

# Topics in Macroeconomics and Labor Economics

## Lecture 1: Introduction

Satoshi Tanaka

University of Queensland

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# Course Information

- ▶ Instructor: Dr. Satoshi Tanaka
  - ▶ PhD at the University of Minnesota
  - ▶ Field of interests: Macro, Labor, and Demographics
  - ▶ E-mail: [s.tanaka.0509@gmail.com](mailto:s.tanaka.0509@gmail.com)
- ▶ Consultation hours: every day after the class

# Assessment

- ▶ Submit your work for a problem set
  - ▶ Data exercise using R
  - ▶ Due by June 29 (50% of the final grade)
- ▶ Submit your research proposal/referee report
  - ▶ Due by June 29 (50% of the final grade)
- ▶ Upload your work to GitHub (<https://github.com/>)
  - ▶ Upload all your works for the problem set to GitHub
  - ▶ Let me know your GitHub site address by email

## Introduction to Wage Inequality

# Motivation

- ▶ Why do we need to study wage/earnings inequality?
- ▶ Understanding the degree and causes of it helps us design
  - ▶ Tax policies (e.g. labor income tax)
  - ▶ Social insurance policies (e.g. unemployment insurance)
  - ▶ Education policies (e.g. educational subsidy)
- ▶ Two important issues
  1. Skill v.s. luck (better to provide insurance for bad lucks)
    - ▶ High salary due to his/her greater skills?
  2. Efficiency v.s. equality (trade-off for a redistributive policy)
    - ▶ Is the wage inequality Pareto optimal?

# Approach

1. Look at the data carefully
  - ▶ Data visualization is important
2. Build an economic model
  - ▶ Write a simplest economic model that explains the relationship
3. Bring the model to data
  - ▶ Reduced-form estimation
    - ▶ Derive a testable equation from the model and estimate it
  - ▶ Structural estimation
    - ▶ Estimate the model's underlying parameters directly

# (Quasi-) Experiment

## 1. Randomized Control Trial (Lab or Field Experiment)

- ▶ Clean comparison of two groups of subjects (treated and control groups), but not easy to create such an environment

## 2. Quasi-Experimental Methods (Check Angrist and Pischke's book)

### 2.1 Differences-in-Differences (DID)

- ▶ Compare changes over time between two groups (treated and control groups)

### 2.2 Instrumental Variables (IV)

- ▶ Use an exogenous variation to assign subjects to the treated group

### 2.3 Regression Discontinuity Design (RDD)

- ▶ Compare two groups (treated and control groups) at a cut-off point

# Data Availability

## 1. Survey Data

- ▶ U.S. Population Census, American Community Survey, 1850 - present
  - ▶ Nationally representative cross-sectional data, decennial
- ▶ Current Population Survey (CPS), March/MORG, 1962 – present
  - ▶ Monthly or annual (March CPS) cross-sectional data. Can be use as short-panel data (MORG)
- ▶ Panel Study of Income Dynamics (PSID), 1968 – present
  - ▶ Nationally representative household panel data
- ▶ National Longitudinal Survey of Youth 1979/1997 (NLSY79/97)

## 2. Administrative Data (matched-employer-employee panel data)

- ▶ Social Security Administration's (SSA) Data, 1957 – present
- ▶ U.S. Census LEHD Data, 1985 – present



# Literature of Wage/Earnings Inequality

## 1. First generation

- ▶ Changes in returns to worker's skills (education and experience)

## 2. Second generation

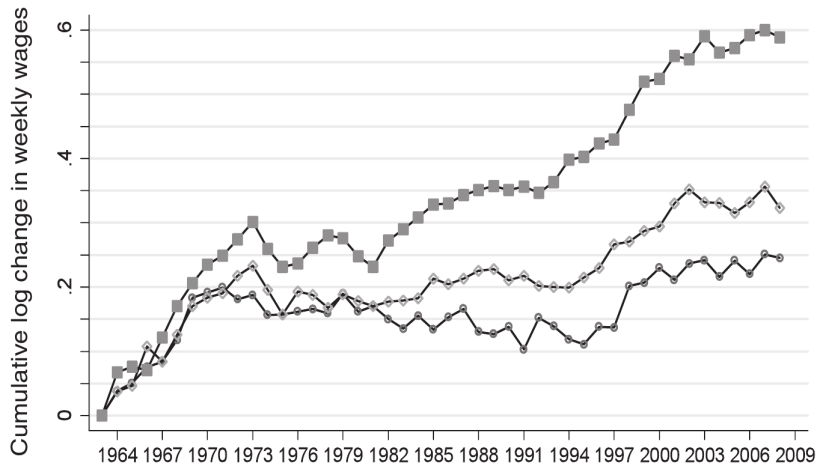
- ▶ Changes in returns to job characteristics (occupation and task)

## 3. Third generation

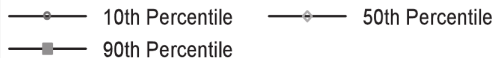
- ▶ Explore the drivers for 1 and/or 2
- ▶ Use between-firm variations

## Empirical Facts on Inequality (Autor and Acemoglu, 2011, March CPS)

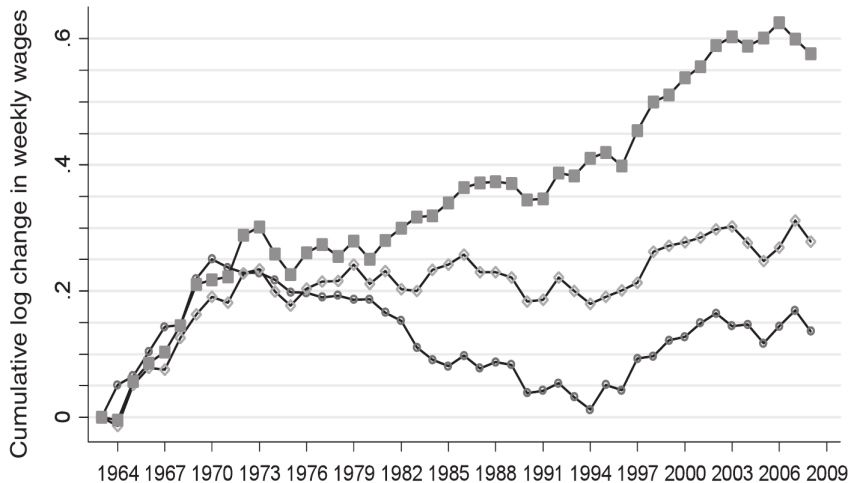
# Change in Real Weekly Earnings, FTFY All



