# Marriage and employment returns to female education 

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August 30, 2023<br>EEA/ESEM 2023, Barcelona

## Motivation

The past century has witnessed a dramatic increase in women's education.
Schooling of female population, 15 years and over

region
-- Advanced Economies
$\simeq$ East Asia and the Pacific

- Europe and Central Asia
- Latin America and the Caribbean
\#- Middle East and North Africa
* South Asia
- Sub-Saharan Africa


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marriage
- Intensive margin:
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job
wage \& salary
- transfer
spouse type
surplus share
unobservable


## Measuring marriage return to education

Wage premium is usually used to measure labor market return to education But in marriage markets, surplus and transfer are not observable

Estimation of surplus using matching patterns (who marries whom)

- Choo and Siow (2006): frictionless matching with transferable utility
- Recovering deterministic gains from marriage patterns


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- Recovering deterministic gains from marriage patterns

Chiappori et al. (2017) use deterministic utilities to measure marriage return to education level 2 compared to 1 as

$$
\underbrace{U_{2}^{\text {married }}-U_{2}^{\text {single }}}_{\text {marriage gain from edu } 2}-(\underbrace{\left.U_{1}^{\text {married }}-U_{1}^{\text {single }}\right)}_{\text {marriage gain from edu } 1})
$$

## This paper

Extends the framework of Choo and Siow (2006); Chiappori et al. (2017)

- Two bilateral matching markets: job market, marriage market
- Matching based on partner's quality (job \& spouse types)
- Estimation using 3-way empirical distribution table of women's education $\times$ employment $\times$ marital status.
- Using transfer in labor market (earnings) as OID restrictions for estimation.

Yet, no joint estimation for the marriage and employment return to education and their difference.

Sign-based identification with no distributional assumption.
Documenting the U.S. trends for 1960-2019 (and many other countries).

## Basic matching model: extending Choo-Siow

There are a large number of women, firms, and men belonging to a small number of observable categories.

- $I \in\left\{1, \ldots, N_{I}\right\}$ women's education
- $J \in\left\{0,1, \ldots, N_{J}\right\}$ job classification ( $0=$ not working)
- $K \in\left\{0,1, \ldots, N_{K}\right\}$ men's education ( $0=$ single)


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- In the job (marriage) market, the marriage (employment) category of the woman is important for firms (men).



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## Extending Choo-Siow model

Woman's utility is quasi-linear in the payoffs from the two markets

$$
u_{i}=\underbrace{x_{i}}_{\substack{\text { marriage } \\ \text { transfer }}}+\Phi(\underbrace{w_{i}}_{\text {earnings }}), \quad \Phi \text { is a strictly increasing }
$$

By assuming separability of unobservable factors in observable categories,
$U^{I J K}=$ average utility of women with education $I$, job $J$, spouse $K$
is estimated up to a constant for each $I$ using a distributional assumption for those unobservable terms.

## Conditional returns to education $I_{2}$ from $I_{1}$ - extensive margin

Marriage return of marrying husband $K$ conditional on employment $J$ :

$$
r_{I_{1} I_{2} J K}^{m}=U^{I_{2} J K}-U^{I_{2} J 0}-\left(U^{I_{1} J K}-U^{I_{1} J 0}\right), \quad K \geq 1
$$

Employment return of getting job $J$ conditional on marriage $K$ :

$$
r_{I_{1} I_{2} J K}^{e}=U^{I_{2} J K}-U^{I_{2} 0 K}-\left(U^{I_{1} J K}-U^{I_{1} 0 K}\right), \quad J \geq 1
$$

The conditional difference between marriage $K$ and employment $J$ returns

$$
\delta_{I_{1} I_{2} J K}^{m e}=\underbrace{U^{I_{2} 0 K}-U^{I_{1} 0 K}}_{\text {married to } K \text { not-working }}-(\underbrace{U^{I_{2} J 0}-U^{I_{1} J 0}}_{\text {single working in } J}) \quad J, K \geq 1
$$

## Returns with logit distribution

Let $n(I J K)$ be the population corresponding to education $I$, occupation $J$, and spouse $K$.

## Proposition

If difference in unobservable terms has logit distribution

$$
\begin{array}{ll}
r_{I_{1} I_{2} J K}^{m}=\ln \frac{n\left(I_{2} J K\right) \times n\left(I_{1} J 0\right)}{n\left(I_{2} J 0\right) \times n\left(I_{1} J K\right)}, & K \geq 1 \\
r_{I_{1} I_{2} J K}^{e}=\ln \frac{n\left(I_{2} J K\right) \times n\left(I_{1} 0 K\right)}{n\left(I_{2} 0 K\right) \times n\left(I_{1} J K\right)}, & J \geq 1 \\
\delta_{I_{1} I_{2} J K}^{r}=\ln \frac{n\left(I_{2} 0 K\right) \times n\left(I_{1} J 0\right)}{n\left(I_{1} 0 K\right) \times n\left(I_{2} J 0\right)}, & J, K \geq 1
\end{array}
$$

