# Investment Networks, Sectoral Comovement, and the Changing U.S. Business Cycle

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Bundesbank Macro Workshop, October 17th 2019

#### Motivation

- Want to understand sources of business cycle fluctuations
- Motivation: change in cyclicality of aggregate labor productivity
  - Pre-1984: highly procyclical
  - Post-1984: roughly acyclical
- Post-1984 period inconsistent with benchmark RBC model driven by aggregate TFP shocks
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- Literature has suggested changes in the shock process or in propagation mechanisms
- Our paper: sectoral investment network crucial to understand declining cyclicality of labor productivity
  - Changing cyclicality of labor productivity reflects shocks to "investment hubs" become more important

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#### Our Contributions

#### New empirical facts using sector-level BEA data 1947 - 2017

- 1. Cyclicality of labor productivity is stable within sectors
- 2. Entire decline is due to changes in covariances across sectors
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## **Multisector business cycle model** driven by observed series of sector-level productivity

- Shocks become less correlated post-1984 ("Great Moderation")
- Matches new empirical facts only w/ realistic investment network
- Post-1984: shocks to investment hubs relatively more important and aggregate labor productivity countercyclical in response
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  - Generate large changes in employment across sectors
- Hubs' value added predicts agg. employment better than GDP and targeting hubs can improve cost-effectiveness of stimulus

## **Empirical Results**

#### Data Source

#### BEA industry database, 1947 - 2017 annual

extended to include finer disaggregation of manufacturing Details

Mining

Construction

Non-metallic minerals Fabricated metals

Computer and electronic manufacturing

Motor vehicles manufacturing

Furniture and related manufacturing Food and beverage manufacturing

Apparel manufacturing

Printing products manufacturing

Chemical manufacturing

Wholesale trade

Transportation and warehousing

Finance and insurance

Management of companies and enterprises

Educational services

Arts, entertainment, and recreation services

Other services

Utilities

Wood products Primary metals

Machinery

Electrical equipment manufacturing Other transportation equipment

Misc. Manufacturing Textile manufacturing

Paper manufacturing

Petroleum and coal manufacturing

Plastics manufacturing

Retail trade

Information

Professional and business services

Administrative and waste management services

Health care and social assistance Accommodation and food services

## Changes in the Aggregate Business Cycle

	Aggregated		Within-Sector	
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.27%	1.36%		
$\rho(y_t - l_t, y_t)$	0.65	0.26		

- $y_t = \log \text{ of value added}$
- $I_t$  = log of employment
- All variables have been HP filtered with smoothing = 6.25
- Within-sector averages weighted by value-added shares

# Cyclicality of Labor Productivity Implied by Rising Volatility of Employment

$$\mathbb{C}orr(y_t, y_t - l_t) = f\left(\mathbb{C}orr(y_t, l_t), \frac{\sigma(l_t)}{\sigma(y_t)}\right)$$

$$= \frac{1 - \frac{\sigma(l_t)}{\sigma(y_t)}\mathbb{C}orr(y_t, l_t)}{\sqrt{1 + \frac{\sigma(l_t)^2}{\sigma(y_t)^2} - 2\frac{\sigma(l_t)}{\sigma(y_t)}\mathbb{C}orr(y_t, l_t)}}$$

#### Components of Labor Productivity

	Pre-1984	Post-1984
$\mathbb{C}orr(y_t - l_t, y_t)$	0.65	0.26
$\mathbb{C}orr(y_t, I_t)$	0.81	0.83
$\mathbb{C}orr(y_t, I_t)$ only	0.65	0.66
$\sigma(l_t)/\sigma(y_t)$	0.76	1.02
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- Within-sector averages weighted by value-added shares
- Inconsistent with RBC model driven by aggregate TFP shocks because aggregate TFP affects output and employment linearly

	Aggregated		Within-Sector	
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.27%	1.36%	3.58%	3.00%
$\rho(y_t - l_t, y_t)$	0.65	0.26	0.73	0.71
$\sigma(l_t)/\sigma(y_t)$	0.75	1.02	0.65	0.65

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## How to Reconcile? Changing Comovement

$$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^{N} (\omega_{jt}^l)^2 \mathbb{V}ar(l_{jt})}{\sum_{j=1}^{N} (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1 - \omega_t) \underbrace{\frac{\sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^l \omega_{ot}^l \mathbb{C}ov(l_{jt}, l_{ot})}{\sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^l \omega_{ot}^l \mathbb{C}ov(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

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	Pre-84	Post-84	Contribution			
			of entire term			
$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(v_t)}$	0.57	0.94	100%			
Within Sector	0.40	0.39	13%			
Between Sector	0.59	1.10	87%			
Within Weight	0.11	0.23				
$(\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$						

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- Comovement of output falls  $\implies$  aggregate volatility falls
- $\cdot$  Comovement of employment stable  $\implies$  agg. volatility stable



► Accuracy

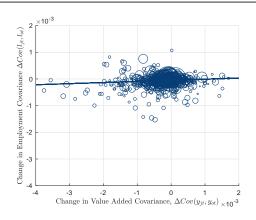
▶ Fixed Weights

▶ First Diffs

▶ Correlations

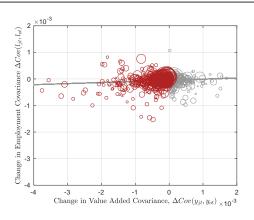
Distribution

## Changes in Covariances, Pre vs. Post 1984



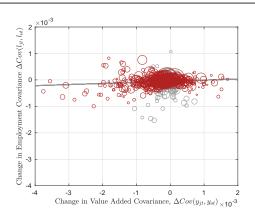
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## Value Added Covariances Fall Substantially



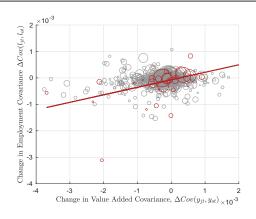
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## 82% of $|\Delta \mathbb{C}ov(I_{jt}, I_{ot})|$ are less than $|\Delta \mathbb{C}ov(y_{jt}, y_{ot})|$



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## Within-Sector Variances Move Together (Coeff $\approx$ .3)



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## Additional Results on the Decomposition

- Results hold at finer disaggregation (450 manufacturing sectors), but not for goods vs. services Details
- 2. Aggregate factor becomes less important for output, but not for employment Details
- 3. Changes in investment volatility and comovement similar to that of employment Details

### Existing Explanations for Changing Business Cycles

#### 1. Changing shock process:

- Aggregate demand shocks: Gali and Gambetti (2009); Barnichon (2010); Sarte, Schwartzman, and Lubik (2015)
- Reallocation shocks become more important: Garin, Pries, and Sims (2018)
- 2. More flexible labor markets: Barnichon (2010), Gali-van Rens (2013)

#### 3. Selective hiring/firing:

- Streamline in recessions: Koenders-Rogerson (2005); Berger (2018)
- · Labor hoarding: Gali-Gambetti (2009); Bachmann (2012)