(a)



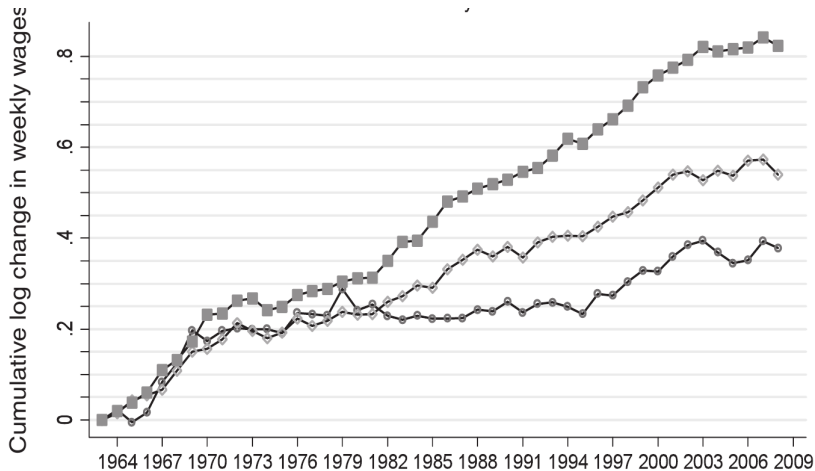
# Change in Real Weekly Earnings, FTFY Males



(b)

—○— 10th Percentile      —◇— 50th Percentile  
—■— 90th Percentile

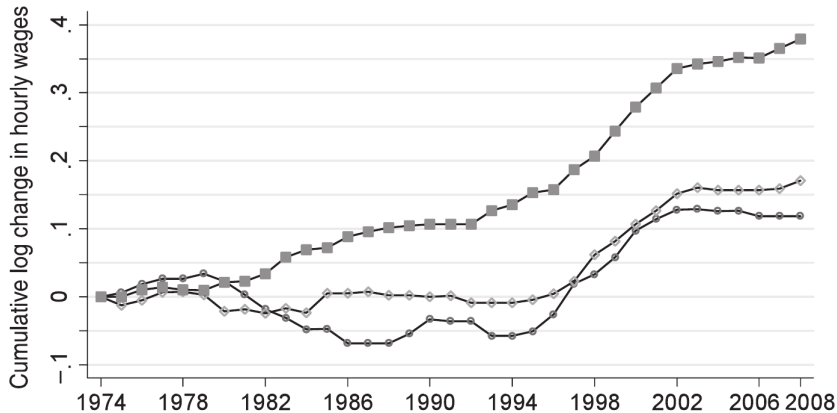
# Change in Real Weekly Earnings, FTFY Females



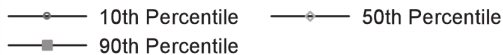
(c)

—○— 10th Percentile      —◇— 50th Percentile  
—■— 90th Percentile

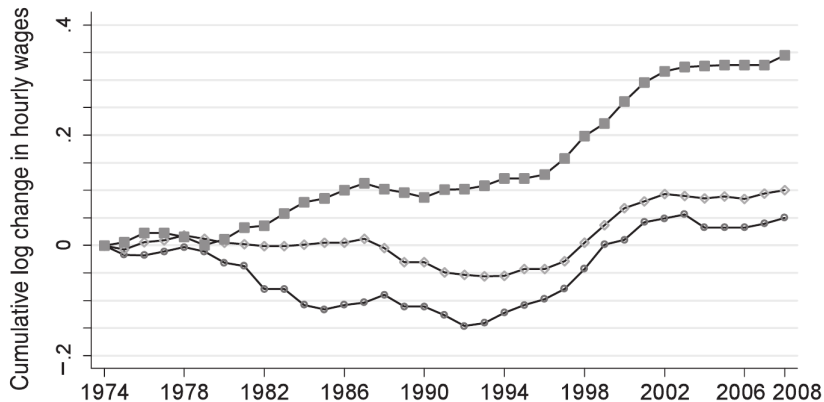
# Change in Real Hourly Earnings, FTFY All



(a)



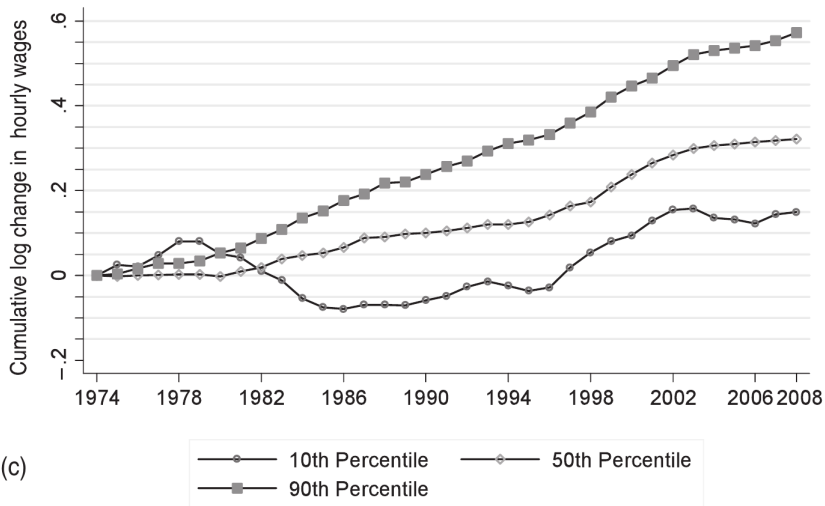
# Change in Real Hourly Earnings, FTFY Males



(b)

