

# ERWIT Lecture

## Globalization and Inequality

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# GLOBALIZATION AND INEQUALITY



# Main Content

- Historical background

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  - Residual inequality

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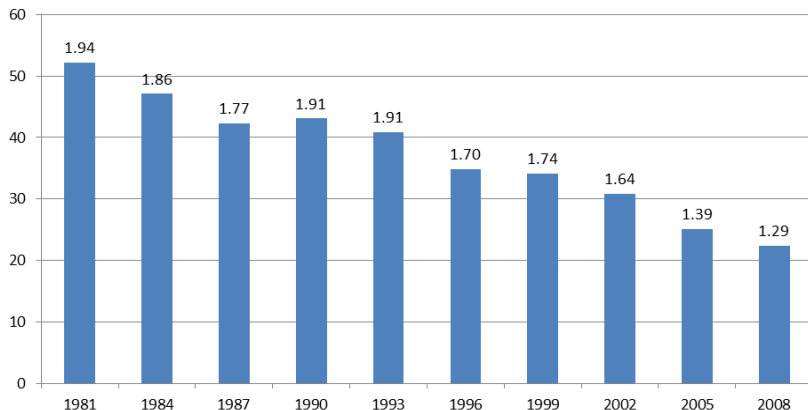
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- Rising inequality **within** many OECD countries

# Declining Poverty

## World poverty:

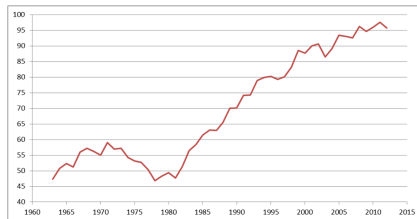
Percent of developing world people with < 1.25 PPP (2005) adjusted dollars per day, and the number of these people in billions



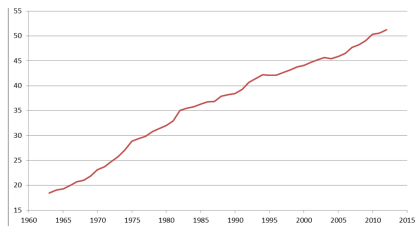
Source: Anand and Segal (2015, Handbook Table 11.8), from Chen and Ravallion (2012).

# College Wage Premium

U.S. college wage premium and share of college hours:



College versus high school measured wage gap. Autor (2014)



College share of hours worked: all working age adults. Autor (2014)

# Rising Wage Inequality

**Table 2.1** Ratio of Wages of the 90th to the 10th Percentile of Male Earners

	<i>Australia</i>	<i>Canada</i>	<i>Finland</i>	<i>France</i>	<i>Italy</i>	<i>Japan</i>	<i>Norway</i>	<i>Sweden</i>	<i>UK</i>	<i>U.S.A.</i>
1979	2.75	3.46	2.44	3.39	2.29	2.59	2.05	2.12	2.46	3.19
1994	2.94	3.78	2.53	3.42	2.64	2.77	1.97	2.20	3.22	4.26

*Source:* Katz and Autor (1999, table 10).

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  - Consistent with **global** technological change

# Factor Content Analysis

**Table 3.1** Contribution of Immigration and Trade with Less-Developed Countries to the US College Wage Premium, 1980–1995 (in log points)

Elasticity of substitution	2	1.41	1
Immigration	0.007	0.009	0.013
Trade	0.007	0.010	0.014
Actual change	0.191	0.191	0.191

*Data source: Borjas, Freeman, and Katz (1997, table 18).*

# Employment

Share of within-industry contribution to the increased percentage of nonproduction workers (Berman, Bound and Machin, 1998):



Similar shifts in less developed countries



- Sectors with faster increases in the demand for nonproduction workers were

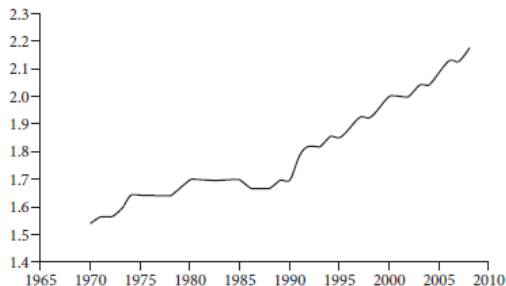
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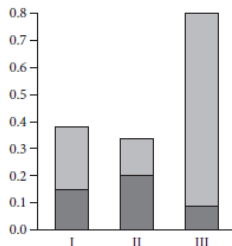
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  - more innovative
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  - more intensive in computer use
  - positive correlation across many countries, rich and poor, in sectoral technology upgradings