#### 4. Mismeasurement of inputs or outputs:

- Utilization less procyclical: Fernald-Wang (2016)
- Non-measured intangible investment is procyclical: McGrattan-Prescott (2007, 2012); McGrattan (2017)

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Existing mechanisms abstract from sectoral heterogeneity,

⇒ need new explanation for falling cyclicality of labor productivity

## Model

### Production

- Fixed number of sectors  $j \in \{1, ..., N\}$
- Gross output  $Q_{it}$  produced according to

$$Q_{jt} = A_{jt} \left( K_{jt}^{\alpha_j} L_{jt}^{1-\alpha_j} \right)^{\theta_j} X_{jt}^{1-\theta_j}$$

Intermediates input-output network

$$X_{jt} = \prod_{i=1}^{N} M_{ijt}^{\gamma_{ij}}$$
, where  $\sum_{i=1}^{N} \gamma_{ij} = 1$ 

TFP shocks

$$\log A_{jt+1} = \rho_j \log A_{jt} + \varepsilon_{jt+1}$$
, where  $(\varepsilon_{1t}, ..., \varepsilon_{Nt})' \sim N(0, \Sigma_t)$ 

### Investment

Capital accumulation technology

$$K_{jt+1} = (1 - \delta_j)K_{jt} + I_{jt}$$

Investment input-output network

$$I_{jt} = \Pi_{i=1}^N I_{ijt}^{\lambda_{ij}}, \text{ where } \sum_{i=1}^N \lambda_{ij} = 1$$

### Household and Equilibrium

Representative household with preferences

$$\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left( \log C_t - L_t \right), \quad \text{where } C_t = \prod_{j=1}^N C_{jt}^{\xi_j} \text{ and } \sum_{j=1}^N \xi_j = 1$$

Output market clearing

$$C_{jt} + \sum_{i=1}^{N} M_{jit} + \sum_{i=1}^{N} I_{jit} = Q_{jt}$$

Labor market clearing

$$\sum_{j=1}^{N} L_{jt} = L_t$$

## Calibration

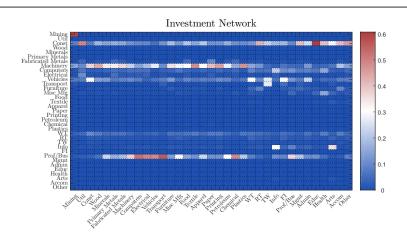
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- Thought experiment: feed in changing shock process, holding structure of the economy fixed
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- **Thought experiment**: feed in changing shock process, holding structure of the economy fixed
  - TFP shocks become less correlated across sectors
  - Main challenge: generate stable comovement of employment
- · Calibrate model in two steps:
  - 1. All parameters other than shocks constant over time Details
  - 2. Feed in measured TFP shocks observed in sectoral data
- Results robust to allowing structure of economy to change
   shock process key change over this period

## **Empirical Investment Network**



- Four investment hubs: construction, machinery, motor vehicles, professional/business services (mostly intellectual property)
- · Supply approximately 2/3 of aggregate investment

### Measurement of Shock Process

$$\log A_{jt+1} = \rho_j \log A_{jt} + \varepsilon_{jt+1}$$
, where  $(\varepsilon_{1t}, ..., \varepsilon_{Nt})' \sim N(0, \Sigma_t)$ 

- Measure sector-level TFP A<sub>jt</sub> as Solow residual, log-polynomially detrended Details
- **Persistence parameters**  $ho_j$ : persistence over whole sample  $ho_j$
- We linearize the model, so  $\Sigma_t$  does not affect decision rules  $\implies$  feed in measured shocks and simulate

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- We linearize the model, so  $\Sigma_t$  does not affect decision rules  $\implies$  feed in measured shocks and simulate
- Robustness: estimate covariance matrix separately for pre vs. post subsamples and compute population moments
  - Empirical estimates not full rank since N=35>T, so collapse number of sectors to N=28< T Details

#### Measured Shock Process

$$\mathbb{V}ar(x_t) = \underbrace{\sum_{j=1}^{N} (\omega_{jt}^{y})^2 \mathbb{V}ar(x_{jt})}_{\text{within-sector}} + \underbrace{\sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^{y} \omega_{ot}^{y} \mathbb{C}ov(x_{jt}, x_{ot})}_{\text{between-sector}}$$

	Measured TFP		HP-Filtered Value Add	
	Pre-84	Post-84	Pre-84	Post-84
100 <i>Var</i> ( <i>x</i> <sub>t</sub> )	0.19	0.10	0.52	0.19
Within Sector	0.03	0.04	0.06	0.05
Between Sector	0.16	0.06	0.46	0.14

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Helpful special case for interpretation:  $\log A_t + \log \widehat{A}_{jt}$ 

- Declining covariances ⇒ aggregate shock less volatile

# Quantitative Results

## Model Matches Aggregate Business Cycle Changes

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$\sigma(l_t)/\sigma(y_t)$	0.76	1.02	0.65	0.65
Model				
$\sigma(y_t)$	2.60%	2.24%	4.03%	4.18%
$\rho(y_t - l_t, y_t)$	0.90	0.45	0.82	0.80
$\sigma(l_t)/\sigma(y_t)$	0.74	0.92	0.48	0.51

- Model generates decline in cyclicality of labor productivity and rise in relative employment volatility
- Model also generates 40% of decline in aggregate GDP volatility ("Great Moderation")

## Model Matches Aggregate Business Cycle Changes



 Model matches timing of change in labor productivity cyclicality (measured using 14-year forward-looking rolling windows)

## Model Consistent with Sectoral Decomposition

$$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)} = \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^{N} (\omega_{jt}^I)^2 \mathbb{V}ar(l_{jt})}{\sum_{j=1}^{N} (\omega_{jt}^Y)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1 - \omega_t) \underbrace{\frac{\sum_{j=1}^{N} \sum_{o \neq i} \omega_{jt}^I \omega_{ot}^J \mathbb{C}ov(l_{jt}, l_{ot})}{\sum_{j=1}^{N} \sum_{o \neq i} \omega_{jt}^J \omega_{ot}^J \mathbb{C}ov(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

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	Data			Model		
	Pre-84	Post-84	Cont.	Pre-84	Post-84	Cont.
$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)}$	0.57	0.94	100%	0.55	0.84	100%
Within Sector	0.40	0.39	13%	0.47	0.47	11%
Between Sector	0.59	1.10	87%	0.56	0.92	89%
Within Weight	0.11	0.23		0.11	0.18	
( $\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$						



► Changing parms

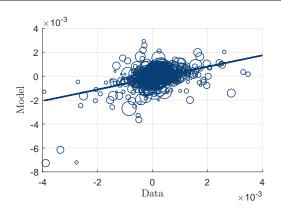
→ GHH

▶ Frisch

Maintananaa

► Capital ACs

#### Model Consistent with Sectoral Decomposition



- Plot sector-pair level "diff-in-diff"  $\Delta \mathbb{C}ov(n_{jt}, n_{ot}) \Delta \mathbb{C}ov(y_{jt}, y_{ot})$
- Model's  $R^2 = 27\%!$

