

STOCHASTIC SORTING

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NBER Summer Institute

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MOTIVATION

- 2-sided matching: labor as an inspection good (Nelson 1970)
- Matching problem (Becker 1973) with stochastic types:
 1. match \rightarrow ex ante characteristics x, y
 2. output \rightarrow ex post realizations ω, σ
- Realistic + can confront model with data:
 1. Attributes change
 2. Account for mismatch
 3. Noise is part of model

EXAMPLES

$$x \rightarrow \omega$$

$$y \rightarrow \sigma$$

	x, y	ω, σ
Marriage	x : man's education y : woman's education	ω : income σ : income

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$$y \rightarrow \sigma$$

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Marriage	x : man's education y : woman's education	ω : income σ : income
Job Market	x : MBA degree y : job level/position	ω : worker productivity σ : realized demand/technology

EXAMPLES

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$$y \rightarrow \sigma$$

	x, y	ω, σ
Marriage	x : man's education y : woman's education	ω : income σ : income
Job Market	x : MBA degree y : job level/position	ω : worker productivity σ : realized demand/technology
Executives	x : past experience y : initial market value	ω : CEO performance σ : stock price change

APPLICATIONS

1. Mismatched CEOs

- Ex post randomness in match
- mismatch + selection:
 - many CEOs are the wrong (wo)man for the job

⇒ Technology: **output** and **distributions**

2. Household Income Inequality: **stochastic** vs. **marital** sorting

RELATED WORK

Mismatch: confronting matching models with reality

- Search Frictions: Shimer and Smith (2000), Cheremukhin, Restrepo, Tutino (2016)
- Learning: Anderson-Smith (2011)
- Matching under uncertainty (Het. pref.): Chiappori-Reny (2005), Legros-Newman (2007) (no mismatch); Chade (2006)
- Unobserved heterogeneity + multidimensional types: Choo-Siow (2006) Galichon-Salanie (2011), Lindenlaub (2012)

MAIN FEATURES OF MODEL

- Endowments when matching: heterogeneous and stochastic
- Who matches with whom?
 - Matching based on ex ante attributes
 - ⇒ Ex ante: no mismatch (Becker)
 - Match value and payoff depend on ex post realization of types
 - ⇒ Ex post: mismatch
- Key assumption: no rematching (same logic if cost)

THE MODEL SETUP

GENERAL FRAMEWORK

- Agents

Workers: $x \rightarrow \omega \sim F(\omega|x)$

Firms: $y \rightarrow \sigma \sim G(\sigma|y)$

→ joint distribution $K(\omega, \sigma|x, y)$

- Output:

$$q(\omega, \sigma)$$

- Competitive equilibrium/stability/efficient matching $\mu(x)$
- Remark:
 - Special Case: Independence $K(\omega, \sigma|x, y) = F(\omega|x)G(\sigma|y)$
 - Assume continuous variables with K, F, G, q smooth

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TRANSFERABLE UTILITY (TU)

- The expected surplus of a match between a type x and y :

$$V(x, y) = \int_{\underline{\omega}}^{\bar{\omega}} \int_{\underline{\sigma}}^{\bar{\sigma}} q(\omega, \sigma) k(\omega, \sigma | x, y) d\omega d\sigma$$

where k is the density of K

- Determinants of equilibrium allocation:
 1. Complementarity of match output $q(\omega, \sigma)$
 2. Distributions $K(\omega, \sigma | x, y) \rightarrow$ stochastic dominance

TRANSFERABLE UTILITY (TU)

PROPOSITION (SORTING)

Optimal sorting patterns are as follows:

1. *PAM if K is spm (sbm) in (x, y) for each (ω, σ) and q is spm (sbm) in (ω, σ) ;*
2. *PAM if $\int_{\underline{\omega}}^{\omega} \int_{\underline{\sigma}}^{\sigma} K(s, t|x, y) d\omega d\sigma$ is spm (sbm) in (x, y) (and marginals independent), if $q_{\omega\sigma}$ is spm (sbm) in (ω, σ) ;*

The conditions on q are also necessary if it is to hold for all K .

→ Condition on q is necessary if result to hold for all K

→ Proof: applying integration by parts iteratively

sketch proof

TRANSFERABLE UTILITY (TU)

PROPOSITION (SORTING)

Optimal sorting patterns are as follows:

1. **NAM** if K is *spm* (*sbm*) in (x, y) for each (ω, σ) and q is **sbm** (**spm**) in (ω, σ) ;
2. **NAM** if $\int_{\underline{\omega}}^{\omega} \int_{\underline{\sigma}}^{\sigma} K(s, t|x, y) d\omega d\sigma$ is *spm* (*sbm*) in (x, y) (and marginals independent), if $q_{\omega\sigma}$ is **sbm** (**spm**) in (ω, σ) ;

The conditions on q are also necessary if it is to hold for all K .

