

The Peer Effect on Future Wages in the Workplace

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Motivation

Workplace interaction is a critical component of human capital accumulation

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- Peer pressure
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Research questions

1. How much do coworkers contribute to future wages?
2. What are the channels through which this contribution is identified?

Literature review

Coworker quality and wage levels

- Specific workplace (via field experiments) [e.g., Mas and Moretti (2009)]
- Local labor market (using population data) [e.g., Cornelissen et al. (2017); Battisti (2017)]

Coworker quality and wage growth

- Herkenhoff et al. (2018); Jarosch et al. (2019); Nix (2020)
- **Limitation:** use observables (wage or education) as a measure of quality

Firms as learning environments

- Gregory (2019); Arellano-Bover and Saltiel (2021)

Estimation

- Arcidiacono et al. (2012)

Contribution

- Go beyond contemporaneous effects and analyze dynamics

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- Go beyond contemporaneous effects and analyze dynamics
- Provide evidence on the channels that identify this contribution
 1. Peers joining the firm
 2. Peers leaving the firm
 3. Worker moving into different peer group

Effect of peer quality on future wages

Data: Veneto Worker History

Worker records

- ~ 3 million **private-sector** workers from 1982 to 2001 → entire working history
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Contribution (wage) records

- wages from each labor contract, without top coding
- weeks worked, basic contract info (e.g., full/part-time), qualification (e.g. blue, white)

Summary statistics

Empirical strategy

Baseline regression:

$$w_{i,t+h} = \alpha_i + \beta \bar{\alpha}_{-i,t} + \mathbf{x}'_{it} \gamma + \psi_{jt} + \eta_{ot} + \theta_{oj} + \varepsilon_{it}, \quad (1)$$

- $w_{i,t+h}$ is the log weekly earnings at time $t + h$, where $h \geq 0$

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 - ▶ **Peer group**: workers employed in same firm & occupation in a year.

Empirical strategy

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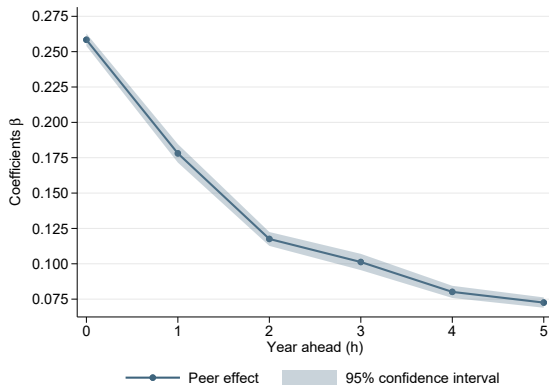
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- $\bar{\alpha}_{-i,t}$ is the *average* coworker's quality at time t
 - ▶ **Peer group:** workers employed in same firm & occupation in a year.
- \mathbf{x}_{it} is a set of individual time-varying characteristics
- $\psi_{jt}, \eta_{ot}, \theta_{oj}$ are firm-year, occupation-year, firm-occupation fixed effects

Baseline results

Effects of coworker's quality on future wages (β)

$$w_{i,t+h} = \alpha_i + \beta \bar{a}_{-i,t} + \mathbf{x}'_{it}\gamma + \phi_{jt} + \delta_{ot} + \theta_{oj} + \varepsilon_{it},$$



Mechanisms that identify β :

An event-study approach

Discussion on potential mechanisms

The effect of coworker quality β is identified through three mechanisms

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We provide an event-study analysis of these job changes

Empirical strategy

When a high/low-quality worker enters/leaves:

- Evolution of coworkers' wages in the **origin** and **destination** firm when worker joins/leaves
- Compare firms hiring/separating from high- and low-quality workers w/ average-quality workers [Diagram](#)
- Ex-ante propensity score matching **at the firm-level** [PS matching hire](#) [PS matching leave](#)

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When a worker joins high-/low-quality peers:

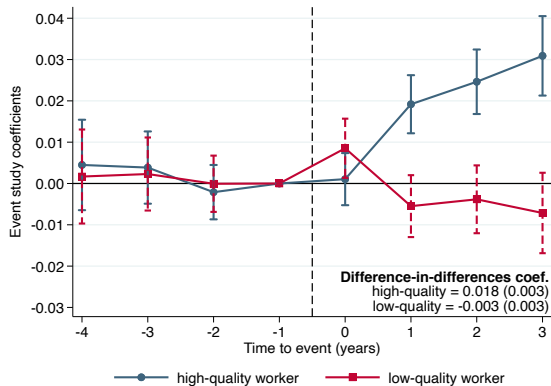
- Evolution of mover's wages when she moves into peer groups of different quality
- Compare workers moving into high- and low-quality peers w.r.t. average-quality peers [Diagram](#)
- Ex-ante propensity score matching **at the worker-level** [PS matching](#)

[Sample sizes](#)

Event-study results

When a high-/low-quality worker enters

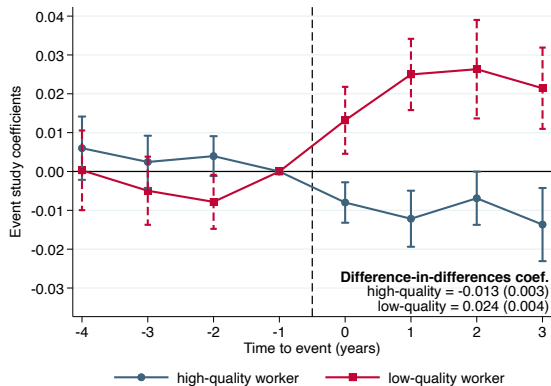
$$w_{-i,t}^{new} = \delta_t + \phi_{j(i)} + \sum_{k \neq -1} \beta_k (\text{Treat}_{j(i)} \times \mathbf{1}\{t = k\}) + \epsilon_{-i,t}$$