#### Main Challenge: Changing Comovement Patterns

$$\rho_{\tau}^{\mathsf{X}} \equiv \frac{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{\mathsf{X}} \omega_{j}^{\mathsf{X}} \mathbb{C}orr(x_{jt}, x_{jt} | t \in \tau)}{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{\mathsf{X}} \omega_{j}^{\mathsf{X}}}$$

- $x_{jt}$  is HP-filtered + logged variable of interest
- $\omega_{i\tau}^{\mathsf{X}} = \mathbb{E}[\frac{\mathsf{X}_{jt}}{\mathsf{X}_{\mathsf{S}}}]$  are sectoral weights
- \*  $\tau \in \{\text{pre 1984, post 1984}\}\$ is time period

	Da	ta	Мо	del
	Employment	Value added	Employment	Value added
1951-1983	0.55	0.36	0.88	0.35
1984-2012	0.51	0.17	0.84	0.19
Difference	-0.04	-0.19	-0.04	-0.17

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- \*  $\tau \in \{\text{pre 1984, post 1984}\}\$ is time period

	Model		Model, no inv	estment net.
	Employment	Value added	Employment	Value added
1951-1983	0.88	0.35	0.39	0.28
1984-2012	0.84	0.19	0.20	0.10
Difference	-0.04	-0.17	-0.19	-0.18

Without investment network, model does not match comovement and produces no change in labor productivity cyclicality (0.87 to 0.91)

# Mechanism

#### Special Case to Explain the Mechanism

- $N = 2 \text{ sectors}, j \in \{1, 2\}$
- Sector j productivity:  $\log A_{jt} = \log A_t + \log \widehat{A}_{jt}$ 
  - Aggregate shock follows:  $\log A_t = \rho \log A_{t-1} + \varepsilon_t$
  - Sector-specific shock follows:  $\log A_{jt} = \rho \log A_{jt-1} + \varepsilon_{jt}$  $\implies \mathbb{C}ov(\log A_{1t}, \log A_{2t}) = \mathbb{V}ar(\log A_t)$
- Changing shock process: aggregate vs. sectoral components
  - Pre-1984:  $\sigma(\varepsilon_t) = 0.01$  and  $\sigma(\varepsilon_{jt}) = 0.00$
  - Post-1984:  $\sigma(\varepsilon_t) = 0.00$  and  $\sigma(\varepsilon_{jt}) = 0.01$
- Network structure mimics calibrated model
  - Sector 1 is investment hub:  $\lambda_{11} = \lambda_{12} = 1$
  - Uniform intermediates network:  $1 \theta_i = 0.4$
- · Less important paramaters set to standard values:

$$\beta = 0.96, \xi = 0.5, \delta = 0.10, \rho = 0.7$$

## Pre-1984 Period: Effect of Aggregate Shock

Value added: generates correlated increase in both sectors

$$Y_{jt} = \frac{1}{\theta_j} \log A_t + \alpha_j \log K_{jt} + (1 - \alpha_j) \log N_{jt}$$

**Employment**: generates correlated increase in both sectors

- · Quantitatively depends on strength of two effects
  - Direct effect: increases  $\Delta MPN_{it} > 0$ , holding  $N_{it}$  fixed
  - Indirect effect: increases consumption  $\Delta C_{jt} > 0$

$$\frac{MPN_{1t}}{C_{1t}} = \chi (N_{1t} + N_{2t})^{\frac{1}{\eta}} = \frac{MPN_{2t}}{C_{2t}}$$

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• Larger investment response  $\implies$  larger employment response (weaker indirect effect  $\Delta C_{it}$ )

**Value added**: uncorrelated shocks  $\implies$  responses less correlated

· Small spillovers through intermediates network, e.g.

$$\frac{1}{C_{1t}} = MPX_{2t} \frac{1}{C_{2t}}$$

**Value added**: uncorrelated shocks  $\implies$  responses less correlated

**Employment**: primarily response to sector 1-specific shock

Sector 1-specific shock ≈ "investment supply shock"

$$\underbrace{\frac{1}{C_{1t}}}_{\text{marginal cost of capital}} = \beta \left( \frac{1}{C_{jt+1}} \frac{MPK_{jt+1}}{MPK_{jt+1}} + (1-\delta) \frac{1}{C_{1t+1}} \right)$$

• Increased consumption  $\Delta C_{1t} > 0$  lowers cost of capital for both sectors  $\implies$  raises investment ( $\Delta MPK_{jt+1} < 0$ )

**Value added**: uncorrelated shocks ⇒ responses less correlated

**Employment**: primarily response to sector 1-specific shock

Sector 1-specific shock ≈ "investment supply shock"

$$\frac{MPN_{1t}}{C_{1t}} = \chi (N_{1t} + N_{2t})^{\frac{1}{\eta}} = \frac{MPN_{2t}}{C_{2t}}$$

- Sector 1 employment increases to supply investment goods
- Sector 2 employment increases to supply intermediates to sector 1

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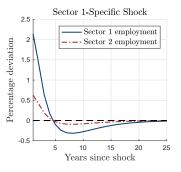
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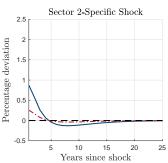
$$\frac{MPN_{1t}}{C_{1t}} = \chi (N_{1t} + N_{2t})^{\frac{1}{\eta}} = \frac{MPN_{2t}}{C_{2t}}$$

- Sector 1 employment increases to supply investment goods
- Sector 2 employment increases to supply intermediates to sector 1
- Sector-2 specific shock ≈ idiosyncratic "investment demand shock" ⇒ small effect on aggregate investment/employment

**Value added**: uncorrelated shocks ⇒ responses less correlated

**Employment**: primarily response to sector 1-specific shock





Also true in full model

#### Changing Business Cycles

	<b>Aggregate shocks</b> (≈ pre-1984)	Sectoral shocks (≈ post-1984)
$Corr(y_{1t}, y_{2t})$	0.99	0.23
$\sigma(y_t)$	1.48%	1.25%
$Corr(n_{1t}, n_{2t})$	1.00	1.00
$\sigma(n_t)$	0.91%	1.04%
$\sigma(n_t)/\sigma(y_t)$	0.62	0.83
$Corr(y_t - n_t, y_t)$	0.96	0.57

- Value added primarily driven by sector-specific shocks
  - · Sector-level value added becomes less correlated
  - Aggregate value added becomes less volatile

#### Changing Business Cycles

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- Employment primarily driven by investment hub shocks
  - Sector-level employment correlations are stable
  - · Aggregate employment volatility is stable

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- Employment primarily driven by investment hub shocks
  - Sector-level employment correlations are stable
  - · Aggregate employment volatility is stable
- Therefore, relative volatility of employment increases
   aggregate labor productivity becomes less cyclical