→ Condition on q is necessary if result to hold for all K

→ Proof: applying integration by parts iteratively

sketch proof

TRANSFERABLE UTILITY (TU)

- Special case: cond. independence: $K = F(\omega|x)G(\sigma|y)$ and FOSD ($F_x < 0, G_y < 0$)
 - If F and G degenerate, then we recover Becker

TRANSFERABLE UTILITY (TU)

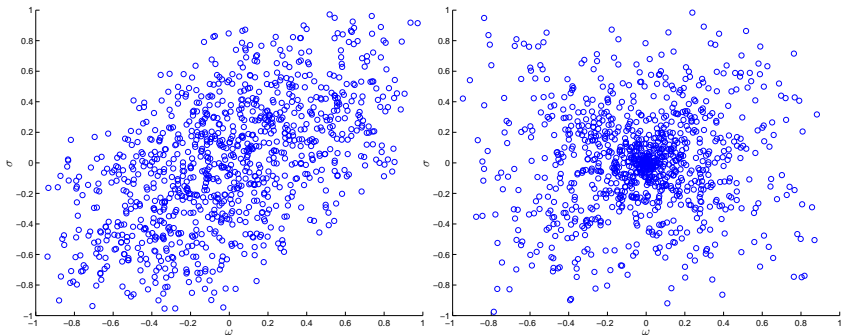
- Special case: cond. independence: $K = F(\omega|x)G(\sigma|y)$ and FOSD ($F_x < 0, G_y < 0$)
 - If F and G degenerate, then we recover Becker
- Selection from matching: even if no correlation in K , there is correlation in observed outcomes across matches of ω, σ

$$\begin{aligned}\text{Cov}[\omega, \sigma] &= \int_{\underline{\omega}}^{\bar{\omega}} \int_{\underline{\sigma}}^{\bar{\sigma}} \omega \sigma k(\omega, \sigma) d\omega d\sigma - \int_{\underline{\omega}}^{\bar{\omega}} \omega k(\omega) d\omega \int_{\underline{\sigma}}^{\bar{\sigma}} \sigma k(\sigma) d\sigma \\ &= \int_{\underline{x}}^{\bar{x}} \mathbb{E}[\omega|x] \mathbb{E}[\sigma|\mu(x)] d\Gamma(x) - \int_{\underline{x}}^{\bar{x}} \mathbb{E}[\omega|x] d\Gamma(x) \int_{\underline{x}}^{\bar{x}} \mathbb{E}[\sigma|\mu(x)] d\Gamma(x) \\ &= \text{Cov}[\mathbb{E}[\omega|x], \mathbb{E}[\sigma|y]]\end{aligned}$$

→ exploit this in decomposition married household inequality

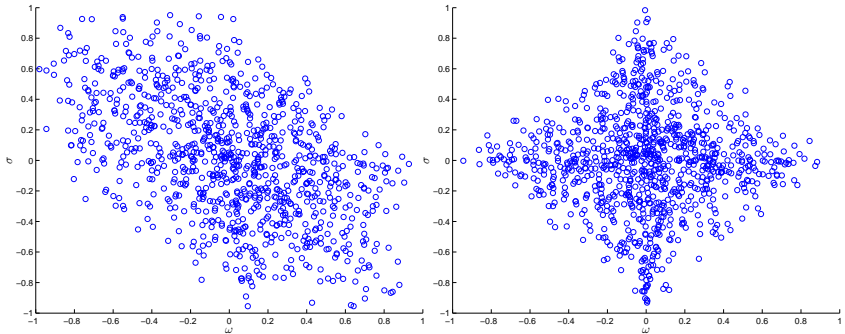
TRANSFERABLE UTILITY (TU)

EXAMPLES OF OBSERVED (ω, σ) – PAM: FOSD vs. MPS



TRANSFERABLE UTILITY (TU)

EXAMPLES OF OBSERVED (ω, σ) – NAM: FOSD vs. MPS



► Technology and Distribution

TRANSFERABLE UTILITY (TU)

