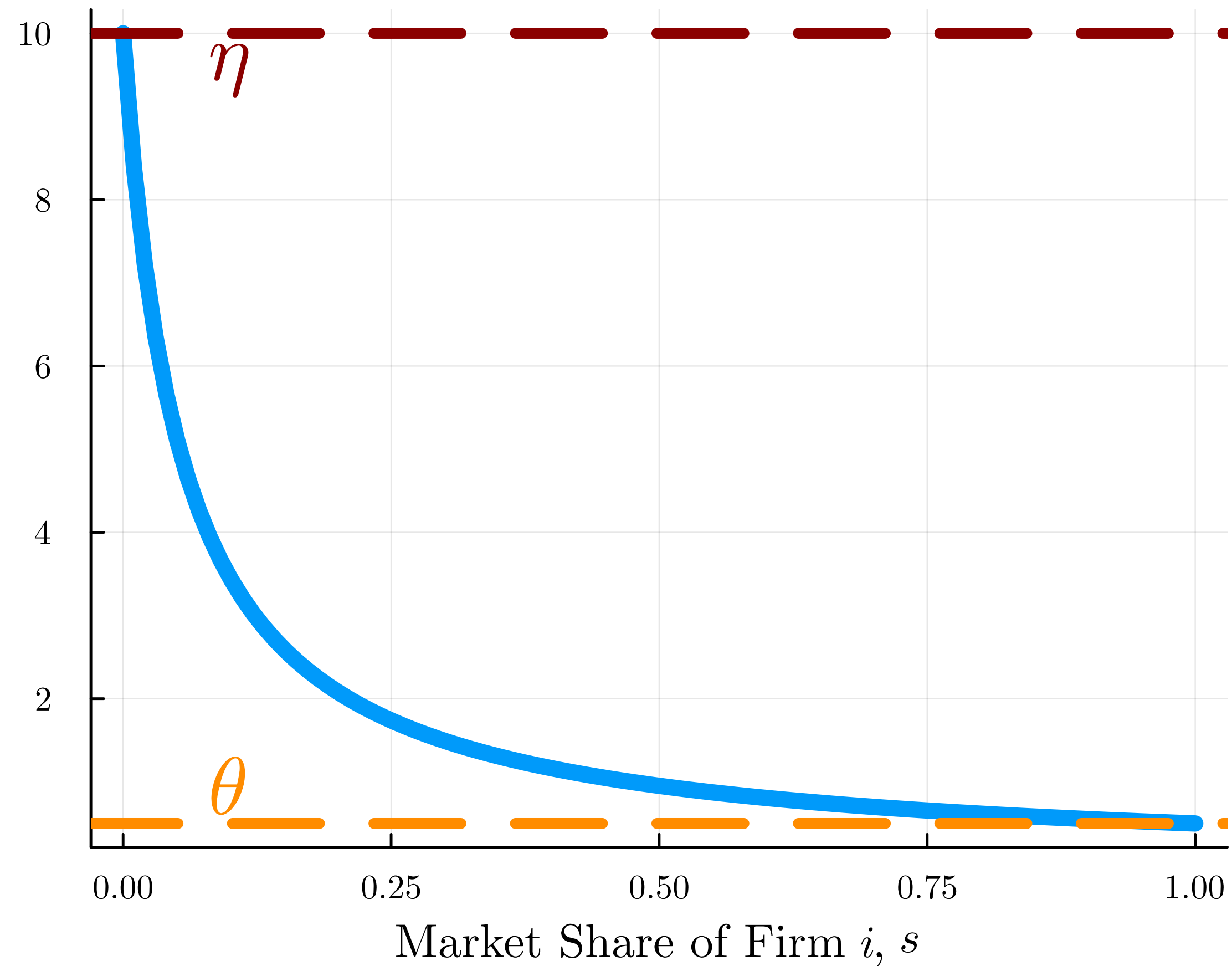
is the labor market share of firm i in market j

1. Competitive labor market: $\theta, \eta \rightarrow \infty \Rightarrow \varepsilon_{ij} \rightarrow \infty$
2. Monopsonistic competition within a market j : $M_j \rightarrow \infty \Rightarrow s_{ij} \rightarrow 0 \Rightarrow \varepsilon_{ij} \rightarrow \eta$
3. Monopsony within a market j : $s_{ij} \rightarrow 1 \Rightarrow \varepsilon_{ij} \rightarrow \theta$

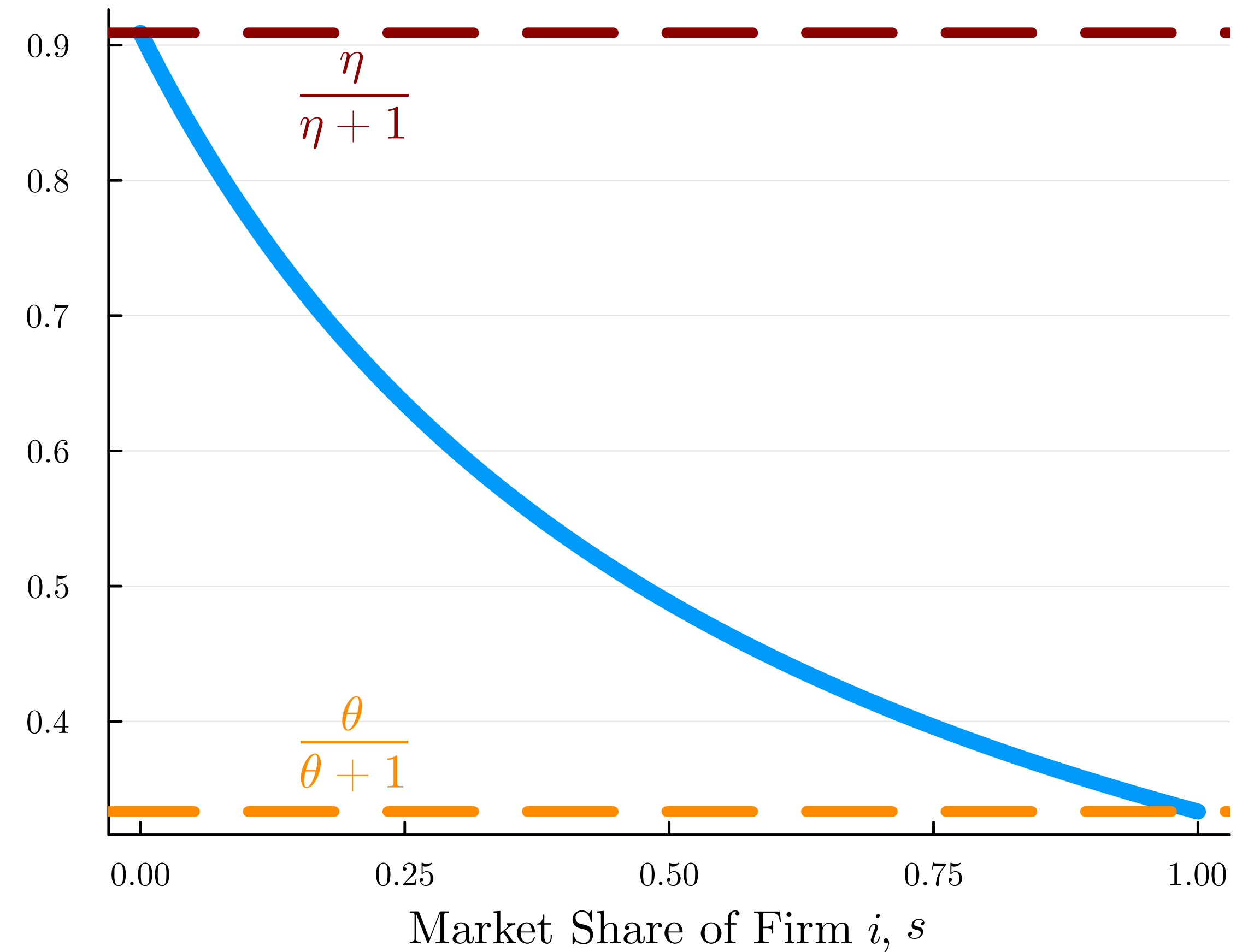
- See also Atkeson-Burstein (2008)

