

# The Allocation of Talent and U.S. Economic Growth

Chang-Tai Hsieh

Erik Hurst

Chad Jones

Pete Klenow

April 2014

# Big changes in the occupational distribution

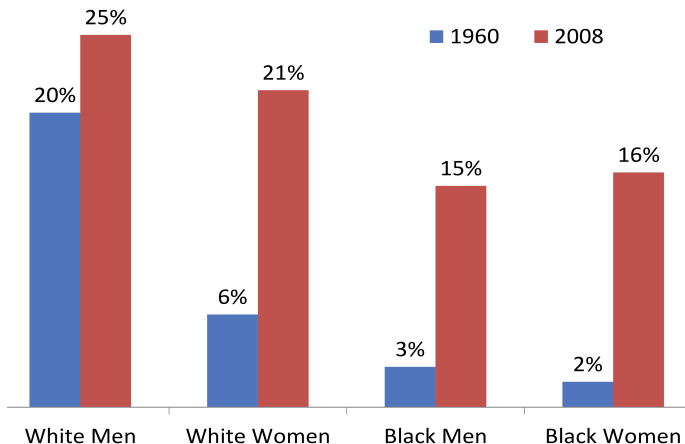
## **White Men in 1960:**

94% of Doctors, 96% of Lawyers, and 86% of Managers

## **White Men in 2008:**

63% of doctors, 61% of lawyers, and 57% of managers

## Share of Each Group in High Skill Occupations



*High-skill occupations* are lawyers, doctors, engineers, scientists, architects, mathematicians and executives/managers.

## Our question

Suppose distribution of talent for each occupation is **identical** for whites, blacks, men and women.

Then:

- Misallocation of talent in both 1960 and 2008.
- But *less* misallocation in 2008 than in 1960.

**How much of productivity growth between 1960 and 2008 was due to the better allocation of talent?**

1. Model
2. Evidence
3. Counterfactuals

# Model

$N$  occupations, one of which is “home”.

Individuals draw talent in each occupation  $\{\epsilon_i\}$ .

Individuals then choose occupation ( $i$ ) and human capital ( $s, e$ ).

Preferences  $U = c^\beta (1 - s)$

Human capital  $h = s^{\phi_i} e^\eta \epsilon$

Consumption  $c = (1 - \tau_w)wh - (1 + \tau_h)e$

# What varies across occupations and/or groups

$w_i$  = the wage per unit of human capital in occupation  $i$  (endogenous)

$\phi_i$  = the elasticity of human capital wrt time invested for occupation  $i$

$\tau_{ig}^w$  = labor market barrier facing group  $g$  in occupation  $i$

$\tau_{ig}^h$  = barrier to building human capital facing group  $g$  for  $i$

# Timing

Individuals draw and observe an  $\epsilon_i$  for each occupation.

They also see  $\phi_i$ ,  $\tau_{ig}^w$ , and  $\tau_{ig}^h$ .

They anticipate  $w_i$ .

Based on these, they choose their occupation, their  $s$ , and their  $e$ .

$w_i$  will be determined in GE (production details later).



# Some Possible Barriers

## Acting like $\tau^w$

- Discrimination in the labor market.

## Acting like $\tau^h$

- Family background.
- Quality of public schools.
- Discrimination in school admissions.

# Identification Problem (currently)

Empirically, we will be able to identify:

$$\tau_{ig} \equiv \frac{(1 + \tau_{ig}^h)^\eta}{1 - \tau_{ig}^w}$$

But not  $\tau_{ig}^w$  and  $\tau_{ig}^h$  separately.

**For now we analyze the composite  $\tau_{ig}$  or one of two polar cases:**

- All differences are from  $\tau_{ig}^h$  barriers to human capital accumulation ( $\tau_{ig}^w = 0$ )
- Or all differences are due to  $\tau_{ig}^w$  labor market barriers ( $\tau_{ig}^h = 0$ ).