EXAMPLE: SUPERMODULARITY OF $q(\omega, \sigma)$ IS NOT SUFFICIENT

- Theorem 1: sorting depends on q and distributions
- Technology:
 - $q = \omega\sigma^2$ supermodular
 - Distributions: F, G conditionally independent:
 - $F_x < 0$ (FOSD)
 - $G = \mathcal{N}(\mu(y), s^2(y))$ with $\mu'(y) > 0$ and $s'(y) < 0$
- Then:

$$\begin{aligned}V(x, y) &= \mathbb{E}[\omega|x]\mathbb{E}[\sigma^2|y] \\ &= \mathbb{E}[\omega|x] (s^2(y) + \mu^2(y))\end{aligned}$$

- The cross partial is:

$$V_{xy} = \mathbb{E}_x[\omega|x] \left(\frac{ds^2}{dy} + \frac{d\mu}{dy} 2\mu \right)$$

$$\therefore V_{xy} < 0 \iff \frac{ds^2}{dy} < -\frac{d\mu}{dy} 2\mu: \text{NAM with } q \text{ supermodular}$$

TRANSFERABLE UTILITY (TU)

SOME OBSERVATIONS

- TU: simple and tractable
- But: ex post payoffs not pinned down (\exists continuum of splits)
- Most applications: information on ex post payoffs
- Non-linear preferences: pins down ex post payoffs
 1. Risk Sharing
 2. Contracting under moral hazard

NON TRANSFERABLE UTILITY (NTU)

RISK SHARING

- Stochastic characteristics \Rightarrow Uncertainty \Rightarrow Risk sharing

$$\begin{aligned}\Phi(x, y, v) &= \max_{c_x, c_y} \int_{\underline{\omega}}^{\bar{\omega}} \int_{\underline{\sigma}}^{\bar{\sigma}} u(c_y(\omega, \sigma)) k(\omega, \sigma | x, y) d\omega d\sigma \\ \text{s.t.} \quad &c_x(\omega, \sigma) + c_y(\omega, \sigma) = q(\omega, \sigma) \quad \forall (\omega, \sigma) \\ &\int_{\underline{\omega}}^{\bar{\omega}} \int_{\underline{\sigma}}^{\bar{\sigma}} u(c_x(\omega, \sigma)) k(\omega, \sigma | x, y) d\omega d\sigma \geq v\end{aligned}$$

- Pins down consumption and thus ex post payoffs

NON TRANSFERABLE UTILITY (NTU)

RISK SHARING

- NTU matching problem \Rightarrow Legros and Newman (2007)
- PAM (NAM) \Leftrightarrow Generalized Increasing (Decr.) Differences
- Differential version of their condition (Spence-Mirrlees):
 - PAM if and only if

$$\Phi_{xy} > \frac{\Phi_x}{\Phi_v} \Phi_{vy}$$

- Focus on $K(\omega, \sigma|x, y) = F(\omega|x)G(\sigma|y)$ and FOSD

NON TRANSFERABLE UTILITY (NTU)

RISK SHARING: OVERVIEW MAIN RESULTS

1. Sorting pattern only depends on q , not on distributions
 - if one side is risk neutral (e.g. firm); or

→ PAM if $q_{\omega\sigma} > 0$

NON TRANSFERABLE UTILITY (NTU)

RISK SHARING: OVERVIEW MAIN RESULTS

2. If u is HARA (CRRA, log, CARA, quadratic,...)

→ PAM if $V = \int \int \hat{q}_{\omega,\sigma} k d\omega d\sigma$ is spm where \hat{q} is a transformation of q : apply Proposition TU

→ FOSD and independence, we verify conditions on \hat{q}

CRRA $u = \frac{c^\alpha}{\alpha}$ then

$$\hat{q} = \frac{q^\alpha}{\alpha(1-\alpha)^{1-\alpha}}$$

α -Root-SPM: can have NAM even if $q_{\omega\sigma} > 0$

LOG $\alpha = 0$: \hat{q} spm or q log-spm

CARA : $\lim \alpha \rightarrow 1$. If $q_{\omega\sigma} = 0$ then matching is independent of distribution (as in one side risk neutral)

MATCHING WITH AGENCY

- Executives match with firms
- NTU \Rightarrow pins down ex post payoffs
- Matching + moral hazard problem
- Variation Holmström-Milgrom linear contracting model
- Objective: sorting due to?
 - match value
 - distribution
- Key assumptions: No rematching, no separation