Numerical Illustration

Labor Supply Elasticity of Firm i , ε



Wage Markdown of Firm i , μ



Equilibrium System

The equilibrium $\{s_{ij}\}$ solve

$$s_{ij} = \frac{\left(\mu_{ij}(s_{ij})z_{ij}^{1-\alpha}\right)^{\frac{1+\eta}{1+\eta(1-\alpha)}}}{\sum_k \left(\mu_{kj}(s_{kj})z_{kj}^{1-\alpha}\right)^{\frac{1+\eta}{1+\eta(1-\alpha)}}}$$

- Proof: Relative employment between i and k :

$$\frac{n_{ij}}{n_{kj}} = \left(\frac{w_{ij}}{w_{kj}}\right)^\eta = \left(\frac{\mu_{ij}(s_{ij})z_{ij}^{1-\alpha}n_{ij}^{\alpha-1}}{\mu_{kj}(s_{kj})z_{kj}^{1-\alpha}n_{kj}^{\alpha-1}}\right)^\eta \Leftrightarrow \frac{n_{ij}}{n_{kj}} = \left(\frac{\mu_{ij}(s_{ij})z_{ij}^{1-\alpha}}{\mu_{kj}(s_{kj})z_{kj}^{1-\alpha}}\right)^{\frac{\eta}{1+\eta(1-\alpha)}}$$

- Substituting into (3) gives the expression
- Given $\{s_{ij}\}$, we can immediately compute $\{\mu_{ij}, n_{ij}, w_{ij}\}$

Sufficient Statistics for Labor Share

- Define the aggregate labor share as

$$LS = \frac{\int_0^1 \sum_{i \in j} w_{ij} n_{ij} dj}{\int_0^1 \sum_{i \in j} y_{ij} dj}$$

- Define the payroll weighted HHI as

$$HHI = \int_0^1 s_j HHI_j dj, \quad s_j = \frac{\sum_{i \in j} w_{ij} n_{ij}}{\int_0^1 \sum_{i \in j} w_{ij} n_{ij} dj}, \quad HHI_j = \sum_{i \in j} s_{ij}^2$$

- Result:

$$LS = \alpha \left[(1 - HHI) \left(\frac{\eta}{\eta + 1} \right)^{-1} + HHI \left(\frac{\theta}{\theta + 1} \right)^{-1} \right]^{-1}$$

⇒ Conditional on the knowledge of (α, θ, η) , payroll shares are sufficient to infer LS

Planning Problem

Planning Problem

- The planner maximizes total consumption subject to the resource constraints:

$$\max_{\{n_{ij}, \ell_{ij}\}} \int_0^1 \sum_i z_{ij}^{1-\alpha} n_{ij}^{\alpha} dj$$

$$\text{s.t.} \quad n_{ij} = \ell_{ij} S_{ij}(\{\ell_{ij}\}) L$$

$$\int_0^1 \sum_j \ell_{ij} di = 1$$

- The FOC is

$$\alpha z_{ij}^{1-\alpha} n_{ij}^{\alpha-1} \left[S_{ij} + \ell_{ij} \partial_{\ell_{ij}} S_{ij} \right] = \lambda$$

Equilibrium vs. Planner

Planner:

$$\alpha z_{ij}^{1-\alpha} n_{ij}^{\alpha-1} \left[S_{ij} + \ell_{ij} \partial_{\ell_{ij}} S_{ij} \right] = \lambda$$

Equilibrium:

$$\mu_{ij} \alpha z_{ij}^{1-\alpha} n_{ij}^{\alpha-1} \left[S_{ij} + \ell_{ij} \partial_{\ell_{ij}} S_{ij} \right] = \lambda$$

- Is the equilibrium efficient? – No, as long as $\{\mu_{ij}\}$ vary in equilibrium
 - If $\mu_{ij} = \mu$ for all i, j , then μ is indistinguishable from λ
- In what way?
 - Firms with low μ_{ij} (high labor market power) are too small in eqm!
 - Firms with high μ_{ij} (low labor market power) are too large in eqm!
- Minimum wage reallocates workers from high μ_{ij} to low μ_{ij}
⇒ can improve efficiency (but Berger-Herkenhoff-Mongey (2025) say it's small)

Bringing the Model to the Data

Identification

- Key parameters: (θ, η)
- Labor supply equation with potential labor supply shifter ξ_{ij}

$$n_{ij}(\{w_{ij}\}_{ij}) = \xi_{ij} \left(\frac{w_{ij}}{\mathbf{w}_j} \right)^\eta \left(\frac{\mathbf{w}_j}{\mathbf{W}} \right)^\theta L$$

- Taking log,

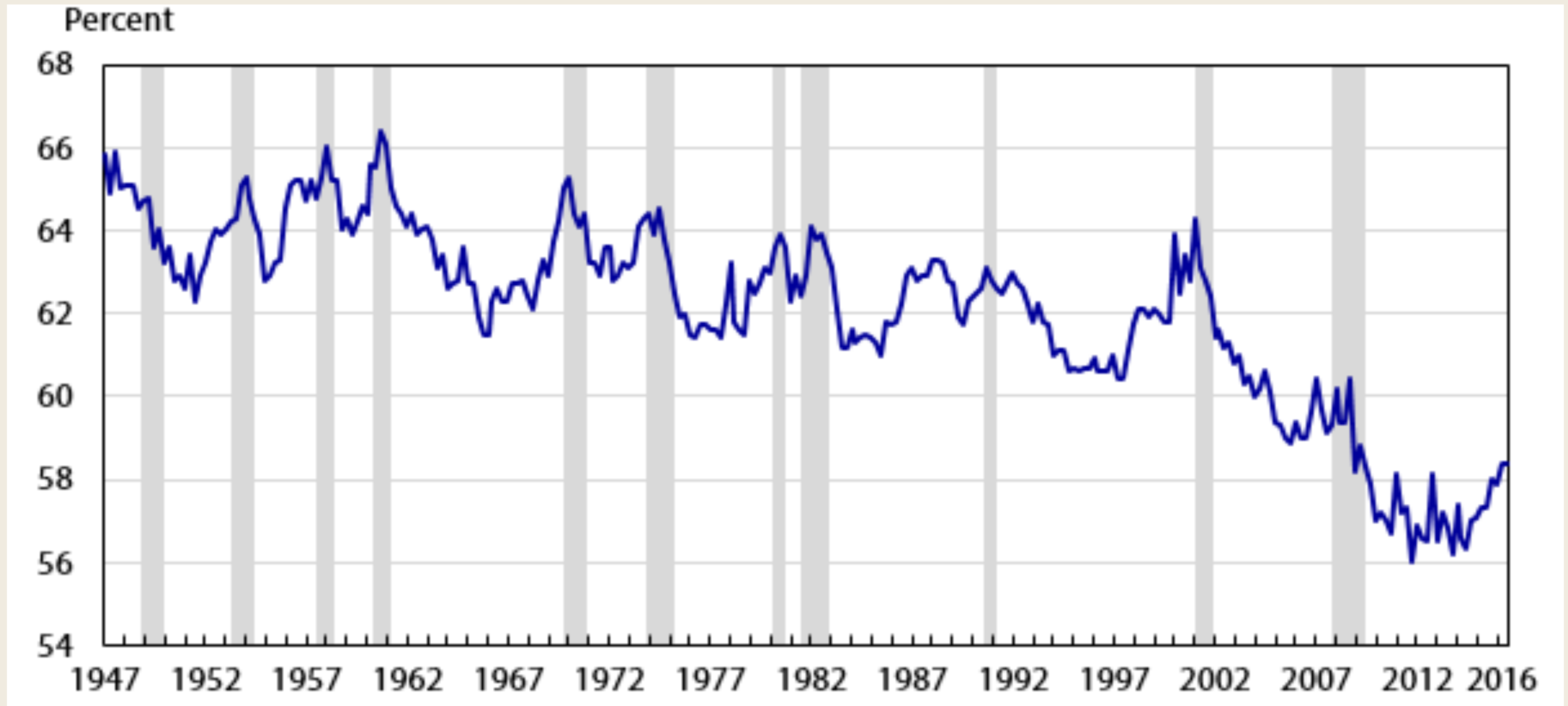
$$\log n_{ij} = \eta \log(w_{ij}) + (\theta - \eta) \log \mathbf{w}_j - \theta \log \mathbf{W} + \log L + \log \xi_{ij}$$

