

# Quantifying Racial Disparities Using Consecutive Employment Spells

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## Motivation and question

Black-white disparities in unemployment and earnings [e.g., Bound and Freeman (1992), Bayer and Charles (2018)]

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How large are the gaps among equally-productive workers?

- ▶ Hard to get to **quantities** from audit studies
- ▶ Hard to generate (natural) experiments to study race [e.g., Charles and Guryan (2011)]
- ▶ Hard to deal with **unobservables** in observational data

## Idea of the paper

Use employer learning to match Black and white workers on unobservables:

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Develop an equilibrium model of learning and turnover

- ▶ Discuss mechanisms/assumptions/how to label this gap



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- ▶ Separations: first spell gaps re-emerge in the second spell
  - ▶ Over 20% of **unconditional** separation gaps
  - ▶ About 0.6% of lifetime consumption



# Outline

1. Data description, coding, and samples
2. Separation gap: heterogeneity and by tenure
3. A model of employer learning + turnover
4. Outcomes in the second spell
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- ▶ Race/ethnicity from Census and ACS
  - ▶ Non-hispanic Black and white
- ▶ Age: 18 - 61 (inclusive)

## Coding separations

- ▶ Separate from  $j$  if earnings in quarter  $t$  and no earnings in quarter  $t + 1$  to  $t + 4$
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### Other data handling steps:

- ▶ Use *full* quarter employment relationships (employed in  $t - 1$ ,  $t$ , and  $t + 1$ )
  - ▶ Quarter  $t$  also known as a *sandwich* quarter
- ▶ Impose earnings floor (annualized \$3250 in \$2011 using CPI-U)

## Match on tenure and earnings

**Matched sample:** Black workers who have white co-workers in bins defined by interaction of:

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Other matching steps:

- ▶ Reweight white workers to match Black workers' distribution (nonparametric propensity score)

# Samples

Sample	Age	Female (%)	Tenure	Earn	Workers	Firms	Worker-Q
<b>All workers</b>							
All	40.6	49.4	22.2	10.74	44,960,000	2,540,000	1,140,000,000

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White	41.2	48.9	23.5	10.79	31,200,000	2,270,000	844,000,000

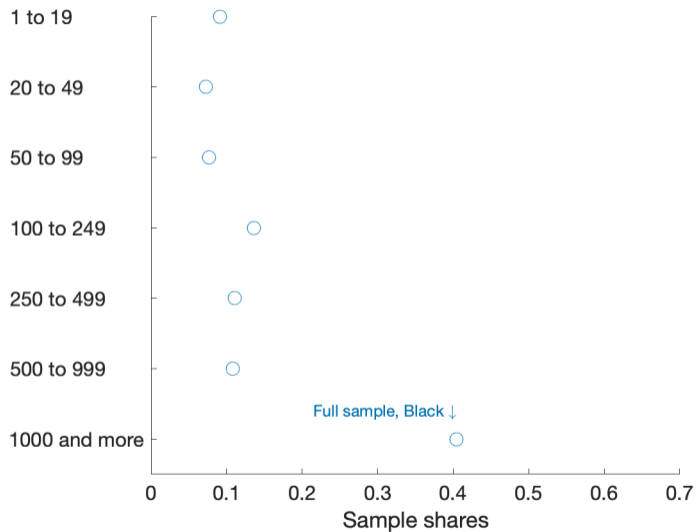
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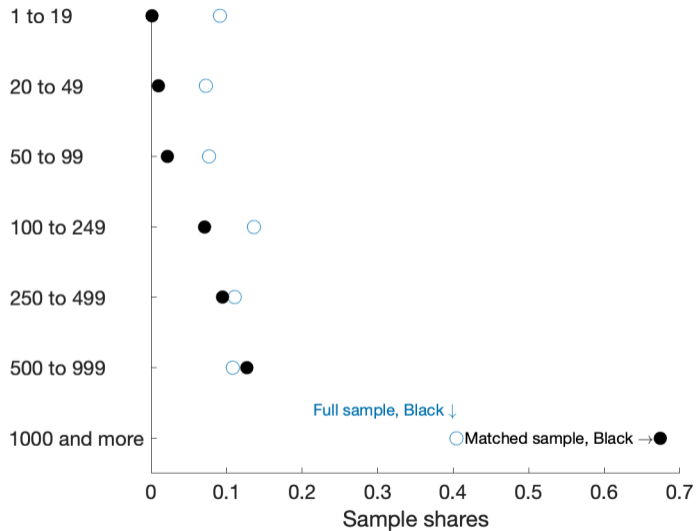
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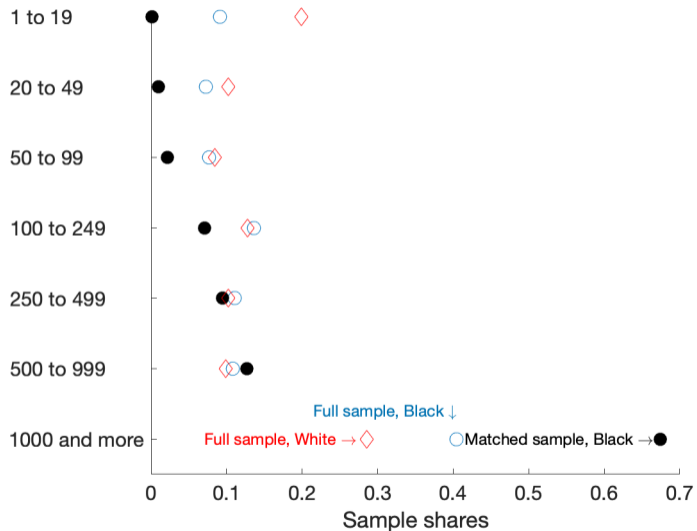