HOLMSTRÖM-MILGROM WITH MATCHING

- Large number of risk averse CEOs, risk neutral firms
- CEO-firm pair (x, y) match. Timing:
 1. Firm offers output-contingent (q) contract
 2. CEO type ω realized (public); CEO chooses effort e
 3. Firm type σ realized (not observed) \rightarrow output q observed
 4. Payments as specified in the contract
- Technology: output and distributions

$$q = \omega \cdot (e + \sigma)$$

$$\omega \sim \mathcal{LN}(k(x), u(x)^2)$$

$$\sigma \sim \mathcal{N}(t(y), s(y)^2)$$

- Linear contracts (α, β) : $w(q, \omega) = \beta(\omega) + \alpha(\omega)q$
- CEO: CARA preferences $-e^{-r(w - \frac{e^2}{2})}$; Reservation wage $a(x)$

OPTIMAL CONTRACTING PROBLEM

- Principal's problem is (where β, α, e depend on ω):

$$\begin{aligned} \max_{\beta, \alpha, e} \quad & \int (\mathbb{E}[q|e] - (\beta + \alpha \mathbb{E}[q|e])) dF(\omega|x) \\ \text{s.t.} \quad & \int \left(\mathbb{E} \left[-e^{-r(\beta + \alpha q - \frac{e^2}{2})} \right] \right) dF(\omega|x) \geq -e^{-ra} \quad (\mathbf{PC}) \\ & e \in \arg \max_{\hat{e}} \int -e^{-r(\beta + \alpha q - \frac{\hat{e}^2}{2})} dG(\sigma|y), \forall \omega \quad (\mathbf{IC}) \end{aligned}$$

where $q = q(\omega, \sigma, y), \alpha(\omega), \beta(\omega), e(\omega)$

- Remark:** (PC) is ex ante, before ω is revealed, while (IC) must hold for each realization of ω

SKETCH DERIVATION AND OPTIMAL CONTRACT

- **(IC)** $\Rightarrow \alpha(\omega) = e(\omega)/\omega$ for all ω
- Insert into objective function and **(PC)**
- Optimal Contract $(\alpha(\cdot), \beta(\cdot), e(\cdot))$ is

$$\alpha(\omega) = \frac{1}{1 + rs^2(y)}$$

$$\beta(\omega) = a(x) - \frac{\omega t(y)}{1 + rs^2(y)} + \frac{\omega^2}{2(1 + rs^2(y))^2} (rs^2(y) - 1)$$

$$e(\omega) = \frac{\omega}{1 + rs^2(y)}$$

OPTIMAL CONTRACT

- Equilibrium:

$$w = a + \frac{\omega^2}{2(1+rs^2)} + \frac{\omega}{1+rs^2}\sigma$$

$$\pi = \omega t - a + \frac{\omega^2}{2(1+rs^2)} + \frac{rs^2}{1+rs^2}\omega\sigma$$

$$q = \omega(t + \sigma) + \frac{\omega^2}{1+rs^2}$$

- Ex ante Match Surplus:

$$\begin{aligned} V(x, y) &= \int \int q(\omega, \sigma, x, y) dF(\omega|x) dG(\sigma|y) \\ &= \int \left[\omega t + \frac{\omega^2}{1+rs^2} \right] dF \end{aligned}$$

OPTIMAL CONTRACT

With ω lognormal:

$$\mathbb{E}w(x) = a(x) + \frac{e^{2(k+u^2)}}{2(1+rs^2)}$$

$$\mathbb{E}\pi(y) = e^{k+\frac{u^2}{2}}t - a(x) + \frac{e^{2(k+u^2)}}{2(1+rs^2)}$$

$$V(x,y) = e^{k+\frac{u^2}{2}}t + \frac{e^{2(k+u^2)}}{1+rs^2}.$$

ENDOGENOUS OUTSIDE OPTION $a(x)$

- Ex post wages w : pinned down by optimal contract
- Ex ante compensation determines $a(x)$
- From FOC:

$$\max_x V(x, y) - a(x) \Rightarrow a'(x) = V_x(x, x)$$

and therefore $a(x) = a(\underline{x}) + \int_{\underline{x}}^x V_x(\tau, \tau) d\tau$ or:

$$a(x) = a(\underline{x}) + \int_{\underline{x}}^x \left(e^{k(z) + \frac{u(z)^2}{2}} t(z) (k'(z) + u(z) u'(z)) + \frac{e^{2(k(z) + u(z)^2)} (k'(z) + 2 u(z) u'(z))}{1 + r s(z)^2} \right) dz + a(\underline{x})$$

where $a(\underline{x}) \in [0, V(\underline{x}, \underline{x})]$.