—○— 10th Percentile      —◇— 50th Percentile  
—■— 90th Percentile

# Change in Real Hourly Earnings, FTFY Females



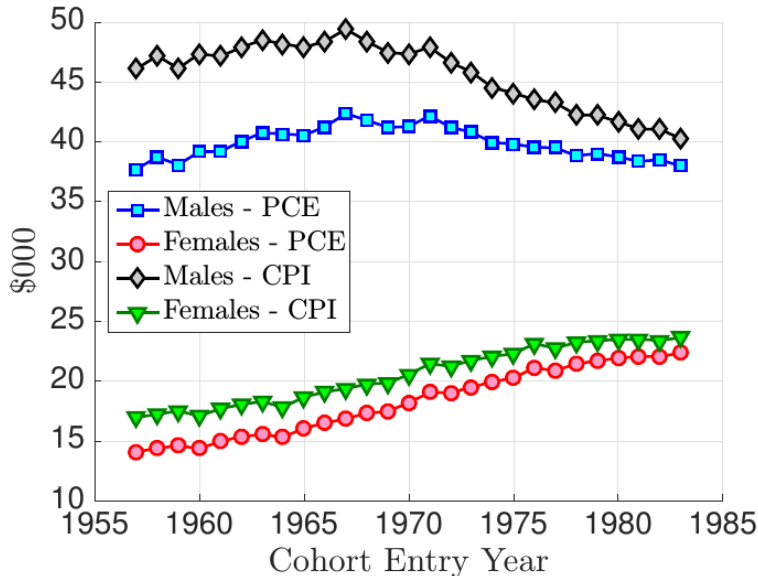


# International Comparison

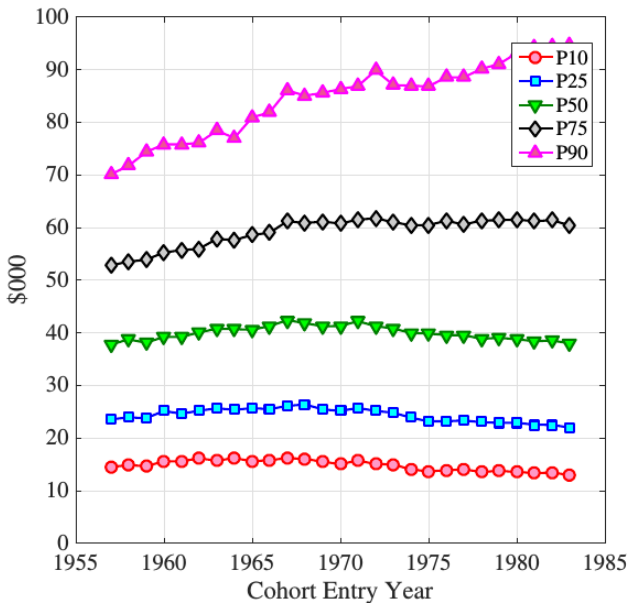
- ▶ Very large increases: US, UK, Germany
- ▶ Modest increases: Australia, Canada, Japan, Spain, and Sweden
- ▶ No noticeable changes: France and Italy
- ▶ Modest falls: Netherlands
- ▶ Large falls: South Korea

## Lifetime Income Inequality (Guvenen et al, 2017, SSA Data)

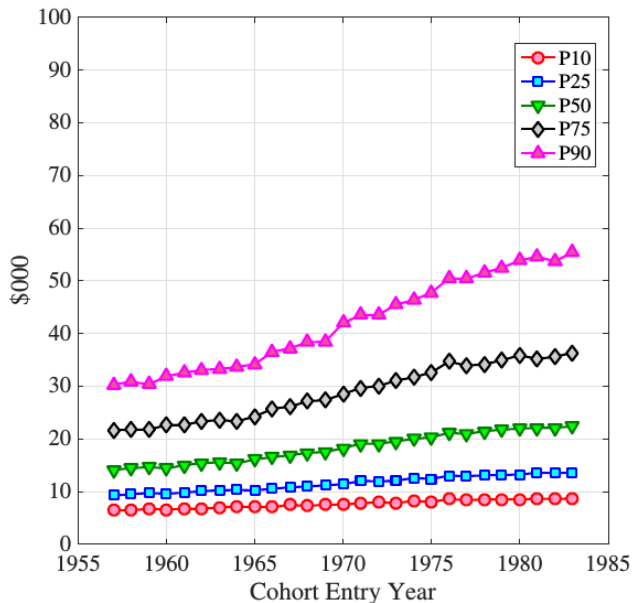
## Median Lifetime Income by Cohort



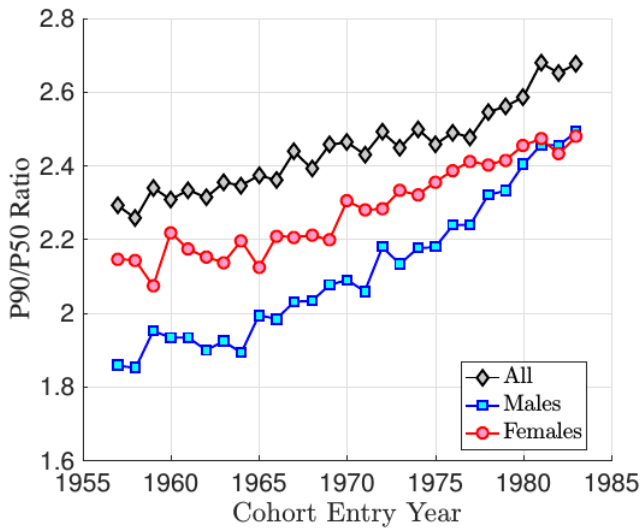
## Percentiles of Lifetime Income by Cohort, Males



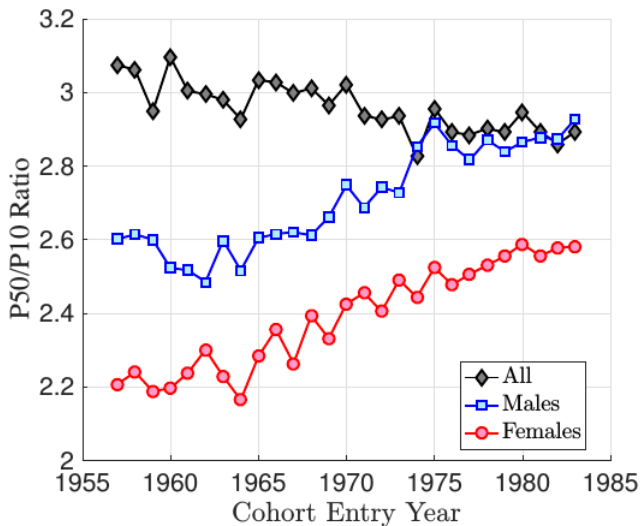
# Percentiles of Lifetime Income by Cohort, Females