Event-study results

When a high-/low-quality worker leaves

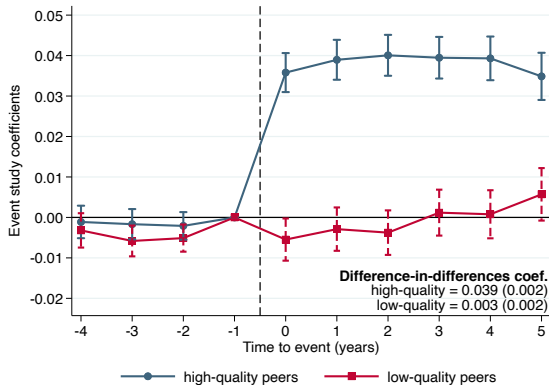
$$w_{-i,t}^{old} = \delta_t + \phi_{j(i)} + \sum_{k \neq -1} \beta_k (\text{Treat}_{j(i)} \times \mathbf{1}\{t = k\}) + \epsilon_{-i,t}$$



Event-study results

When a worker moves into high-/low-quality peer group

$$w_{i,t} = \delta_t + \eta_i + \sum_{k \neq -1} \gamma_k (\text{Treat}_i \times \mathbf{1}\{t = k\}) + \epsilon_{i,t}$$



Conclusion

Key takeaways

- Explore an under-studied component of wage growth: coworker quality
- Coworkers play an important role in generating future wages
- Better peers are associated w/higher wages even after 5 years
- Hiring high-quality workers, separating from low-quality workers and moving into high-quality peers imply the highest wage gains

Next steps: Disentangle the mechanisms (structural)

Data: Summary statistics

	(1) Mean	(2) S.D.	(3) Median
Annual earnings	33350.06	40250.33	31730
Weekly wage	744.38	1652.81	652
Weeks worked	42.41	15.24	52
Age	34.54	10.69	32
Tenure	2.45	2.58	2
Firm size	17	75	6
Movers per firm	4	26	1
Peer group size	12	54	4
Mover	0.61	0.49	
Woman	0.36	0.48	
Blue-collar	0.70	0.46	
Manufacturing	0.53	0.50	
Person-year observations		17,723,260	
Number of workers		2,531,411	
Number of firms		168,613	

Estimation

We follow the methodology developed by Hong and Sørlevstuen (in progress)

Rewrite Equation 1 in a matrix form:

$$w = X\delta + C\delta\beta + \varepsilon$$

where

- $w \in \mathbb{R}^n$ and $X, C \in \mathbb{R}^{n \times k}$ are observed: the construction of the coworker matrix C
- $\delta \in \mathbb{R}^k$ and $\beta \in \mathcal{B}$, where \mathcal{B} is compact.

Assumptions

- exogeneity, $\mathbb{E}[\varepsilon|X, C] = \mathbf{0}$,
- homoskedasticity, $\mathbb{E}[\varepsilon\varepsilon'|X, C] = \sigma^2 I_n$ where $\sigma^2 > 0$ is unknown.

Estimation

The non-linear least squares estimator β is as follows.

$$\hat{\beta} = \min_{\beta \in B} Q_n(\beta) = \arg \min_{\beta \in B} \left\{ \min_{\delta \in \mathbb{R}^k} \|w - X\delta - C\delta\beta\|^2/n \right\} \quad (2)$$

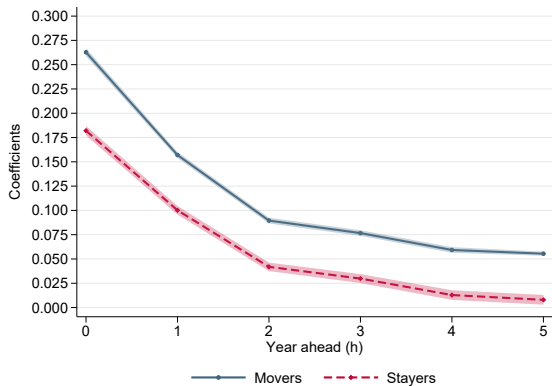
- β is the unique minimizer of sample analogue to $Q_n \rightarrow \sqrt{\cdot}$ consistency (ensured by the exogeneity and homoskedasticity assumptions)
- Taking FOC of δ and β in Equation 2 \Rightarrow moment condition for β