# Individual Consumption and Schooling

The solution to an individual's utility maximization problem, given an occupational choice:

$$s_i^* = \frac{1}{1 + \frac{1-\eta}{\beta\phi_i}}$$

$$e_{ig}^*(\epsilon) = \left( \frac{\eta w_i s_i^{\phi_i} \epsilon}{\tau_{ig}} \right)^{\frac{1}{1-\eta}}$$

$$c_{ig}^*(\epsilon) = \bar{\eta} \left( \frac{w_i s_i^{\phi_i} \epsilon}{\tau_{ig}} \right)^{\frac{1}{1-\eta}}$$

$$U(\tau_{ig}, w_i, \epsilon_i) = \bar{\eta}^\beta \left( \frac{w_i s_i^{\phi_i} (1-s_i)^{\frac{1-\eta}{\beta}} \epsilon_i}{\tau_{ig}} \right)^{\frac{\beta}{1-\eta}}$$

# The Distribution of Talent

We assume **Fréchet** for analytical convenience:

$$F_i(\epsilon) = \exp(-T_{ig}\epsilon^{-\theta})$$

- McFadden (1974), Eaton and Kortum (2002)
- $\theta$  governs the dispersion of skills
- $T_{ig}$  scales the supply of talent for an occupation

**Benchmark case:**  $T_{ig} = T_i$  — identical talent distributions

$T_i$  will be observationally equivalent to production technology parameters, so we normalize  $T_i = 1$ .

## Result 1: Occupational Choice

$$U(\tau_{ig}, w_i, \epsilon_i) = \bar{\eta}^\beta \left( \frac{w_i s_i^{\phi_i} (1 - s_i)^{\frac{1-\eta}{\beta}} \epsilon_i}{\tau_{ig}} \right)^{\frac{\beta}{1-\eta}}$$

**Extreme value theory:**  $U(\cdot)$  is Fréchet  $\Rightarrow$  so is  $\max_i U(\cdot)$

Let  $p_{ig}$  denote the fraction of people in group  $g$  that work in occupation  $i$ :

$$p_{ig} = \frac{\tilde{w}_{ig}^\theta}{\sum_{s=1}^N \tilde{w}_{sg}^\theta} \quad \text{where} \quad \tilde{w}_{ig} \equiv \frac{T_{ig}^{1/\theta} w_i s_i^{\phi_i} (1 - s_i)^{\frac{1-\eta}{\beta}}}{\tau_{ig}}.$$

Note:  $\tilde{w}_{ig}$  is the reward to working in an occupation for a person with average talent

## Result 2: Wages and Wage Gaps

Let  $\overline{\text{wage}}_{ig}$  denote the average earnings in occupation  $i$  by group  $g$ :

$$\overline{\text{wage}}_{ig} \equiv \frac{(1 - \tau_{ig}^w) w_i H_{ig}}{q_g p_{ig}} = (1 - s_i)^{-1/\beta} \gamma \bar{\eta} \left( \sum_{s=1}^N \tilde{w}_{sg}^\theta \right)^{\frac{1}{\theta} \cdot \frac{1}{1-\eta}}$$

The wage gap between groups is the **same** across occupations:

$$\frac{\overline{\text{wage}}_{i,women}}{\overline{\text{wage}}_{i,men}} = \left( \frac{\sum_s \tilde{w}_{s,women}^{-\theta}}{\sum_s \tilde{w}_{s,men}^{-\theta}} \right)^{\frac{1}{\theta} \cdot \frac{1}{1-\eta}}$$

- Selection exactly offsets  $\tau_{ig}$  differences across occupations because of the Fréchet assumption
- Higher  $\tau_{ig}$  barriers in one occupation reduce a group's wages proportionately in **all** occupations.

Therefore:

$$\frac{P_{ig}}{P_{i,wm}} = \frac{T_{ig}}{T_{i,wm}} \left( \frac{\tau_{ig}}{\tau_{i,wm}} \right)^{-\theta} \left( \frac{\overline{\text{wage}}_g}{\overline{\text{wage}}_{wm}} \right)^{-\theta(1-\eta)}$$

Misallocation of talent comes from **dispersion** of  $\tau$ 's across occupation-groups.

# Inferring Barriers

$$\frac{\tau_{ig}}{\tau_{i,wm}} = \left( \frac{T_{ig}}{T_{i,wm}} \right)^{\frac{1}{\theta}} \left( \frac{p_{ig}}{p_{i,wm}} \right)^{-\frac{1}{\theta}} \left( \frac{\overline{\text{wage}}_g}{\overline{\text{wage}}_{wm}} \right)^{-(1-\eta)}$$

We infer high  $\tau$  barriers for a group with low average wages.

We infer particularly high barriers when a group is underrepresented in an occupation.

We pin down the *levels* by assuming  $\tau_{i,wm} = 1$ . The results are similar if we instead impose a zero average  $\tau$  in each occupation.