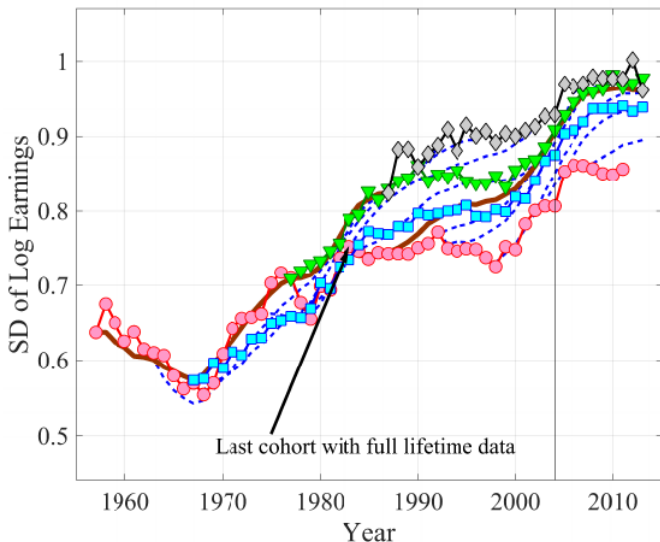
# P90/P50 Ratio



## P50/P10 Ratio

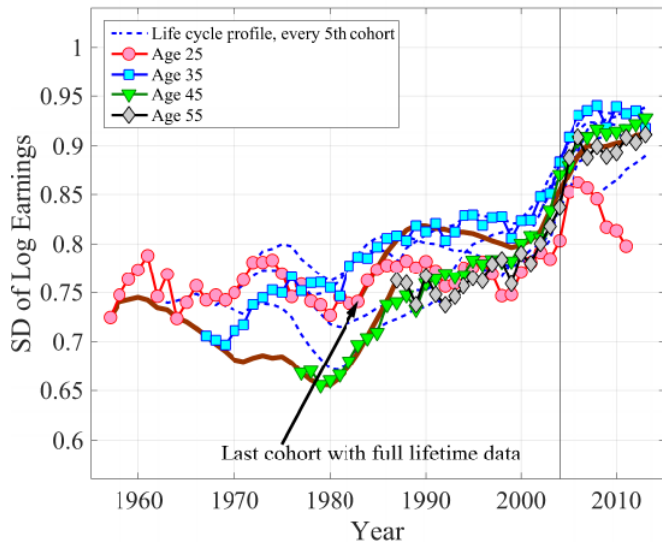


# Standard Deviation of Log Earnings by Age, Males

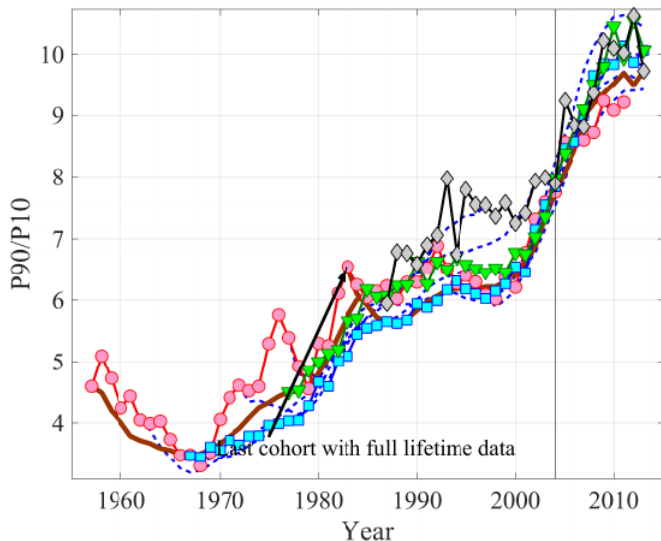




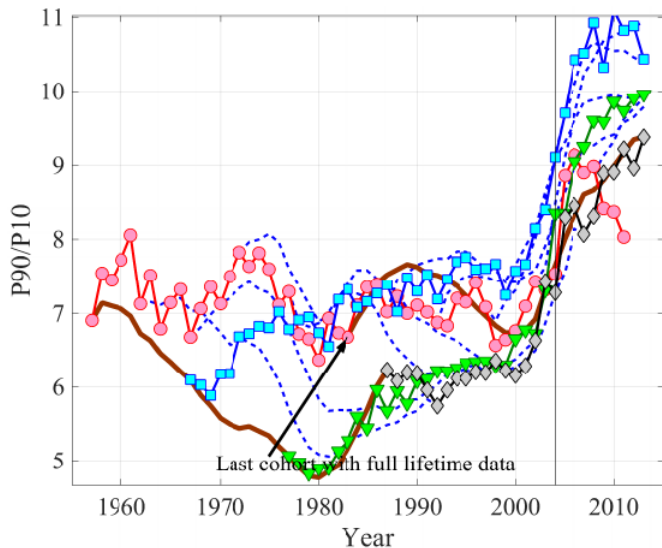
# Standard Deviation of Log Earnings by Age, Females



# P90/P10 Ratio of Log Earnings by Age, Males



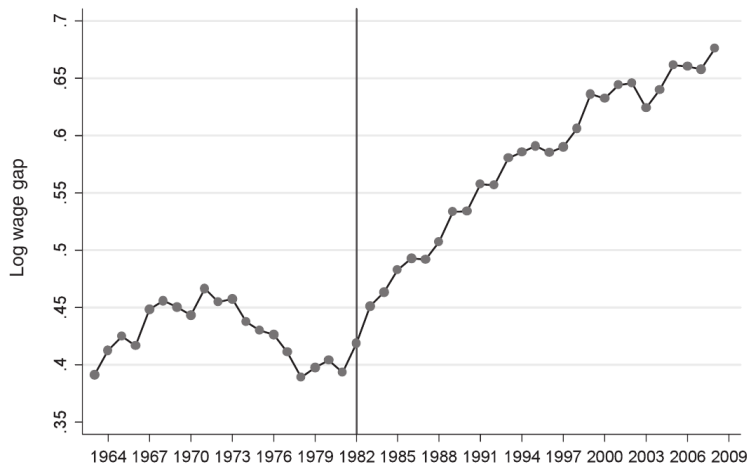
## P90/P10 Ratio of Log Earnings by Age, Females