$$S_n(\beta) = w' M C (R' R)^{-1} R' w / n = 0,$$

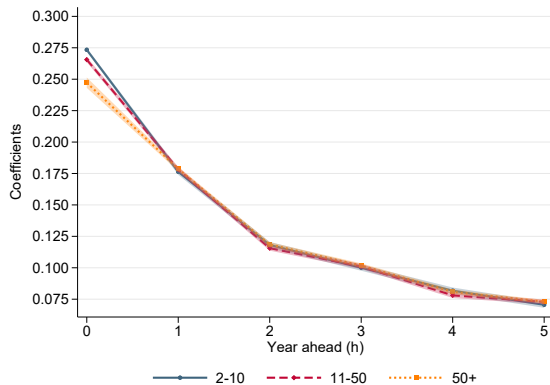
$$R = X + C\beta$$

$$M = I_n - P, \text{ and } P = R (R' R)^{-1} R'$$

Heterogeneous effects/1

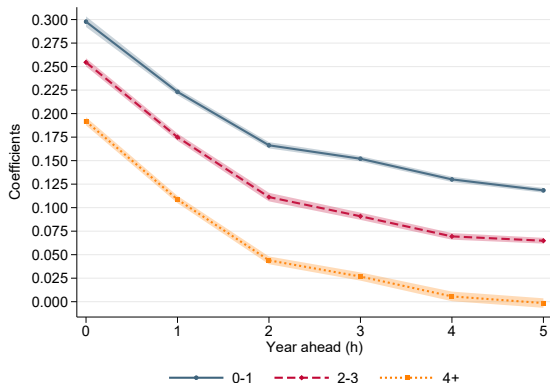


(a) Movers and stayers

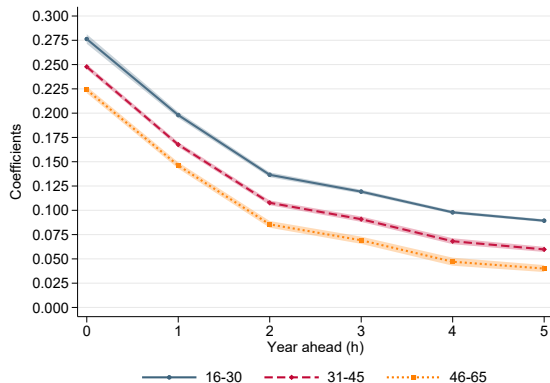


(b) Firm size

Heterogeneous effects/2



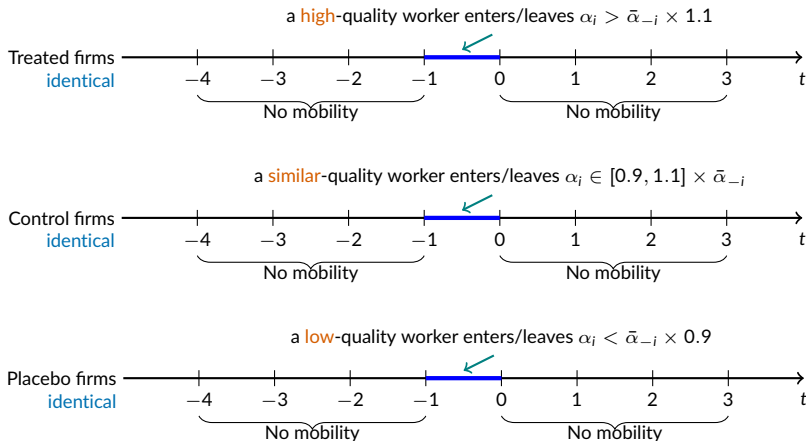
(a) Tenure brackets



(b) Age brackets

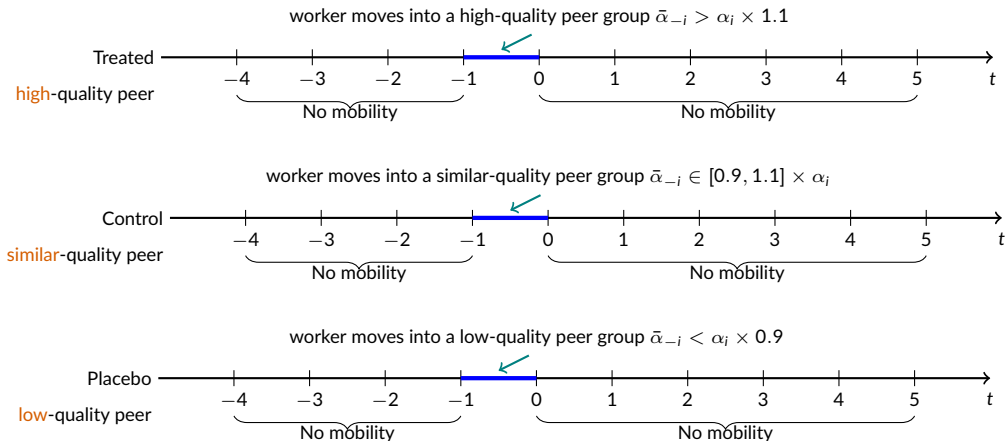
Empirical strategy

When a high/low-quality worker enters/leaves



Empirical strategy

When a worker joins high-/low-quality peers



Event-study

Sample sizes

	Treatment	Matched control	Total observations
Panel A: Hiring design	(# of firms)	(# of firms)	(person-year)
<i>High-quality worker</i>	2,164	2,164	285,350
<i>Low-quality worker</i>	1,848	1,848	238,046
Panel B: Leaver design	(# of firms)	(# of firms)	(person-year)
<i>High-quality worker</i>	2,905	2,905	390,135
<i>Low-quality worker</i>	1,885	1,885	234,016
Panel C: Mover design	(# of workers)	(# of workers)	(person-year)
<i>High-quality peers</i>	15,551	15,551	310,220
<i>Low-quality peers</i>	12,778	12,778	255,560

The construction of the coworker matrix C

A (simple) example of the coworker matrix C :

- 5 workers, 2 peer groups, and 1 period: data = $\begin{pmatrix} 1 & 1 \\ 2 & 1 \\ 3 & 2 \\ 4 & 2 \\ 5 & 2 \end{pmatrix}$
- The averaging matrix $\tilde{C} = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0.5 & 0.5 \\ 0 & 0 & 0.5 & 0 & 0.5 \\ 0 & 0 & 0.5 & 0.5 & 0 \end{pmatrix}$
- $C = [\tilde{C}, \mathbf{0}]$ so that C has the same dimension as X .

See the next (backup) slide for more details.

The construction of the coworker matrix C

For each worker i at time t (for convenience, we suppress it to i)

$$c_i = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \\ \vdots \\ 1 \\ \vdots \end{pmatrix} - \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \\ \vdots \\ 0 \\ \vdots \end{pmatrix}$$

The averaging matrix \tilde{C} and the coworker matrix C are constructed as follows.

