

Adoption of industrial robots and fertility

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Labour force participation (LFP) and fertility

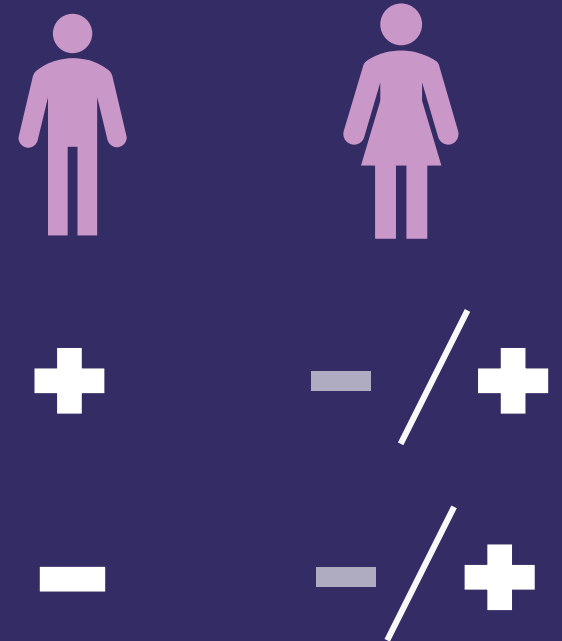
- source of income
- opportunity costs / work-family conflict
- social context: family policies, social norms

Employment / income

Unemployment

Temporary employment

Employment uncertainty





- Labour augmenting / displacing effects?

Our 1000-Year Struggle Over
Technology & Prosperity

POWER AND PROGRESS

DARON ACEMOGLU

Co-author of WHY NATIONS FAIL

SIMON JOHNSON

Co-author of 13 BANKERS

- Technological innovations facilitate development but may lead to substantial social inequalities
- Labour augmenting or displacing
 - Marginal productivity of workers
 - Institutions

Labour augmenting effects

Labour augmenting effects

complementing human labour



marginal productivity of workers ↑



output ↑



labour demand & wages ↑



Electric assembly line

- Reduction in production time and costs
- Production tripled
- Larger demand for lower skilled workers
- Working hours declined from 9 to 8 hours
- Wages doubled
- Better working conditions (safety and health)
- Expansion of trade unions

Labour displacing effects?

Labour displacing effects

Certain job and work tasks get replaced by machines



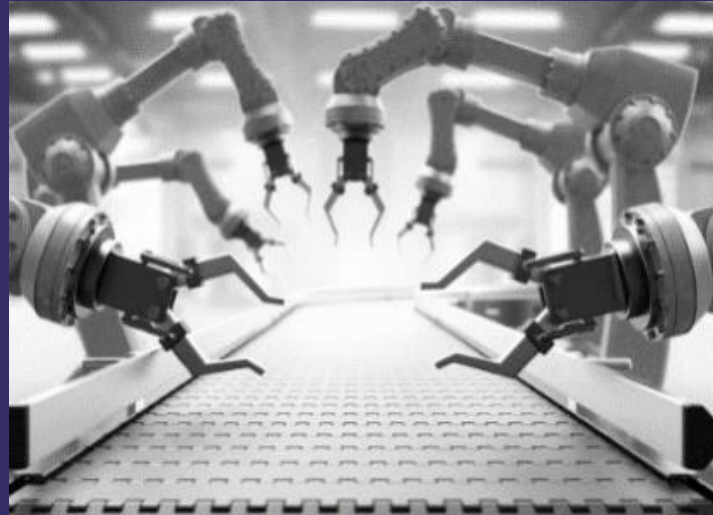
production costs ↓

average productivity ↑

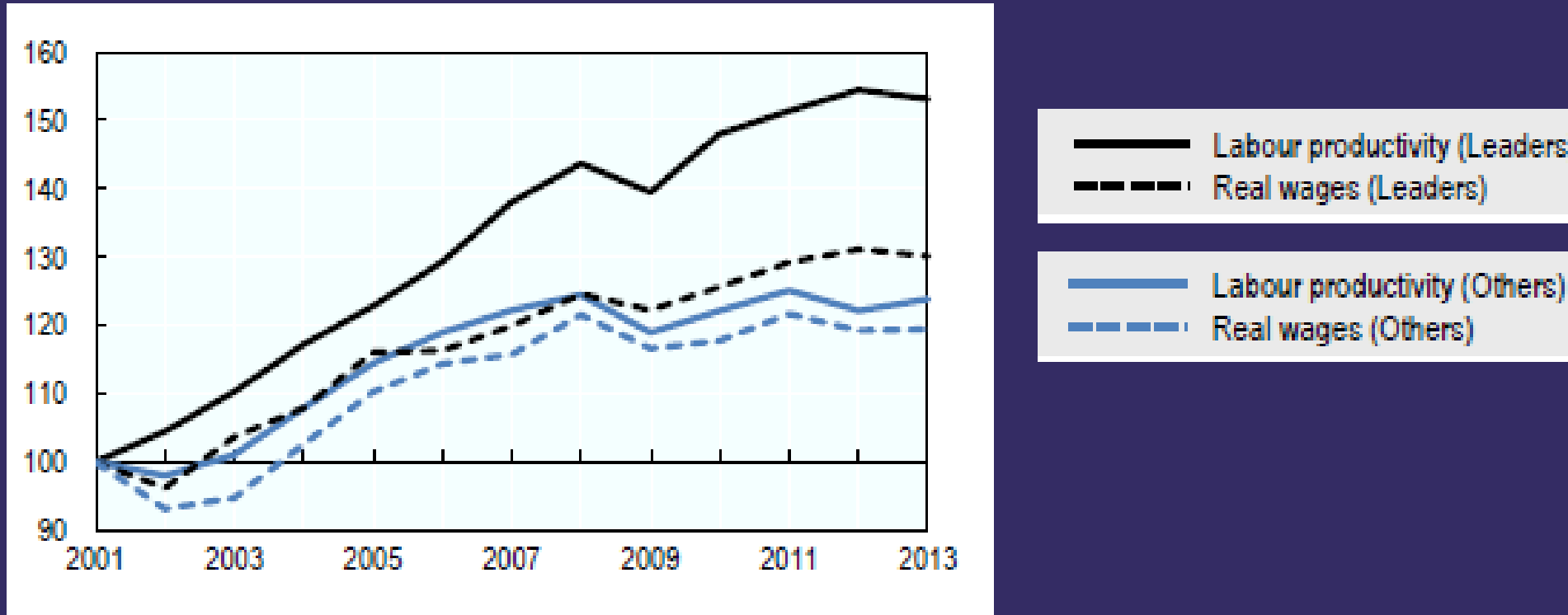
marginal productivity of workers ↓



Output? Labour demand? Wages?