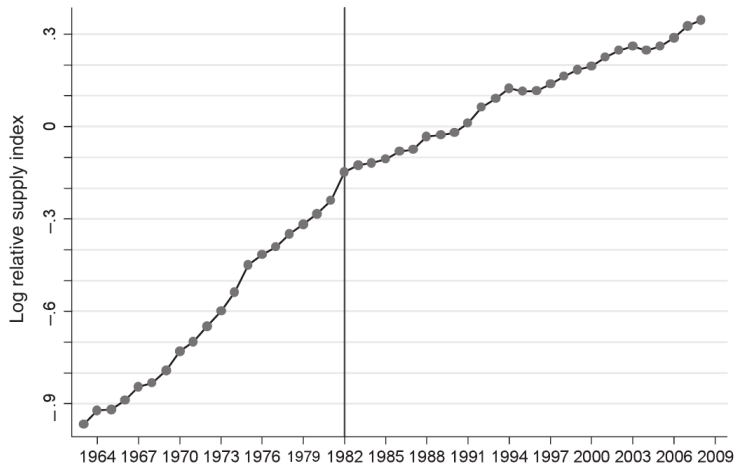
## Skill Premium

(Autor and Acemoglu, 2011, March CPS)

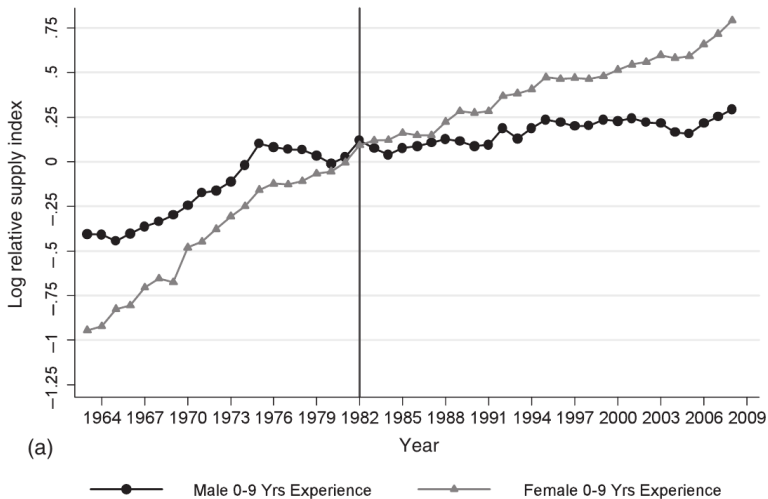
# College/High-School Log Weekly Wage Ratio



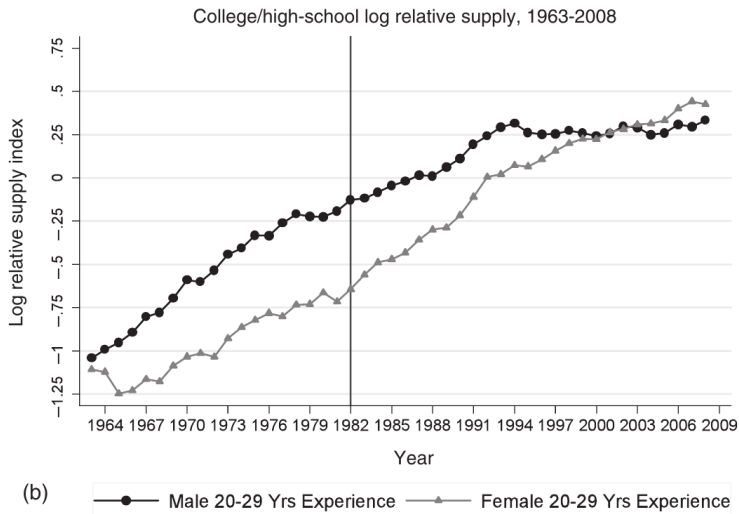
# College/High-School Log Relative Supply



# College/High-School Log Relative Supply

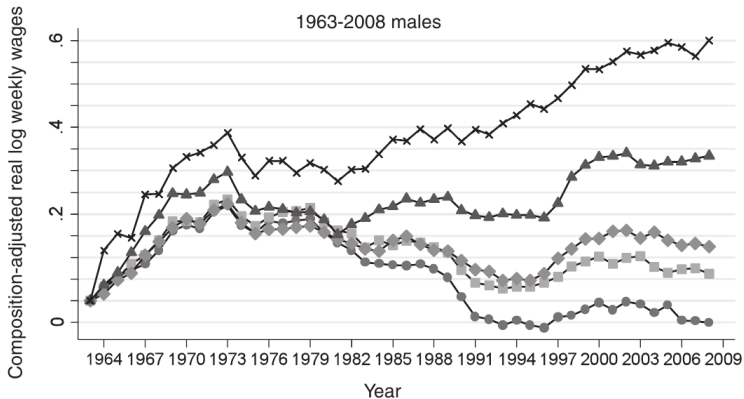


# College/High-School Log Relative Supply

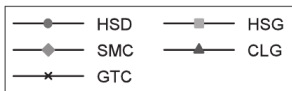




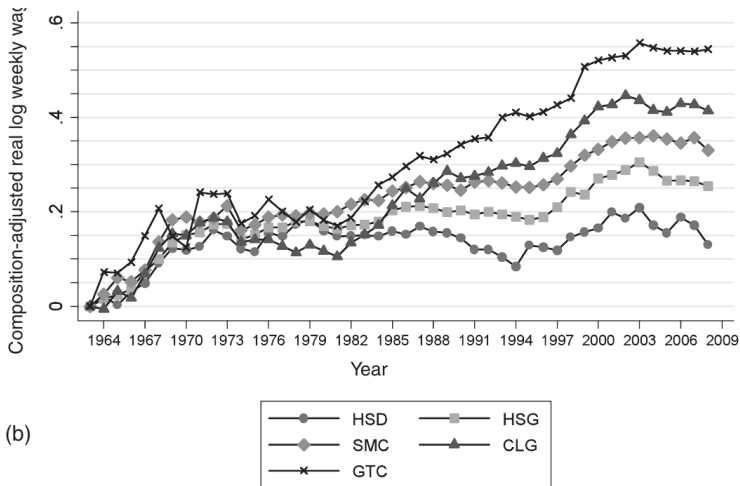
# Real Log Weekly Wages, FTFY Male



(a)

