International Outsourcing, Labor Unions, and Job Stability: Evidence from U.S. Manufacturing in the 1980s

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Abstract

After the results from Feenstra and Hanson (1999) and Diebold et al. (1997) are combined, three questions arise: Did international outsourcing lead to a deterioration in the job stability of workers in manufacturing industries in the 1980s? What was the impact on workers of different skill levels? Can labor unions moderate the impact of international outsourcing on workers? This study employs CPS data, the NBER Manufacturing Productivity Database (Bartelsman and Gray, 1996), and the outsourcing data in Feenstra and Hanson (1999) to analyze the impact of international outsourcing, labor unions, and the interaction between outsourcing and labor unions on workers' job retention rates. The results of this study show that international outsourcing decreases blue-collar but not white-collar workers' job retention rates. Unions, however, can mitigate the negative impact of international outsourcing on the loss of blue-collar workers' job stability. An increase in R&D expenditure enhances the job stability of white-collar workers.

Jel codes: *F16, J51, J63* Copyright © 2014 JAEBR

Key words: International Outsourcing, Labor Union, Job Stability.

1. Introduction

In recent years, a great deal of concern has been raised regarding the negative impact of international outsourcing on less-skilled labor. Debates such as whether international outsourcing causes wage inequality between less-skilled workers and skilled workers have attracted economists' attention.² In comparison with the effects on wage inequality, however, the issue of job security or stability has been

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² See, e.g., Lawrence and Slaughter (1993), Berman et. al. (1994), Slaughter (1995), Feenstra and Hanson (1996, 1999), Jones and Kierzkowski (2001), Egger and Kreickemeier (2008), Sayek and Sener (2006), Hsu (2011), and Hsu and Chiang (2014).

less mentioned. Most studies agree that it takes less time to see the effects of a structural change on workers' jobs than on their wages.³ Therefore, the first form of adversity that less-skilled workers are likely to encounter after their employers outsource their jobs to foreign countries is unemployment and job turnover.

Job stability has always been a frequently discussed topic in labor economics literature.⁴ Concerns over the influence of international outsourcing on workers' jobs also have been the focus of much discussion, but the conclusions are mixed. Some empirical studies such as Egger et al. (2007),⁵ Geishecker (2008),⁶ and Munch (2010) have concluded that international outsourcing does have a negative impact on workers.⁷ All these three papers look at the adjustment process in employment in response to international outsourcing. The differences between Munch (2010), Egger et al. (2007), and Geishecker (2008) are with respect to the country considered, the methodology (single risk vs. competing risk models), and the modeling of unobserved heterogeneity. Geishecker (2008) found international outsourcing measured narrowly has a negative impact on all workers' employment security, but the results of Munch (2010) indicated that international outsourcing, when broadly defined, increases only less-skilled workers' unemployment risk.

Some research papers, however, have found evidence for a positive impact of internationalization on workers' jobs. Becker and Muendler (2008) analyzed the impact of foreign direct investment (FDI) on job security and found that multinational enterprises (MNEs) that expand abroad retain more domestic jobs than competitors without foreign expansion. Maertz et al. (2010) investigated the reactions of survivors of downsizing through layoffs, offshoring, and outsourcing and found that layoffs and offshoring lowered organizational performance, reduced job security, lowered affective and calculative attachment, and raised turnover intention in remaining employees more than outsourcing did. Their explanation is that survivors of outsourcing feel that they could potentially work with outsourcing recipients.

Relevant topics such as globalization have also been addressed in relation to the issue of job stability. Kletzer (2001) examined the relationship between

³ See, e.g., Geishecker (2008).

⁴ See, e.g., Hall (1982), Leighton and Mincer (1982), Ureta (1992), Diebold et al. (1997), Farber (1998), Marcotte (1999), Bernhardt et al. (1999), and Heisz (2005).

⁵ Egger et al. (2007) analyzed the impact of international forces, including trade and outsourcing, on employment based on a sample of individual Austrian male workers over the period 1988-2001. By employing dynamic fixed effects multinomial logit models proposed by Honoré and Kyriazidou (2000), they found a significant negative impact of international outsourcing on the labor market turnover in manufacturing sectors with a comparative disadvantage.