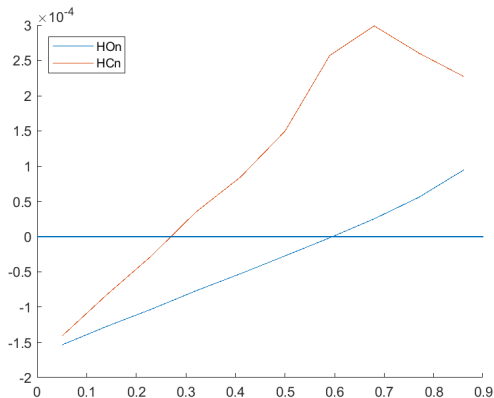
$$\tilde{c}_i = \frac{c_i}{c_i' \cdot \vec{1}} \Rightarrow \tilde{C} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix} \Rightarrow C = \begin{pmatrix} \tilde{C} \\ 0 \end{pmatrix}$$

where the matrix 0 has the same size as F and other covariates

Heteroskedasticity estimator

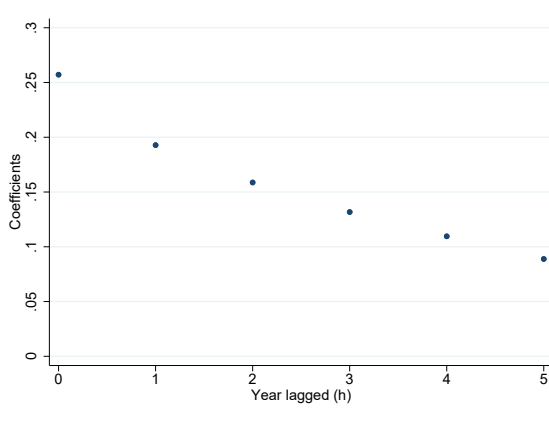
$$w_{it} = \alpha_i + \beta \bar{\alpha}_{-i,t} + \phi_j + \mathbf{x}_{it}'\gamma + \varepsilon_{it},$$

Figure 1: Behaviors of the objective functions of HOn and HCn



Alternative baseline specification

$$w_{i,t} = \alpha_i + \beta \bar{\alpha}_{-i,t-h} + \mathbf{x}_{it}'\gamma + \phi_{jt} + \delta_{ot} + \theta_{oj} + \varepsilon_{it},$$



Occupations with low/high learning content

based on Cornelissen et al. (2017) Table F.3

Occupations with low learning content

- Land transport drivers
- Cargo handling and storage
- Laundry, dry cleaning
- Waste disposal
- Packers
- Typists
- Sales persons
- Ready-meal preparers
- Restaurant waiters
- Construction machine attendants
- ...

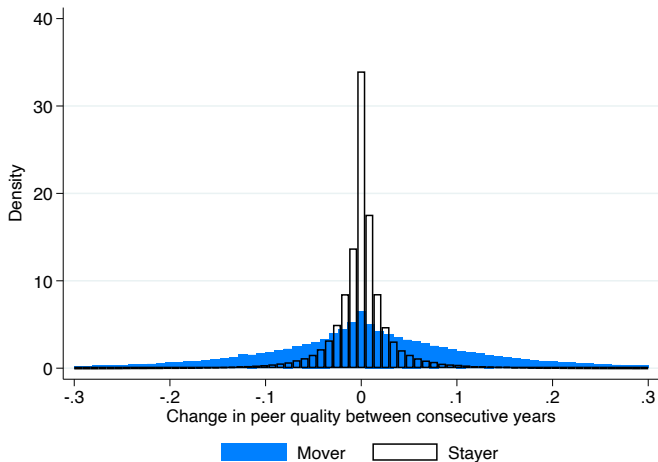
Occupations with high learning content

- Research in sciences/engineering
- Legal services
- Accounting and tax consultants
- Architects
- Entrepreneurs
- Medical services
- Dentists
- Pharmacist
- Civil engineers
- ...

Summary statistics after estimation

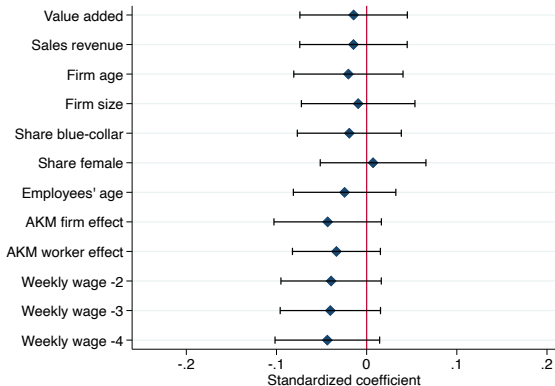
Statistic	Value
Standard deviation log weekly wages	0.436
Standard deviation worker fixed effect	0.269
Standard deviation peer fixed effect	0.178
Standard deviation occupation-time fixed effect	0.065
Standard deviation firm-occupation fixed effect	0.103
Standard deviation firm-time fixed effect	0.137
Standard deviation change of peer fixed effect between t and $t - 1$	0.090
Standard deviation change of peer fixed effect between t and $t - 1$ for movers	0.173
Standard deviation change of peer fixed effect between t and $t - 1$ for stayers	0.066
Correlation worker fixed effect/peer fixed effect	0.551

Summary statistics after estimation

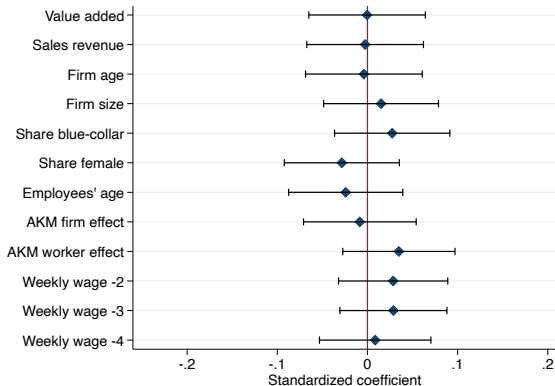


Propensity score matching

Balance test of covariates ($t = -3$) - Hiring design



Matched Control and Treatment 1



Matched Control and Treatment 2

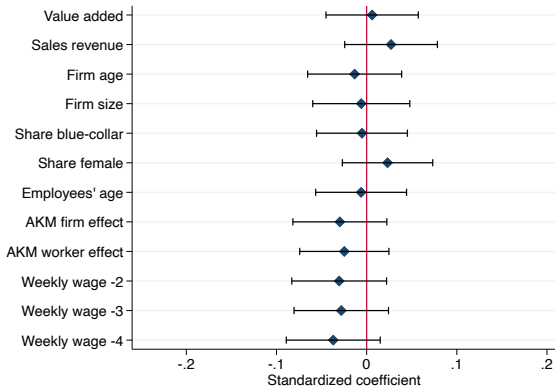
[Balance test unmatched](#)