- With suitable instruments (labor demand shifter), one can identify (θ, η)
 1. Berger-Mongey-Herkenhoff (2021): changes in state corporate taxes
 2. Felix (2023): changes in tariffs

Estimation Results

- BHM's implementation: US Census LBD data
- Market: 3-digit NAICS \times commuting zone
- Estimates: $\eta = 10.85$, $\theta = 0.42$
- With $HHI = 0.11$ in 2014, the model implies **30%** aggregate wage markdown

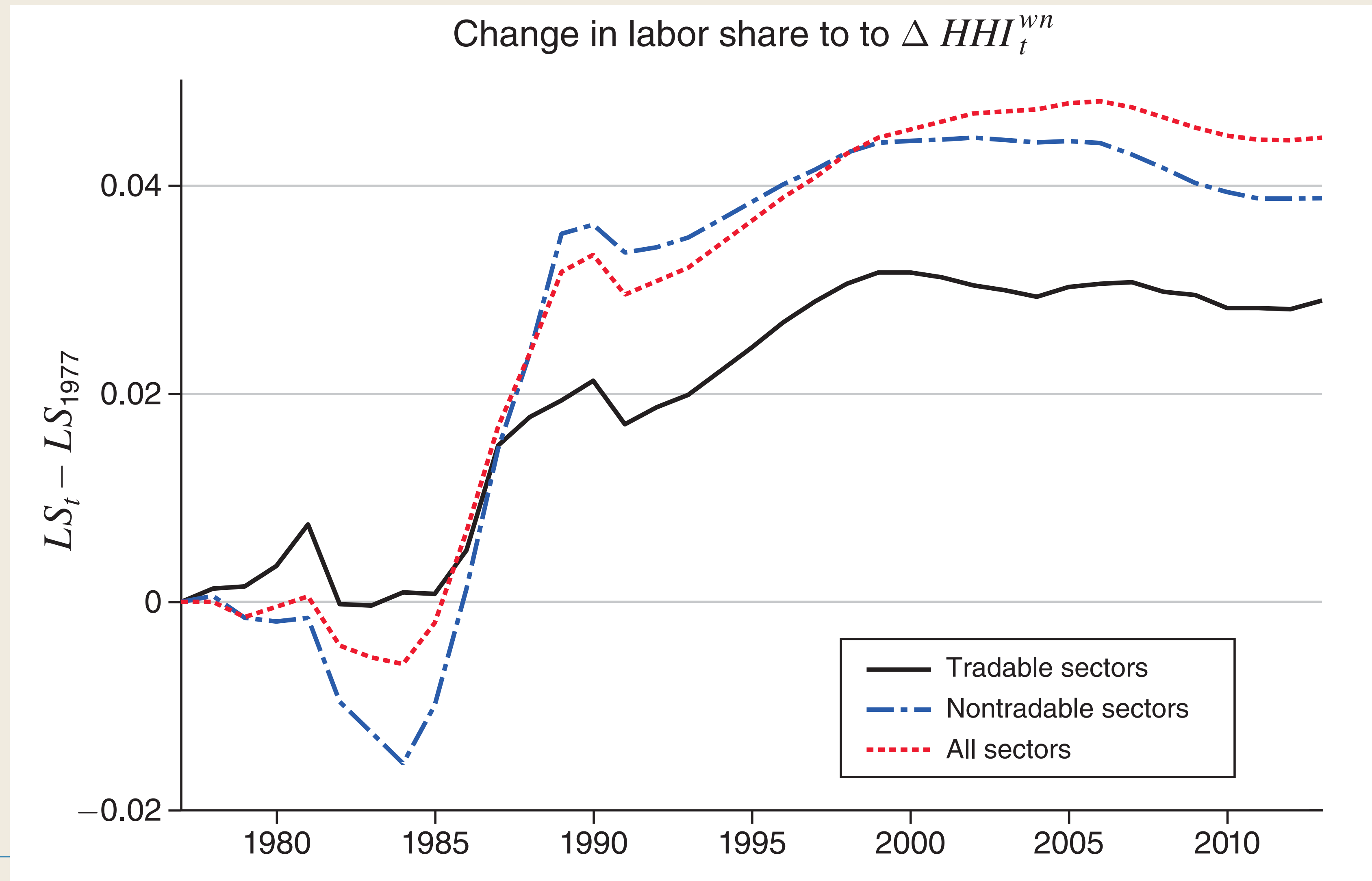
Labor Share



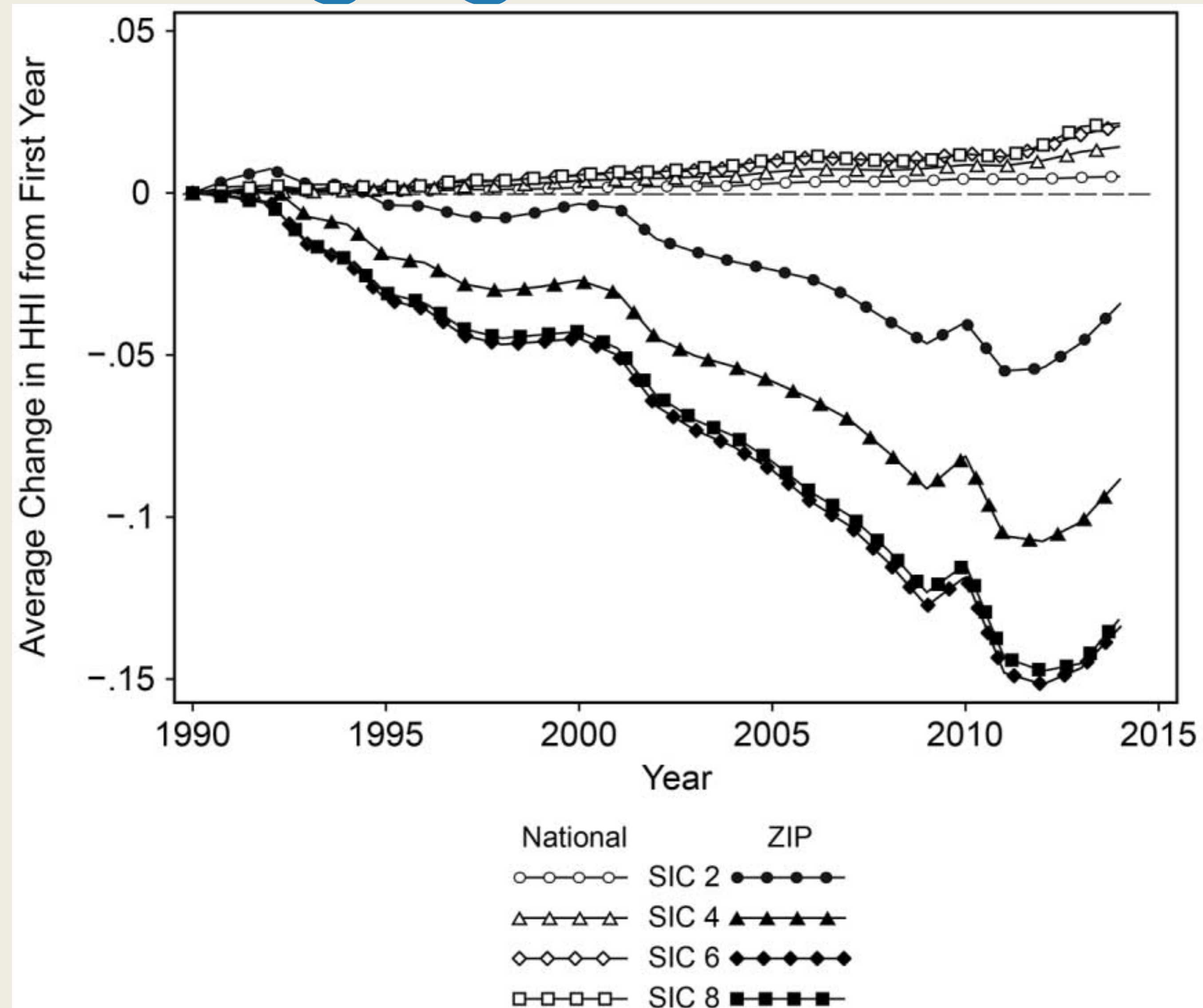
- Can the changes in concentration explain the changes in labor share?