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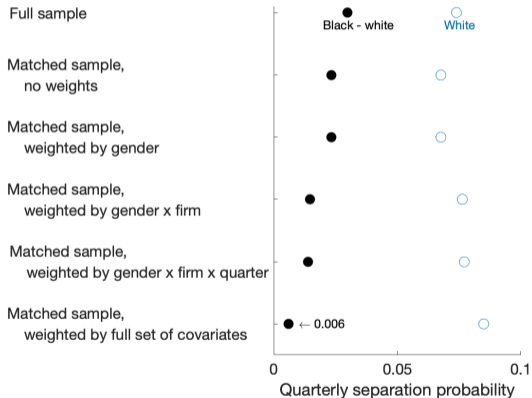
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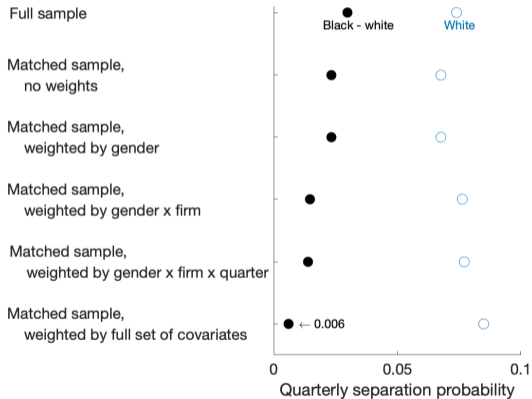
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# Separation gaps: full sample to matched and weighted

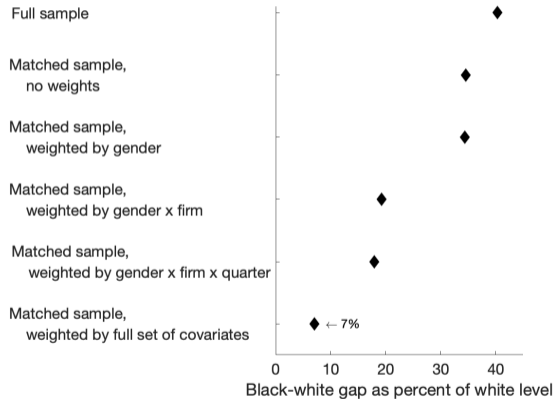


Full set of covariates: firm-quarter and state-quarter earnings and tenure deciles, and firm x gender x quarter

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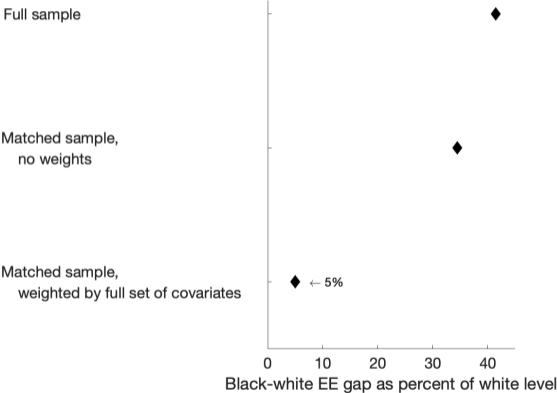


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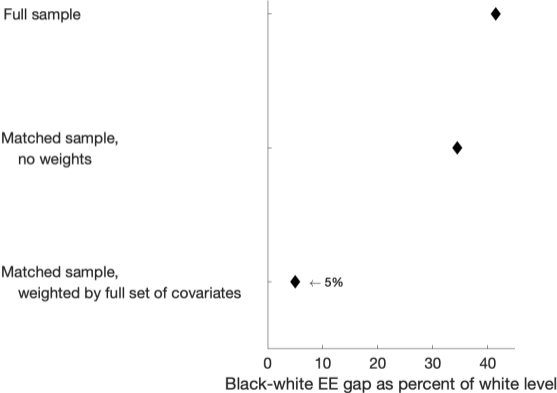
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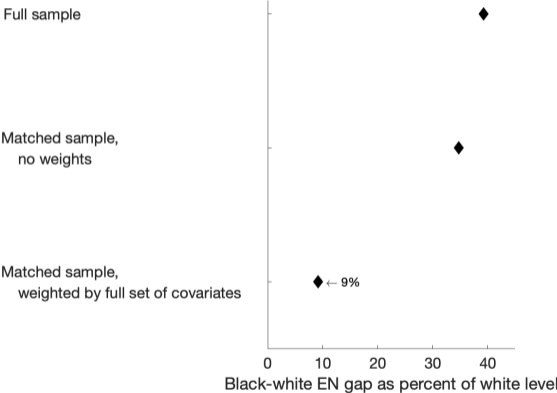
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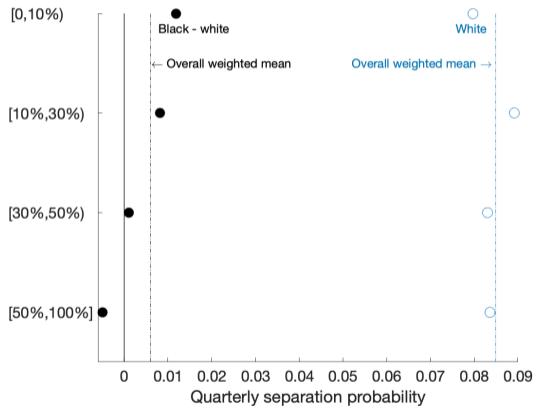
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## Employer to nonemployment (EN)