Wages have decoupled from productivity in technologically advanced firms...

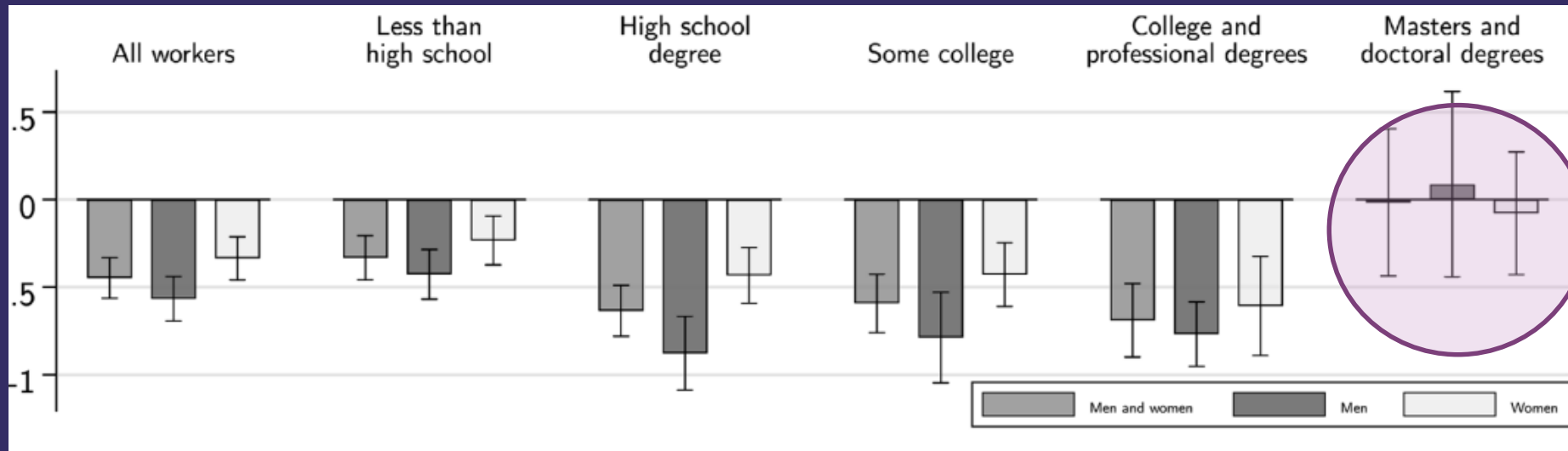


Countries: Belgium, Denmark, Germany, Ireland, Japan, Korea, Sweden, UK, US

Source: OECD Employment Outlook 2018

Automation and employment

- **US:** 1 robot / 1000 workers reduces the employment rate by 0.2 pp. and wages by about 0.42% (Acemoglu and Restrepo 2020)



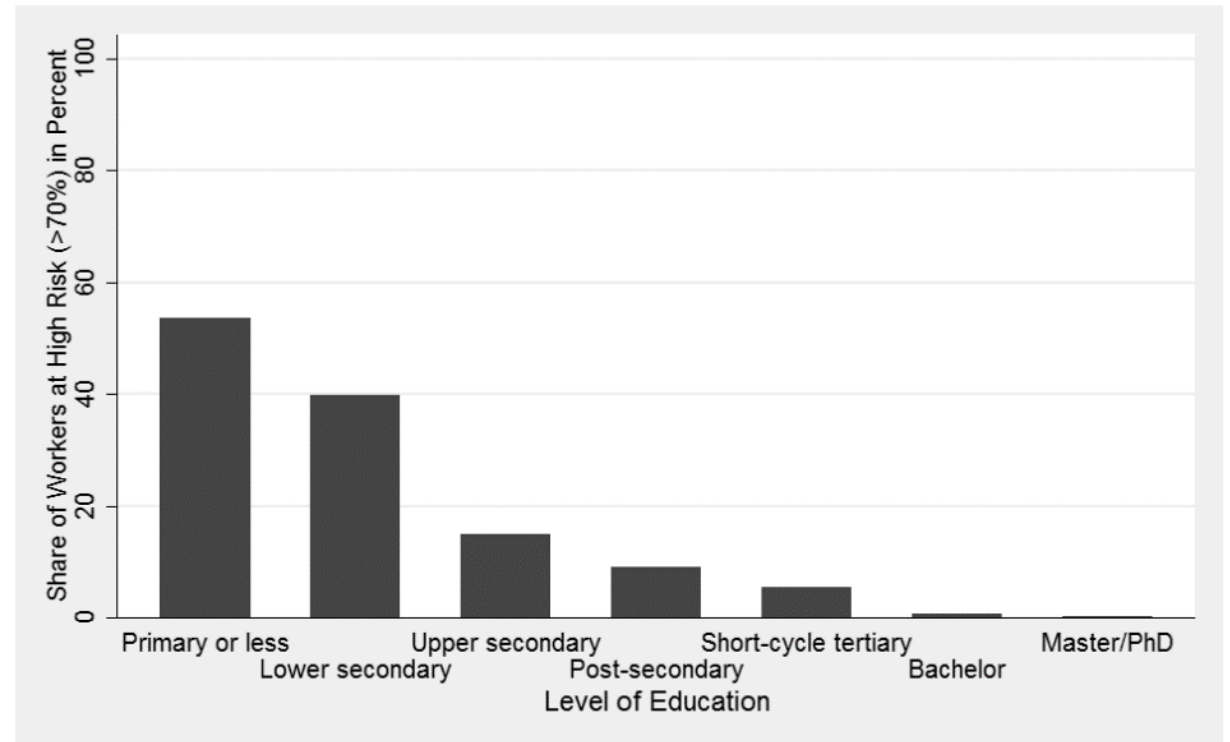
Source: Acemoglu and Restrepo 2020

- **Europe:** null overall effect, but negative effects on employment of low and middle educated workers (Graetz and Michaels 2018)

Automation and employment

- 9-14% of jobs at high risk of full automation (more than 70% of tasks automatable)
- 25-32% jobs at medium risk (50-70% of tasks automatable)

Figure 6. Share of Workers with High Automatability by Education



Source: Authors' calculation based on the Survey of Adult Skills (PIAAC) (2012)

Source: Arntz et al. (2017), Nedelkoska and Quintini (2018)



- Labour augmenting / displacing effects
- Changing structure of the labour demand / growing disparities between high and low-to-middle skilled
- Effects on wages, employability, stability and certainty of employment
- Structural LM change (not cyclical!)