# Aggregates

Human Capital  $H_i = \sum_{g=1}^G \int h_{jgi} dj$

Production  $Y = \left( \sum_{i=1}^I (A_i H_i)^\rho \right)^{1/\rho}$

Expenditure  $Y = \sum_{i=1}^I \sum_{g=1}^G \int (c_{jgi} + e_{jgi}) dj$

# Competitive Equilibrium

1. Given occupations, individuals choose  $c, e, s$  to maximize utility.
2. Each individual chooses the utility-maximizing occupation.
3. A representative firm chooses  $H_i$  to maximize profits:

$$\max_{\{H_i\}} \left( \sum_{i=1}^I (A_i H_i)^\rho \right)^{1/\rho} - \sum_{i=1}^I w_i H_i$$

4. The occupational wage  $w_i$  clears each labor market:

$$H_i = \sum_{g=1}^G \int h_{jgi} dj$$

5. Aggregate output is given by the production function.

## Solution in a Special Case

- $\rho = 1$  so that  $w_i = A_i$
- 2 groups, men and women
- $\phi_i = 0$  (no schooling time),  $\tau^h = 0$
- $A$  and  $\tau^w$  are joint lognormal

**Then:**

$$\overline{wage}_m = \left( \sum_{i=1}^N A_i^\theta \right)^{\frac{1}{\theta} \cdot \frac{1}{1-\eta}}$$

$$\ln \frac{\overline{wage}_w}{\overline{wage}_m} = \frac{1}{1-\eta} \left( \ln(1 - \bar{\tau}^w) - \frac{1}{2}(\theta - 1)\text{Var}(\ln(1 - \tau_i^w)) \right).$$

1. Model

2. Evidence

3. Counterfactuals

- U.S. Census for 1960, 1970, 1980, 1990, and 2000
- American Community Survey for 2006-2008
- 67 consistent occupations, one of which is the “home” sector.
- Look at full-time and part-time workers, hourly wages.
- Prime-age workers (age 25-55).

# Examples of Baseline Occupations

## **Health Diagnosing Occupations**

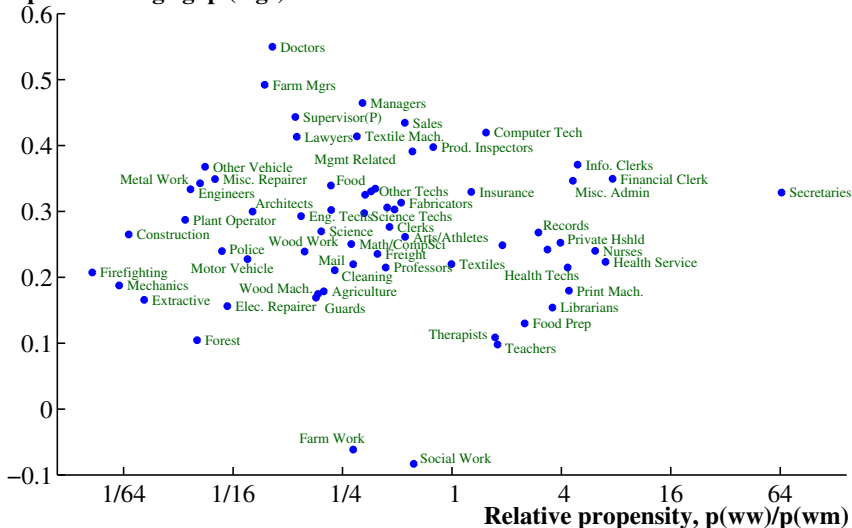
- Physicians
- Dentists
- Veterinarians
- Optometrists
- Podiatrists
- Health diagnosing practitioners, n.e.c.

## **Health Assessment and Treating Occupations**

- Registered nurses
- Pharmacists
- Dietitians

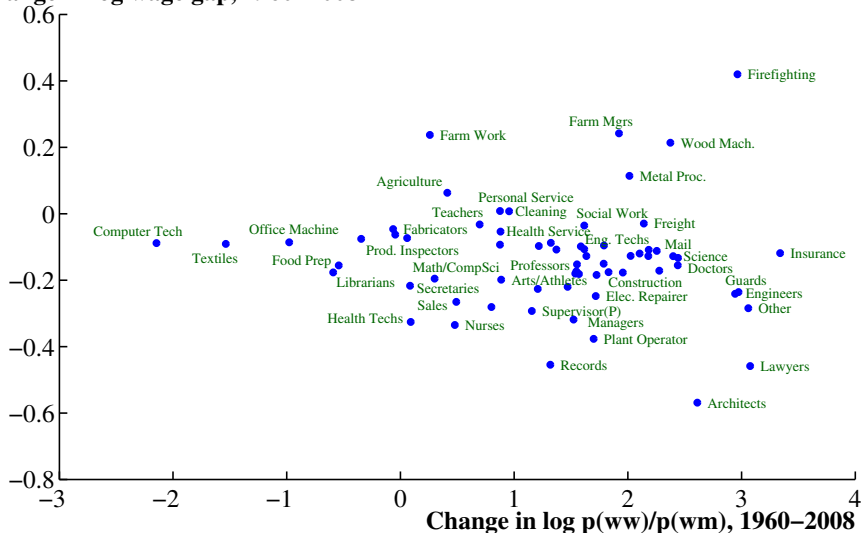
# Occupational Wage Gaps for White Women in 1980