#### Supporting Evidence of Mechanism

- Volatility of aggregate investment rises relative to output in the post-1984 period Details
- 2. Investment comovement is stable post-1984 and accounts for rise in relative volatility of investment Details

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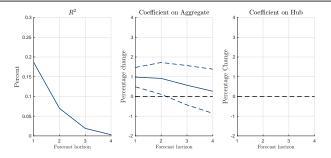
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- 3. Investment hub shocks become more volatile and more correlated post-1984 Details

#### Supporting Evidence of Mechanism

- 2. Investment comovement is stable post-1984 and accounts for rise in relative volatility of investment Details
- 3. Investment hub shocks become more volatile and more correlated post-1984 Details
- 4. Spillovers from investment hubs onto aggregate employment stronger than spillovers for non-hubs Details

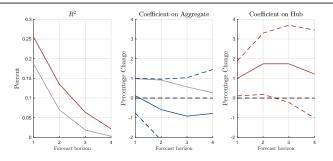
# More Aggregate Implications Of Investment Network

## Forecasting Aggregate Employment



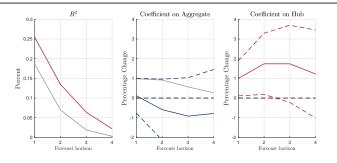
$$\log N_{t+h} - \log N_t = \alpha + \gamma (\log Y_t - \log Y_{t-1}) + \varepsilon_{t+h}$$
 GDP growth rate is standardized

#### Forecasting Aggregate Employment



$$\log N_{t+h} - \log N_t = \alpha + \gamma (\log Y_t - \log Y_{t-1}) + \beta (\log y_{st} - \log y_{st-1}) + \varepsilon_{t+h}$$
  
 $\log y_{st} - \log y_{st-1} = \text{growth rate of hubs' value added}$   
 $(y_{st} = \text{aggregated across hubs, RHS variables standardized})$ 

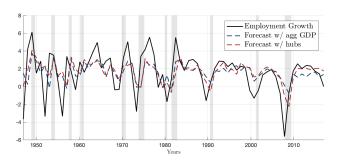
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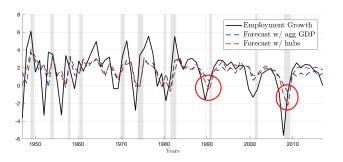
Despite the fact that hubs are 10% of aggregate GDP!

## Fitted Values From Forecasting Regression



$$\log N_{t+1} - \log N_t = \alpha + \beta (\log y_{hubs,t} - \log y_{hubs,t-1}) + \varepsilon_{t+h} \text{ vs.}$$
  
 
$$\log N_{t+1} - \log N_t = \alpha + \beta (\log Y_t - \log Y_{t-1}) + \varepsilon_{t+h}$$

#### Fitted Values From Forecasting Regression



$$\log N_{t+1} - \log N_t = \alpha + \beta (\log y_{hubs,t} - \log y_{hubs,t-1}) + \varepsilon_{t+h} \text{ vs.}$$

$$\log N_{t+1} - \log N_t = \alpha + \beta (\log Y_t - \log Y_{t-1}) + \varepsilon_{t+h}$$

 Hubs especially improve forecasts in post-1984 recessions (and subsequent "jobless recoveries")

#### Improving Cost-Effectiveness of Stimulus Policies

- Goal of many countercyclical stimulus policies is to generate broad-based increase in aggregate employment
- Often work by increasing aggregate demand for goods
- Our model: resources spent on hubs have larger bang-for-the-buck than resources spent at non-hubs
- Back of the envelope (in two-sector model for now): production subsidy  $\tau_t$  financed lump-sum from own-sector output

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	$%\Delta N_t$	$\Delta Y_t$
Blanket 1% subsidy	1.8	1.1
Cost-equivalent hub subsidy	3.5	8.0

⇒ targeting hubs doubles bang-for-the-buck

# Conclusion

#### Our contributions

 Decline in cyclicality of aggregate labor productivity driven by changes in sectoral comovement, not changes within sectors

2. Rising importance of investment hubs accounts for declining cyclicality and changing comovement

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1. Decline in cyclicality of aggregate labor productivity driven by changes in sectoral comovement, not changes within sectors

Rising importance of investment hubs accounts for declining cyclicality and changing comovement

#### Investment network important for aggregate dynamics

- Investment hubs' value added predicts agg. employment better than aggregate GDP
- 2. Stimulus directed toward hubs more cost-efficient

# Appendix

#### Construction of the Data Set Data

- 1. **Value added** from BEA industry database, 1947 2017 (35 NAICS sector level)
- 2. **Investment** and capital stocks from BEA fixed asset tables, aggregated to sector level using shares of capital types, 1947 2017 (35 NAICS sector level)
- 3. **Employment** from two sources, harmonized using Fort-Klimek (2016) crosswalk
  - BEA industry database, 1977 2017 (35 NAICS sector level)
  - Historical supplements, 1948 1977 (SIC codes)

# Average Within-Sector Cycles Using Different Weights

	Time-Varying (Baseline)		Fixed Weights	
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	3.58%	3.00%	3.32%	3.23%
$\sigma(l_t)/\sigma(y_t)$	0.65	0.64	0.65	0.65
$\rho(y_t - l_t, y_t)$	0.73	0.71	0.72	0.73

- $y_t = \log \text{ of value added}$
- $I_t$  = log of employment
- All variables have been HP filtered with smoothing = 6.25



# Divergence of Aggregate and Within-Sector Cycles in First Differences

	Aggregated		Within-Sector	
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	3.39%	2.30%	5.71%	5.01%
$\rho(y_t - l_t, y_t)$	0.68	0.40	0.77	0.74
$\sigma(l_t)/\sigma(y_t)$	0.74	0.93	0.62	0.63

- $y_t = \log \text{ of value added}$
- $I_t$  = log of employment
- · All variables have been first-differenced
- · Within-sector averages weighted by value-added shares



$$\mathbb{V}ar(x_t) = \underbrace{\sum_{j=1}^{N} (\omega_{jt}^{\mathsf{X}})^2 \mathbb{V}ar(x_{jt})}_{\text{within-sector}} + \underbrace{\sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^{\mathsf{X}} \omega_{ot}^{\mathsf{X}} \mathbb{C}ov(x_{jt}, x_{ot})}_{\text{between-sector}}$$



$$\mathbb{V}ar(x_t) = \sum_{j=1}^{N} (\omega_{jt}^{x})^2 \mathbb{V}ar(x_{jt}) + \sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^{x} \omega_{ot}^{x} \mathbb{C}ov(x_{jt}, x_{ot})$$

$$\mathbb{V}ar(y_t) = \sum_{j=1}^{N} (\omega_{jt}^{y})^2 \mathbb{V}ar(y_{jt}) + \sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^{y} \omega_{ot}^{y} \mathbb{C}ov(y_{jt}, y_{ot})$$