Past research



Anelli et al. (2021):

- regional study (commuting zones in the US)
- adoption of industrial robots → more cohabitation and divorce, decline in marital fertility, increase in non-marital fertility



MACRO-LEVEL STUDY

Co-authors: D. Bellani & H. Bogusz

Countries: DE, IT, FR, UK, PL & CZ

Period: 1993-2017

MICRO-LEVEL STUDY

Co-authors: L. Andersson, W. Hardy

Countries: Sweden

Period: 1993-2017

MACRO-LEVEL STUDY

Co-authors: D. Bellani & H. Bogusz

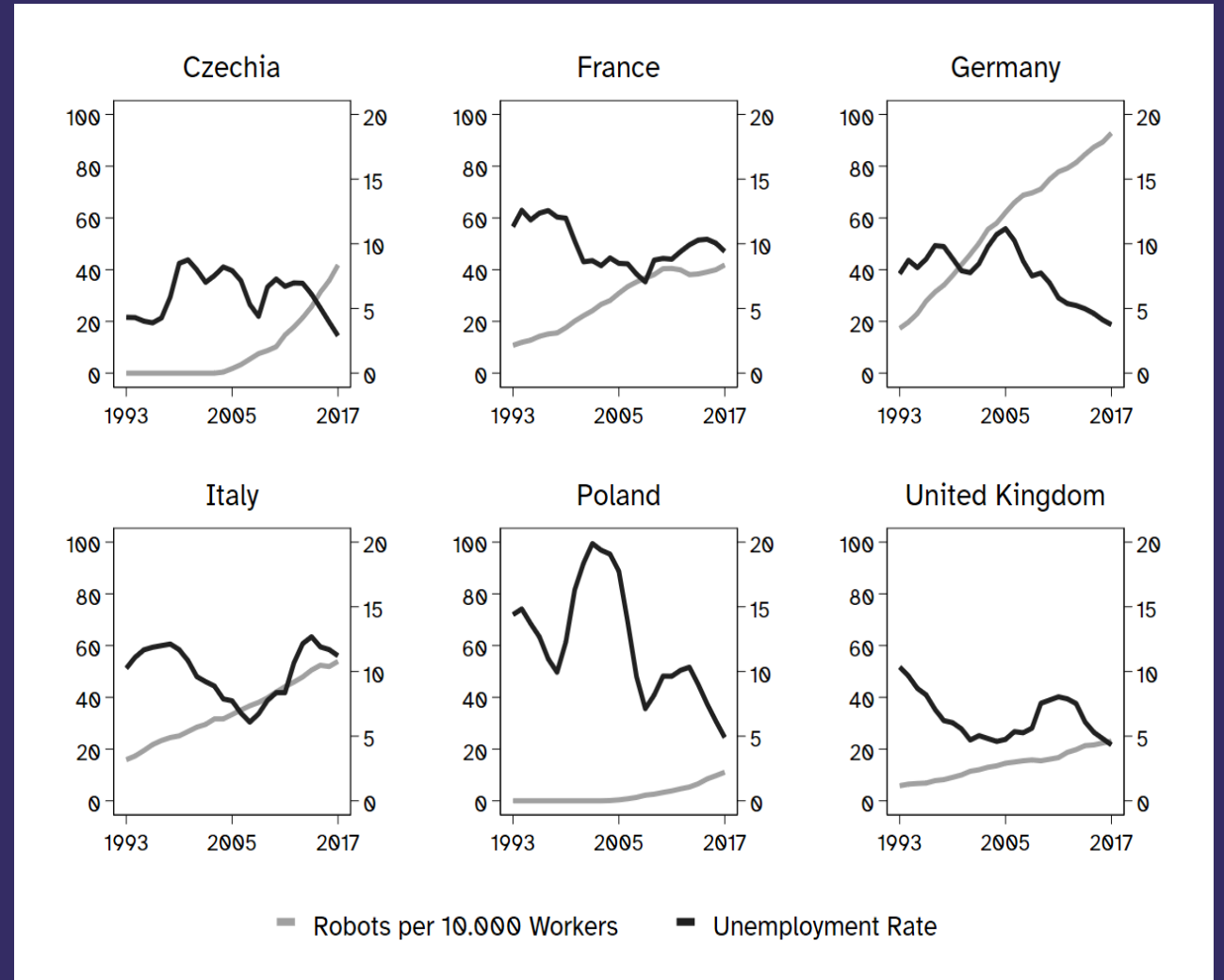
Countries: DE, IT, FR, UK, PL & CZ

Period: 1993-2017

Fertility effects of automation less pronounced in regions with:

H1: better educated populations

H2: more technologically advanced



MACRO LEVEL STUDY: Data (1993-2017)

EUROSTAT:

- Regional NUTS-2 fertility rates (total and age-specific)
- Regional employment structures by industry (NACE 2-digit)

INTERNATIONAL FEDERATION OF ROBOTICS (IFR)

- Robot stocks (country and industry-specific) at 3-digit since 1993

fully autonomous machines that do not require a human operator

MACRO LEVEL STUDY: Measurement

$$\text{Exposure to robots}_{r,t} = \sum_{i=1}^N \left(\frac{\text{empl}_{r,i,t_0}}{\text{empl}_{r,t_0}} \right) \left(\frac{\text{robots}_{i,t}}{\text{empl}_{i,t_0}} \right)$$

distribution of initial employment at t0 across regions

replacement of initial employment (at t0) in industry i by robots

MACRO LEVEL STUDY: Modelling

$$fertility_{r,t} = \alpha \cdot Exposure\ to\ robots_{r,t-2} +$$

$$+ \beta \cdot Controls_{r,t-1} + \eta_r + v_t + \varepsilon_{r,t}$$

Controls:

- population age structure
- % highly educated
- ratio highly educated women to men
- women's economic activity rate

Year fixed effects

Regional fixed effects

MACRO LEVEL STUDY: Modelling

$$fertility_{r,t} = \alpha \cdot Exposure\ to\ robots_{r,t-2} + \beta \cdot Controls_{r,t-1} + \eta_r + v_t + \varepsilon_{r,t}$$

$$\sum_{i=1}^N \frac{empl_{r,i,t_0}}{empl_{r,t_0}} \left(\frac{robots_{i,t}^c}{empl_{i,t_0}} \right)$$

Overidentified IV model:

- Robot stocks instrumented with robots in {Germany, France, UK, Italy, Spain, Sweden, Norway and Finland} excluding the studied country
- In models for Czechia and Poland we additionally use US as an instrument

MACRO LEVEL STUDY: Modelling

$$\text{fertility}_{r,t} = \alpha \cdot \text{Exposure to robots}_{r,t-2} + \gamma \cdot \text{Moderator}_{r,t-1} + \beta \cdot \text{Controls}_{r,t-1} + \eta_r + \nu_t + \varepsilon_{r,t}$$

Fertility effects less pronounced if:

H1: better educated populations

H2: region more technologically advanced



Moderators:

% highly educated

% empl in techn and knowledge sector

MACRO-LEVEL STUDY: Results

Country	TFR	FR 20-24	FR 25-29	FR 30-34	FR 35-39	FR 40-44	FR 45+
Germany	ns	ns	ns	ns	-0.00011***	-0.00005***	ns
France	ns	ns	ns	ns	ns	ns	ns
Italy	-0.00118*	ns	-0.00090***	ns	ns	ns	ns
UK	ns	ns	ns	ns	ns	0.00039*	ns
Czechia & Poland	ns	ns	ns	ns	0.00025*	ns	ns

*** 1% ** 5% * 10%. Sample sizes: 680 observations for Germany, 440 for France, 400 for Italy, 700 for the UK, and 240 for Poland and Czechia jointly.