PAM

- Match Value is separable

$$\Phi(x, y, \bar{v}) = \int \int \left(\frac{\omega^2}{1 + rs^2} + \omega(t + \sigma) \right) dF dG - \frac{1}{r} \log(-\bar{v}(x))$$

→ from CARA, quadratic cost, normal distribution

⇒ If FOSD: PAM $\iff \frac{\partial^2}{\partial x \partial y} \left(\frac{\omega^2}{1 + rs^2} + \omega t \right) > 0$
(from Proposition 1, TU)

APPLICATION 1. MISMATCHED CEOs

EMPIRICAL EXERCISE

Objective: illustrate the model can be applied to analyze mismatch

1. Use US data CEO compensation and firm profits to estimate:
 - Match value function
 - CEO and firm type distributions
2. Quantify mismatch in market for CEOs
3. Decompose value loss due to mismatch
 - Forgone complementarities
 - Changes in effort (incentives)

MISMATCHED CEOs

DATA

- Data sources:
 - Wages: Execucomp (Compustat) – total compensation: TDC1
 - Profits: Compustat: change in MkVal
- Constructing the variables:
 1. Newly hired 2010 (4 separations, 53 missing obs.)
 2. Rank firms by 2010 market value: $y = \log(\text{MkVal})$
 3. Rank workers: $x = y$
 4. w : $\text{TDC1}(2011) + \text{TDC1}(2012)$
 5. π : $\text{MkVal}(2012) - \text{MkVal}(2010)$

MISMATCHED CEOs

IMPLEMENTING THE MODEL

- Parametric form for k, u, t, s : e.g. $k(x) = k_0 + k_1x + k_2x^2$
- Calculate $a(x)$ from integral expression
- Distributions

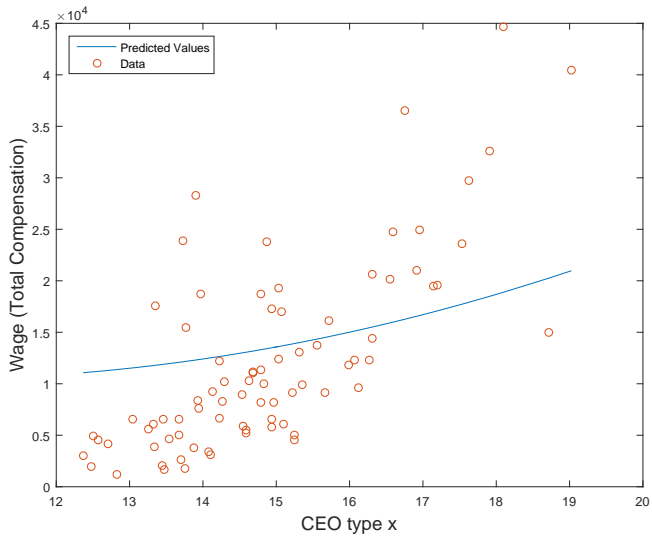
$$F(\omega|x) = \mathcal{LN}(k(x), u(x)^2)$$