Labor Share Increases due to ΔHHI

- Fix (η, θ, α) and feed the changes in HHI over time



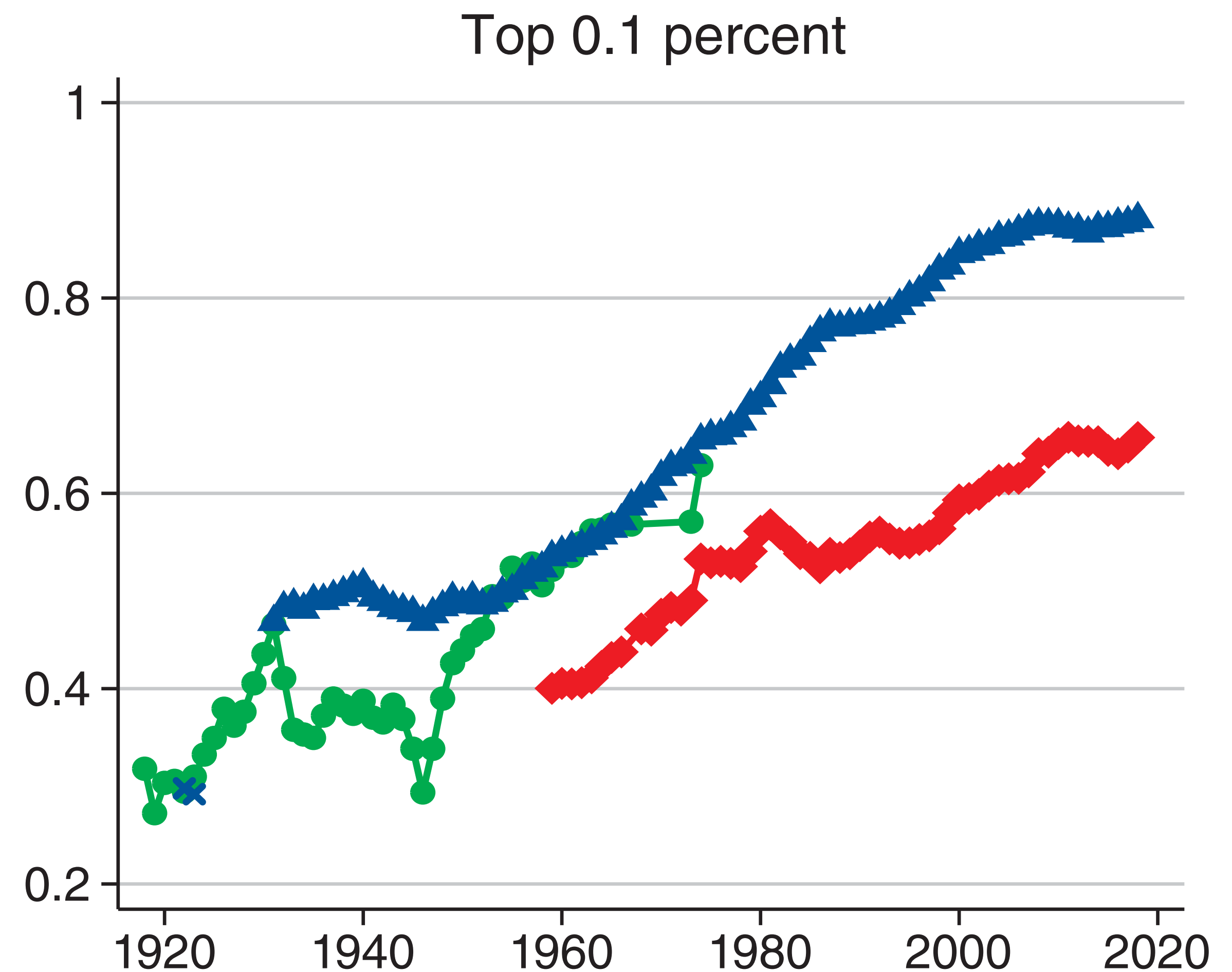
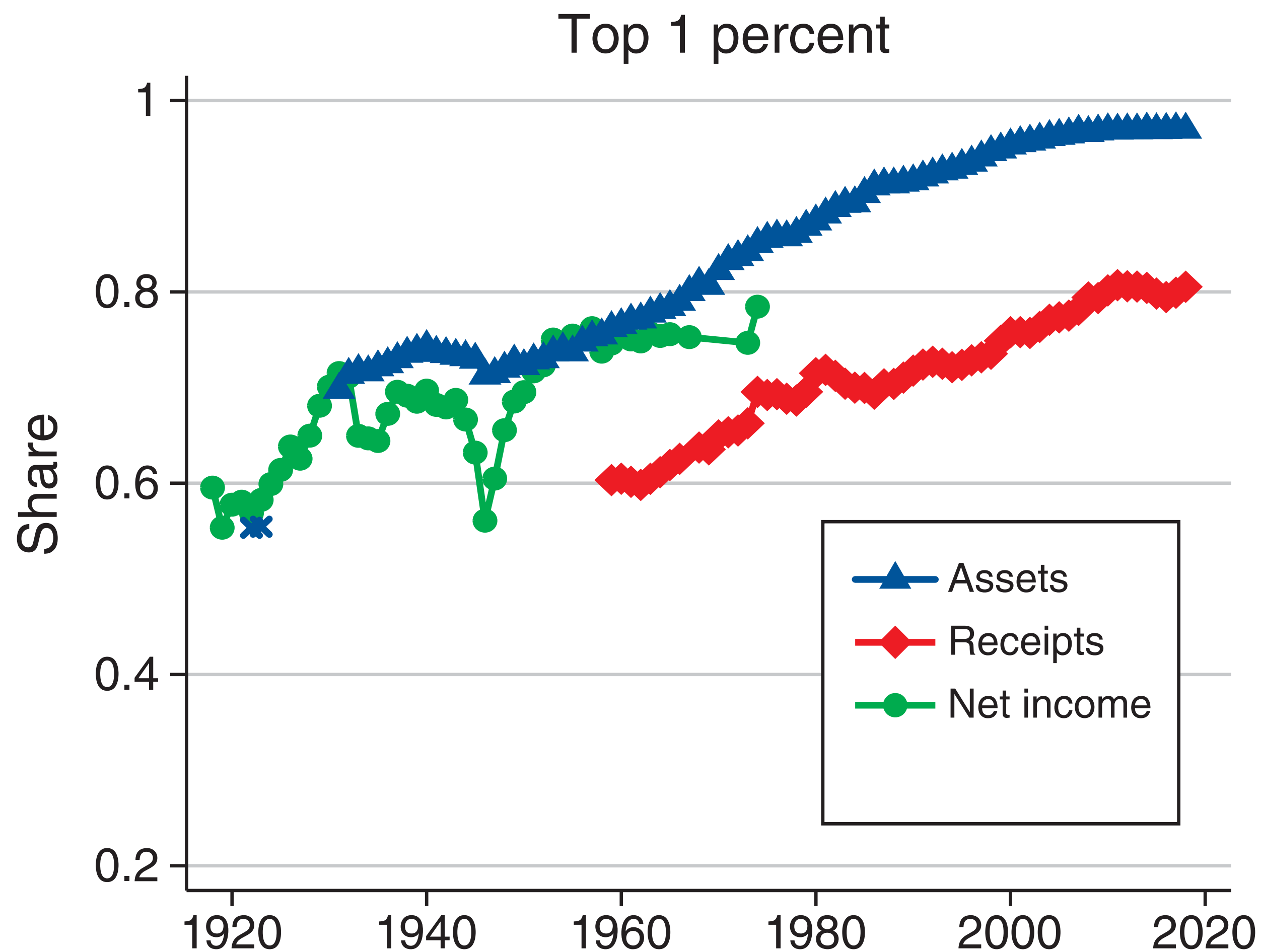
Diverging Trends in HHI