[Common support](#)

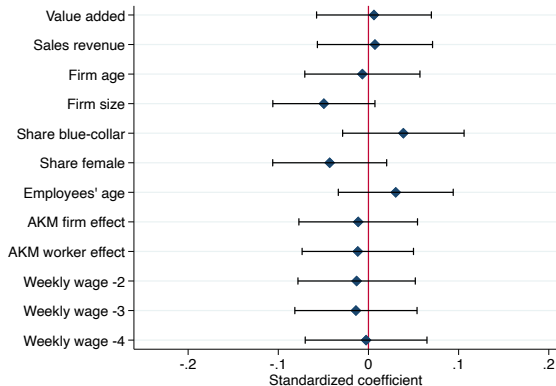
[Back to ES Intro](#)

Propensity score matching

Balance test of covariates ($t = -3$) - Leaver design



Matched Control and Treatment 1



Matched Control and Treatment 2

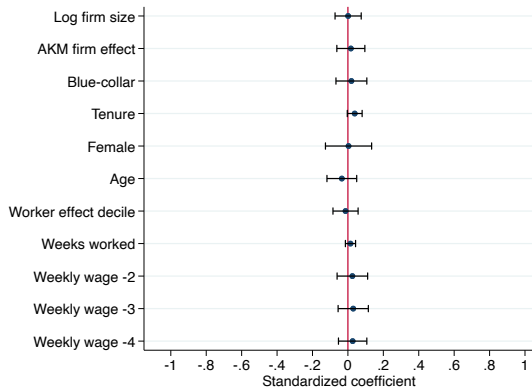
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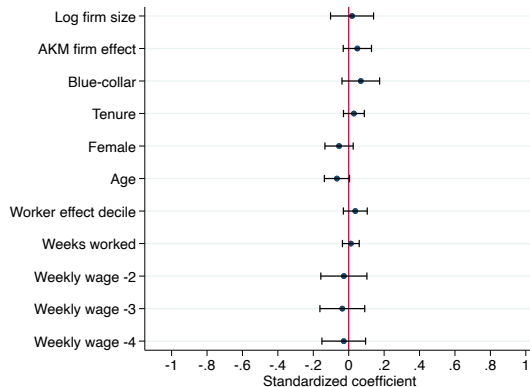
[Back to ES Intro](#)

Propensity score matching

Balance test of covariates ($t = -3$) - Mover design



Matched Control and Treatment 1



Matched Control and Treatment 2

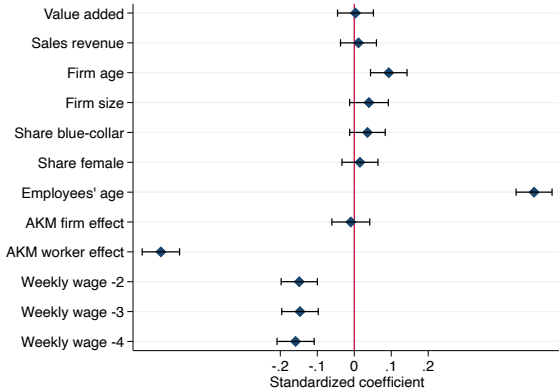
[Balance test unmatched](#)

[Common support](#)

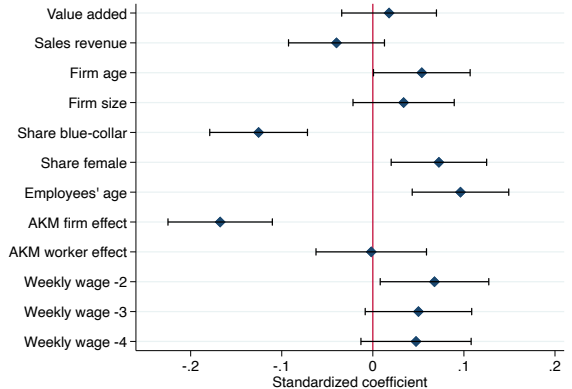
[Back to ES worker](#)

Propensity score matching

Hiring design



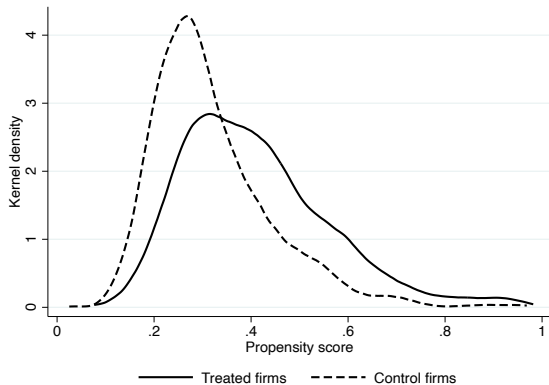
[Unmatched Control and Treatment 1]



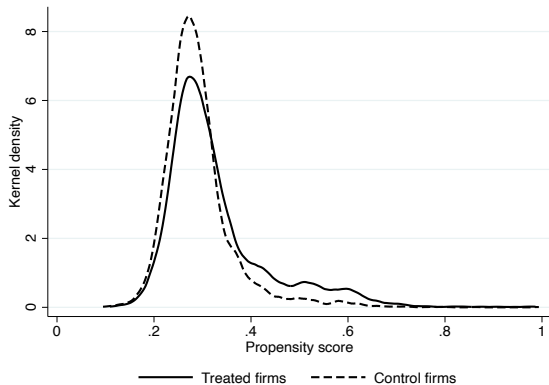
[Unmatched Control and Treatment 2]