$$G(\sigma|y) = \mathcal{N}(t(y), s(y)^2)$$

- Solve for ω, σ from w, π expressions
- Estimate k_0, k_1, k_2, \dots with MLE

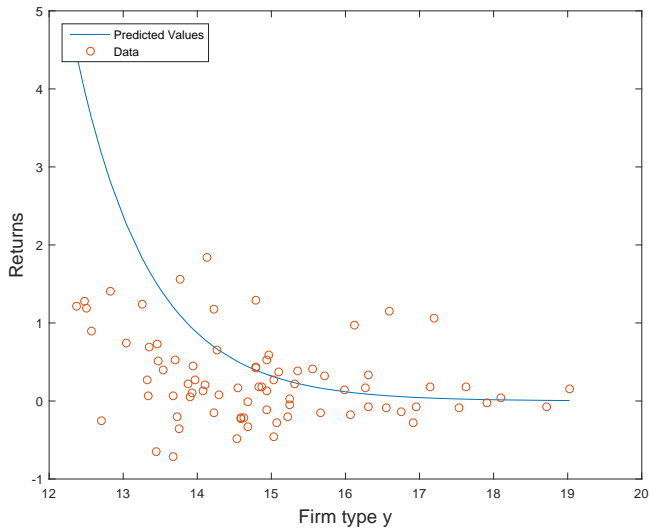
MISMATCHED CEOs

WAGES



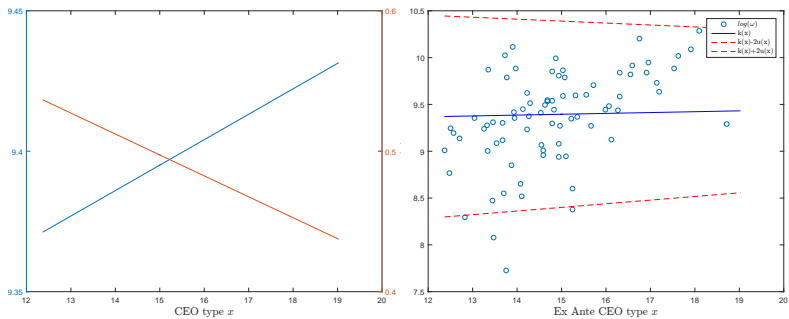
MISMATCHED CEOs

$$\text{RETURN} = \pi/V_0$$



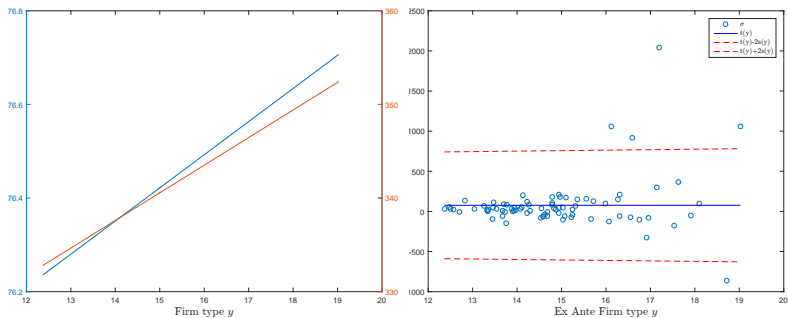
MISMATCHED CEOs

ESTIMATED PARAMETERS k, u



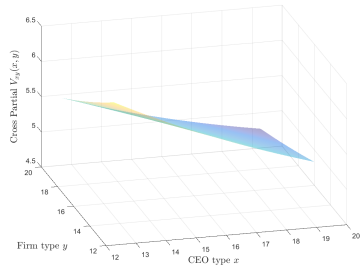
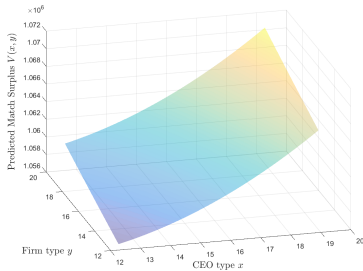
MISMATCHED CEOs

ESTIMATED PARAMETERS t, s



MISMATCHED CEOs

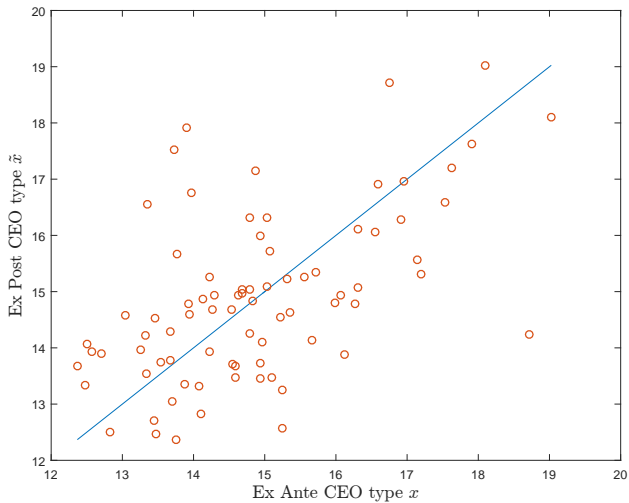
MATCH VALUE: $V(x, y)$ AND V_{xy}



⇒ Justifies identifying assumption of PAM!