MACRO-LEVEL STUDY: Results

% highly educated

Country	TFR <i>main effect</i>	TFR <i>interaction effect</i>
Germany	-0.0016***	0.00005***
France	0.0015**	-0.00058**
Italy	-0.00292*	0.0001
UK	ns	ns
Czechia & Poland	ns	ns

*** 1% ** 5% * 10%. Sample sizes: 680 observations for Germany, 440 for France, 400 for Italy, 700 for the UK, and 240 for Poland and Czechia jointly.

MACRO-LEVEL STUDY: Results

empl in technology and knowledge sectors

Country	TFR <i>main effect</i>	TFR <i>interaction effect</i>
Germany	ns	ns
France	ns	ns
Italy	-0.00116*	0.000005
UK	ns	ns
Czechia & Poland	ns	ns

*** 1% ** 5% * 10%. Sample sizes: 680 observations for Germany, 440 for France, 400 for Italy, 700 for the UK, and 240 for Poland and Czechia jointly.

MICRO-LEVEL STUDY

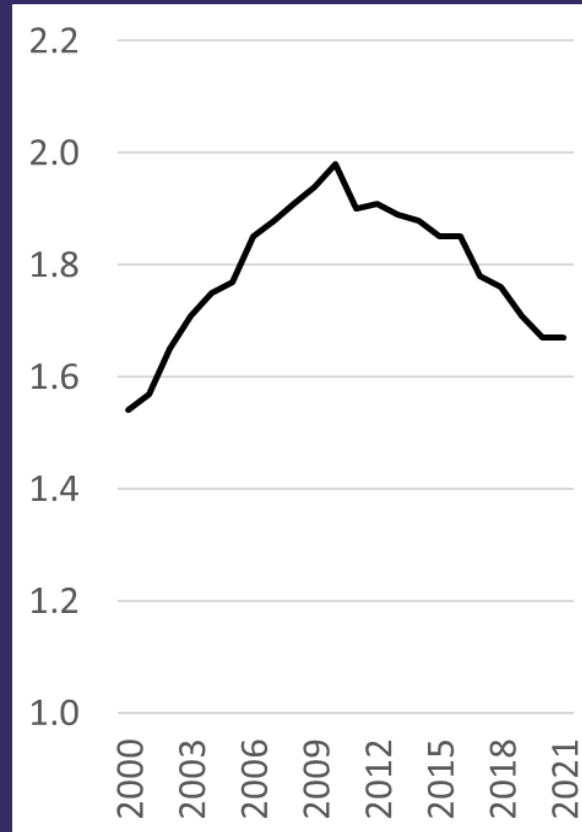
Co-authors: L. Andersson, W. Hardy

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Period: 1993-2017

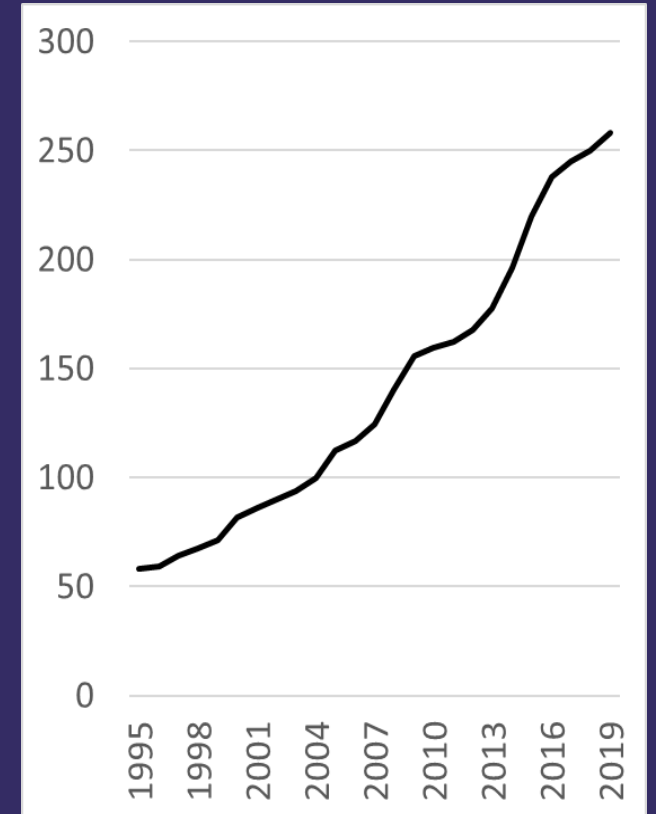


Total fertility rate



Robot density

Number of industrial robots per 10,000 workers in manufacturing



MICRO-LEVEL STUDY: Data and Method



Data:

- Swedish register data
- IFR robot stocks (industry-specific) at 3 digit since 1993

Period: 1993-2017

Method:

- Event history models

Events:

- Marriage
- 1st, 2nd, 3rd birth
- Divorce

MICRO-LEVEL STUDY: Data and Method



Measure:

- Exposure to robots

$$\text{Exposure to robots}_{i,t} = \frac{\text{robots}_{i,t}}{\text{empl}_{i,t_0}}$$