## Occupational wage gap (logs)



# Change in Wage Gaps for White Women, 1960–2008

Change in log wage gap, 1960–2008





## Test of Model Implications: Changes by Schooling

Occupational Similarity to White Men	1960	2008	1960–2008
High-Educated White Women	0.38	0.59	0.21
Low-Educated White Women	0.40	0.46	0.06

Wage Gap vs. White Men	1960	2008	1960–2008
High-Educated White Women	-0.50	-0.24	-0.26
Low-Educated White Women	-0.56	-0.27	-0.29

## Estimating $\theta(1 - \eta)$

$$\frac{\tau_{ig}}{\tau_{i,wm}} = \left( \frac{T_{ig}}{T_{i,wm}} \right)^{\frac{1}{\theta}} \left( \frac{p_{ig}}{p_{i,wm}} \right)^{-\frac{1}{\theta}} \left( \frac{\overline{\text{wage}}_g}{\overline{\text{wage}}_{wm}} \right)^{-(1-\eta)}$$

Under Fréchet, wages within an occupation-group satisfy

$$\frac{\text{Variance}}{\text{Mean}^2} = \frac{\Gamma(1 - \frac{2}{\theta(1-\eta)})}{\left( \Gamma(1 - \frac{1}{\theta(1-\eta)}) \right)^2} - 1.$$

- Assume  $\eta = 1/4$  for baseline (midway between 0 and 1/2).
- Then use this equation to estimate  $\theta$ .
- Attempt to control for “absolute advantage” as well (next slide).

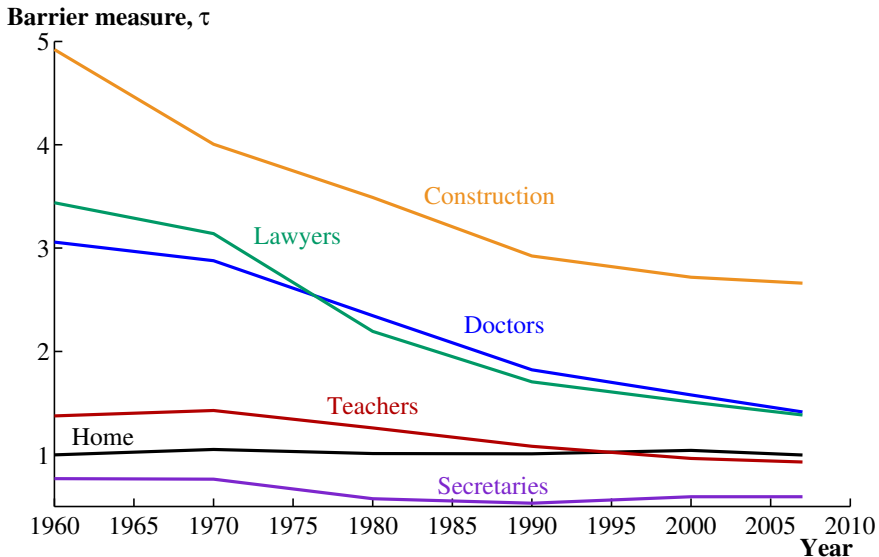
## Estimating $\theta(1 - \eta)$ (continued)

<b>Adjustments to Wages</b>	<b>Estimates of <math>\theta(1 - \eta)</math></b>
Base controls	3.11
Base controls + Adjustments	<b>3.44</b>
Wage variation due to absolute advantage:	
25%	<b>3.44</b>
50%	4.16
75%	5.61
90%	8.41