$$\begin{split} \frac{\mathbb{V}ar(\mathbf{x}_t)}{\mathbb{V}ar(\mathbf{y}_t)} &= \frac{\sum_{j=1}^N (\omega_{jt}^{\mathsf{v}})^2 \mathbb{V}ar(\mathbf{x}_{jt})}{\sum_{j=1}^N (\omega_{jt}^{\mathsf{v}})^2 \mathbb{V}ar(\mathbf{y}_{jt}) + \sum_{j=1}^N \sum_{o \neq j} \omega_{jt}^{\mathsf{v}} \omega_{ot}^{\mathsf{v}} \mathbb{C}ov(\mathbf{y}_{jt}, \mathbf{y}_{ot})} \\ &+ \frac{\sum_{j=1}^N \sum_{o \neq j} \omega_{jt}^{\mathsf{v}} \omega_{ot}^{\mathsf{v}} \mathbb{C}ov(\mathbf{x}_{jt}, \mathbf{x}_{ot})}{\sum_{j=1}^N (\omega_{jt}^{\mathsf{v}})^2 \mathbb{V}ar(\mathbf{y}_{jt}) + \sum_{j=1}^N \sum_{o \neq j} \omega_{jt}^{\mathsf{v}} \omega_{ot}^{\mathsf{v}} \mathbb{C}ov(\mathbf{y}_{jt}, \mathbf{y}_{ot})} \end{split}$$

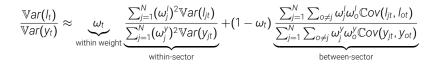


$$\begin{split} \frac{\mathbb{V}ar(x_{t})}{\mathbb{V}ar(y_{t})} &= \frac{\sum_{j=1}^{N} (\omega_{jt}^{y})^{2} \mathbb{V}ar(y_{jt})}{\mathbb{V}ar(y_{t})} \frac{\sum_{j=1}^{N} (\omega_{jt}^{x})^{2} \mathbb{V}ar(x_{jt})}{\sum_{j=1}^{N} (\omega_{jt}^{y})^{2} \mathbb{V}ar(y_{jt})} \\ &+ \frac{\sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^{x} \omega_{ot}^{x} \mathbb{C}ov(x_{jt}, x_{ot})}{\sum_{j=1}^{N} (\omega_{jt}^{y})^{2} \mathbb{V}ar(y_{jt}) + \sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^{y} \omega_{ot}^{y} \mathbb{C}ov(y_{jt}, y_{ot})} \end{split}$$



$$\begin{split} \frac{\mathbb{Var}(x_{t})}{\mathbb{Var}(y_{t})} &= \frac{\sum_{j=1}^{N} (\omega_{jt}^{y})^{2} \mathbb{Var}(y_{jt})}{\mathbb{Var}(y_{t})} \frac{\sum_{j=1}^{N} (\omega_{jt}^{x})^{2} \mathbb{Var}(x_{jt})}{\sum_{j=1}^{N} (\omega_{jt}^{y})^{2} \mathbb{Var}(y_{jt})} \\ &+ \frac{\sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^{y} \omega_{ot}^{y} \mathbb{Cov}(y_{jt}, y_{ot})}{\mathbb{Var}(y_{t})} \frac{\sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^{x} \omega_{ot}^{x} \mathbb{Cov}(x_{jt}, x_{ot})}{\sum_{j=1}^{N} \sum_{0 \neq j} \omega_{jt}^{y} \omega_{ot}^{y} \mathbb{Cov}(y_{jt}, y_{ot})} \end{split}$$

# Accuracy of Decomposition •••••



	Pre-84	Post-84
Actual, variance	0.58	1.04
Approximation, variance	0.57	0.94
Actual, standard deviation	0.76	1.02
Approximation, standard deviation	0.75	0.97

# Decomposition with Fixed Weights ••••

$$\frac{\mathbb{V}ar(I_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^N (\omega_j^I)^2 \mathbb{V}ar(I_{jt})}{\sum_{j=1}^N (\omega_j^V)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1-\omega_t) \underbrace{\frac{\sum_{j=1}^N \sum_{0 \neq j} \omega_j^I \omega_0^J \mathbb{C}\text{ov}(I_{jt}, I_{ot})}{\sum_{j=1}^N \sum_{0 \neq j} \omega_j^V \omega_0^V \mathbb{C}\text{ov}(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

	Pre-84	Post-84	Contribution of entire term		
$\overline{\mathbb{V}ar(l_t)}$ $\overline{\mathbb{V}ar(y_t)}$	0.60	0.81	100%		
Within Sector	0.44	0.32	8%		
Between Sector	0.62	0.93	92%		
Within Weight	0.11	0.20			
( $\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$					

# Decomposition of First Differences

$$\frac{\mathbb{V}ar(I_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^N (\omega_j^I)^2 \mathbb{V}ar(I_{jt})}{\sum_{j=1}^N (\omega_j^V)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1-\omega_t) \underbrace{\frac{\sum_{j=1}^N \sum_{0 \neq j} \omega_j^I \omega_o^J \mathbb{C}\text{ov}(I_{jt}, I_{ot})}{\sum_{j=1}^N \sum_{0 \neq j} \omega_j^V \omega_o^V \mathbb{C}\text{ov}(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

	Pre-84	Post-84	Contribution of entire term		
$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)}$	0.55	0.87	100%		
Within Sector	0.35	0.39	15%		
Between Sector	0.58	1.01	85%		
Within Weight	0.12	0.23			
$(\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$					

# Measuring Comovement with Correlations •••••

$$\rho_{\tau}^{\mathsf{X}} \equiv \frac{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{\mathsf{X}} \omega_{j}^{\mathsf{X}} \mathbb{C}orr(\mathsf{x}_{it}, \mathsf{x}_{jt} | t \in \tau)}{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{\mathsf{X}} \omega_{j}^{\mathsf{X}}}$$

- x<sub>it</sub> is logged + HP-filtered variable of interest
- $\tau \in \{\text{pre 1984, post 1984}\}\)$  is time period
- $\omega_{i\pi}^{x}$  are sectoral shares

Employment	Value added
0.55	0.36
0.51	0.17
-0.04	-0.18
	0.55 0.51





#### Correlations of First Differences

$$\rho_{\tau}^{x} \equiv \frac{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{x} \omega_{j}^{x} \mathbb{C}orr(x_{it}, x_{jt} | t \in \tau)}{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{x} \omega_{j}^{x}}$$

- $x_{it}$  is logged + HP-filtered variable of interest
- $\tau \in \{\text{pre 1984, post 1984}\}\$ is time period
- $\omega_{i\tau}^{x}$  are sectoral shares

	Employment	Value added
1951 - 1983	0.49	0.31
1984 - 2014	0.52	0.18
Difference	0.03	-0.13

# 



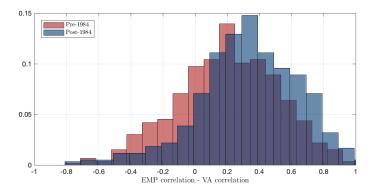
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- $x_{it}$  is logged + HP-filtered variable of interest
- $\tau \in \{\text{pre } 1984, \text{post } 1984\}$  is time period
- $\omega_i^X$  are fixed sectoral shares