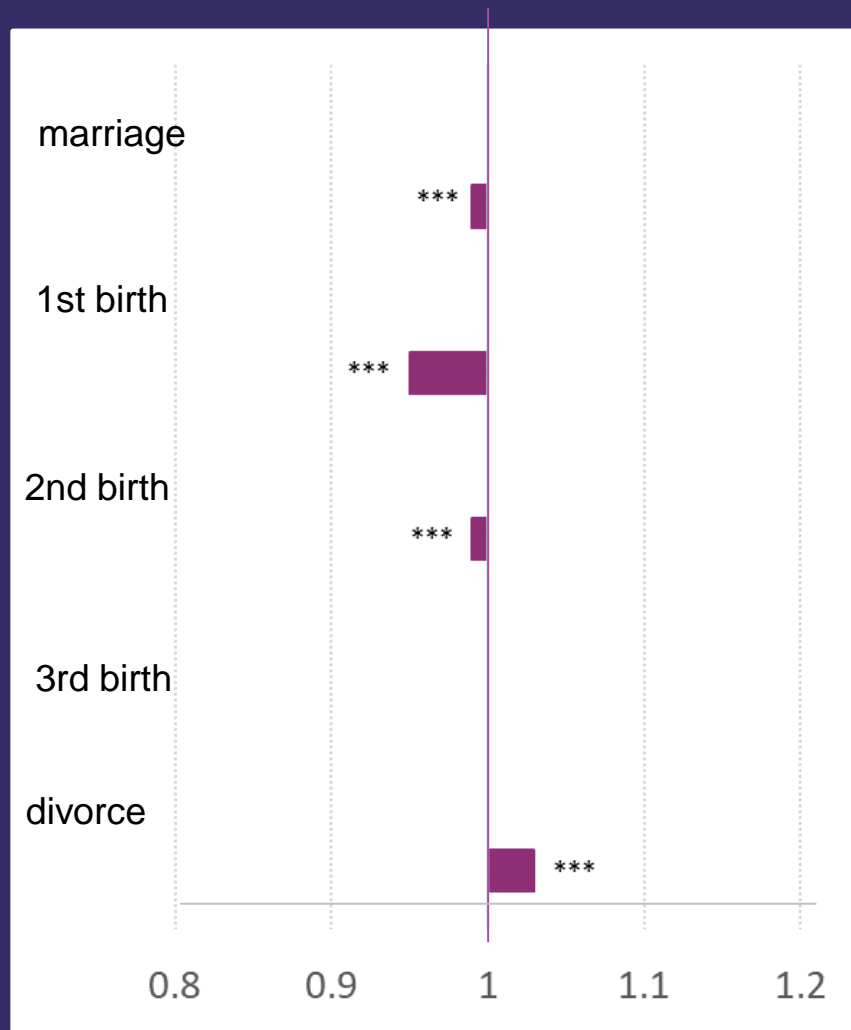
IV: Using stock of robots in other countries which are similarly (Finland, Denmark)

Controls:

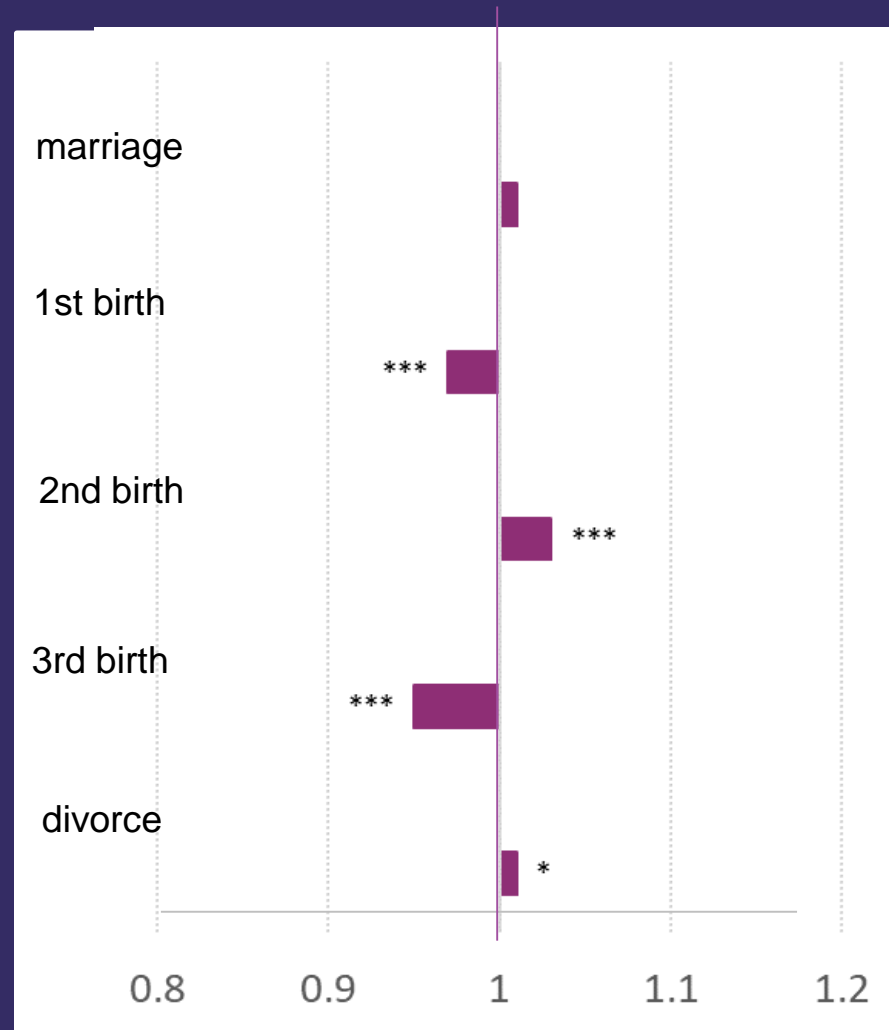
- Age
- (Age of the previous child)
- Calendar year
- Education
- Employment status (works in a sector with / without robotisation, no work)
- Firm size
- Seniority status