The Rise of Large Firms

– Ma, Zhang, and Zimmermann (2025)

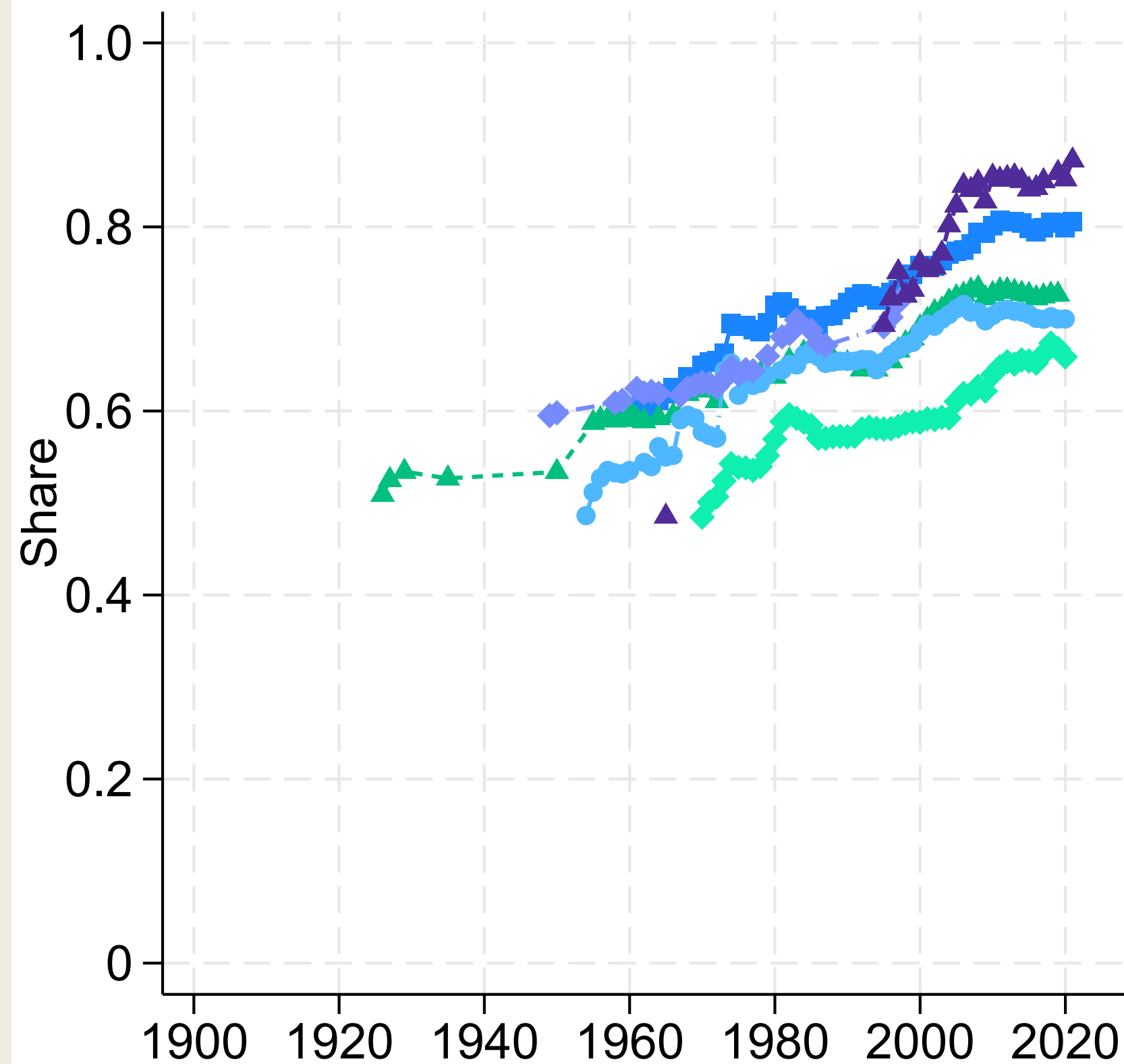
100 Years of Rising Concentration in the US



Source: Kwon, Ma, & Zimmermann (2024)