	Employment	Value added
1951 - 1983	0.56	0.37
1984 - 2014	0.47	0.14
Difference	-0.09	-0.23

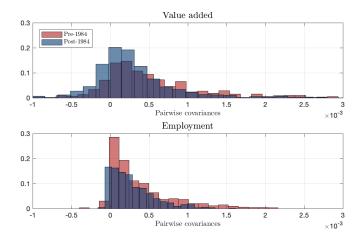
# Distribution of Changes in Correlations •••••

$$\rho_{\tau}^{\mathsf{X}} \equiv \frac{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{\mathsf{X}} \omega_{j}^{\mathsf{X}} \mathbb{C}orr(x_{it}, x_{jt} | t \in \tau)}{\sum_{i=1}^{N} \sum_{j=i+1}^{N} \omega_{i}^{\mathsf{X}} \omega_{j}^{\mathsf{X}}}$$



# Change in Covariances is Broad-Based





# Decomposition at 450 Sector Level (NBER-CES Manufacturing Data)

$$\frac{\mathbb{V}ar(I_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\frac{\omega_t}{\mathbb{V}ar(y_t)}}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^N (\omega_{jt}^J)^2 \mathbb{V}ar(I_{jt})}{\sum_{j=1}^N (\omega_{jt}^V)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1 - \omega_t) \underbrace{\frac{\sum_{j=1}^N \sum_{0 \neq j} \omega_{jt}^J \omega_{ot}^J \mathbb{C}ov(I_{jt}, I_{ot})}{\sum_{j=1}^N \sum_{0 \neq j} \omega_{jt}^J \omega_{ot}^J \mathbb{C}ov(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

	Pre-84	Post-84	Contribution		
			of entire term		
$rac{\mathbb{V}ar(I_t)}{\mathbb{V}ar(y_t)}$	0.40	0.57	100%		
Within Sector	0.34	0.20	1.4%		
Between Sector	0.37	0.60	92.6%		
Within Weight	0.03	0.06			
( $\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$					

# Decomposition At Goods vs. Services Level



$$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^N (\omega_{jt}^l)^2 \mathbb{V}ar(l_{jt})}{\sum_{j=1}^N (\omega_{jt}^v)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1 - \omega_t) \underbrace{\frac{\sum_{j=1}^N \sum_{o \neq j} \omega_{jt}^J \omega_{ot}^J \mathbb{C}ov(l_{jt}, l_{ot})}{\sum_{j=1}^N \sum_{o \neq j} \omega_{jt}^J \omega_{ot}^V \mathbb{C}ov(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

	Pre-84	Post-84	Contribution of entire term		
$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)}$	0.58	1.05	100%		
Within Sector	0.56	0.96	51%		
Between Sector	0.61	1.17	49%		
Within Weight	0.57	0.58			
$(\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$					

## Factor Analysis of Sectoral Comovement



- Study changes in aggregate shock process using factor analysis (e.g. Garin-Pries-Sims 2011)
  - Let  $X_t = (\Delta \log x_{1t}, ..., \Delta \log x_{nt})'$  be a vector of sector-level value added or employment
  - Denote  $V = \text{variance/covariance matrix of } X_t$
  - Decompose as  $V = \Gamma \Lambda \Gamma'$  where  $\Lambda$  is matrix of eigenvalues
  - "Aggregate" factor is first principle component:  $F_t = X_t \Gamma_1$
- Investigate how much variation F<sub>t</sub> explains pre vs. post 1984
- Interpret F<sub>t</sub> as combination of
  - 1. Aggregate shocks which affect all sectors
  - 2. Sectoral shocks propagated across sectors through linkages

# Factor Analysis of Sectoral Comovement ••••

Sample period	$1000 \mathbb{V}ar(\Delta \log X_t)$	Due to 1st component	Residual
Value added			
1951-2014	0.80	0.63 (79%)	0.17 (21%)
1951-1983	1.12	0.97 (86%)	0.15 (14%)
1984-2014	0.46	0.26 (57%)	0.20 (43%)
Employment			
1951-2014	0.51	0.47 (93%)	0.03 (7%)
1951-1983	0.61	0.57 (93%)	0.04 (7%)
1984-2014	0.40	0.38 (94%)	0.02 (6%)

- · Our model's interpretation:
  - 1. Aggregate shocks became less volatile post 1984
  - 2. But sectoral shock spillovers still strong for employment

# Divergence of Aggregate and Within-Sector Cycles Including Investment

	Aggregated		Within-Sector	
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.27%	1.36%	3.58%	3.00%
$\rho(y_t - l_t, y_t)$	0.65	0.26	0.73	0.71
$\sigma(l_t)/\sigma(y_t)$	0.75	1.02	0.65	0.64
$\sigma(i_t)/\sigma(y_t)$	1.94	2.91	2.76	2.84

- $y_t$  = log of value added
- $I_t$  = log of employment
- $i_t = \log \text{ of investment}$
- All variables have been HP filtered with smoothing = 6.25
- Within-sector averages weighted by value-added shares

# Decomposition of Investment Volatility

$$\frac{\mathbb{V}ar(i_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^{N} (\omega_{jt}^i)^2 \mathbb{V}ar(i_{jt})}{\sum_{j=1}^{N} (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1 - \omega_t) \underbrace{\frac{\sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^i \omega_{ot}^i \mathbb{C}ov(i_{jt}, i_{ot})}{\sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^y \omega_{ot}^y \mathbb{C}ov(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

	Pre-84	Post-84	Contribution of entire term
$rac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(y_t)}$	3.77	8.49	100%
Within Sector	4.89	6.14	19%
Between Sector	3.64	9.18	81%
Within Weight	0.11	0.23	
$(\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt}) / \mathbb{V}ar(y_t))$			

### Calibration of Production Parameters Residual Comparison Compari



$$Q_{jt} = A_{jt} (K_{jt}^{\alpha_j} L_{jt}^{1-\alpha_j})^{\theta_j} X_{jt}^{1-\theta_j} \quad \text{where } X_{jt} = \Pi_{i=1}^N M_{ijt}^{\gamma_{ij}}$$

1. **Value added shares**  $\theta$ : average value added as share of gross output (BEA I-O database 1947 - 2017) Details

#### Calibration of Production Parameters Residual Comparison Compari

$$Q_{jt} = A_{jt} (K_{jt}^{\alpha_j} L_{jt}^{1-\alpha_j})^{\theta_j} X_{jt}^{1-\theta_j} \quad \text{where } X_{jt} = \Pi_{i=1}^N M_{ijt}^{\gamma_{ij}}$$

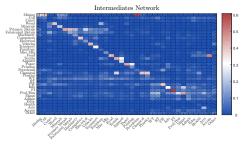
- 1. Value added shares  $\theta$
- 2. **Labor shares**  $\alpha$ : average labor compensation as share of total costs adjusted for taxes and self-employment (BEA I-O database extended back to 1947 - 2017) Details