# Separation gap heterogeneity

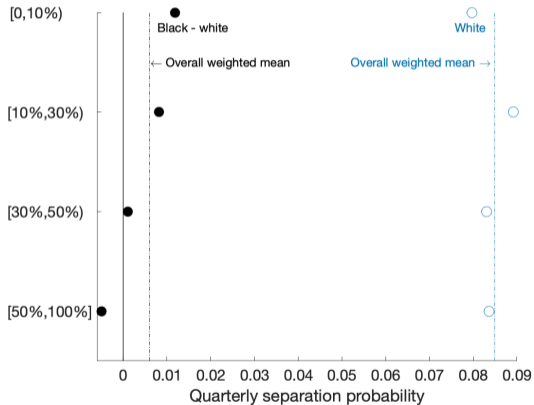
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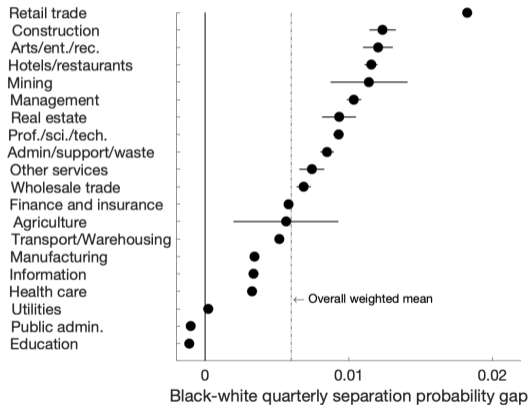
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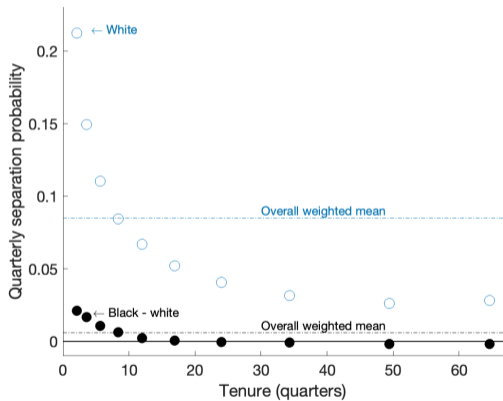
▶ Employment shares    ▶ By employer size

Sector

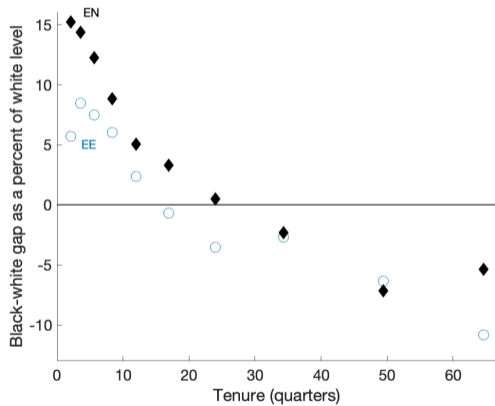
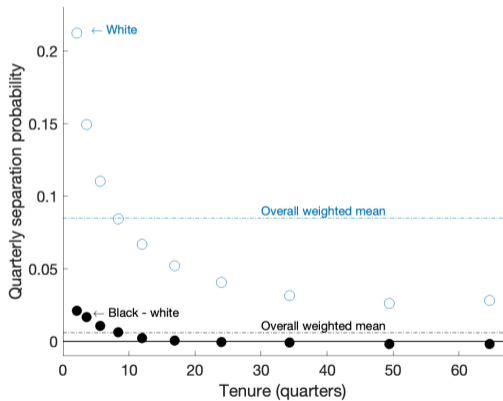


▶ Employment shares    ▶ By gender

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## Set-up

Group differences:

- ▶ Observable groups:  $g \in \{c, d\}$
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To keep (the model solver's) life simple:

- ▶ Workers live for two employment spells, terminal payoff is market's view of their type



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First employer:

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Market/second employer:

- ▶ Observe tenure (and employer identity) with the first employer
  - ▶ Use tenure (along with group identity) to infer worker type

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Conjecture: if (history of) wages do not fully reveal firm's information, results go through

- ▶ Hard to generate (conditional) separation gaps if wages convey firm's information

## Separation rates by type and group in the first spell

Separation rate for worker known to be high/low productivity:

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- ▶  $\Rightarrow$  High-enough tenure = group separation rates converge

## Implications for the second spell

Market's belief given  $t$  periods of tenure in the first spell and group  $g$ :

$$\tilde{p}(g, t) = \frac{\alpha^g (1 - s_h)^{t-1} s_h}{\alpha^g (1 - s_h)^{t-1} s_h + (1 - \alpha^g) (1 - s_l)^{t-1} s_l}$$

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For “high enough” tenure:

- ▶ Workers are matched on unobservables
- ▶  $\Rightarrow$  in the “second spell,” outcomes shouldn't depend on group identity
- ▶ Any gaps in earnings or separations are among equally productive workers

## Key assumptions

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What happens if these assumptions do not hold?

- ▶ Failure of first: conceptual idea falls apart
  - ▶ Heterogeneity/mediation is hard to reconcile with simple “second unobservable” stories

## Key assumptions

1. No “second unobservable”: what first employer learns is relevant to second employer
2. Equal treatment: first firm is only engaged in rational employer learning

What happens if these assumptions do not hold?

- ▶ Failure of first: conceptual idea falls apart
  - ▶ Heterogeneity/mediation is hard to reconcile with simple “second unobservable” stories
- ▶ Failure of the second:
  - ▶ Upper or lower bound: if employer acts biased against Black (lower) or white workers (upper)

# Outline

1. Data description, coding, and samples
2. Separation gap: heterogeneity and by tenure
3. A model of employer learning + turnover
4. **Outcomes in the second spell**
5. Welfare calculation



## Summary statistics

Sample	Age	Female (%)	Tenure	Earn	Workers
<b>All workers</b>					
Black	39.8	55.5	19.2	10.54	5,145,000
White	41.2	48.9	23.5	10.79	31,200,000
<b>High-tenure* workers</b>					
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<b>High-tenure matched and reweighted separators, mass layoff</b>					
Black				11.04	2,600
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\* Top 3 deciles of state-year-quarter distribution, AND 20 or more quarters of tenure

\*\* Match on current and lagged quarter

\*\*\* If a worker goes A to B, then only a separation if no more than 20% of workers at A go to B AND no more than 20% of B's workers joined from A

## Methods: how this displaced worker analysis differs from others

$$y_{ik} = \beta_{0,k} + \beta_{1,k}Black_{ik} + \epsilon_{ik},$$

- ▶  $k$  is horizon relative to separation (negative), and finding post-separation job (positive)
- ▶ Post-period is only in first post-separation job
- ▶ Matched sample (reweight at each horizon), conditional on mass layoff in same quarter

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Four ways in which this differs from conventional displaced worker papers:

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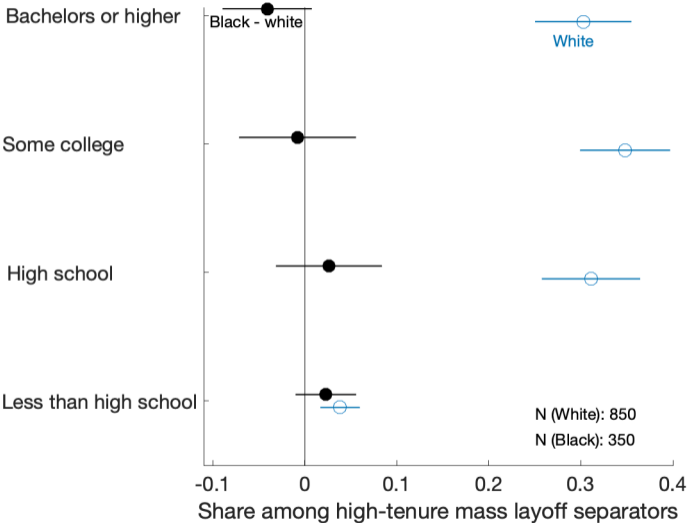
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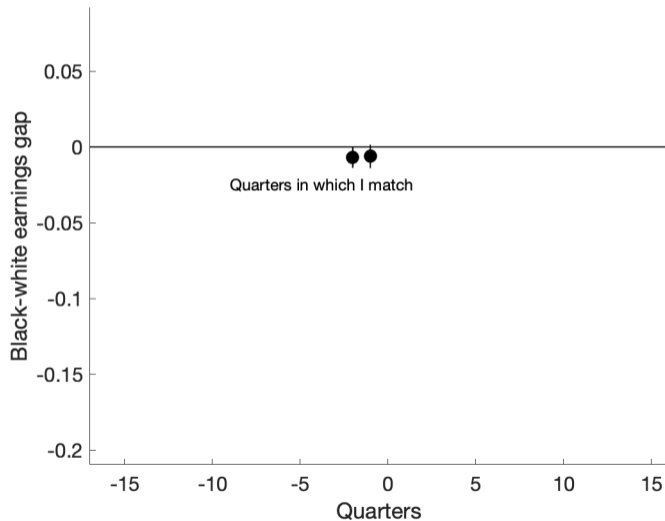
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4. Timing in the post-period is relative to getting first post-separation job

# Separators are approximately balanced on (non-imputed) education

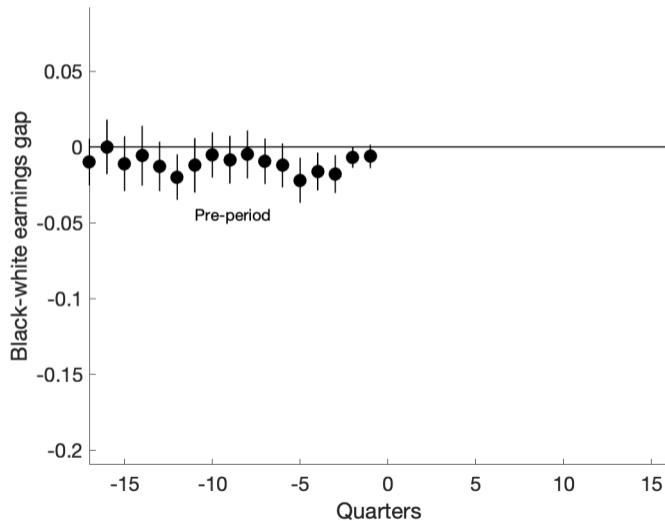




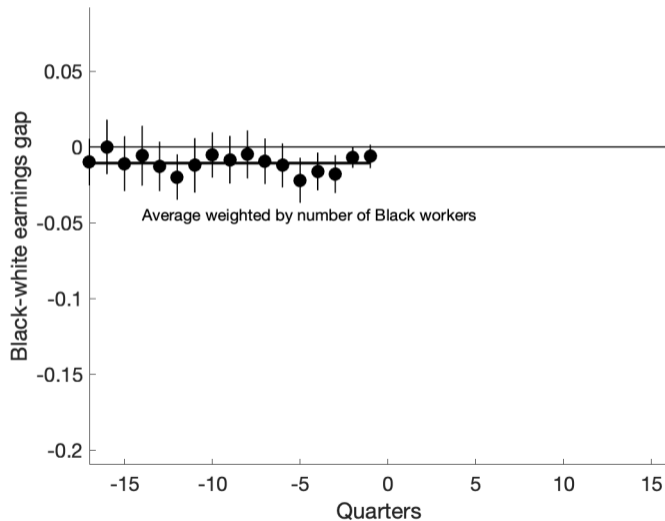
## Earnings gaps among high-tenure matched mass layoff separators



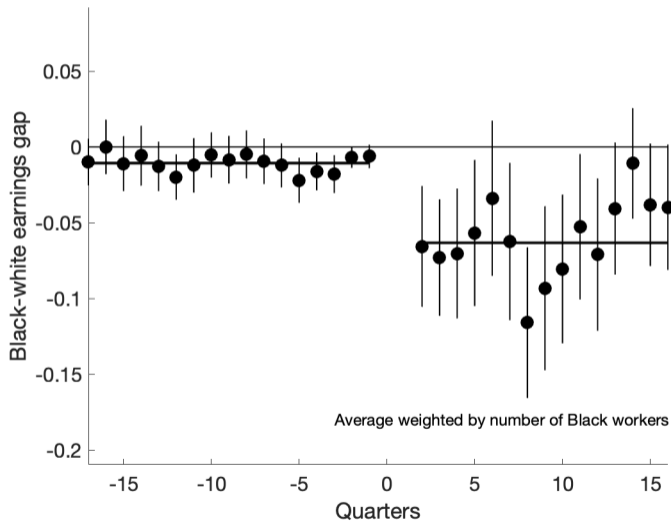
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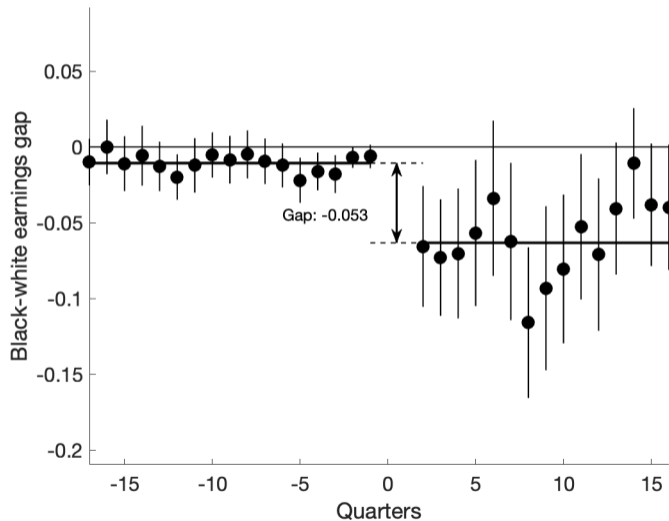
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## Levels and gaps of mediating outcomes