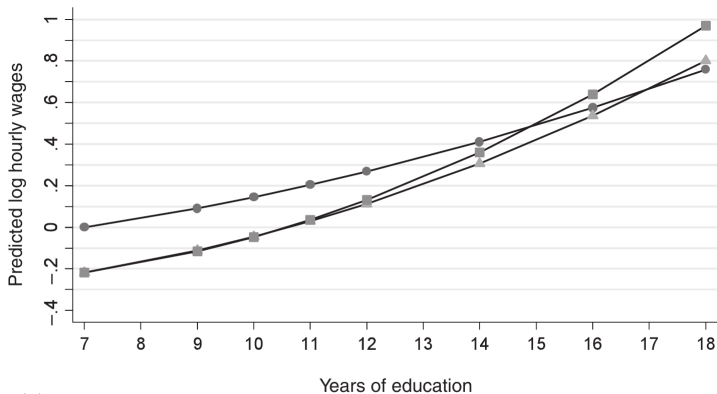


# Real Log Weekly Wages, FTFY Female

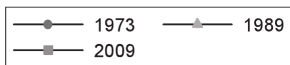


(b)

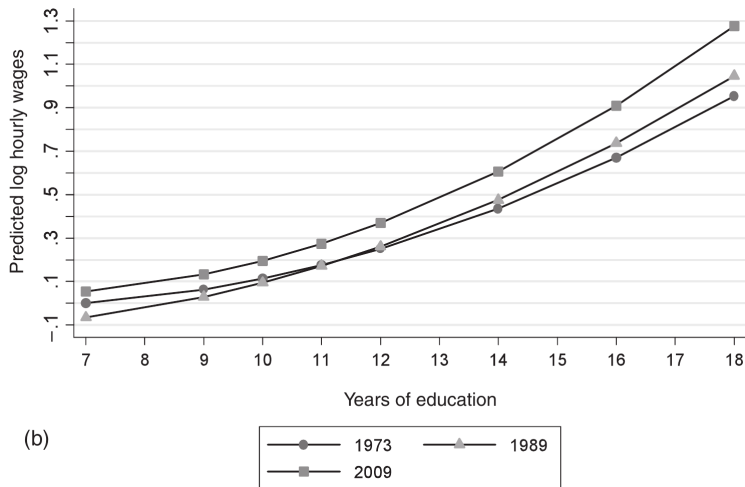
# Predicted Log Hourly Wages by Yrs of Education, Males



(a)



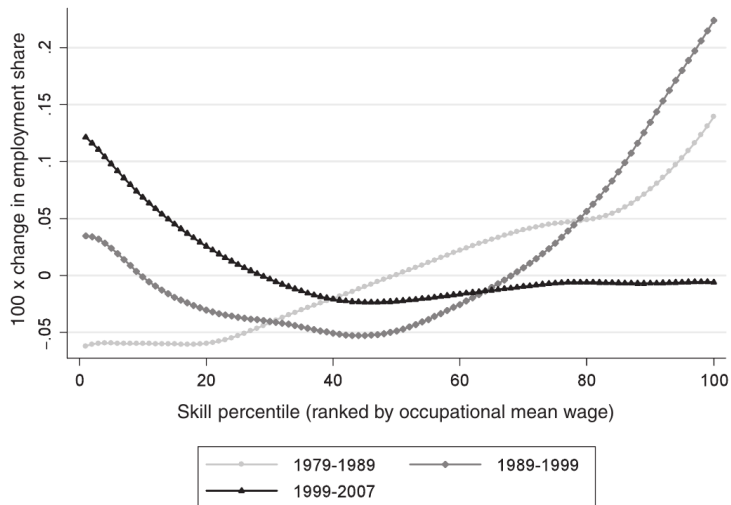
# Predicted Log Hourly Wages by Yrs of Education, Females



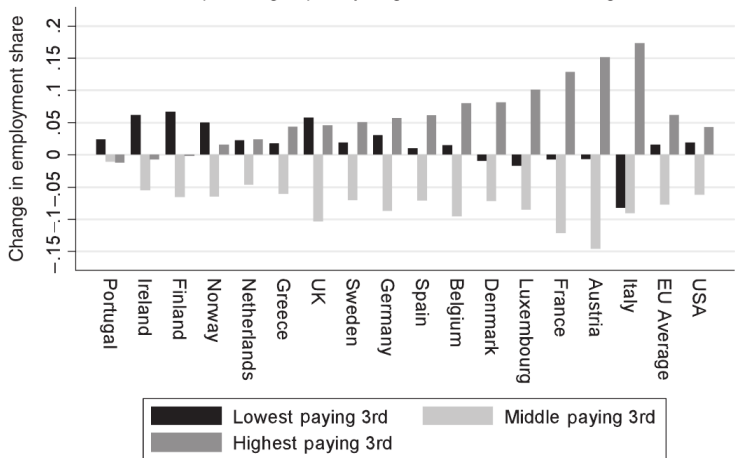
## Task Approach and Job Polarization

(Autor and Acemoglu, 2011, Census, March CPS)

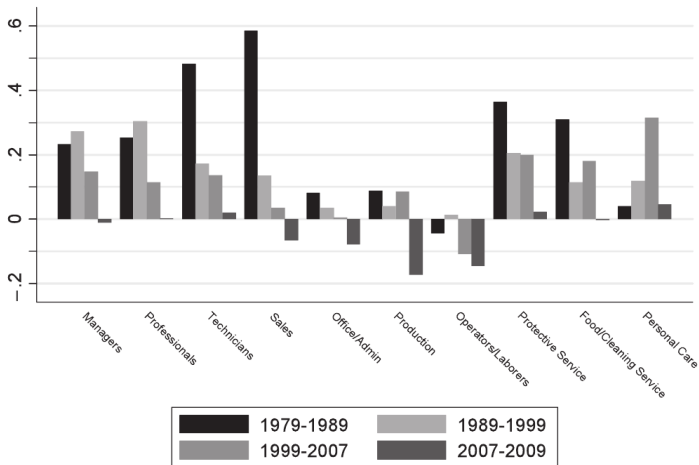
# Changes in Employment by Occupational Skill Percentile