#### Calibration of Production Parameters Residual Comparison Compari



$$Q_{jt} = A_{jt} (K_{jt}^{\alpha_j} L_{jt}^{1-\alpha_j})^{\theta_j} X_{jt}^{1-\theta_j} \quad \text{where } X_{jt} = \Pi_{i=1}^N M_{ijt}^{\gamma_{ij}}$$

- 1. Value added shares  $\theta$
- 2. Labor shares  $\alpha$
- 3. Intermediates input-output network  $\Gamma$ : average intermediates cost as share of total costs (BEA I-O database 1947-2017)



#### Calibration of Investment Parameters



$$K_{jt+1} = (1 - \delta_j)K_{jt} + I_{jt}$$
 where  $I_{jt} = \prod_{i=1}^{N} I_{ijt}^{\lambda_{ij}}$ 

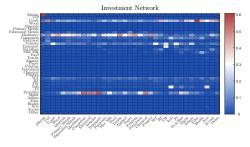
1. **Depreciation rate**  $\delta_i$ : average annual depreciation (BEA fixed assets 1947 - 2017) Details

#### Calibration of Investment Parameters



$$K_{jt+1} = (1 - \delta_j)K_{jt} + I_{jt}$$
 where  $I_{jt} = \prod_{i=1}^N I_{ijt}^{\lambda_{ij}}$ 

- 1. Depreciation rate  $\delta_i$
- 2. **Investment input-output network** Λ: average investment cost from j as share of total investment cost (constructed from BEA capital flows + fixed assets 1947 - 2017)



### Calibration of Preference Parameters Residual Company (1988)



$$\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left( \log C_t - L_t \right), \quad \text{where } C_t = \Pi_{j=1}^N C_{jt}^{\xi_j} \text{ and } \sum_{j=1}^N \xi_j = 1$$

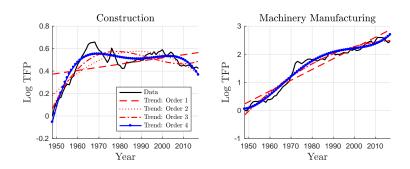
1. **Discount factor**  $\beta = 0.96$  (annual)

### Calibration of Preference Parameters Back

$$\mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \left( \log C_t - L_t \right), \quad \text{where } C_t = \prod_{j=1}^N C_{jt}^{\xi_j} \text{ and } \sum_{j=1}^N \xi_j = 1$$

- 1. **Discount factor**  $\beta = 0.96$  (annual)
- 2. **Consumption shares**  $\xi_j$ : average consumption expenditure on j as share of total consumption expenditure (BEA I-O database 1947 2017) Details

# Detrending Sector-Level Data Back



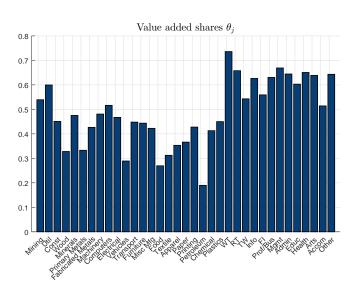
- Sector-level data is not well-described by linear trend
- Choose log-polynomial trend with order = 4 in order to balance:
  - 1. Flexibility of the trend ( $\implies$  higher order)
  - 2. Overfitting of the data ( $\Longrightarrow$  lower order)

# Collapsing Sectors Back

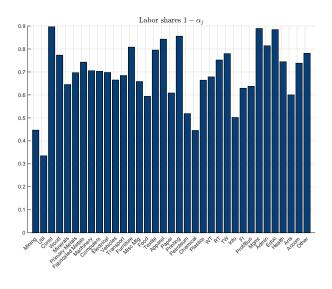
- Need N = 30 to estimate full-rank covariance matrix
- · Collapse all of non-durable manufacturing together because:
  - 1. Not investment hubs, so not central to our main results
  - 2. More similar to each other than other sectors (e.g. services)
  - 3. Readily available from BEA

Utilities	
Wood products	
Primary metals	
Machinery	
Electrical equipment manufacturing	
Other transportation equipment	
Misc. Manufacturing	
Retail trade	
Information	
Professional and business services	
Administrative and waste management services	
Health care and social assistance	
Accommodation and food services	
Non-durable manufacturing	

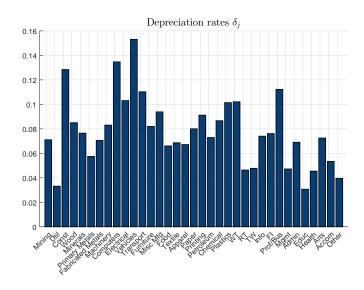
#### Measured Value Added Shares • Back



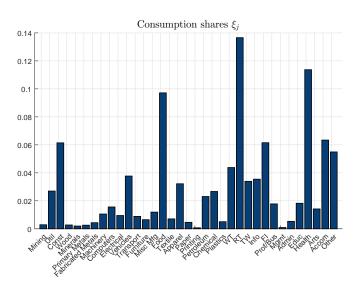
#### Measured Labor Shares Pack



# Measured Depreciation Rates • Back

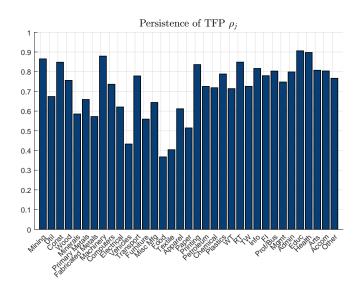


#### Measured Consumption Shares Pack



#### Measured TFP Persistence





#### Interpretation of Change in Shock Process



Helpful special case to interpret change in shock process:

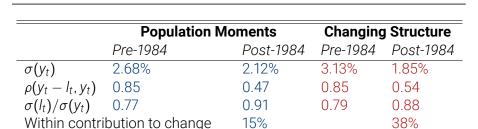
$$\log A_{jt} = \underbrace{\log A_t}_{\text{aggregate shock}} + \underbrace{\log \widehat{A}_{jt}}_{\text{sector-specific shock}}$$

 Characterize using principal components analysis: (on collapsed N = 28 sector data)

Sample period	$1000 \mathbb{V}ar(\Delta \log A_t)$	Due to 1st component	Residual
1949-1983	0.40	0.32 (81%)	0.08 (19%)
1984-2017	0.27	0.15 (56%)	0.12 (44%)

 Volatility of aggregate factor falls in half, but volatility of idiosyncratic factor stable

#### Robustness of Main Results Pack



• Population moments is long simulation for N = 28 < T partition

85%

- Changing structure computes population moments and allows following parameters to differ pre vs. post 1984: Measurement Details
  - Value added shares  $\theta_j$ , labor shares  $\alpha_j$ , intermediates network  $\Gamma_{ij}$
  - Depreciation rates  $\delta_{j}$ , investment network  $\Lambda_{ij}$
  - Consumption shares  $\xi_i$