Propensity score matching

Hiring design



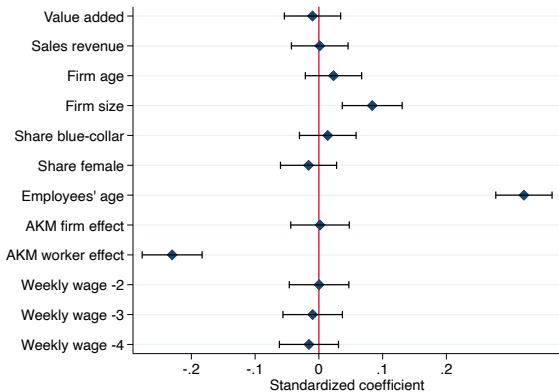
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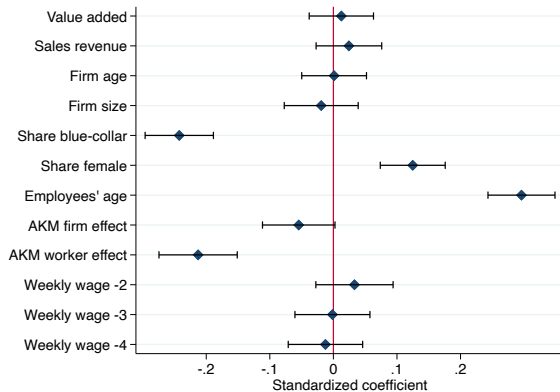
[Matched Control and Treatment 2]

Propensity score matching

Leaver design



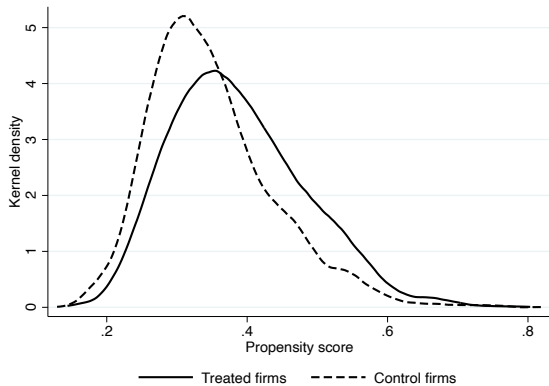
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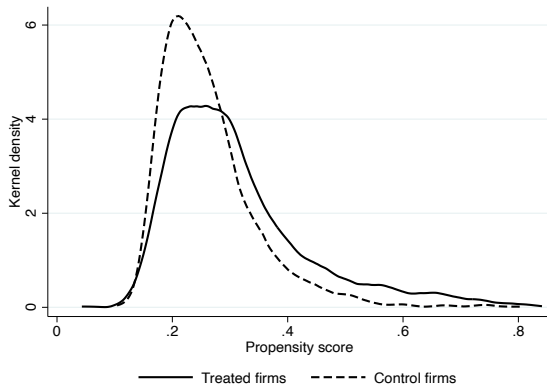
[Unmatched Control and Treatment 2]

Propensity score matching

Leaver design



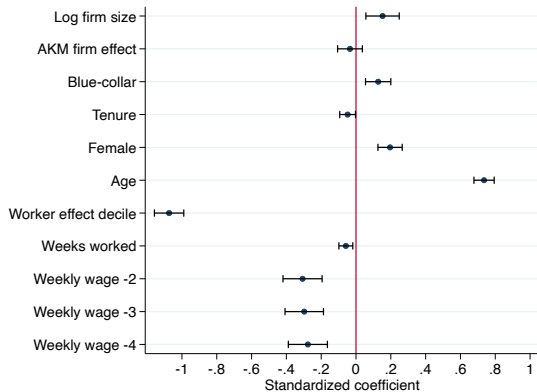
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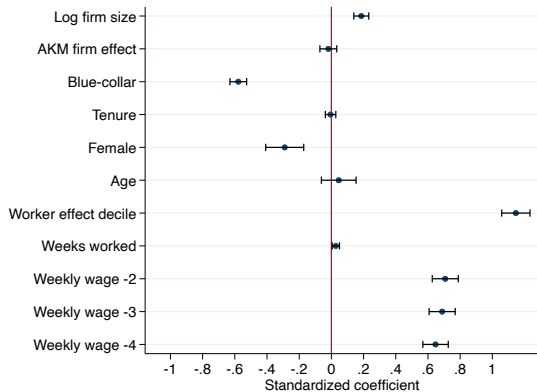
[Matched Control and Treatment 2]

Propensity score matching

Worker-level design



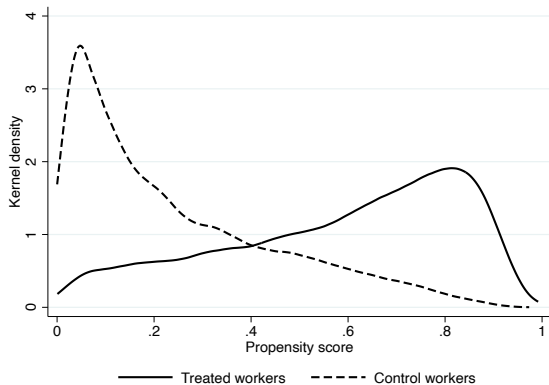
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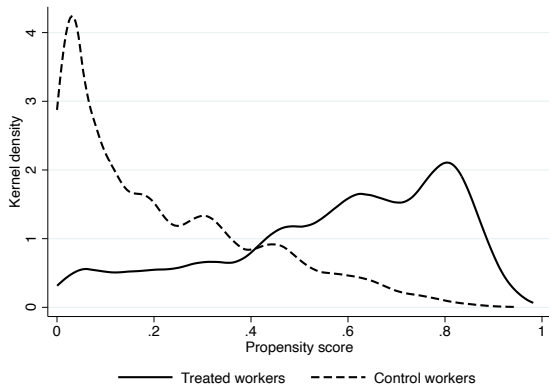
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Propensity score matching