# Change in Employment Shares by Occupation, Europe

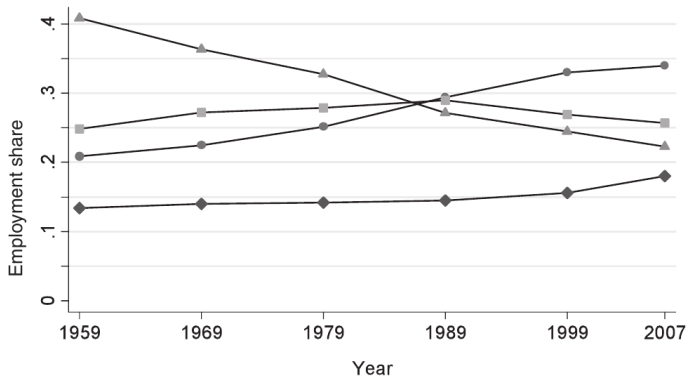


# Percent Change in Employment by Occupation





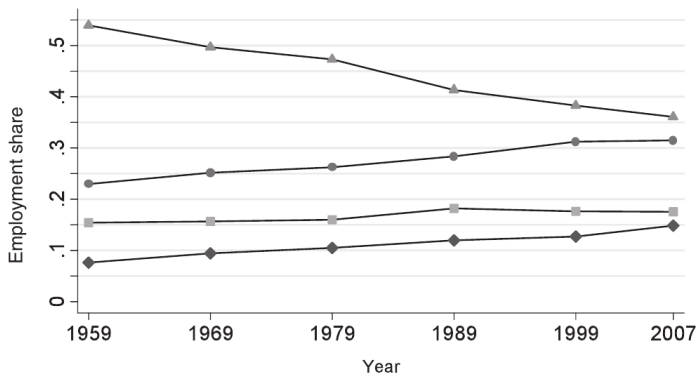
# Employment Shares by Major Occupation Groups, All



(a)



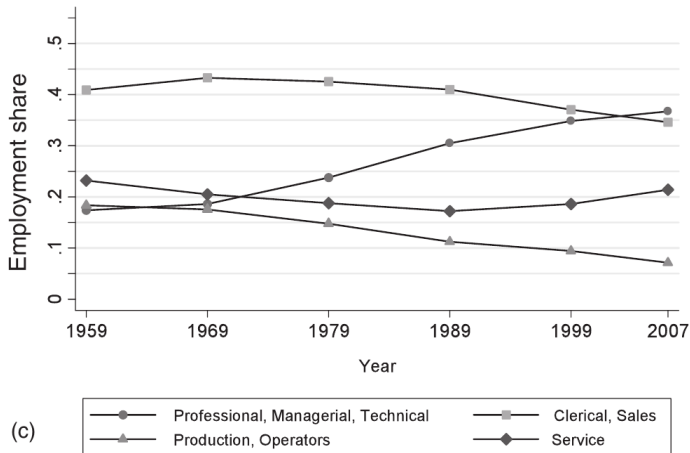
# Employment Shares by Major Occupation Groups, Male



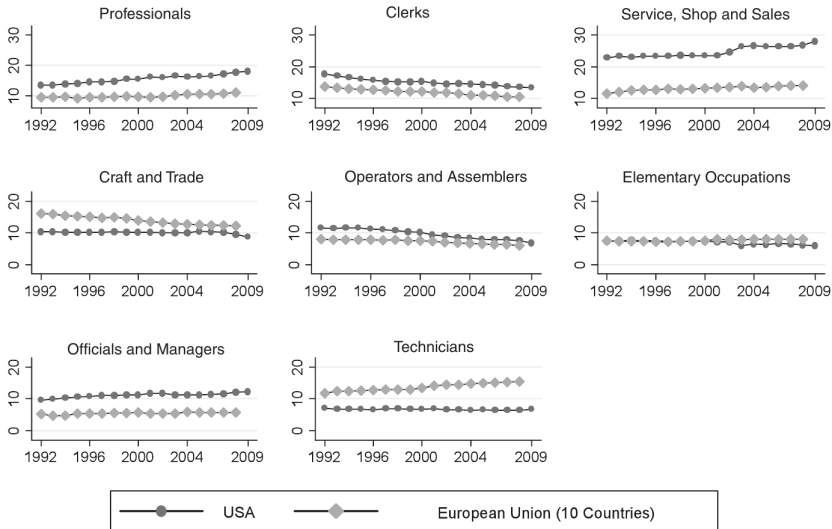
(b)



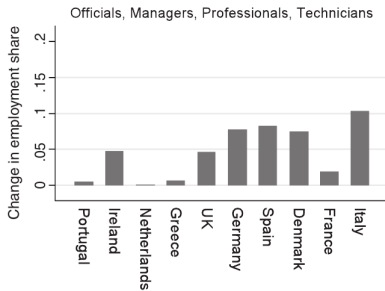
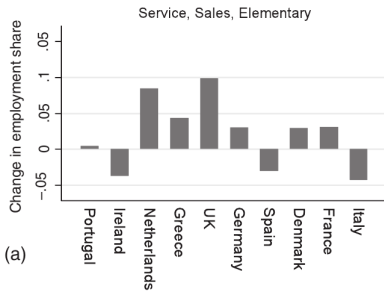
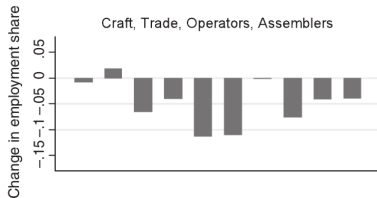
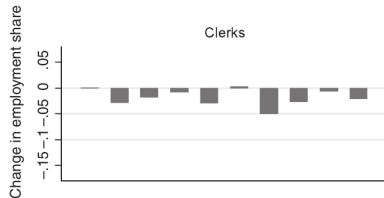
# Employment Shares by Major Occupation Groups, Female