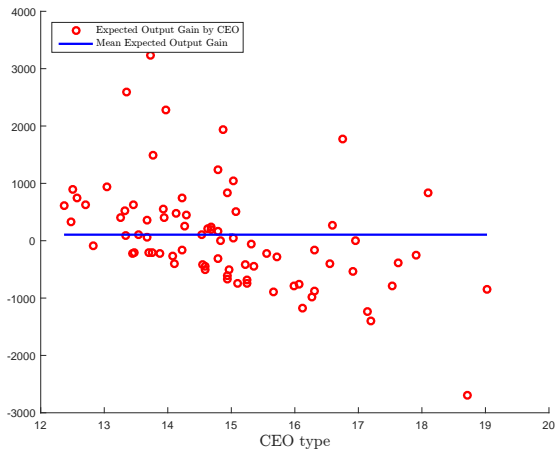
MISMATCHED CEOs

MISMATCH



AN EXPERIMENT: REASSIGN CEOs AFTER ω

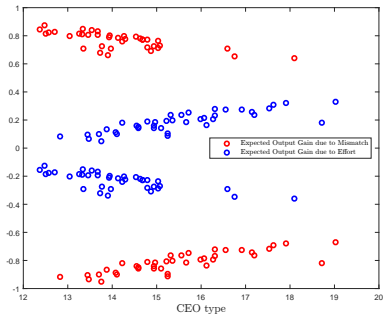
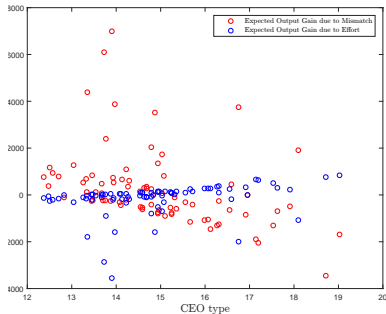
OUTPUT GAIN



- Given $V_{xy} > 0$, aggregate output gain positive

AN EXPERIMENT: REASSIGN CEOs AFTER ω

DECOMPOSITION: MISMATCH AND EFFORT



- Mainly mismatch, not effort
- Mismatch biggest at bottom

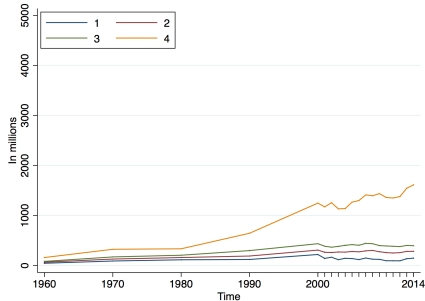
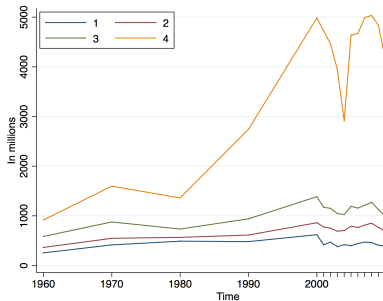
APPLICATION 2. HOUSEHOLD INCOME INEQUALITY

MARITAL VS. STOCHASTIC SORTING

- Household income inequality \uparrow between 1960-2014: $\text{Var} \times 8$
- Census + ACS data on education, earnings of married couples
- 4 education categories
- No data on ex post consumption
- Accounting exercise (like Greenwood e.a. (2015); Lam (1997))
- Inequality due to:
 1. Marital Sorting: ex ante \rightarrow education
 2. Stochastic Sorting: ex post \rightarrow earnings

HOUSEHOLD INCOME INEQUALITY

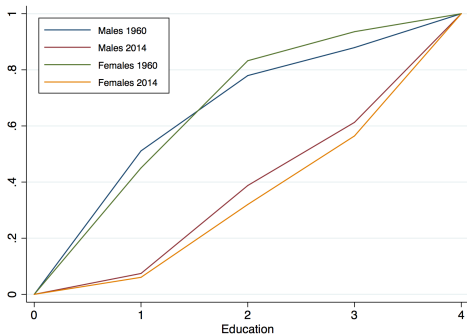
VARIANCE OF MALE AND FEMALE EARNINGS



- Correlation from 13.4% in 1960 to 23.4% in 2014

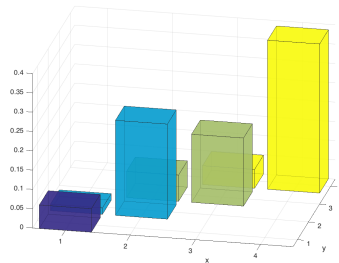
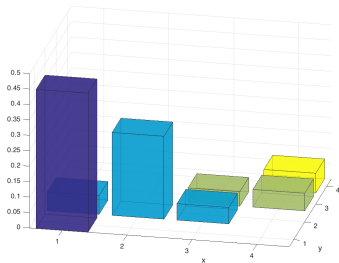
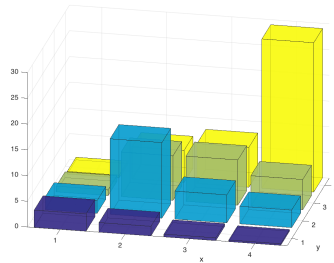
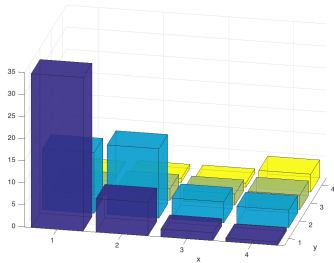
HOUSEHOLD INCOME INEQUALITY