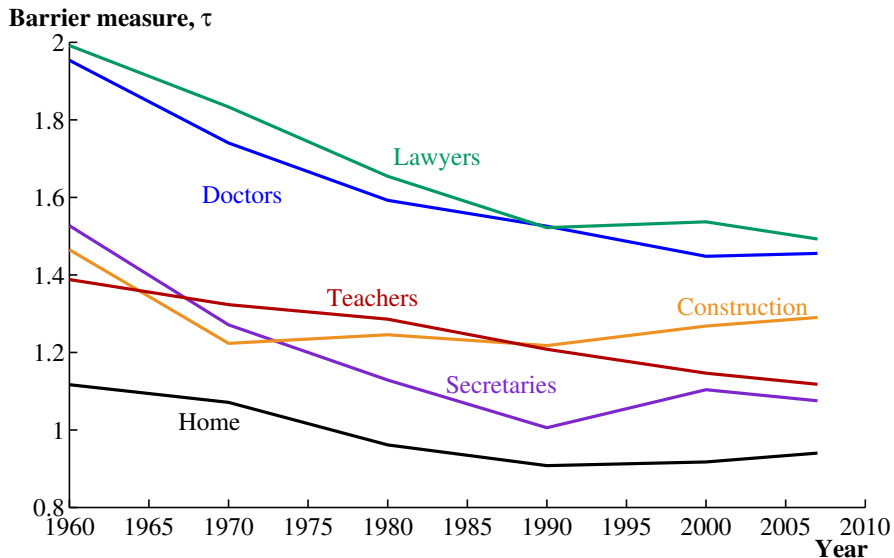
**Base controls** = potential experience, hours worked, occupation-group dummies

**Adjustments** = transitory wages, AFQT score, education

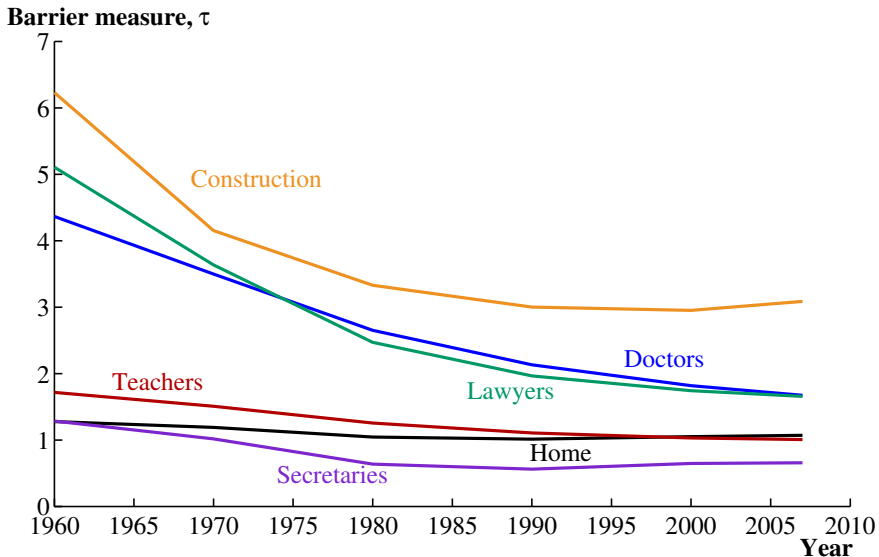
# Estimated Barriers ( $\tau_{ig}$ ) for White Women