Worker-level design



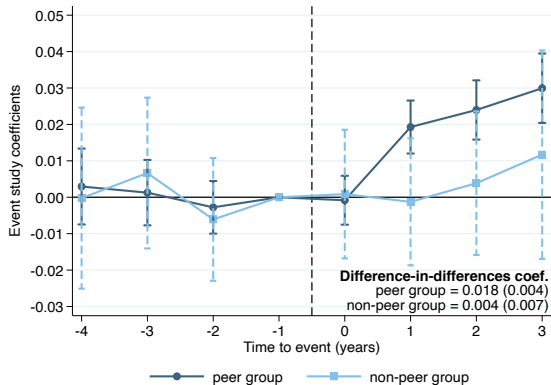
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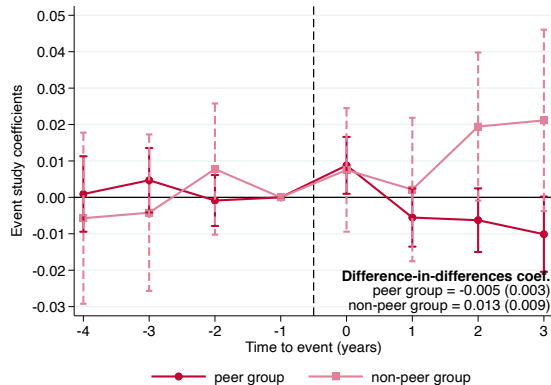
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Heterogeneity - Hire design