# US and EU Occupational Employment Shares, Age < 40

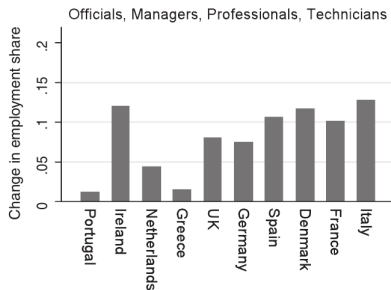
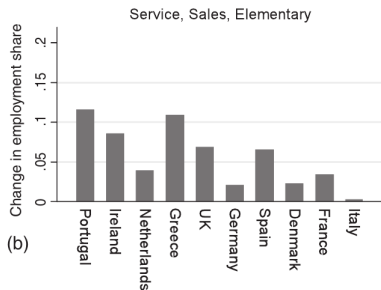
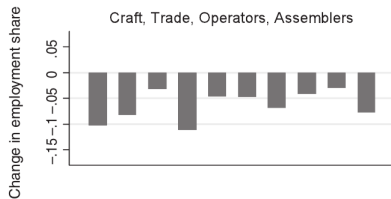
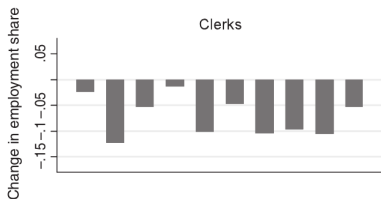


# Change in Emp. Shares, Males (Age<40) by Country



(a)

# Change in Emp. Shares, Females (Age<40) by Country



(g)

Between-Firm

(Autor and Acemoglu, 2011, March CPS)

# Simple Framework to Decompose Earnings

- ▶ Let  $w_t^{i,j}$  the log of real earnings for a worker  $i$  at firm  $j$  in period  $t$ .

- ▶ We can write

$$w_t^{i,j} \equiv \bar{w}_t^j + \left( w_t^{i,j} - \bar{w}_t^j \right),$$

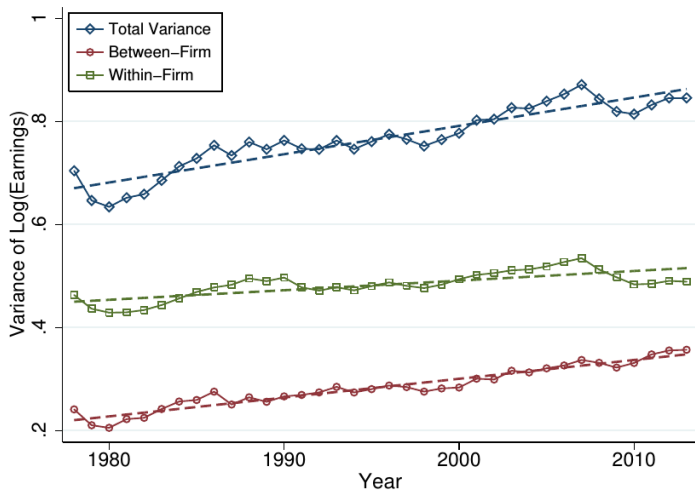
where  $\bar{w}_t^j$  is the average wage at firm  $j$  in period  $t$ .

- ▶ Then, we have

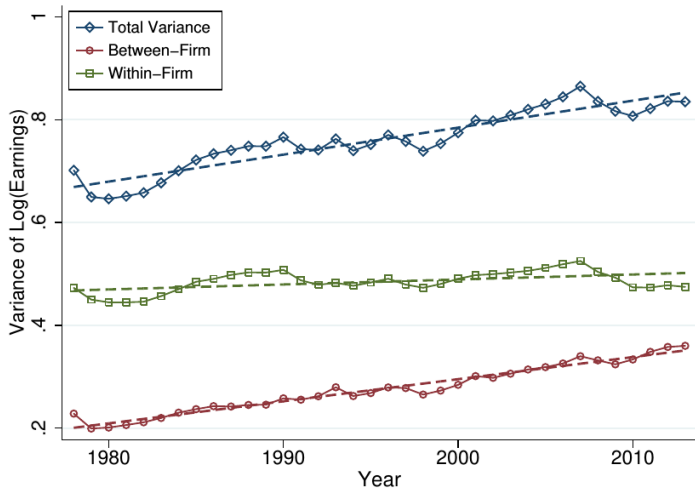
$$\text{Var}_t \left( w_t^{i,j} \right) = \underbrace{\text{Var}_t \left( \bar{w}_t^j \right)}_{\text{Between-firm variance}} + \underbrace{\text{Var}_t \left( w_t^{i,j} - \bar{w}_t^j \right)}_{\text{Within-firm variance}}.$$



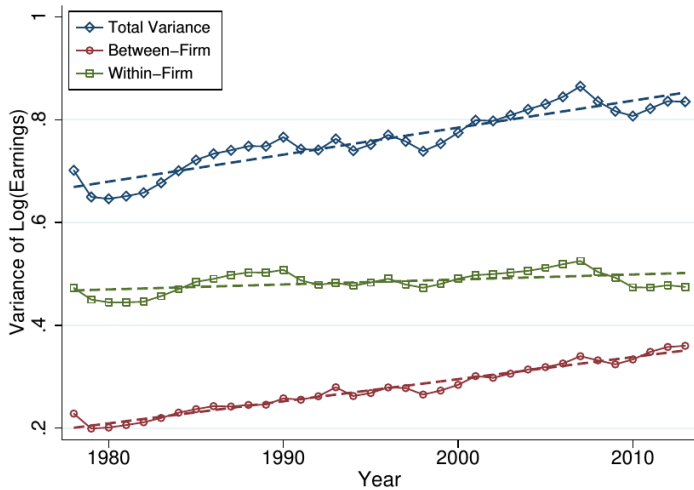
# Variance in Annual Earnings Within and Between Firms



## Firms with 20 to 10,000 Employees



# Firms with 10,000+ Employees



# Conclusion

1. Wage/earnings inequality rose sharply in the U.S. starting in the early 1980s
2. Returns to education also increased sharply despite the increasing supply of college graduates
3. The employment shares of the middle-skilled occupations dropped dramatically
4. The rise of wage/earnings inequality happened between firms