# Growth of Offshoring



**Figure 5.1** Index of offshoring manufacturing products in the world economy, 1970–2008. *Data source:* The inverse of VAX for manufacturing in Johnson and Noguera (2016, table D1).

# Offshoring vs. Technology (US)



**Figure 5.4** Estimated annual rates of increase in wages of nonproduction relative to production workers in percent for different measures of computers and high-tech capital: United States, 1979–1990. The lower bar represents the impact of offshoring while the upper bar represents the impact of computers and other high-tech capital. *Data source: Feenstra and Hanson (1999, table VI).*

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- **Residual wage inequality** was large and contributed significantly to rising wage inequality

# Response: Broadening the Canvas

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# Sorting and Matching

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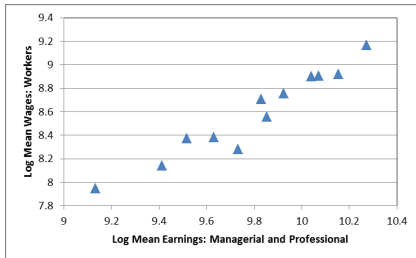
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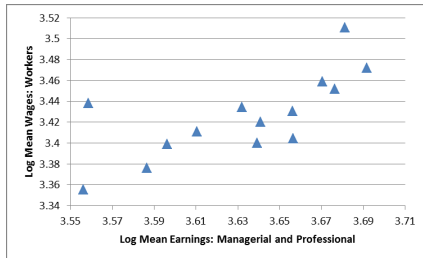
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- Helps explaining changes in inequality between sectors, within sectors, different segments of the distribution, inequality correlation between inputs
- Generates predictions about growth, trade and inequality: Grossman and Helpman (2018)

# Correlated Compensation Levels

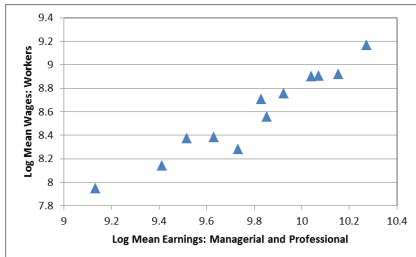


Brazil 1994

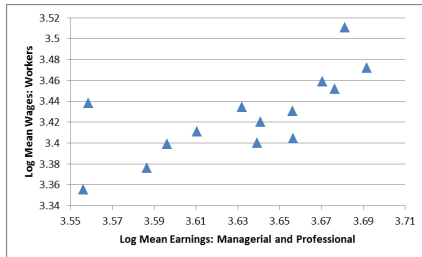


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# Correlated Compensation Levels



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- In Brazil, dispersion of managerial compensation was negatively correlated with dispersion of workers' wages across industries

# Evidence on Inequality with Matching: Lee (2017)

Table: Rise in U.S. Real Wages: 2000-2007 (in percent)

Worker type	HD	HG	SC	CG	AD
Decline in trade costs	1.15	1.15	1.49	1.62	1.81
Rise in China's productivity	0.06	0.09	0.13	0.14	0.17

Worker types: high school dropouts (HD), high school graduates (HG), some college education (SC), college graduates (CG), and advanced degrees (AD).

# Regional Disparity: Brazil

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- Dix-Carneiro and Kovak (2015) extended the analysis to include skilled and unskilled workers and found that skill premia declined, but the declines can explain at most 14% of the 1991-2000 actual declines



# The China Shock

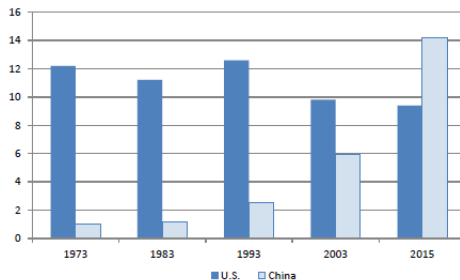


Figure 8.1: Shares of China and the United States in the world's merchandise exports (in percent). Source: World Trade Organization (2016, Table A4).