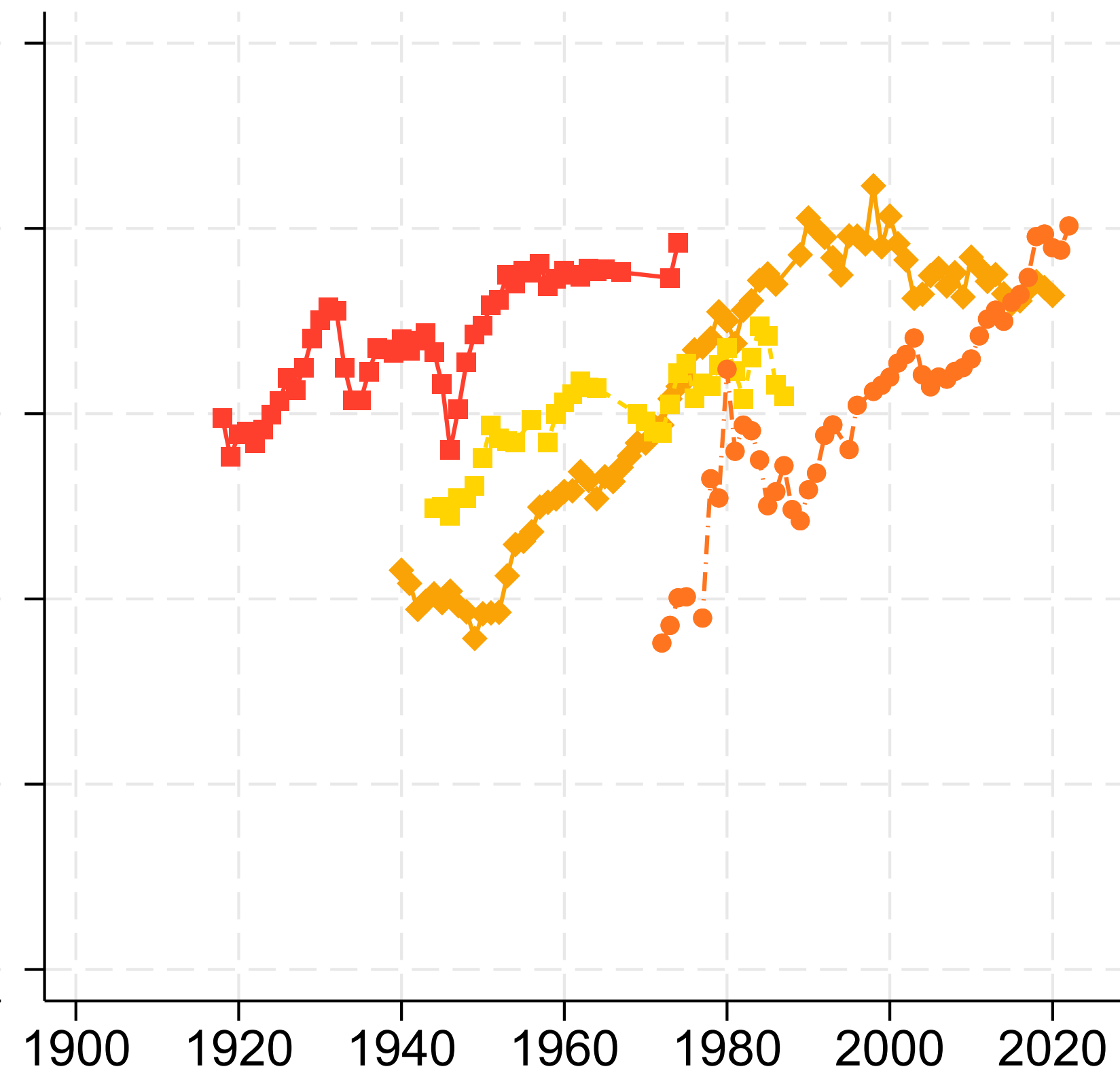
Top 1% Share in the World

Top 1% by Sales



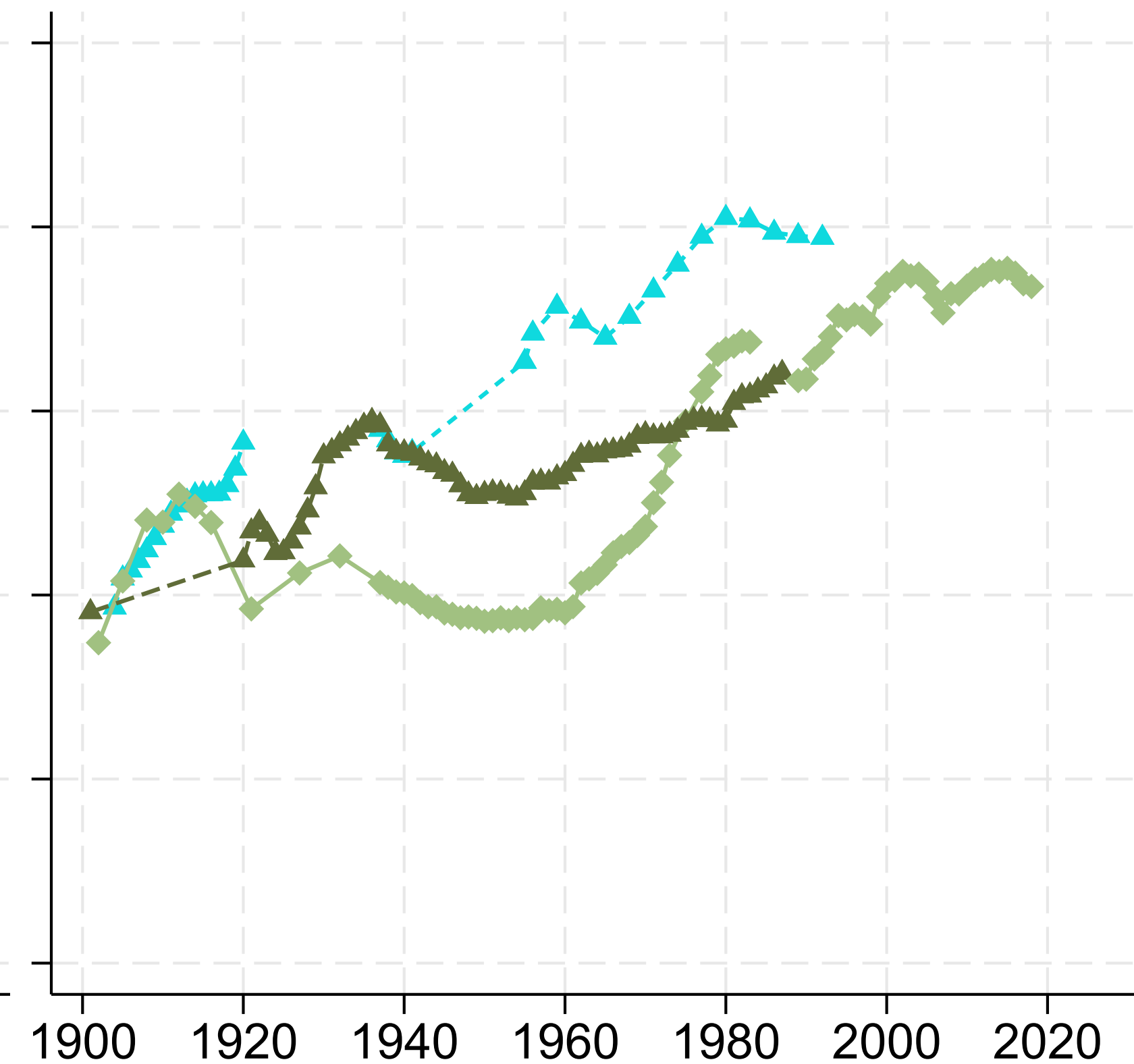
Germany Austria France
Switzerland Denmark US

