# Occupational routine intensity and the adjustment to job loss

Evidence from mass layoffs

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#### THIS PAPER

#### Overview



- How does a worker's ability to adjust to economic shocks vary with the occupational intensity of routine tasks?
  - Exposure to changing environment due to technological progress.
  - Employment share of routine-intensive occupations has been falling over the past decades.
  - Effect of job loss on future employment and wage earnings.
- To ensure an **exogenous source** of unemployment we use data from mass layoffs:
  - Compare workers with identical careers but who work in occupations with different degrees of routine intensity.
  - Use of a difference-in-differences approach.

#### THIS PAPER

# Findings



- All workers affected by a mass layoff suffer from persistent negative effects on subsequent employment and earnings.
- These effects are **considerably more pronounced** for workers that were formerly employed in routine-intensive occupations.
- Negative earnings effects can be decomposed into similarly sized effects on employment duration and wages.
- Chance of re-employment in higher-quality jobs reduced.
- Transitions into other occupations or industries more likely.
- Adjustment more difficult in light of falling employment shares of routine-intensive occupations and devaluation of human capital.

## Mass layoff sample



- Identification of mass layoff workplaces based on the Establishment History Panel (BHP):
  - Annual dataset of all establishments in Germany.
  - Mass layoffs take place between 1980 and 2010.
- Match with the full employment biographies of affected workers:
  - Taken from Integrated Employment Biographies (IEB).
  - Only those with at least 3 years of tenure in the establishment.
  - 12 quarters before and up to 24 quarters after the mass layoff.
- In total 9,365 establishments and 342,045 workers.

# Descriptive statistics



	1980-89		1990-99		2000-10	
	ML sample	Random	ML sample	Random	ML sample	Random
Manufacturing	73.41	48.37	65.68	39.27	56.25	37.58
	(44.18)	(49.97)	(47.48)	(48.83)	(49.61)	(48.43)
Electricity, gas	0.34	1.75	2.81	1.72	0.89	1.28
	<i>(</i> 5.80 <i>)</i>	<i>(13.11)</i>	<i>(16.54)</i>	(13.02)	(9.38)	<i>(11.22)</i>
Construction	2.64	6.44	1.20	7.23	1.40	3.83
	(16.04)	<i>(24.54)</i>	<i>(10.87)</i>	(25.90)	<i>(11.74)</i>	(19.20)
Wholesale/retail trade	12.84	8.70	13.86	9.52	18.74	11.20
	<i>(</i> 33. <i>4</i> 5 <i>)</i>	(28.18)	<i>(34.55)</i>	(29.35)	(39.02)	<i>(31.53)</i>
Hotels and restaurants	0.31	0.43	0.76	0.70	1.07	1.11
	<i>(5.60)</i>	(6.55)	(8.67)	(8.32)	(10.28)	<i>(10.49)</i>
Transport, storage	3.10	4.58	8.77	4.83	6.85	5.95
	<i>(17.34)</i>	(20.91)	(28.29)	(21.45)	(25.27)	(23.66)
Financial intermed.	1.41	5.11	1.51	5.88	3.03	5.52
	<i>(11.79)</i>	<i>(</i> 22.01)	(12.18)	(23.52)	(17.15)	(22.83)
Real estate, rental	5.95	3.48	5.41	4.92	11.77	10.35
	<i>(</i> 23.65 <i>)</i>	(18.33)	(22.63)	(21.62)	(32.23)	<i>(30.47)</i>
50-99	25.09	16.59	28.08	21.28	32.20	22.71
	(43.36)	<i>(37.20)</i>	(44.94)	(40.93)	(46.72)	(41.90)
100-199	24.90	14.49	26.60	16.71	25.97	18.93
	(43.24)	<i>(</i> 35.20 <i>)</i>	(44.19)	<i>(</i> 37.31 <i>)</i>	(43.85)	<i>(39.18)</i>
200-499	28.60	19.84	25.54	21.03	21.15	22.23
	(45.19)	<i>(39.88)</i>	(43.61)	(40.75)	(40.83)	(41.58)
500+	21.41	49.08	19.79	40.99	20.68	36.13
	(41.02)	<i>(49.99)</i>	<i>(</i> 39.84 <i>)</i>	(49.18)	(40.50)	(48.04)
East	2.41	3.53	17.86	16.80	14.61	16.09
	(15.35)	<i>(18.45)</i>	(38.30)	(37.38)	(35.32)	(36.75)