### Post-separation firm characteristics

---

	Quarters b/w jobs
Black-white gap	<b>0.16</b> (0.09)
White	2.19 (0.08)
<hr/>	
N (Black)	2,600
N (White)	6,100

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		Post-separation firm characteristics
	Quarters b/w jobs	Same sector
Black-white gap	<b>0.16</b> (0.09)	<b>-0.015</b> (0.009)
White	2.19 (0.08)	0.475 (0.012)
N (Black)	2,600	2600
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Characteristics computed the quarter before the worker joins.

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Characteristics computed the quarter before the worker joins.



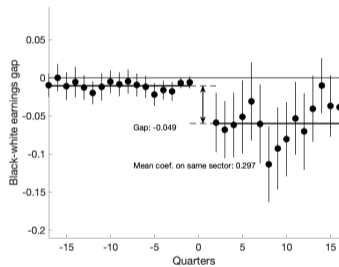
## Levels and gaps of mediating outcomes

		Post-separation firm characteristics		
	Quarters b/w jobs	Same sector	Mean firm earnings	Share Black
Black-white gap	<b>0.16</b> (0.09)	<b>-0.015</b> (0.009)	<b>-0.023</b> (0.014)	<b>0.075</b> (0.004)
White	2.19 (0.08)	0.475 (0.012)	10.76 (0.015)	0.136 (0.003)
N (Black)	2,600	2600	2600	2600
N (White)	6,100	6100	6100	6100

Characteristics computed the quarter before the worker joins.

# Event study: controlling for...

Same sector

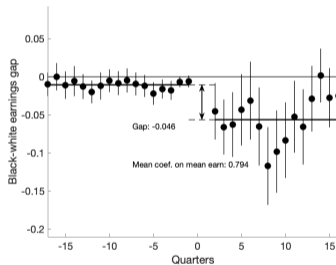
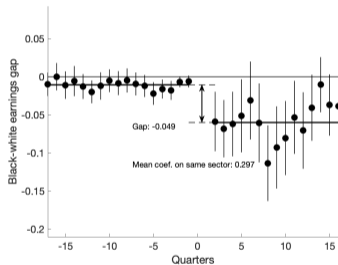


Gap: -0.049

# Event study: controlling for...

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Mean firm earnings

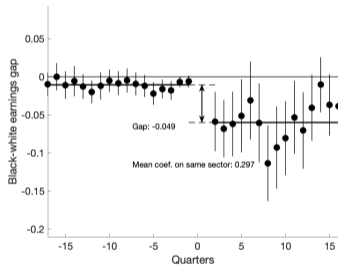


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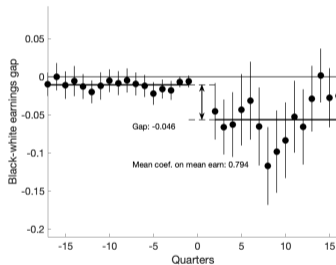
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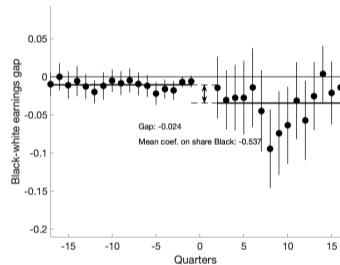
► Coefficients on controls

## Mean firm earnings



Gap: -0.046

## Share black



Gap: -0.024

## Taking stock

What does share of Black workers proxy for?

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What is the remaining half?

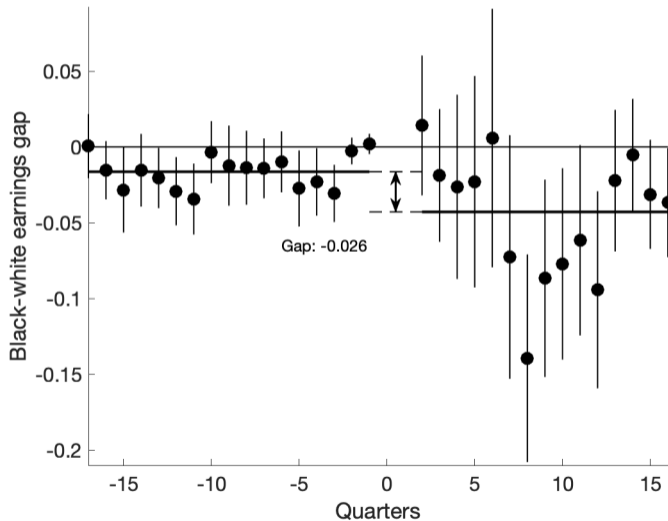
- ▶ Other between-firm sorting?
- ▶ Within-firm?



# Samples

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<b>High-tenure matched separators, mass layoff, same second firm (same quarter)</b>					
Black				11.16	800
White				11.16	2,700

## Event study: same second firm (in the same quarter)



# Interpretation

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Summing up:

- ▶ 5.3 log point gap
  - ▶ Closes to 2.4 log points controlling for share of Black workers (“half is between firm, mediated by share of Black workers”)
  - ▶ Gap of 2.6 log points among workers joining same second firm (“half is within firm”)

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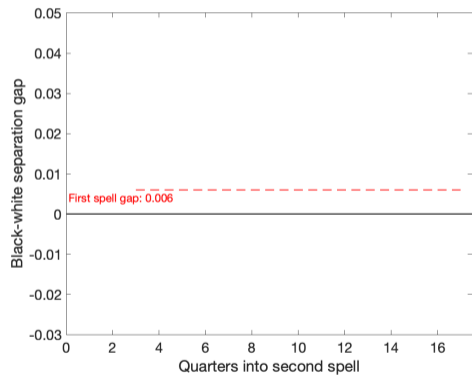
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Second unobservable:

- ▶ Same sector rules out large role for one observable form
- ▶ Share of Black workers labels between-firm—not obviously about productivity
- ▶ Within-firm holds fixed technology

# Gaps in separations

All

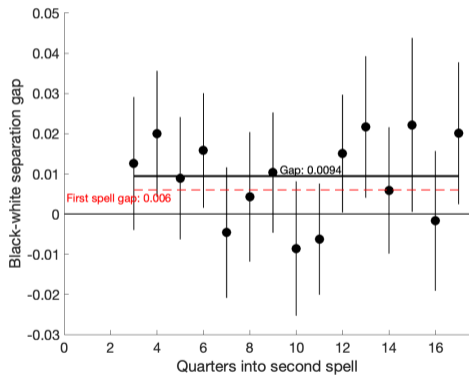


▶ Sample counts



# Gaps in separations

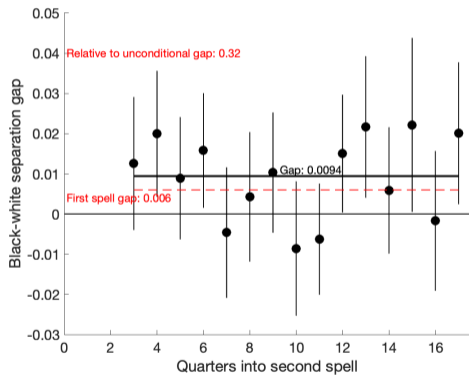
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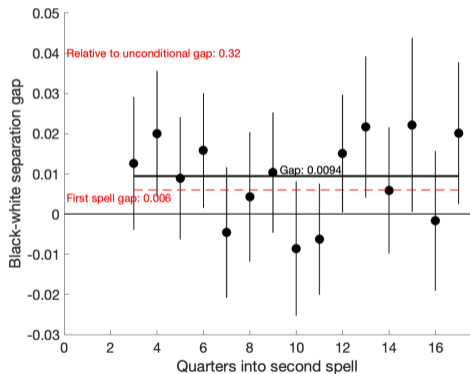
All



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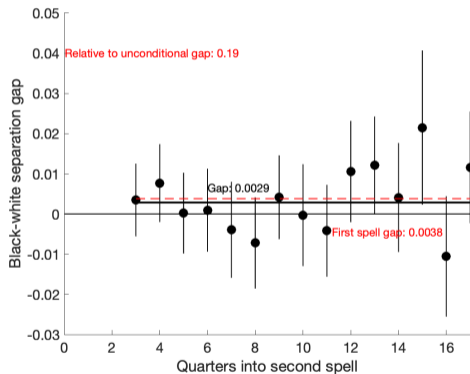
# Gaps in separations

All



▶ Sample counts

EN



▶ White level

# Outline

1. Data description, coding, and samples
2. Separation gap: heterogeneity and by tenure
3. A model of employer learning + turnover
4. Outcomes in the second spell
5. **Welfare calculation** [▶ Skip to end](#)

# Set-up

Search block:

- ▶ Workers are born unemployed
- ▶ Flow payoff to unemployment is  $b$
- ▶ Find a job with probability  $\lambda$
- ▶ All jobs pay  $w$
- ▶ Workers live for  $A$  periods

Consumption (and curvature) block:

- ▶ No borrowing or saving
- ▶ Period utility is  $u(\cdot)$  (CRRA, with coefficient  $\gamma$ )

Group difference:

- ▶ Probability that a job is destroyed depends on group membership and tenure:  $\delta_t^g$

## Value functions and welfare calculation

Employed worker:

$$\underbrace{W(g, t, a)}_{\text{value of a job}} = \underbrace{u(w)}_{\text{flow payoff}} + \underbrace{\beta \delta_t^g U(g, a + 1)}_{\text{lose job}} + \underbrace{\beta (1 - \delta_t^g) W(g, t + 1, a + 1)}_{\text{keep job}}$$

Unemployed worker:

$$\underbrace{U(g, a)}_{\text{value of u/e}} = \underbrace{u(b)}_{\text{flow payoff}} + \underbrace{\beta \lambda W(g, 0, a + 1)}_{\text{find a job}} + \underbrace{\beta (1 - \lambda) U(g, a + 1)}_{\text{remain u/e}}$$

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Consumption equivalent, solve for  $c^g$  such that:

$$U(g, 0) = \sum_{a=1}^A \beta^{a-1} u(c^g)$$

Compare  $c^w$  and  $c^b$

## Model parameters

Parameter	Description	Value	Source
$\beta$	Discounter	$0.95^{\frac{1}{4}}$	Convention
$\gamma$	CRRA curvature	1.5	Low, Meghir, Pistaferri (2010)
$b$	Flow value of u/e	0.4	Chodorow-Reich and Karabarbounis (2016)
$\lambda$	0.61	Job finding	Black rate, 2003-2019
$\delta^g$	Job loss probability		This paper ("second spell")



## Model results

		Black-white gaps	
	Unemployment (p.p)	PDV of cons. (%)	Certain cons.-equivalent (%)
Baseline (EN only)	0.5	-0.3	-0.6
All separations	2.0	-1.2	-2.1
$b = 0.65$	0.5	-0.2	-0.2
$b = 0.9$	0.5	-0.2	-0.2
$f = .3757$	0.4	-0.3	-0.5
$\gamma = 4$	0.5	-0.3	-1.3

## Summary

Some of Black-white disparities in unemployment and earnings are discrimination

- ▶ Hard to get to **quantities** from audit studies
- ▶ Hard to deal with **unobservables** in observational data

This paper: at high-enough tenure, firms have learned about worker unobservables

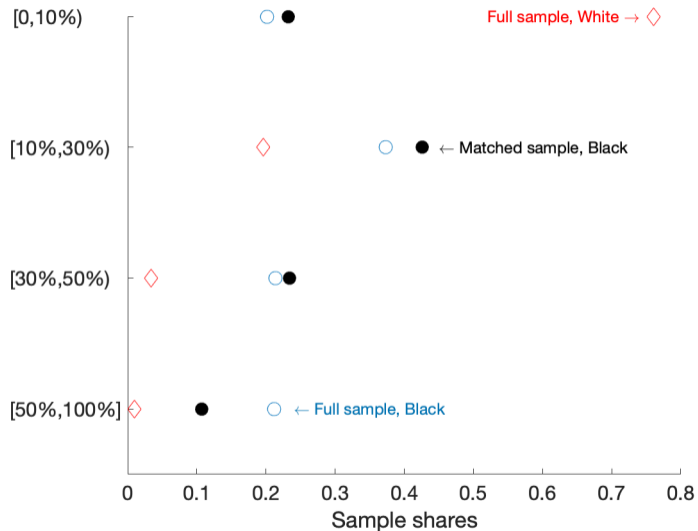
- ▶ What happens in the next job?