Between contribution to change

• Persistence of TFP  $\rho_j$ 

62%

#### GHH Preferences Back

	Baseline Res	Changing Structure		
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.60%	2.24%	X%	X%
$\rho(y_t - l_t, y_t)$	0.90	0.45	Χ	X
$\sigma(l_t)/\sigma(y_t)$	0.74	0.92	Χ	X
Within contribution to change		11%		X%
Between contribution to change		89%		Χ%

Description

### Frisch Elasticity of Labor Supply = 4 Pack

	Baseline Res	Changing Structure		
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.60%	2.24%	2.21%	1.84%
$\rho(y_t - l_t, y_t)$	0.90	0.45	0.96	0.8
$\sigma(l_t)/\sigma(y_t)$	0.74	0.92	0.61	0.77
Within contribution to change		11%		21%
Between contribution to change		89%		79%

#### 25% Maintenance Investment • Back

	Baseline Res	Changing Structure		
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.60%	2.24%	2.58%	2.06%
$\rho(y_t - l_t, y_t)$	0.90	0.45	0.93	0.6
$\sigma(l_t)/\sigma(y_t)$	0.74	0.92	0.73	0.88
Within contribution to change		11%		10%
Between contribution to change		89%		90%

#### Capital Adjustment Costs Capital Adjustment Costs

	Baseline Res	Changing Structure		
	Pre-1984	Post-1984	Pre-1984	Post-1984
$\sigma(y_t)$	2.60%	2.24%	2.43%	2.05%
$\rho(y_t - l_t, y_t)$	0.90	0.45	0.92	0.65
$\sigma(l_t)/\sigma(y_t)$	0.74	0.92	0.68	0.85
Within contribution to change		11%		7%
Between contribution to change		89%		93%

- Each sector faces quadratic capital adjustment cost  $\varphi$
- Choose large adjustment cost parameter  $\varphi = 4$

#### Measurement of Parameter Changes over Time



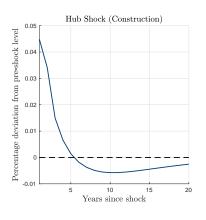
- Most parameters based on moments that are available year-by-year: value added shares, intermediates network, depreciation rates, consumption shares
- **Persistence of TFP** estimated via MLE on two subsamples
- Labor shares combines two data sources (harmonized using Fort-Klimek crosswalk):
  - 1. BEA industry database 1987 2017 on payroll, value added, indirect taxes, and self-employment (NAICS)
  - 2. Historical data on payroll, value added, and indirect taxes 1948 - 1987 (SIC)
  - 3. Self-employment back-casted using average ratio from NAICS data

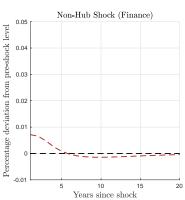
#### Measurement of Parameter Changes over Time



- See sector's total investment expenditure year-by-year, but need to allocate across sectors using bridge file
- All structures produced by construction, except for mining (following BEA practice)
- Intellectual property also follows BEA practice:
  - Pre-packed software and most artistic originals from info
  - Other software and R&D investment from prof/technical
  - Misc. other small allocations.
- Equipment production combines three BEA datasets:
  - 1997 2017 census year: BEA provides bridge file
  - 1987 and 1992: BEA provides SIC bridge file, harmonized using Fort-Klimek
  - 1948 1987: interpolate based on observed bridge files

### Effects of Sectoral Shocks on Aggregate Employment in Full Model





## Divergence of Aggregate and Within-Sector Cycles Including Investment

	Aggr	egated	Within-Sector		
	Pre-1984 Post-1984		Pre-1984	Post-1984	
$\sigma(y_t)$	2.27%	1.36%	3.58%	3.00%	
$\rho(y_t - l_t, y_t)$	0.65	0.26	0.73	0.71	
$\sigma(l_t)/\sigma(y_t)$	0.75	1.02	0.65	0.64	
$\sigma(i_t)/\sigma(y_t)$	1.94	2.91	2.76	2.84	
$\sigma(i_t)/\sigma(y_t)$ model	Χ	Χ	Χ	Χ	

- $y_t = \log \text{ of value added}$
- $I_t$  = log of employment
- $i_t = \log \text{ of investment}$
- All variables have been HP filtered with smoothing = 6.25
- Within-sector averages weighted by value-added shares
- Model = model with capital adjustment costs

#### Decomposition of Investment Volatility

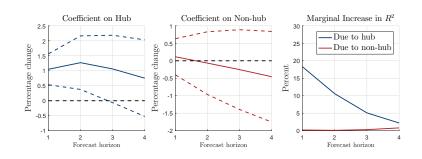
$$\frac{\mathbb{V}ar(i_t)}{\mathbb{V}ar(y_t)} \approx \underbrace{\omega_t}_{\text{within weight}} \underbrace{\frac{\sum_{j=1}^{N} (\omega_{jt}^i)^2 \mathbb{V}ar(i_{jt})}{\sum_{j=1}^{N} (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})}}_{\text{within-sector}} + (1 - \omega_t) \underbrace{\frac{\sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^i \omega_{ot}^j \mathbb{C}ov(i_{jt}, i_{ot})}{\sum_{j=1}^{N} \sum_{o \neq j} \omega_{jt}^y \omega_{ot}^y \mathbb{C}ov(y_{jt}, y_{ot})}}_{\text{between-sector}}$$

	Data			Model		
	Pre-84	Post-84	Cont.	Pre-84	Post-84	Cont.
$\frac{\mathbb{V}ar(l_t)}{\mathbb{V}ar(v_t)}$	3.77	8.49	100%	Χ	Χ	100%
Within Sector	4.89	6.14	19%	Χ	Χ	42%
Between Sector	3.64	9.18	81%	X	Χ	58%
Within Weight	0.11	0.23		Χ	Χ	
$(\omega_t = \sum_{j=1}^N (\omega_{jt}^y)^2 \mathbb{V}ar(y_{jt})/\mathbb{V}ar(y_t))$						

# Rising Importance of Investment Hub Shocks (Unweighted Averages)

	Pre-84	Post-84	Percentage Change
$rac{\mathbb{E}[\sigma(A_{jt}) hubs]}{\mathbb{E}[\sigma(A_{it}) non-hubs]}$	1.13	1.27	12%
$\mathbb{E}[\mathbb{C}orr(A_{it}, A_{ot}) $ hubs]	0.25	0.27	8%
$\mathbb{E}[\mathbb{C}orr(A_{jt}, A_{ot}) $ non-hubs]	0.17	0.06	-65%

### Spillovers from Sector-Level Shocks Onto Aggregate Employment



$$\log N_{t+h} - \log N_t = \alpha + \gamma (\log y_{hub,t} - \log y_{hub,t-1}) + \beta (\log y_{non,t} - \log y_{non,t-1}) + \varepsilon_{t+h}$$

 $y_{st} = \text{aggregated across } s \in \{\text{hub, non-hub}\}\ \text{in year } t \\ \log y_{s,t} - \log y_{s,t-1} = \text{is standardized}$