Top 1% by Net Income



Australia Canada Singapore
US

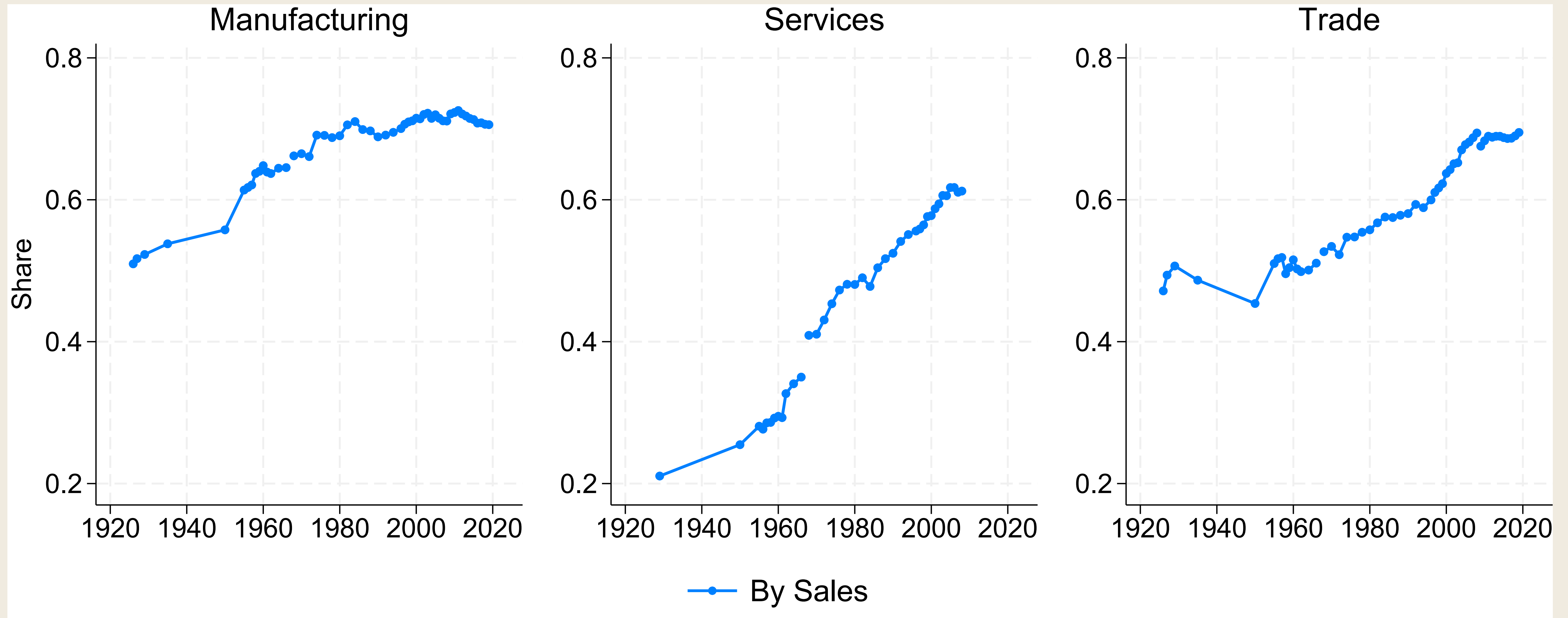
Top 1% by Equity Capital



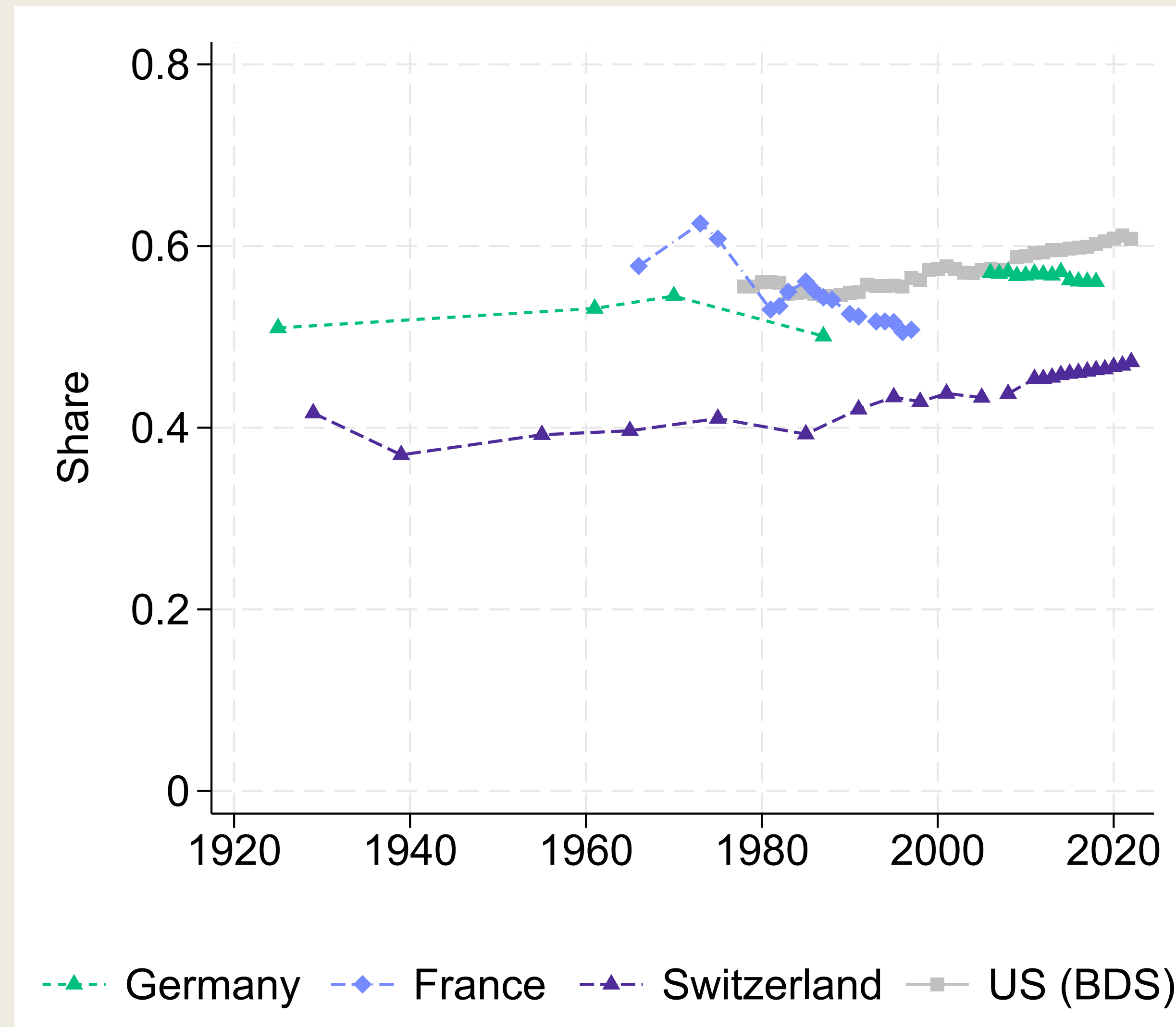
Germany Denmark Switzerland

Top 1% Sales Share by Sectors

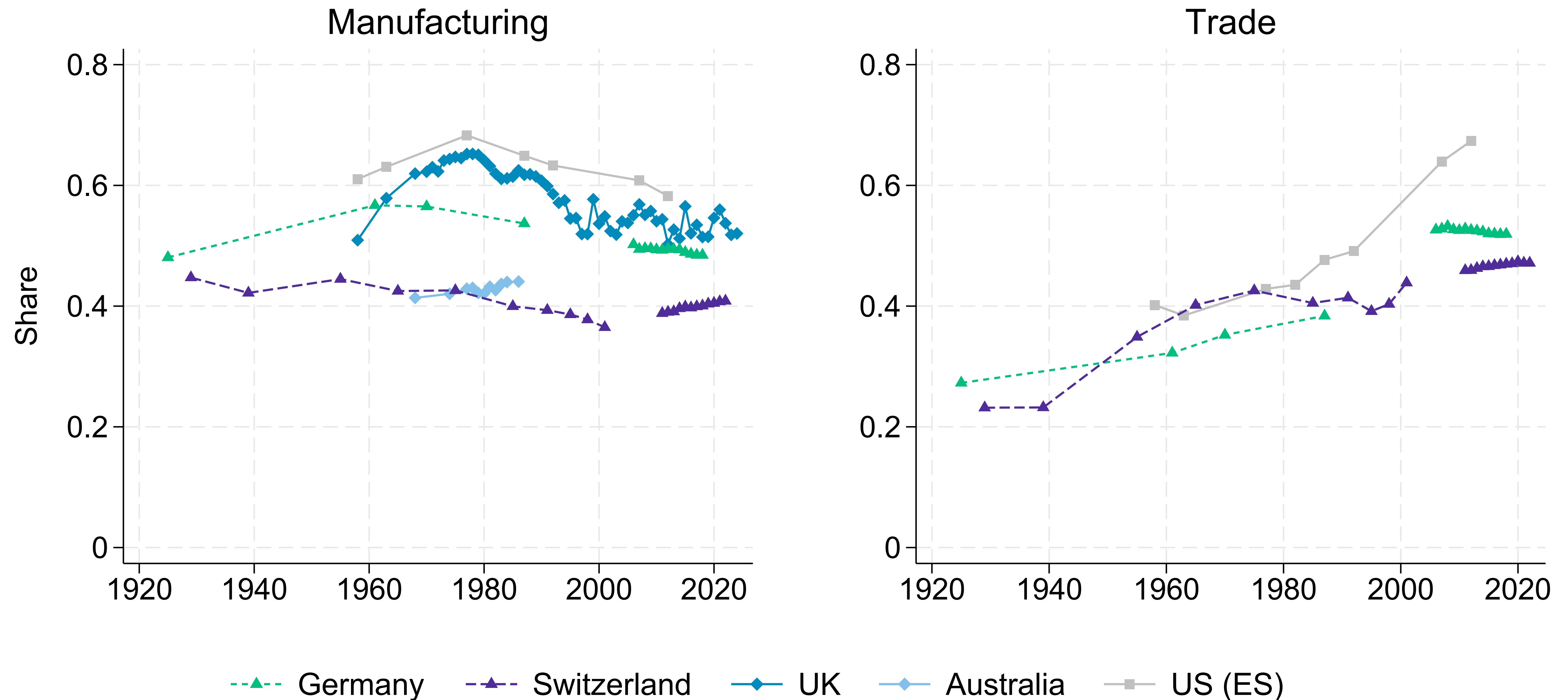
Panel A. Top 1% Share



Top 1% Employment Share Has Been Stable