Results:

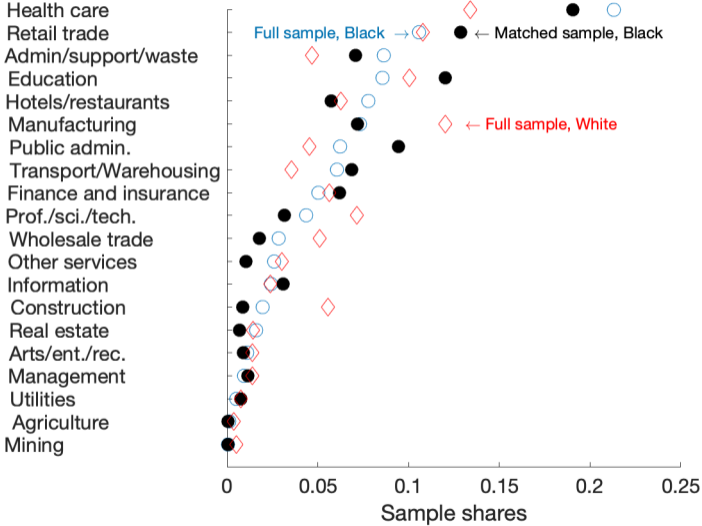
- ▶ Earnings gaps: 5.3 log points (compared to 16 log point gap among high-tenure workers)
  - ▶ About half is between-firm, mediated by share of Black workers
  - ▶ About half is within-firm
- ▶ Separation gaps: first spell gaps re-emerge
  - ▶ About 0.6% of lifetime consumption
  - ▶ 20% of **unconditional** separation gaps

**Thank You**

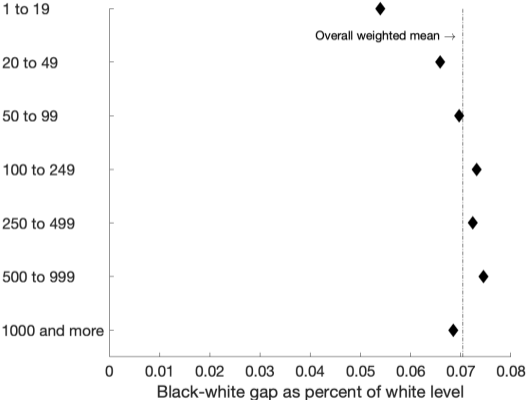
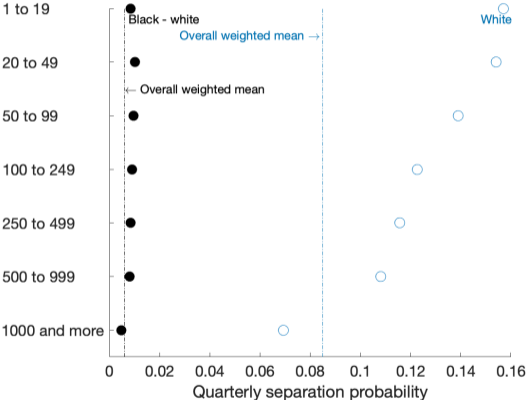
# Sample shares: share of workers that are Black



# Sample shares: sectors

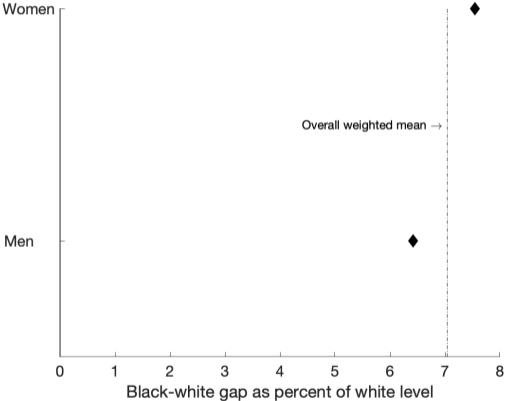
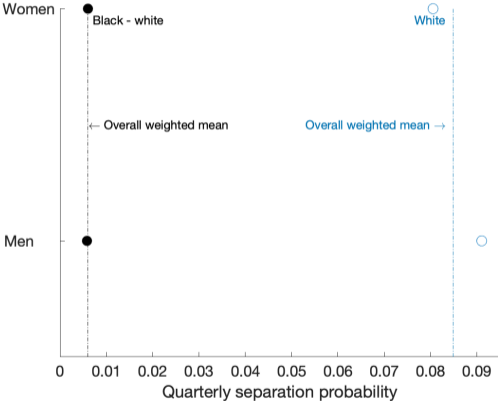


# Separation gap heterogeneity: employer size



▶ [Back to gaps](#)