MICRO-LEVEL STUDY: Results

MEN



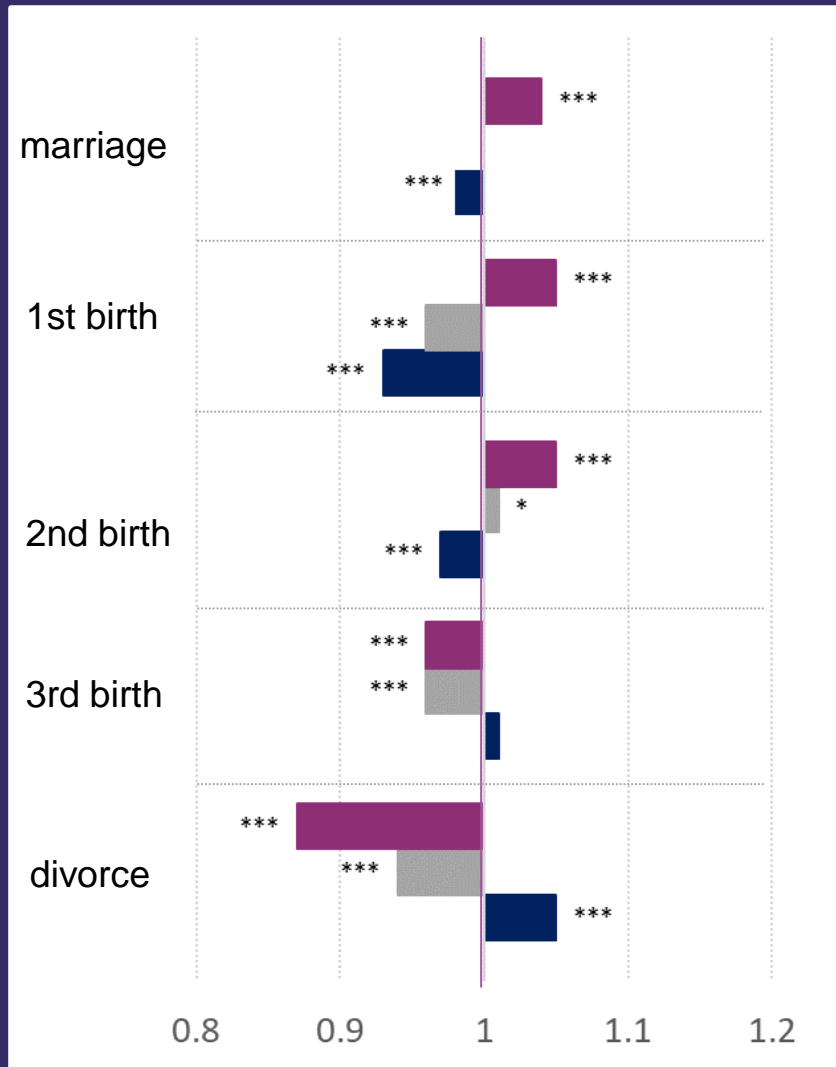
WOMEN



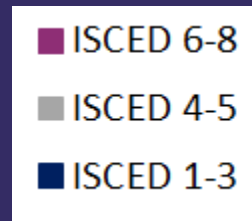
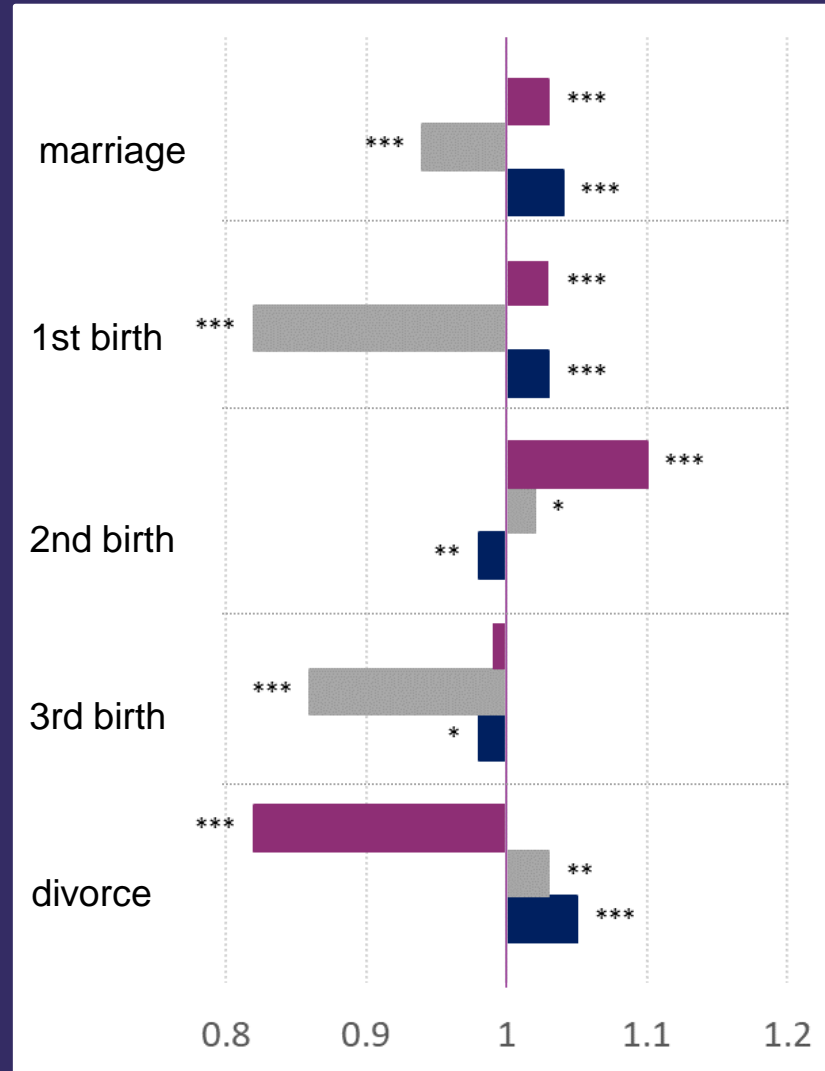
Note: A change in the risk of an event due to an increase in robot adoption in an industry by one standard deviation