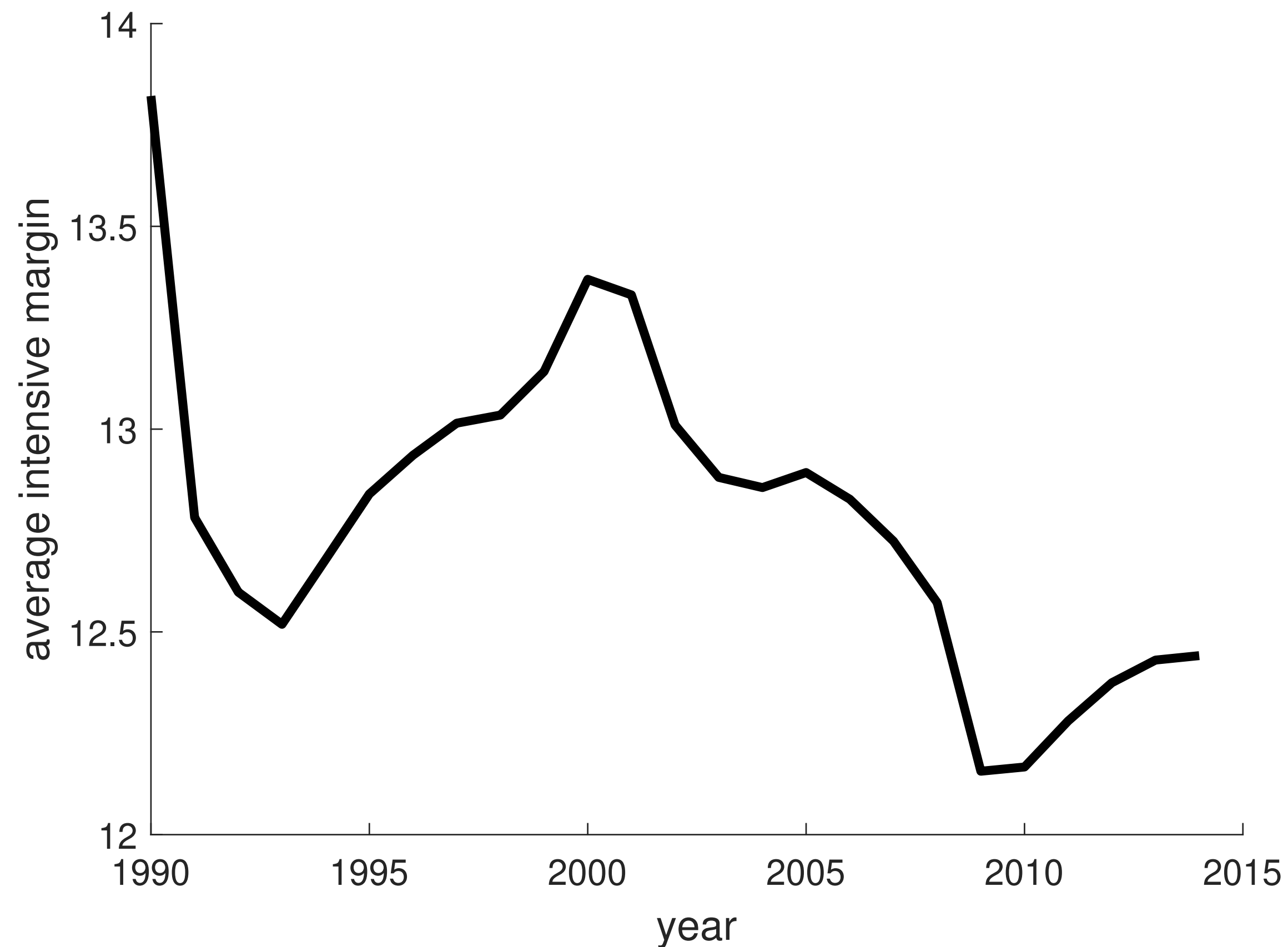


Top 1% Emp. Share in Manufacturing & Trade

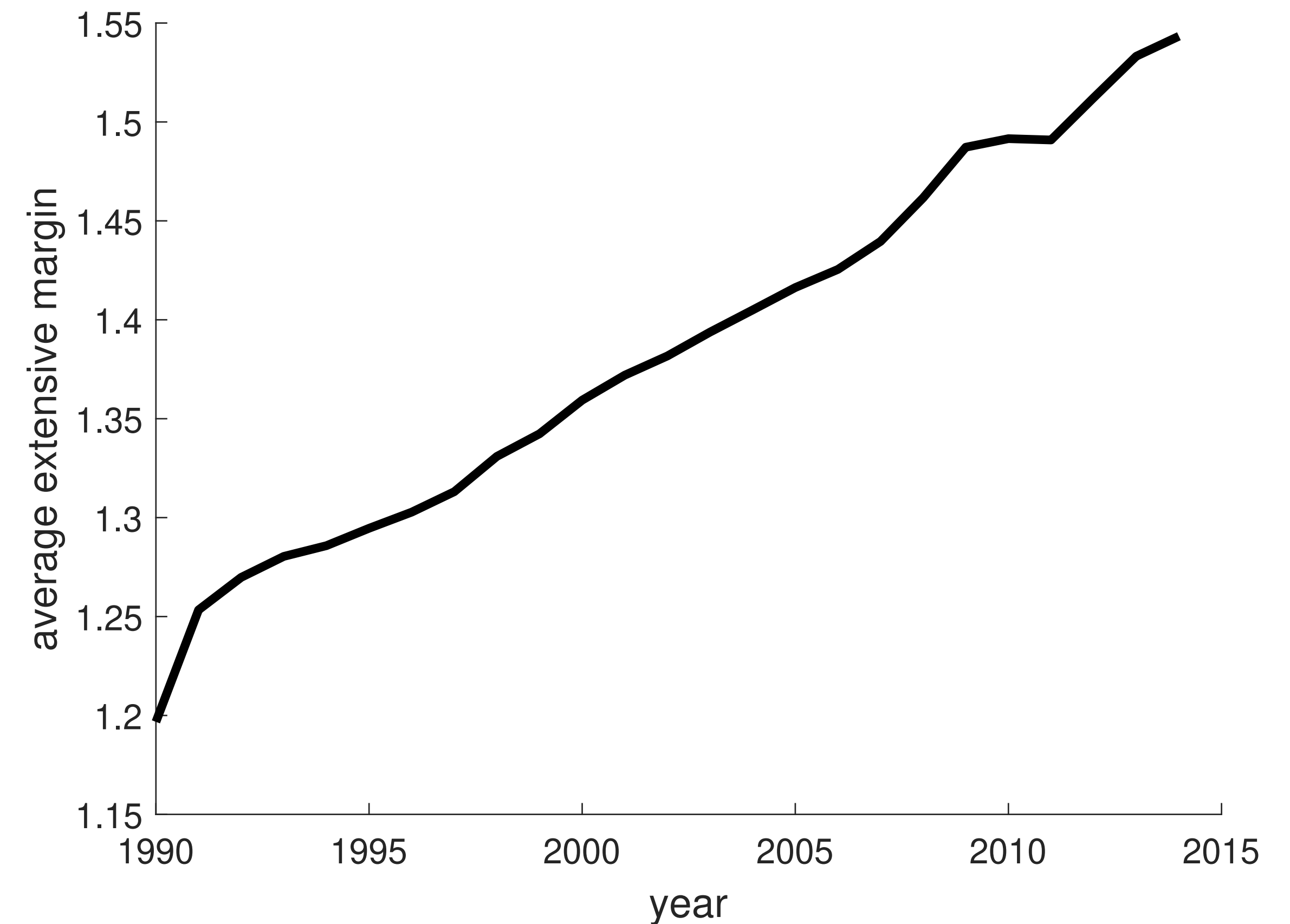


Firm Growth Through Establishments

Average size of establishment



Number of establishments



Source: Cao, Hayyatt, Mukoyama, Sager (2022)

Wrapping Up

1. Problem set 2 is due Dec 21
2. I also look forward to reading your research proposal!