## Aggregate returns

Aggregate returns at the extensive margin

$$
\begin{aligned}
& \hat{r}_{I_{1} I_{2}}^{m}=E\left[r_{I_{1} I_{2} J K}^{m} \mid I_{1}, I_{2}\right] \\
& \hat{r}_{I_{1}}^{e} I_{2}=E\left[r_{I_{1} I_{2} J K} \mid I_{1}, I_{2}\right] \\
& \hat{\delta}_{I_{1} I_{2}}^{m e}=E\left[\delta_{I_{1} I_{2} J K}^{m e} \mid I_{1},\right. \\
& \hline
\end{aligned}
$$

## Empirical methodology

The basic model is just-identified and throws out earnings information.
Adding more moments from average earnings $W^{I J K}$ :

$$
U^{I J K}-U^{I J^{\prime} K}=\rho_{I K}\left(\Phi\left(W^{I J K}\right)-\Phi\left(W^{I J^{\prime} K}\right)\right), \quad J, J^{\prime}>0
$$

$\rho_{I K}$ is the sharing rule from just-identified model.
A heteroskedastic $\Phi(\cdot)$ as

$$
\Phi(W)=\frac{1}{1-\phi_{I}} W^{1-\phi_{I}}
$$

Finding $U^{I J K}$ and $\phi_{I}$ using Minimum Distance Estimator.

## Dollar equivalent of spouse education

$$
u_{i}=\underbrace{x_{i}}_{\substack{\text { marriage } \\ \text { trancfor }}}+\Phi(\underbrace{w_{i}}_{\text {earnings }})
$$

$\Phi^{-1}(\cdot)$ converts utility units to earnings units
$r=\Delta U_{2}-\Delta U_{1}$ and from inverse function theorem

$$
\Phi^{-1}(r) \approx \frac{r}{\Phi^{\prime}\left(\Phi^{-1}(\bar{U})\right)}=\frac{r}{\Phi^{\prime}(\bar{W})}=r \bar{W}^{\phi}
$$

## Data

USA: IPUMS USA, Version 12.0

- Census 1960 (5\%), 1970 (1\%), 1980 (5\%), 1990 (5\%), 2000 (5\%)
- American Community Surveys (ACS): 5-year averages $(2007,2012,2017)$
- 3-D discrete distribution for women between 35 and 50
- Average number of women per round: $1,270,000$


## Classifications

- Education

1. Dropouts: have no high school qualification
2. High school: finished high school
3. Some college: attend 1 to 3 years of college
4. Bachelor: bachelor's degree
5. Graduate: higher education than bachelor's degree

- Occupation (ISCO code)

1. Unskilled: elementary occupations (code 9)
2. Skilled: skilled/semi-skilled workers (codes 0, 4 to 8 )
3. High skilled: technicians and associate professionals (code 3)
4. Professional: managers, professionals (codes 1, 2)

## Conditional marriage returns to female education in the US


spouse dropout $\square$ high school $\square$ some college
bachelor $\square$ graduate

## Conditional employment returns to female education in the US


occupation - unskilled $*$ skilled $\cdot$ highskilled $\cdot \theta$ professional

## Equivalent remuneration of marrying different men in 2017

|  |  | equivalent annual worth (in 2023 dollars) |  |  |  | \% of women's yearly earnings |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  |  |  |  |  |  | women's education |  |  |
|  |  |  | $$ |  |  |
| \% | dropout |  |  |  |  | -1607 | -4630 | -8215 | -7510 | -4.91 | -11.52 | -13.76 | -9.05 |
| $\cdots$ | high school | 1014 | 858 | 3 | -1230 | 3.10 | 2.13 | 0.00 | -1.48 |
| $\stackrel{\square}{2}$ | some college | 965 | 3421 | 4445 | 3157 | 2.95 | 8.51 | 7.45 | 3.81 |
| $\stackrel{0}{0}$ | bachelor | 1119 | 3767 | 10357 | 9580 | 3.42 | 9.37 | 17.35 | 11.55 |
| $\stackrel{\square}{3}$ | graduate | 883 | 3500 | 10949 | 15056 | 2.70 | 8.71 | 18.34 | 18.15 |

## Aggregate extensive margin return indices in the US










## Conclusion

Building a method for the joint estimation of marriage and employment returns to female education.

- Enabling comparison between the two returns
- Feasible in widely available cross-sectional household surveys
- Sign-based identification with no distributional assumption
- Separate estimations for extensive and intensive margins of the returns


## Thank you!

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## References:

Chiappori, P.-A., Salanié, B., and Weiss, Y. (2017). Partner Choice, Investment in Children, and the Marital College Premium. American Economic Review, 107(8):2109-2167.
Choo, E. and Siow, A. (2006). Who Marries Whom and Why. Journal of Political Economy, 114(1):175-201.