## Routine-intensity measure



- Occupations differ with respect to their contents and specifically to the extent that they contain routine components:
  - Use of machines easier to implement in jobs characterised by routines.
- To obtain information on job contents we use data from an employee survey (Erwerbstätigenbefragung):
  - Information on job characteristics at the worker level.
  - Conducted by the Federal Institute for Vocational Education and Training (BIBB) and the Institute for Employment Research (IAB).
  - Use data from the waves 1985, 1991, 1999.

## Routine-intensity measure



- Focus on two items in order to assess the extent of an occupation's routine intensity:
  - Are the contents of your job minutely described by the employer?
  - Does your job sequence repeat itself regularly?
  - Possible answers: 'almost always', 'often', 'occasionally', 'rarely', 'hardly anytime'.
- The **routine-intensity variable** is defined as the fraction of workers reporting both items to be the case 'almost always'.

# Descriptive statistics



	1980-89		1990-99		2000-10	
	ML sample	Random	ML sample	Random	ML sample	Random
Routine	12.03	11.56	13.48	13.03	12.33	12.33
	(9.69)	(9.73)	(11.61)	(11.18)	<i>(10.66)</i>	(10.87)
Earnings	8,536.50	8,787.09	9,893.81	9,671.34	11,134.60	10,642.98
	(3,966.27)	(4,369.54)	(6,281.68)	(5,938.22)	(9,389.24)	(8,148.71)
Duration	91.04	90.59	91.08	90.65	91.11	90.66
	<i>(4.84)</i>	(4.72)	<i>(4.89)</i>	(4.32)	<i>(4.56)</i>	(4.23)
Wage	93.72	96.91	108.54	106.58	122.07	117.23
	(43.12)	<i>(47.78)</i>	(68.41)	(65.10)	(102.57)	(89.42)
Female	27.79	30.80	30.54	33.56	27.18	29.21
	(44.80)	(46.17)	(46.06)	(47.22)	(44.49)	<i>(45.47)</i>
Foreign	16.13	12.40	11.59	8.92	8.25	7.97
	<i>(</i> 36.78 <i>)</i>	(32.96)	(32.01)	(28.51)	<i>(</i> 27.51 <i>)</i>	(27.09)
Low skill	28.45	26.39	15.98	13.80	13.22	9.39
	(45.12)	(44.08)	(36.65)	<i>(34.49)</i>	(33.87)	<i>(</i> 29.17)
Medium skill	68.15	68.41	75.93	77.08	75.17	75.65
	<i>(46.59)</i>	<i>(46.49)</i>	(42.75)	(42.03)	(43.20)	(42.92)
High skill	3.40	5.19	8.08	9.13	11.60	14.96
	(18.13)	<i>(22.19)</i>	<i>(</i> 27.26 <i>)</i>	<i>(</i> 28.80)	(32.03)	<i>(35.67)</i>
Tenure	7.91	6.87	9.49	7.41	10.14	7.35
	<i>(</i> 2.80 <i>)</i>	(2.72)	<i>(</i> 5.39 <i>)</i>	(4.69)	<i>(</i> 5.98 <i>)</i>	(4.77)
Observations	95,529	191,058	137,929	275,858	108,587	217,174