Peers vs non-peers



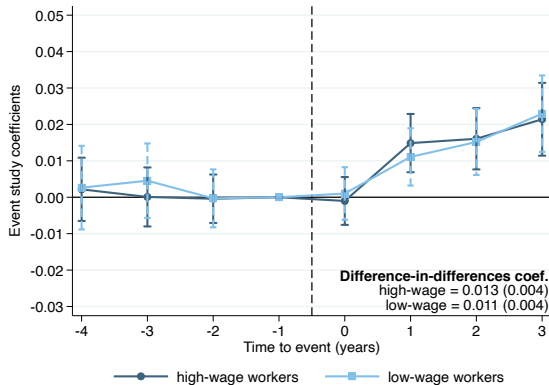
(a) High-quality entry



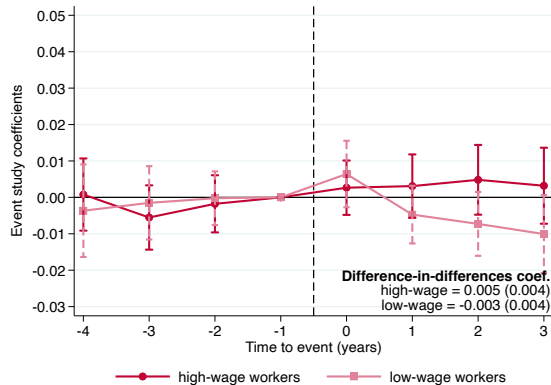
(b) Low-quality entry

Heterogeneity - Hire design

High- vs low-wage workers



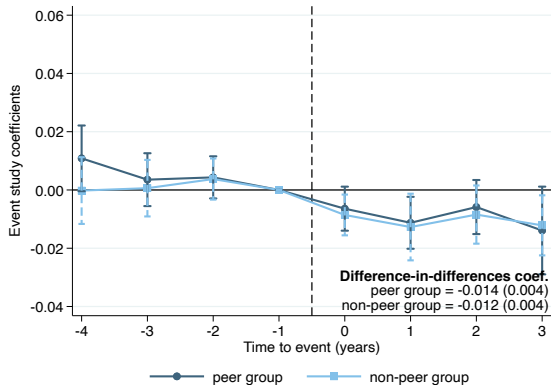
(c) High-quality entry



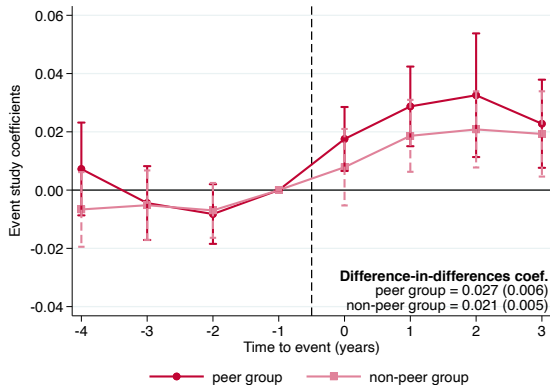
(d) Low-quality entry

Heterogeneity - Leaver design

Peers vs non-peers



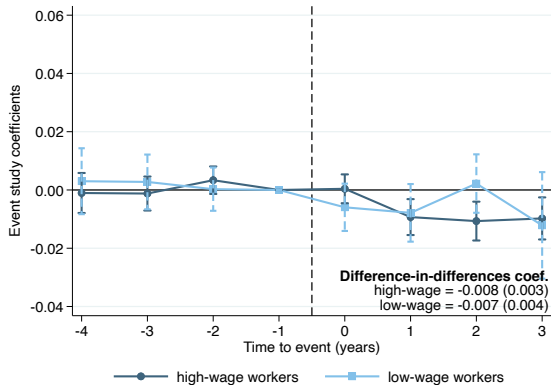
(a) High-quality leave



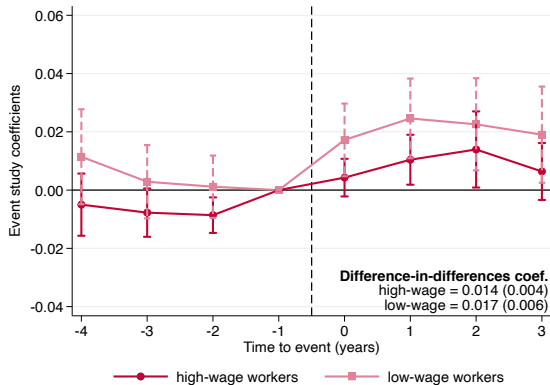
(b) Low-quality leave

Heterogeneity - Leaver design

High- vs low-wage workers



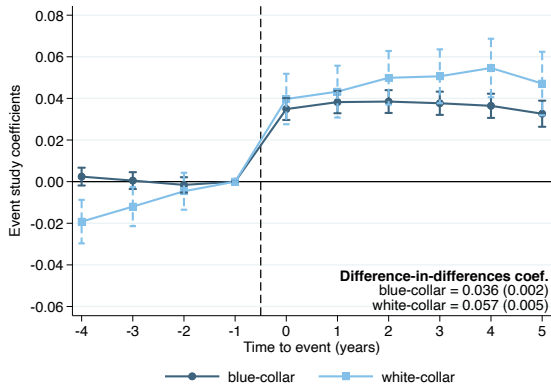
(c) High-quality leave



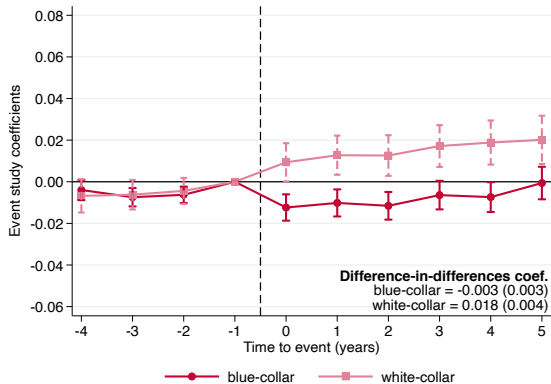
(d) Low-quality leave

Heterogeneity - Mover design

Blue- vs white-collar workers



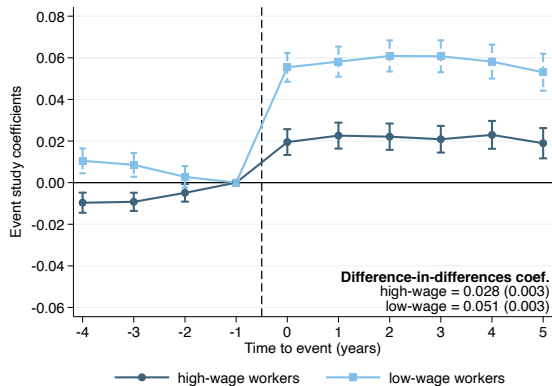
(a) High-quality mover



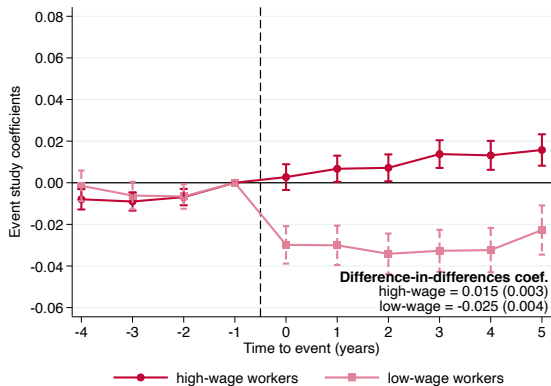
(b) Low-quality mover

Heterogeneity - Mover design

High- vs low-wage workers



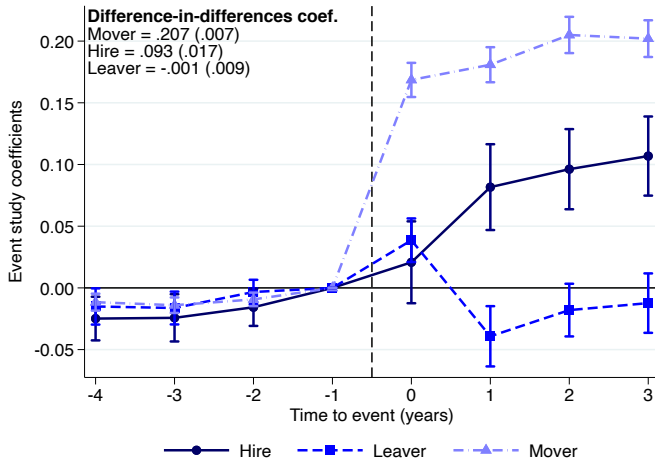
(a) High-quality mover



(b) Low-quality mover

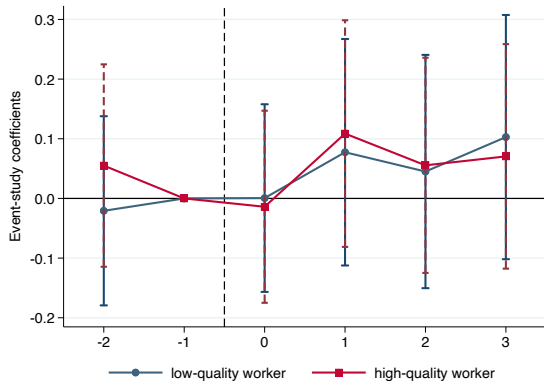
Event study - Robustness

Continuous treatment

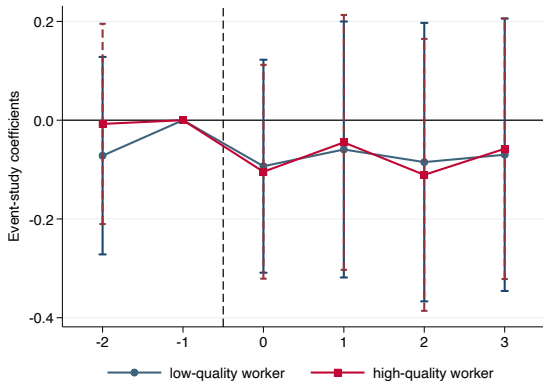


Event study - Robustness

Firms' value added and sales



(a) Sales per worker



(b) Value added per worker