MICRO-LEVEL STUDY: Results

MEN



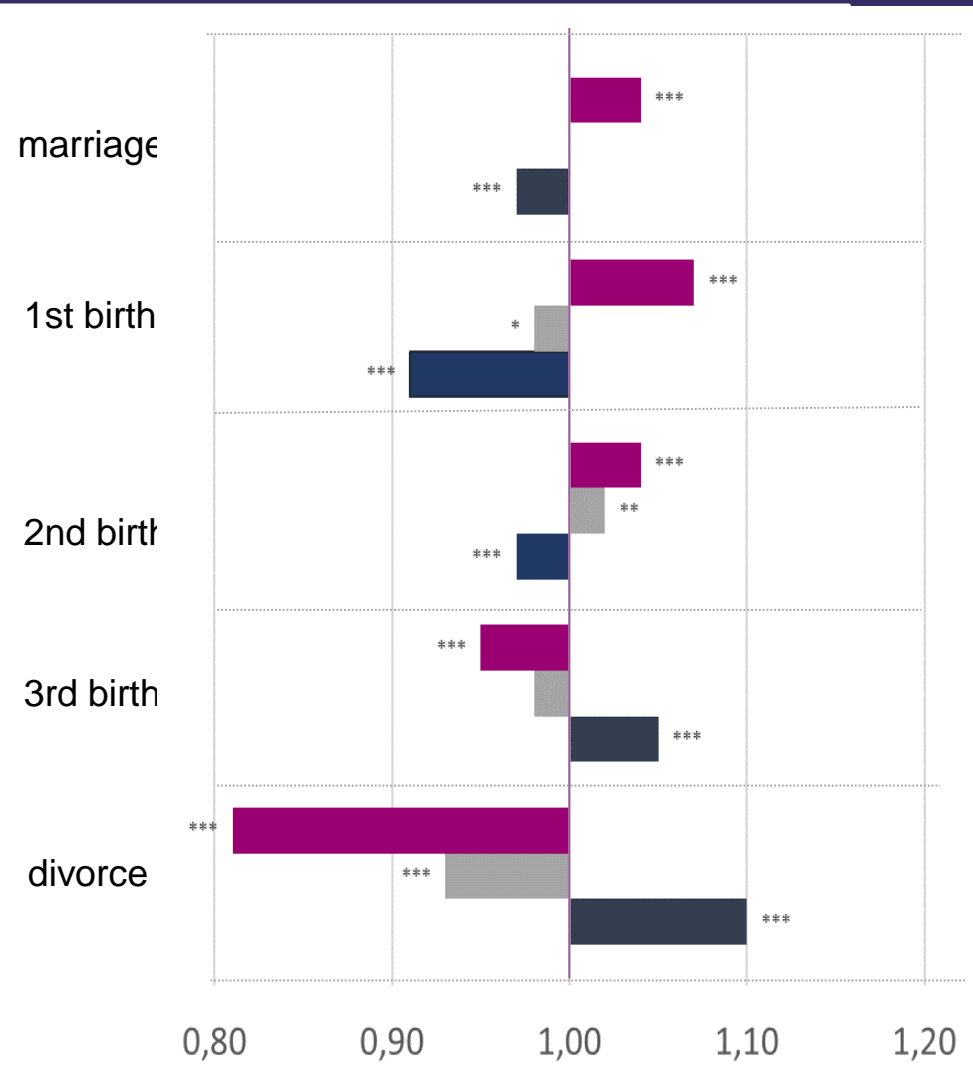
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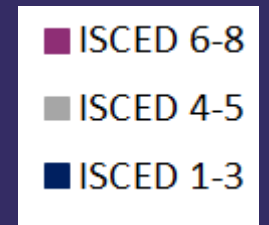
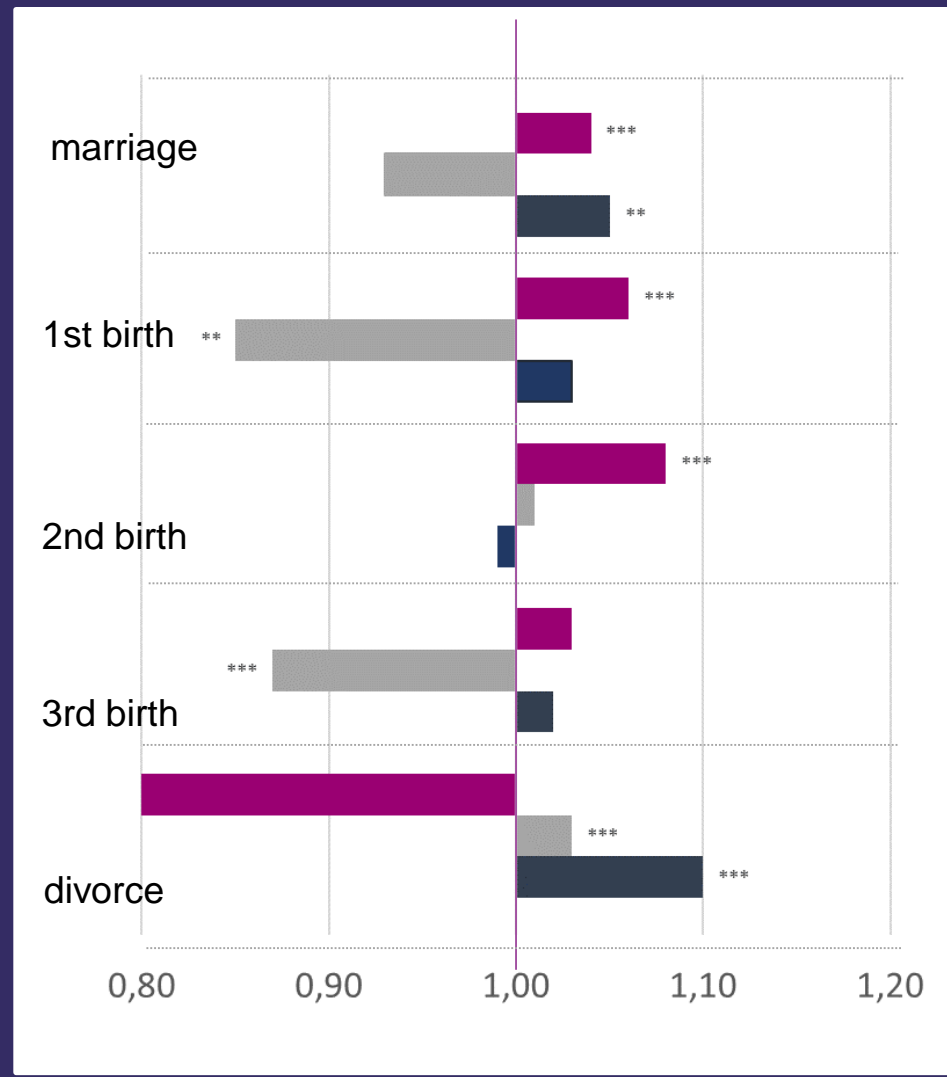
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MICRO-LEVEL STUDY: IV Results