# Estimated Barriers ( $\tau_{ig}$ ) for Black Men

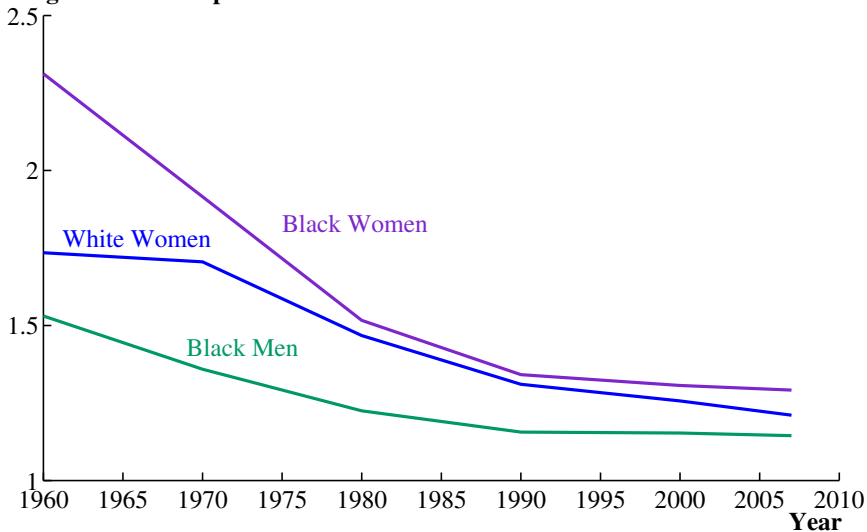


# Estimated Barriers ( $\tau_{ig}$ ) for Black Women



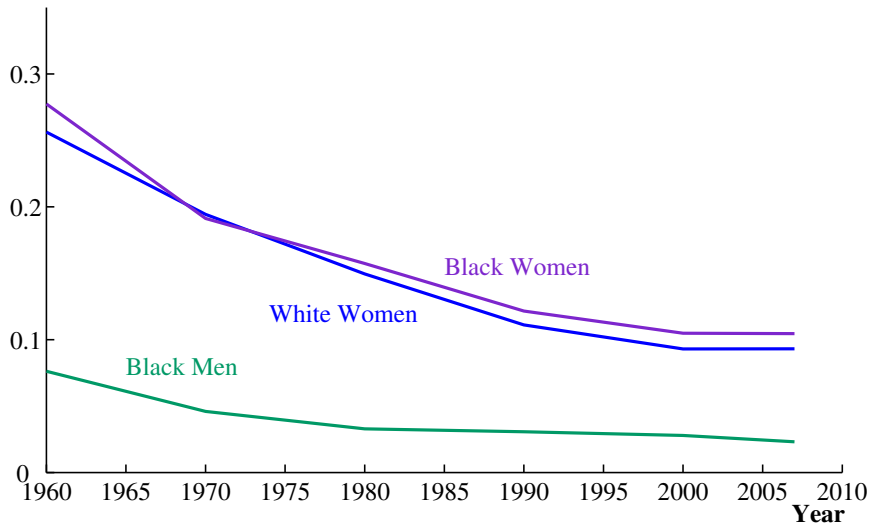
# Average Values of $\tau_{ig}$ over Time

Average  $\tau$  across occupations



# Variance of $\log \tau_{ig}$ over Time

Variance of  $\log \tau$





# Driving Forces

Allow  $A_i$ ,  $\phi_i$ ,  $\tau_{ig}$ , and population to vary across time to fit observed employment and wages by occupation and group in each year.

$A_i$ : Occupation-specific productivity

Average size of an occupation

Average wage growth

$\phi_i$ : Occupation-specific return to education

Wage differences across occupations

$\tau_{ig}$ : Occupational sorting

Trends in  $A_i$  could be skill-biased and market-occupation-biased.

# Baseline Parameter Values

---

Parameter	Value	Target
$\theta(1 - \eta)$	3.44	wage dispersion within occupation-groups
$\eta$	0.25	midpoint of range from 0 to 0.5
$\beta$	0.693	Mincerian return across occupations
$\rho$	2/3	elasticity of substitution b/w occupations of 3
$\phi_{min}$	by year	schooling in the lowest-wage occupation

---

1. Model

2. Evidence

3. Counterfactuals

## What share of labor productivity growth is explained by changing barriers?

	$\tau^h$ case	$\tau^w$ case
Frictions in all occupations	20.4%	15.9%
No frictions in “brawny” occupations	18.9%	14.1%
No frictions in 2008	20.4%	12.3%
Market sector only	26.9%	23.5%

# Potential Remaining Productivity Gains

	$\tau^h$ case	$\tau^w$ case
Cumulative gain, 1960–2008	15.2%	11.3%
Remaining gain from zero barriers	14.3%	10.0%

# Sources of productivity gains in the model

## **Better allocation of human capital investment:**

- White men over-invested in 1960
- Women, blacks under-invested in 1960
- Less so in 2008

## **Better allocation of talent to occupations:**

- Dispersion in  $\tau$ 's for women, blacks in 1960
- Less in 2008

# Back-of-the-envelope calculation

## The calculation:

- Take wages of white men as exogenous.
- Growth from faster wage growth for women and blacks?

**Answer = 12.8%**

Versus 20.4% gains in our  $\tau^h$  case, 15.9% in our  $\tau^w$  case.

## Why do these figures differ?

- We are isolating the contribution of  $\tau$ 's.
- We take into account GE effects.

# Gains when changing only the dispersion of ability

Value of $\theta(1 - \eta)$	$\tau^h$ case	$\tau^w$ case
3.44	20.4%	15.9%
4.16	18.6%	15.1%
5.61	9.5%	8.0%
8.41	8.4%	3.9%



# Summary of Other Findings

## **Changing barriers also led to:**

- 40+ percent of WW, BM, BW wage growth
- A 6 percent reduction in WM wages
- Essentially all of the narrowing of wage gaps
- 70+ percent of the rise in female LF participation
- Substantial wage convergence between North and South

**Extensive range of robustness checks in paper...**

## **Distinguishing between $\tau^h$ and $\tau^w$ empirically:**

- Assume  $\tau^h$  is a cohort effect,  $\tau^w$  a time effect.
- Early finding: mostly  $\tau^h$  for white women, a mix for blacks.