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- Accounting for exports was important in a study of Germany by Dauth, Findeisen and Suedekum (2014)
  - There the China shock had much smaller effects, while the rise of Eastern Europe had bigger effects

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  - Lazear and Spletzer (2012): in the last quarter of 2007 there were over 12 million hires and separations, other estimates for different period find 9 million hires and separations

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- Employment problems are highly localized, not **macro** problems

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- A theory of trade with heterogeneous firms and monopolistic competition has been developed by Melitz (2003), emphasizing [selection into exporting](#)

# Firm Characteristics

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  - globalization induces convergence of growth and inequality in the presence of *international R&D spillovers*
  - in the absence of international R&D spillovers globalization raises growth:

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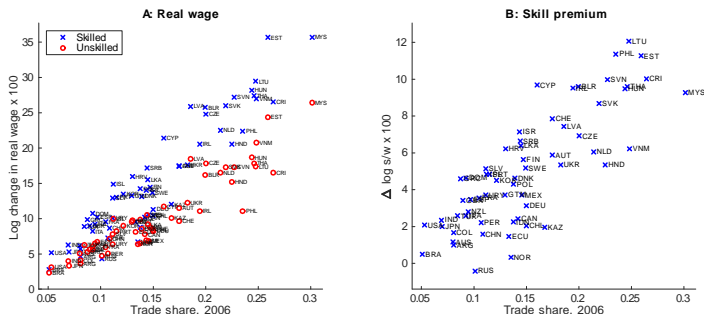
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- Factor proportions magnify the inequality in rich countries, moderate inequality in developing countries

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# Quantitative Evaluation

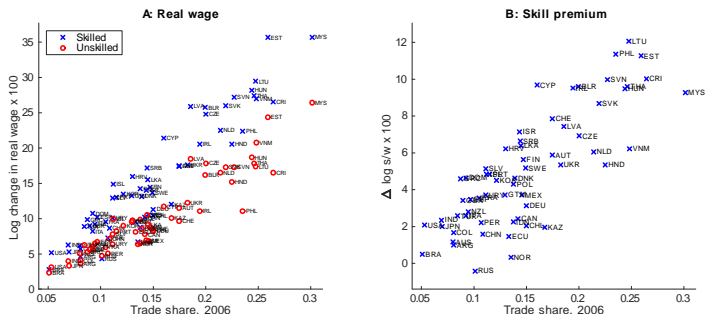
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- These combined effects explain only a fraction of the rise in the college wage premium

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- Calibrating the model to US data, he found that trade can explain 20% of the rise in the skill premium between 1980 and 1995, half of it due to *directed technical change*

# Residual Wage Inequality

- **Brazil:**

	<b>Level 1990</b>	<b>Change 1986-95</b>
Residual wage inequality	57	48
—within sector-occupation	88	91

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- **Sweden:**

	<b>Level 2001</b>	<b>Change 2001-7</b>
Residual wage inequality	70	87
—within sector-occupation	83	79

Akerman, Helpman, Itskhoki, Muendler and Redding (2013)

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# Trade Liberalization

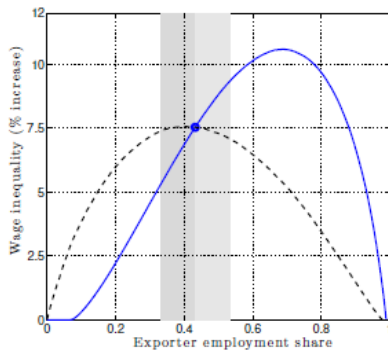


Figure 1: Counterfactual wage inequality

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- 5 While the novel mechanisms find support in the data, better explaining the link between trade and inequality, their *quantitative* impact is not overwhelming
- 6 Based on available evidence, foreign trade does not appear to be the main driver of inequality