## **DESCRIPTIVE ANALYSIS**

# The effects of mass layoffs

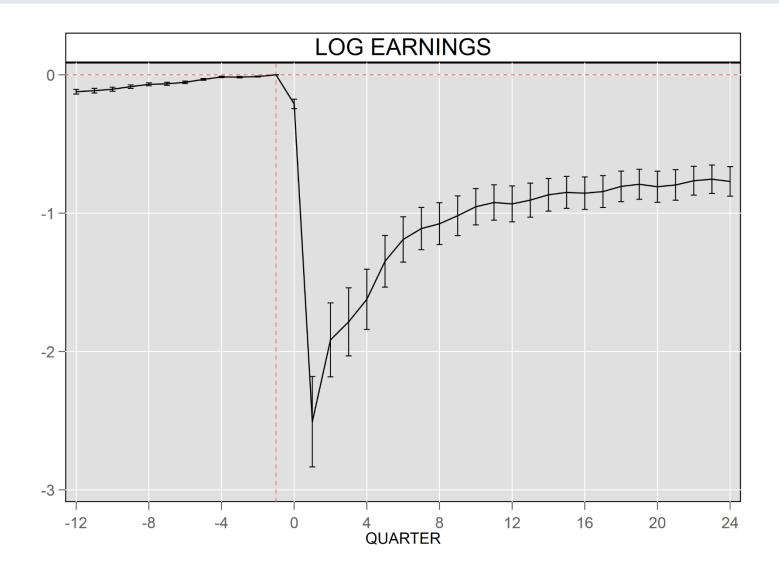


	Quarter -1 (before ML)	Quarter 1 (after ML)	% change			
Quarterly Earnings						
All	9,908.76	7,028.40	-29.07%			
Low routine	13,367.68	10,623.45	-20.81%			
High routine	7,217.33	3,960.71	-45.39%			
Quarterly Employment duration						
All	91.08	61.72	-32.23%			
Low routine	91.21	71.42	-21.57%			
High routine	90.83	48.76	-46.06%			
Average kalender daily wages						
All	108.69	82.03	-24.53%			
Low routine	146.44	121.21	-17.53%			
High routine	79.41	48.62	-39.10%			

### **DESCRIPTIVE ANALYSIS**

# The effects of mass layoffs

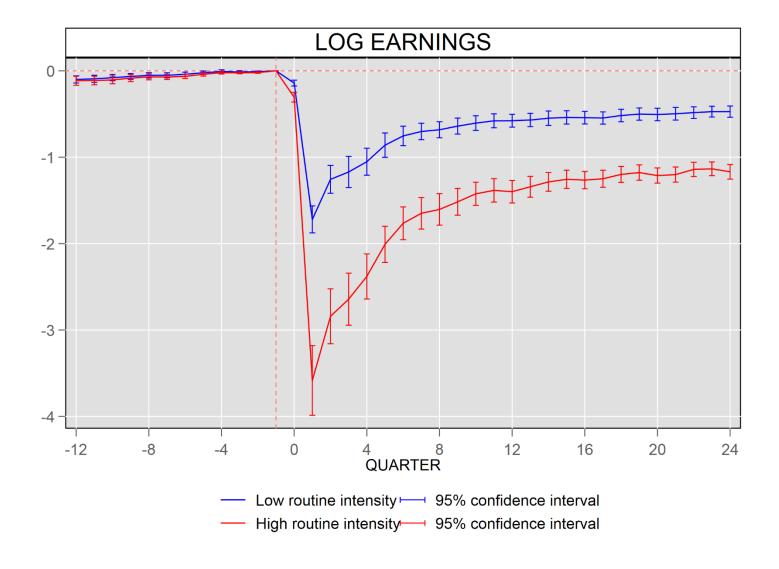




### **DESCRIPTIVE ANALYSIS**

## The effects of mass layoffs





#### **EMPIRICAL ANALYIS**

## Identification strategy and model



### Event-study approach:

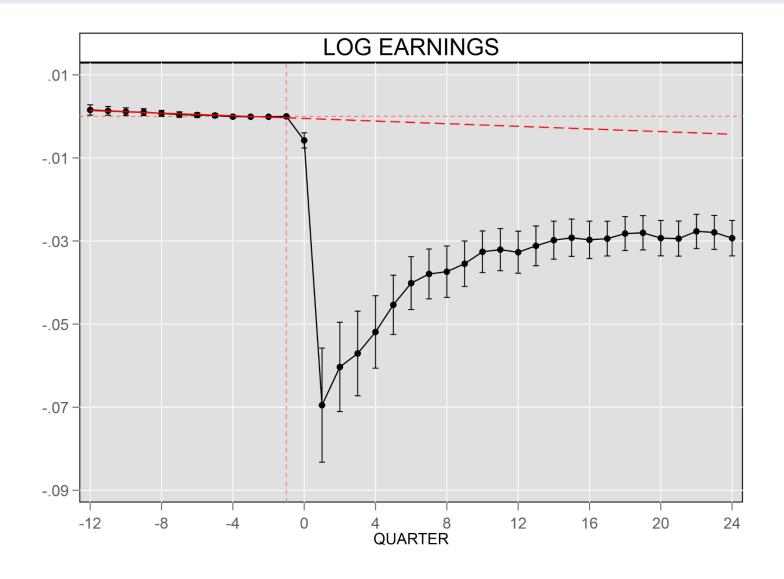
$$- y_{it} = \alpha + \sum_{k \neq -1} [\beta_k R I_i \times I[t = k] + \delta_k I[t = k]] + \mu_i + \varphi_t + w_{it}$$