EDUCATION 1960-2014



1. Huge right-shift for males and females
2. More for females

HOUSEHOLD INCOME INEQUALITY



DECOMPOSITION

Baseline	Sample	Normal Model*		
$\text{Var}_{2014}(\omega + \sigma)$	4.01×10^9	3.92×10^9		
$\text{Var}_{1960}(\omega + \sigma)$	0.49×10^9	0.42×10^9		
$\frac{\text{Var}_{2014}(\omega + \sigma)}{\text{Var}_{1960}(\omega + \sigma)}$	8.18	9.33		
	$X_{1960}[x, y]$	$\text{Var}_{2014}(\omega + \sigma; X_{1960})$	$\frac{\text{Var}_{2014}(\omega + \sigma; X_{1960})}{\text{Var}_{1960}(\omega + \sigma)}$	X_{2014} explains
A. Stochastic Sorting (ex post)				
1. Marginals Earnings*	F_{1960}, G_{1960}	0.75×10^9	1.78	81%
2. Correlation Earnings*	ρ_{1960}	3.50×10^9	8.33	11%
Total (Normal model)*	$F_{1960}, G_{1960}, \rho_{1960}$	0.68×10^9	1.39	85%
Total	K_{1960}	0.79×10^9	1.61	80%
B. Marital Sorting (ex ante)				
1. Marginals Education	$\Gamma_{1960}, \Psi_{1960}$	1.70×10^9	3.47	58%
Allocation: Gender Equal.	$\Gamma_{2014}, \tilde{\Psi}_{2014}, \mu_{1960}^f$	3.49×10^9	7.12	13%
All more educated	$\Gamma_{1960}, \tilde{\Psi}_{1960}, \mu_{2014}^f$	2.29×10^9	4.67	43%
2. Assortativeness Education	d_{1960}	3.97×10^9	8.10	1%
Total	M_{1960}	1.69×10^9	3.45	58%

* Assumes (ω, σ) is normally distributed.

CONCLUDING REMARKS

- Stochastic Sorting: Becker with realistic types
- Appealing:
 1. Characteristics change
 2. Mismatch in data
 3. “Noise” is integral part

CONCLUDING REMARKS

- Stochastic Sorting: Becker with realistic types
- Appealing:
 1. Characteristics change
 2. Mismatch in data
 3. “Noise” is integral part
- Applications
 1. Mismatched CEO's: loss driven by **mismatch**, not effort provision \Rightarrow focus on *selection*, rather than incentives
 2. Household Ineq.: **80+%** stochastic sorting; little marital sorting

STOCHASTIC SORTING

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TRANSFERABLE UTILITY (TU)

- Sketch of part (i):
 - Integration by parts and marginals function of one type yield

$$V_{xy} = \int_{\underline{\sigma}}^{\bar{\sigma}} \int_{\underline{\omega}}^{\bar{\omega}} q_{\omega\sigma}(\omega, \sigma) K_{xy}(\omega, \sigma | x, y) d\sigma d\omega$$

- If K is supermodular in (x, y) then so is V if and only if q is supermodular. Similar for K submodular
- Sketch for part (ii):
 - More integration by parts and marginals, functions of one type, yield

$$V_{xy} = \int_{\underline{\sigma}}^{\bar{\sigma}} \int_{\underline{\omega}}^{\bar{\omega}} q_{\omega\omega\sigma\sigma}(\omega, \sigma) \left(\int_{\underline{\omega}}^{\omega} \int_{\underline{\sigma}}^{\sigma} K_{xy}(s, t | x, y) ds dt \right) d\sigma d\omega$$

- If $\int_{\underline{\omega}}^{\omega} \int_{\underline{\sigma}}^{\sigma} K$ is supermodular in (x, y) then so is V if and only if q is supermodular. Similar for K submodular

TRANSFERABLE UTILITY (TU)

EXAMPLES OF OBSERVED (ω, σ) – NAM: FOSD vs. MPS

Distribution of matched ω, σ realizations under PAM/NAM and with FOSD and MPS. Simulations with 1000 types. Under FOSD (both for PAM and NAM), x, y uniform on $[-.5, .5]$, and $\omega = x + \varepsilon, \sigma = y + \varepsilon$ where $\varepsilon_\omega, \varepsilon_\sigma$ are conditionally independent uniform draws on $[-.5, .5]$. Under MPS, (both for PAM and NAM), x, y uniform on $[0, 1]$, and $\omega = x \cdot \varepsilon, \sigma = y \cdot \varepsilon$ where $\varepsilon_\omega, \varepsilon_\sigma$ are conditionally independent uniform draws on $[-1, 1]$.