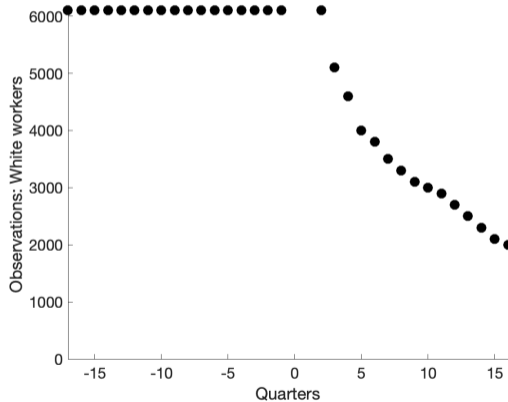
# Separation gap heterogeneity: gender



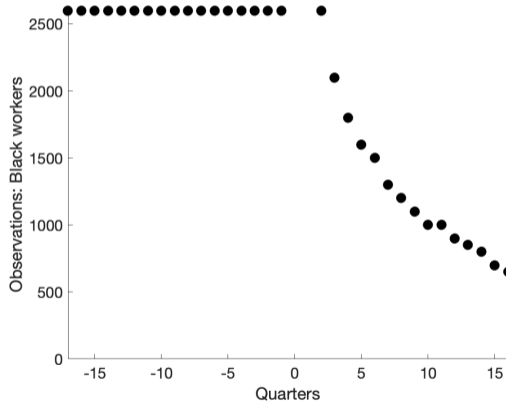
▶ [Back to gaps](#)

# Sample counts: mass layoff

White workers



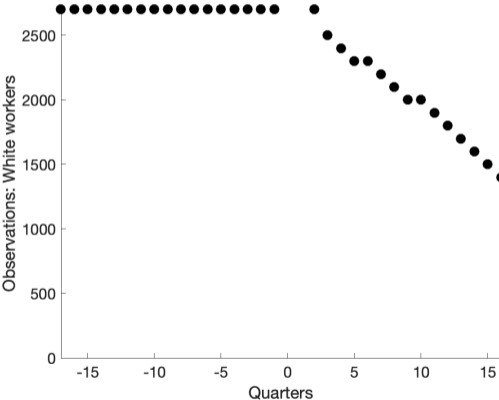
Black workers



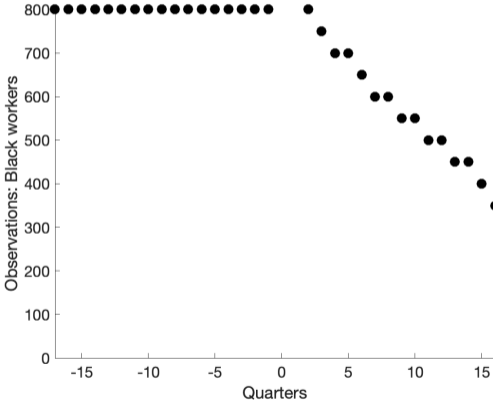
[▶ Back to mass layoff](#)   [▶ Back to separations](#)

# Sample counts: mass layoff, same second firm

White workers



Black workers

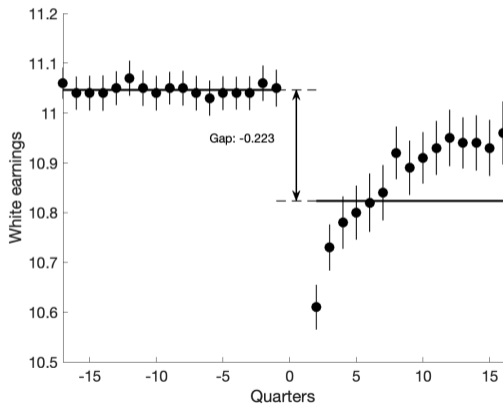


[▶ Back to mass layoff, same second firm](#)

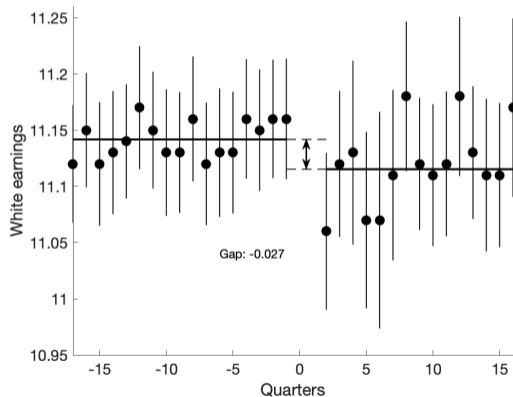


# White level

## Mass layoff



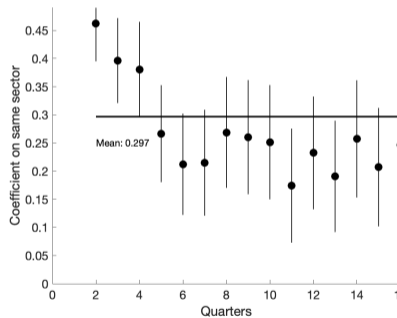
## Mass layoff, same next firm



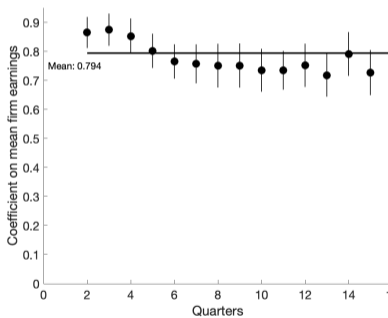
[▶ Back to mass layoff](#)   [▶ Back to mass layoff, same second firm](#)

# Coefficients on controls

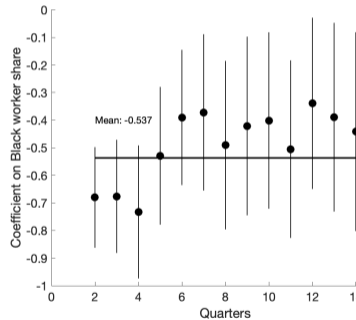
## Same sector



## Mean firm earnings



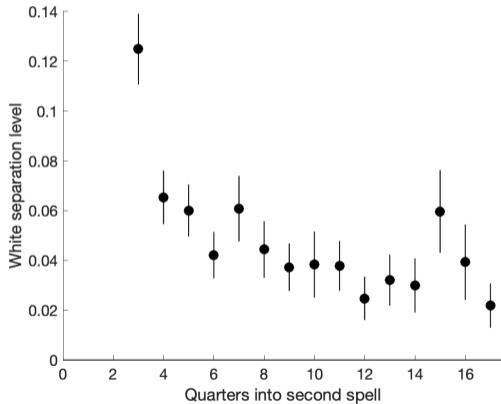
## Share black



[Back to conditional event study](#)

# White level: mass layoff separations

All separations



[▶ Back to separations](#)

EN separations