#### Outcome variables:

- Quarterly earnings (in logs).
- Other outcomes: days in employment per quarter, average daily wage.
- Standard errors clustered at the occupational level.

# **Baseline specification**





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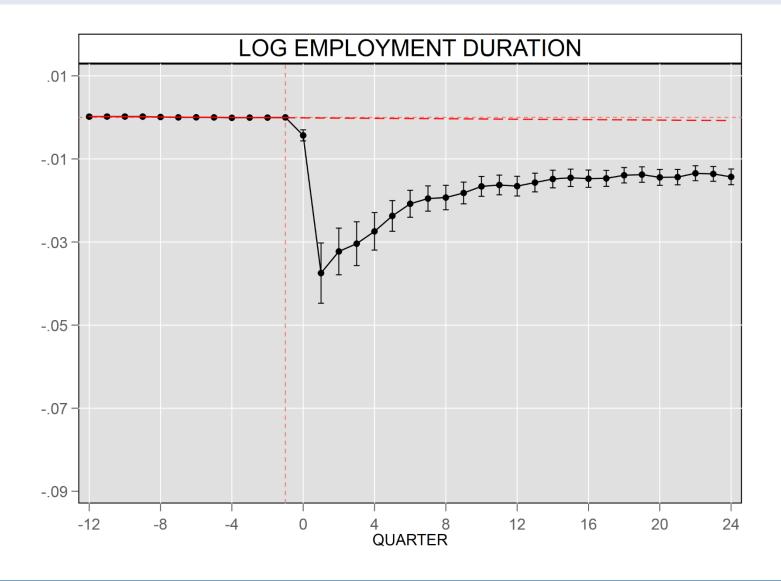


Δ <b>RI</b>	Relative (k=1)	Relative (average)	Absolute (cum.)		
Earnings					
Percentage point	-0.07	-0.03	-3,226.17		
Standard deviation	-0.53	-0.31	-29,797.25		
Interdecile range	-0.84	-0.60	-43,262.00		
Employment					
Percentage point	-0.04	-0.02	-24.79		
Standard deviation	-0.33	-0.18	-244.72		
Interdecile range	-0.63	-0.39	-443.35		
Daily earnings					
Percentage point	-0.03	-0.02			
Standard deviation	-0.29	-0.16			
Interdecile range	-0.57	-0.35			



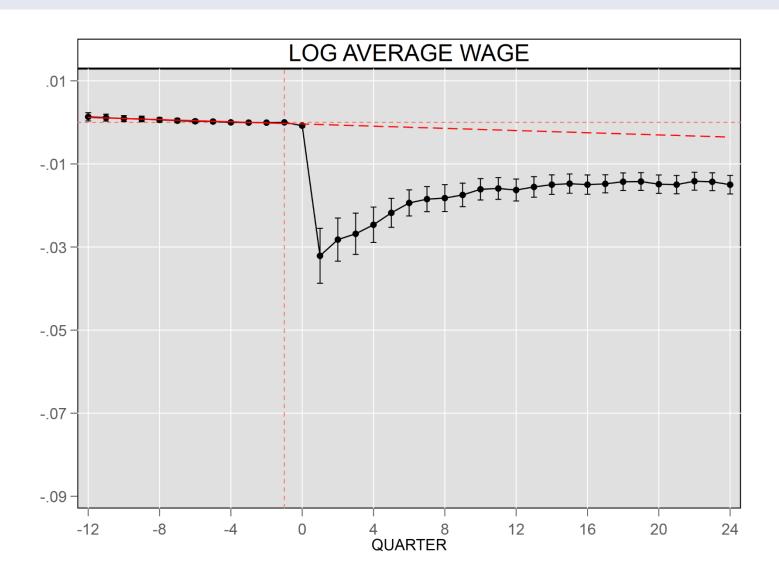
- Can we say anything about the source of these effects?
  - Quarterly earnings are the product of days in employment and an average daily wage.
  - Estimate corresponding models for these variables (in logs).
  - The estimated coefficients add up to those from the earnings model.
- Both components appear equally important in magnitude:
  - Employees from routine-intensive occupations are on average less likely to find employment and are employed in jobs paying lower wages.
- **Differences in pre-treatment trends** are more pronounced for earnings and wages than for employment duration.