## **Absolute advantage correlated with comparative advantage:**

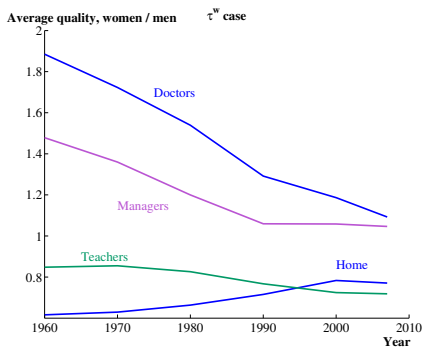
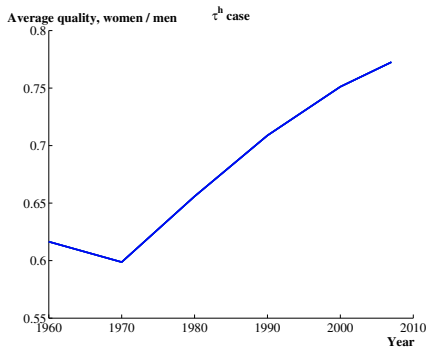
- Talented 1960 women went into teaching, nursing, home sector?
- As barriers fell, lost talented teachers, child-raisers?
- Could explain Mulligan and Rubinstein (2008) facts.

## **Separate paper:**

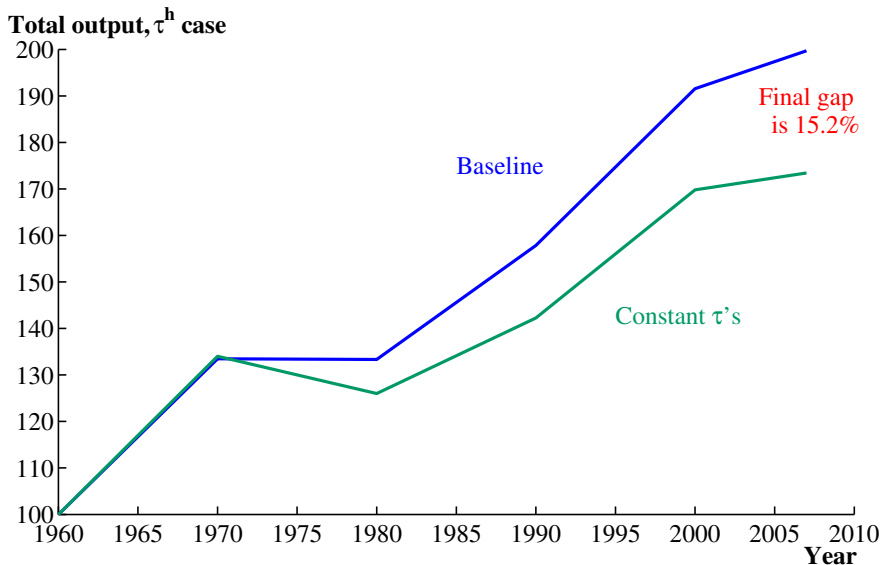
Rising inequality from misallocation of human capital investment?