MEN



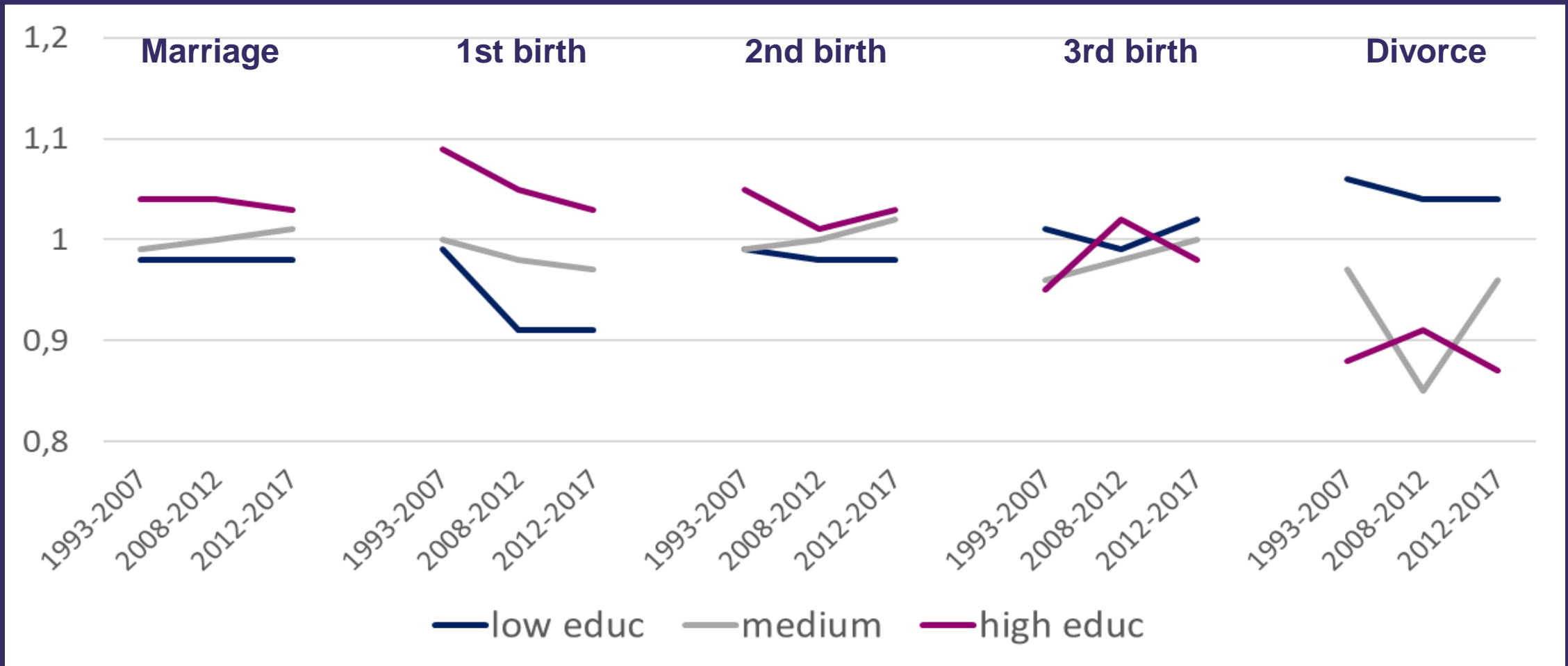
WOMEN



Note: A change in the risk of an event due to an increase in robot adoption in an industry by one standard deviation

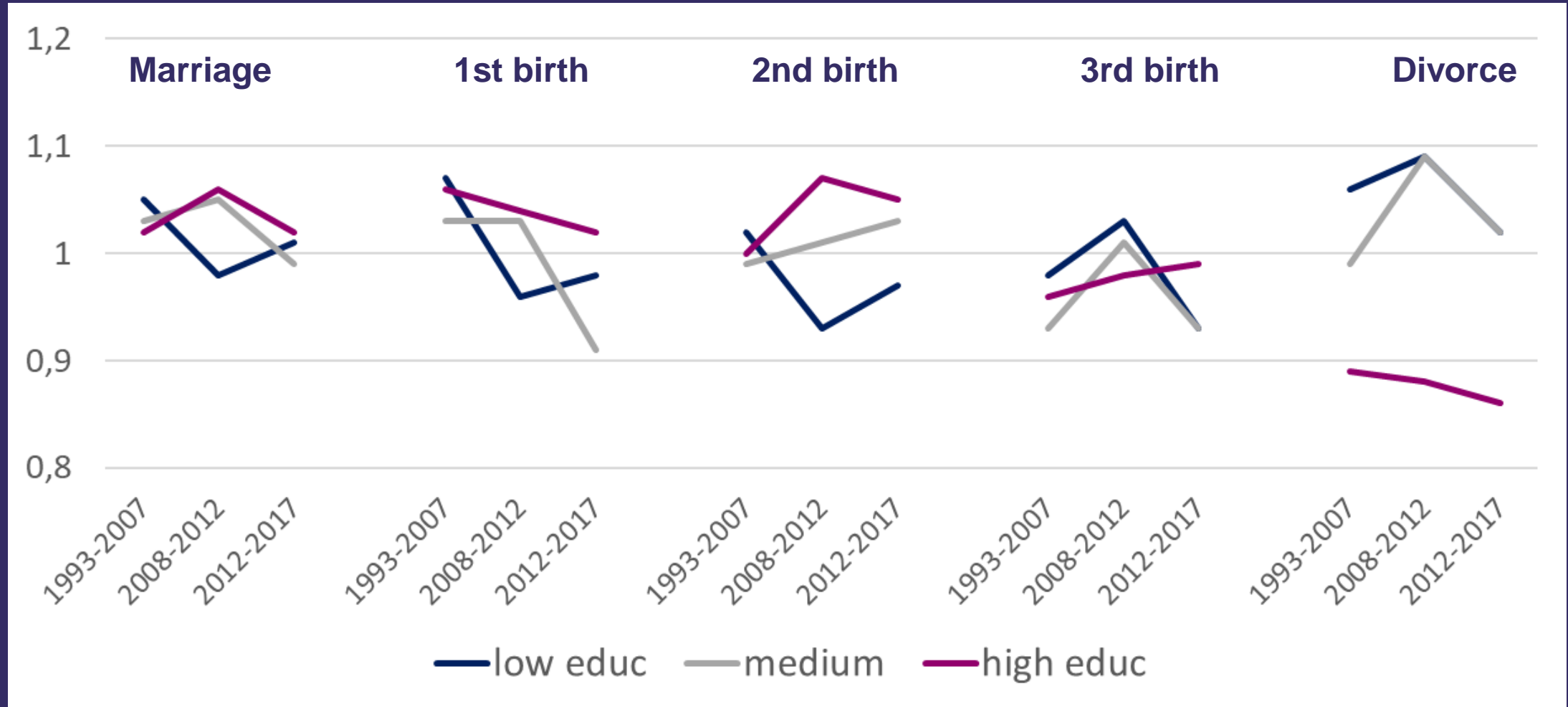
MICRO-LEVEL STUDY: Results

Change in the risk of the event due to an increase in automation by 1 st dev., MEN



MICRO-LEVEL STUDY: Results

Change in the risk of the event due to an increase in automation by 1 st dev., WOMEN



Conclusions

- Rather weak overall effects of robot adoption on fertility / family formation and its stability
- Clear edu differences
 - Negative effects on fertility more pronounced in regions with lower educated populations
 - Negative effects on family formation and stability among low educated workers and positive among highly educated workers (Sweden)
- No intensification of the negative effects of robot adoption over time

Outlook

- Does structural LM change / adoption of robots cause a reversal in educational gradient in fertility?



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POLSKIE POWROTY
POLISH RETURNS



UNIVERSITY OF WARSAW
Faculty of Economic Sciences



European
Commission

Horizon 2020
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for Research & Innovation



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