# **RESULTS**Decomposition I



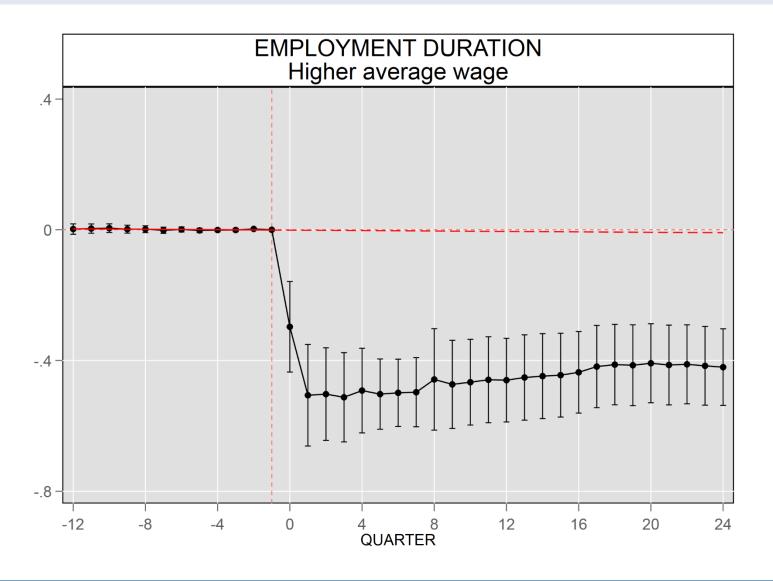




- How do the careers of workers develop after the mass layoff?
  - Does the initial degree of routine intensity affect the type of jobs that are subsequently found?
- Differentiate subsequent employment according to average wages as well as regional and occupational mobility.
- Initial employment in routine-intensive occupations decreases the chance of entering higher-paying jobs.
- It leads to higher occupational, but lower regional mobility.
- Moving into other occupations potentially associated with costs due to loss of human capital.