## **Extra Slides**

# Average quality of white women vs. white men

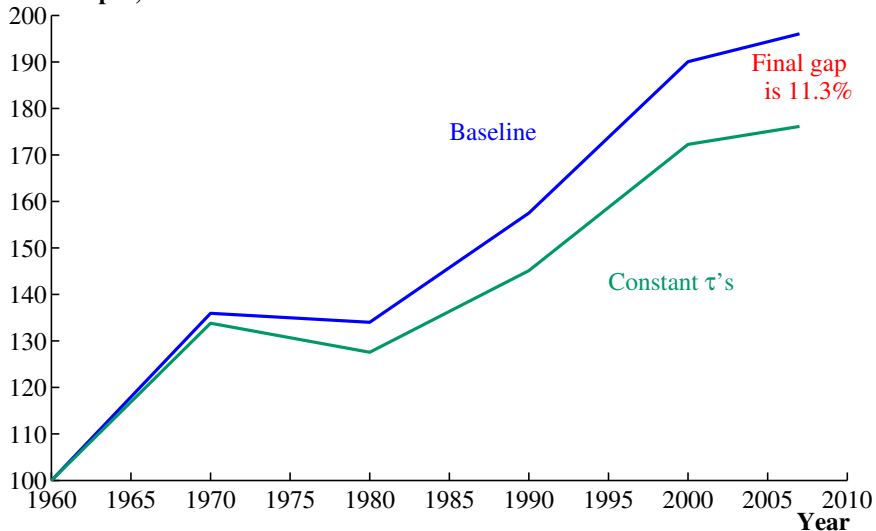


# Counterfactuals in the $\tau^h$ Case



# Counterfactuals in the $\tau^w$ Case

Total output,  $\tau^w$  case



# Sensitivity of Gains to the Wage Gaps

---

	$\tau^h$ case	$\tau^w$ case
Baseline	20.4%	15.9%
Counterfactual: wage gaps halved	12.5%	13.7%
Counterfactual: zero wage gaps	2.9%	11.8%

---

## Wage Growth Due to Changing $\tau$ 's

---

	Actual Growth	Due to $\tau^h$ 's	Due to $\tau^w$ 's
White men	77.0%	-5.8%	-7.1%
White women	126.3%	41.9%	43.0%
Black men	143.0%	44.6%	44.3%
Black women	198.1%	58.8%	59.5%

---

Note:  $\tau$  columns are % of growth explained.



# Decomposing the Gains: Dispersion vs. Mean Barriers

	$\tau^h$ case	$\tau^w$ case
1960 Eliminating Dispersion	22.2%	14.9%
1960 Eliminating Mean and Variance	26.9%	18.6%
2008 Eliminating Dispersion	16.6%	7.8%
2008 Eliminating Mean and Variance	14.3%	10.0%

# Robustness: $\tau^h$ case

	Baseline				
	$\rho = 2/3$	$\rho = -90$	$\rho = -1$	$\rho = 1/3$	$\rho = .95$
Changing $\rho$	20.4%	19.7%	19.9%	20.2%	21.0%
	3.44	4.16	5.61	8.41	
Changing $\theta$	20.4%	20.7%	21.0%	21.3%	
	$\eta = 1/4$	$\eta = 0.01$	$\eta = .05$	$\eta = .1$	$\eta = .5$
Changing $\eta$	20.4%	20.5%	20.5%	20.5%	20.3%

Note: Entries are % of output growth explained.

# Robustness: $\tau^w$ case

	Baseline				
	$\rho = 2/3$	$\rho = -90$	$\rho = -1$	$\rho = 1/3$	$\rho = .95$
Changing $\rho$	15.9%	12.3%	13.3%	14.7%	18.4%
	3.44	4.16	5.61	8.41	
Changing $\theta$	15.9%	14.6%	12.9%	11.2%	
	$\eta = 1/4$	$\eta = 0$	$\eta = .05$	$\eta = .1$	$\eta = .5$
Changing $\eta$	15.9%	13.9%	14.4%	14.8%	17.5%

Note: Entries are % of output growth explained.

## Gains are not sensitive to:

- More detailed occupations (331 for 1980 onward)
- A broader set of occupations (20)
- Weight on consumption vs. time in utility ( $\beta$ )

# Female Labor Force Participation

Data

---

<i>Women's LF participation</i>	1960 = 0.329	2008 = 0.692
<i>Change, 1960 – 2008</i>		0.364

Model

---

Due to changing $\tau^h$ 's	0.235
Due to changing $\tau^w$ 's	0.262

---

# Education Predictions, $\tau^h$ case

	Actual 1960	Actual 2008	Actual Change	Change vs. WM	Due to $\tau$ 's
White men	11.11	13.47	2.35		
White women	10.98	13.75	2.77	0.41	0.63
Black men	8.56	12.73	4.17	1.81	0.65
Black women	9.24	13.15	3.90	1.55	1.17

Note: Entries are years of schooling attainment.

## Gains from white women vs. blacks, $\tau^h$ case

	1960–1980	1980–2008	1960–2008
All groups	19.7%	20.9%	20.4%
White women	11.3%	18.2%	15.3%
Black men	3.3%	0.9%	1.9%
Black women	5.1%	1.9%	3.2%

Note: Entries are % of growth explained. “All” includes white men.

## North-South wage convergence, $\tau^h$ case

---

	1960–1980	1980–2008	1960–2008
Actual wage convergence	20.7%	-16.5%	10.0%
Due to all $\tau$ 's changing	4.9%	1.5%	6.9%
Due to black $\tau$ 's changing	3.6%	1.9%	5.6%

---

Note: Entries are percentage points. “North” is the Northeast.