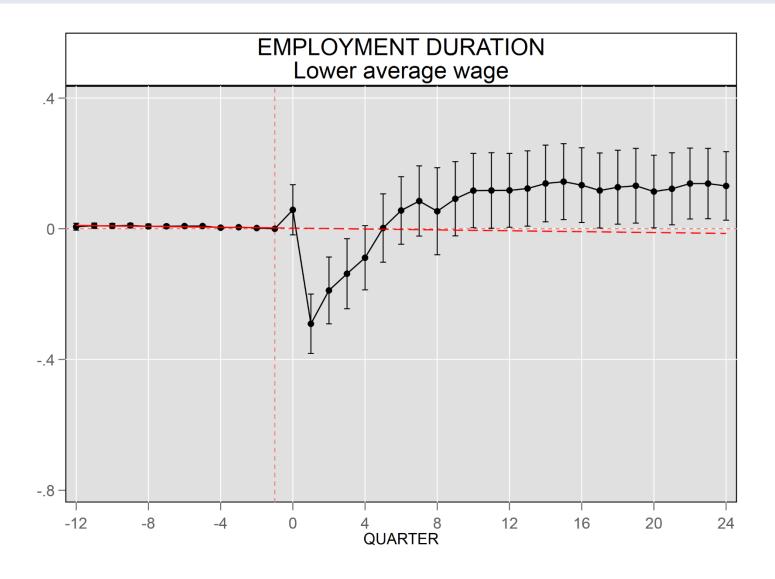
# **RESULTS**Decomposition II



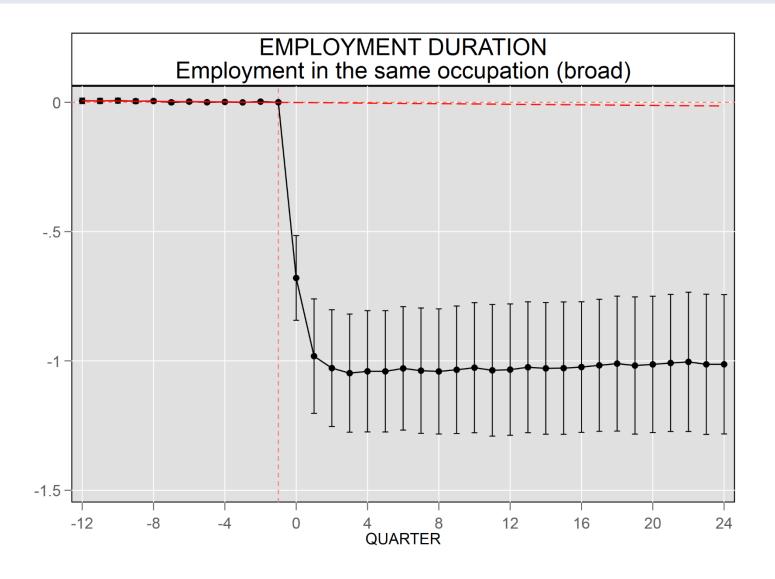


# **RESULTS**Decomposition II

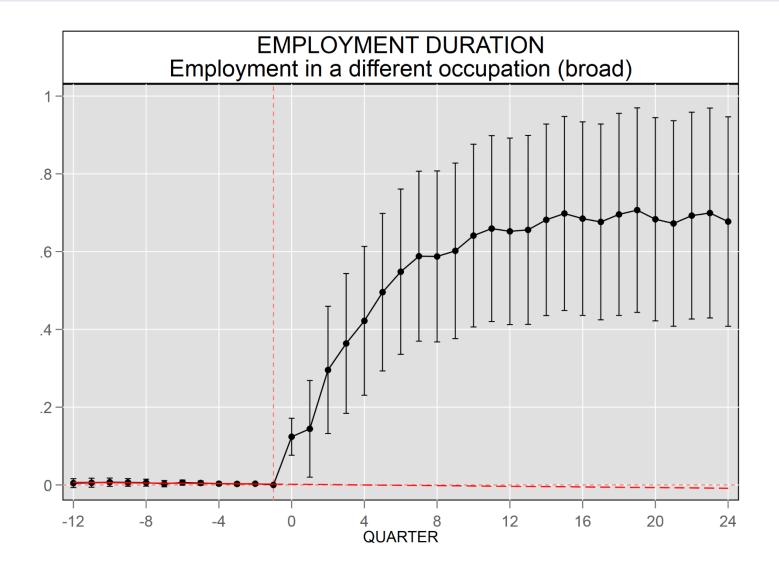




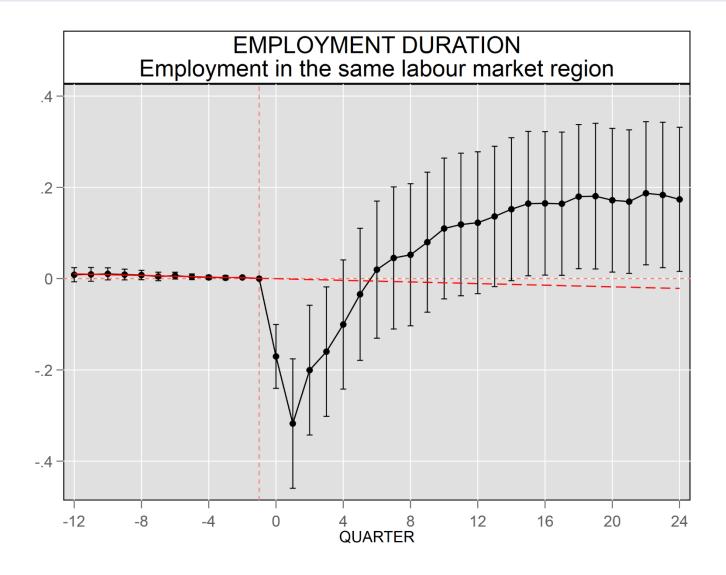




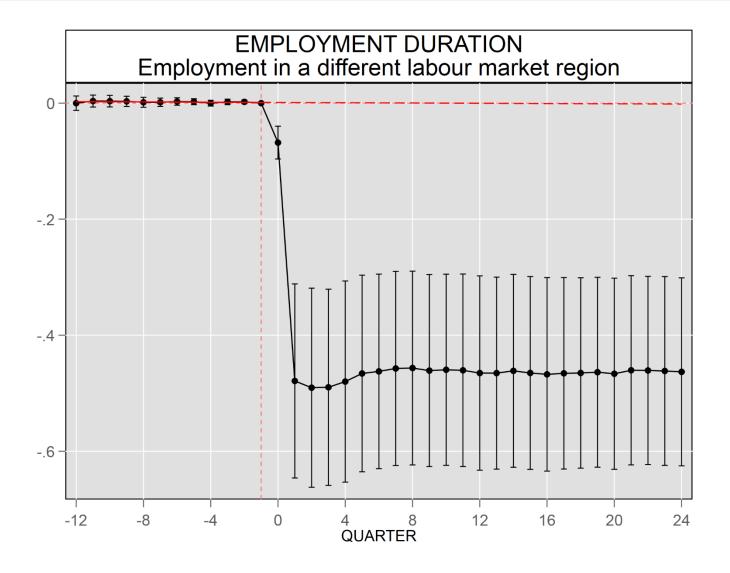












# Effect heterogeneity



Earnings	Relative (k=1)	Relative (average)	Absolute (cum.)
Baseline	-0.07 (0.01)	-0.03	-3,226.17
Unskilled	-0.04 (0.01)	-0.03	-698.46
Vocational	-0.05 (0.01)	-0.02	-2,347.83
College	-0.04 (0.02)	-0.02	-7,109.84
23-29 years	-0.06 (0.01)	-0.03	-2,410.99
30-44 years	-0.07 (0.01)	-0.03	-3,395.97
45-51 years	-0.07 (0.01)	-0.04	-2,883.65
Manufacturing	-0.06 (0.00)	-0.03	-2,710.34
Non-manufacturing	-0.04 (0.02)	-0.03	-4,010.17
Less than 90%	-0.07 (0.01)	-0.04	-3,476.72
More than 90%	-0.06 (0.01)	-0.03	-2,623.26

#### CONCLUSION



- How does recovery from job loss vary with an occupation's routine intensity?
- Employment share of routine-intensive occupations declining.
- Use of mass layoffs to identify exogenous shock.
- Persistent negative effects in terms of subsequent employment duration and earnings.
- Substantially larger for individuals formerly employed in routineintensive occupations.
- Adjustment to shocks more difficult for this group of individuals.
- This form of human capital has become less valuable, potentially due to technological progress.



# Thanks for your attention.

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#### **EMPIRICAL ANALYIS**

## Identification strategy and model



#### Potential-outcomes framework:

- Treatment: mass layoff.
- Treatment measure: routine intensity in the quarter before the layoff.
- The expected marginal effect of the treatment measure on the outcome is:
  - $E[dy_t^1|t \ge z] = E[\beta_t dT|t \ge z]$
  - β<sub>t</sub> is a quarter-specific parameter.

#### Problem of identification:

- What would have been the marginal effect of the treatment measure in the absence of treatment?
- $E[dy_t^0|t \ge z]$  is not observable.

#### **EMPIRICAL ANALYIS**

## Identification strategy and model



- The marginal effect of the treatment measure can be estimated for the pre-treatment period (given that treatment is not active):
  - $E[dy_t^0|t < z] = E[\beta_t dT|t < z]$
  - Assume that the marginal treatment effect **interacts linearly** with time.
- Regress the estimated year effects on a linear time trend:

$$- \hat{\beta}_t = \gamma_0 + \gamma_1 t + v_t$$

- The counterfactual marginal effect of the treatment measure is:
  - $E[dy_t^0|t \ge z] = E[\gamma_0 + \gamma_1 t|t \ge z]dT$
- Accordingly, the treatment effect is given by:
  - $E[dy_t^{\ 1}|t \ge z] E[dy_t^{\ 0}|t \ge z] = E[\beta_t \gamma_0 \gamma_1 t]dT$