⁶ Geishecker (2008) criticized Egger et al. (2007) for controlling only age as the time-varying individual characteristic in their analysis. Geishecker (2008) advanced the work performed by Munch (2010) and Egger et al. (2007) by adding a wider range of time-varying individual and workplace-related characteristics and by using monthly data instead of yearly data. The main finding of Geishecker (2008) was that international outsourcing, when narrowly defined, has a significant negative impact on individual employment security, regardless the level of skill of the workers.

⁷ Munch (2010) employed Danish manufacturing data covering the period 1990-2003 to estimate the impact of international outsourcing on individual job separation risk. Based on an individual-level empirical model with a distinction between job-to-job and job-to-unemployment transitions, his findings suggested that outsourcing, measured broadly, increased the unemployment risk of low-skilled workers, but that the quantitative impact was limited. The probability of the job-to-job changes has been raised for all education groups by international outsourcing.

increasing foreign competition and job displacement in U.S. manufacturing over the 1975-94 period.⁸ They found strong positive relationships between increasing foreign competition and job displacement for some industries that compete with imports, but overall only a small share of job displacement can be explained by increased foreign competition.

In addition to the mixed results in the literature, the role of labor unions or trade unions has not been considered in discussions of the impact of outsourcing on job stability. Previous studies agree that unions decrease the layoff rate (Tinsley, 2003) and unemployment rate (Allen, 1988), as well as enhance job security (Bryson and White, 2006). It is also argued that labor unions motivate outsourcing: studies such as Abraham and Taylor (1996) have argued that since labor unions raise workers' wages, they increase the probability of firms outsourcing their activities. Their argument is supported by the evidence from a survey of 2,700 establishments. However, Braun and Scheffel (2007) concluded that labor unions' collective bargaining leads to a decline in outsourcing.⁹ Similarly, Lommerud et al. (2009) concluded that labor unions can make international outsourcing less profitable by increasing the wage rate of in-house production. Therefore, the ability of labor unions to moderate the negative impact of foreign outsourcing on job stability is also a topic of this study.

For the reasons just stated, we propose two questions. First, how job security is measured plays an important role in the discussion of the impact of international outsourcing on workers' job security. Job retention rate, which is an important measure of workers' job security and stability, has not been measured in the previous studies. Diebold et al. (1997) found that job retention rates were higher for white-collar workers but were lower for blue-collar workers in the U.S. during the 1980s. Studying that same period of time, Feenstra and Hanson (1999) found that international outsourcing decreased the relative wage of production workers to that of non-production workers. Based on the results of Diebold et al. (1997) and Feenstra and Hanson (1999), we infer that international outsourcing deteriorated blue-collar workers' job retention rates in the 1980s. If our inference is true, this is one more finding in the literature that international outsourcing has a negative impact on unskilled workers. The present study tries to provide empirical evidence for our inference.

Second, we are interested in the function and role of labor unions when firms outsource in-house production overseas. Whether labor unions can moderate the impact of international outsourcing on workers' job security is the second focus of our paper.

This study uses a sample period from 1980 to 1990 for two reasons. First, international outsourcing of U.S. manufacturing grew rapidly in the 1980s. This is the same sample period chosen by Feenstra and Hanson (1999) and Diebold et al.

⁸ International trade and international competition are two important factors that induce international outsourcing. See the discussions in McLaren (2000), Grossman and Helpman (2002), Görg and Hanley (2004), and Diaz-Mora (2008).

⁹ The positive effect of labor unions on outsourcing is direct: it induces more outsourcing. An indirect effect, which reduces outsourcing, is based on the premise that the higher wage rate of the remaining in-house workers decreases the marginal benefit from outsourcing. Thus, if the indirect effect suppresses the direct effect, labor unionization can lead to a decline in outsourcing.

(1997). Second, labor union membership dramatically changed during the 1980s. The union membership density of U.S. manufacturing began declining in the 1950s. According to the CPS,¹⁰ private sector union density dropped dramatically during the 1970s and 1980s. In the 1990s, union density still declined, but at a slower pace. Although international outsourcing became more widespread during the 1990s and 2000s, unions in U.S. manufacturing industries became less influential.

The main contribution of this paper is that it is the first of its kind to empirically examine the effects of international outsourcing on workers' job retention rates. Secondly, it also discusses whether labor unions can moderate the impact of international outsourcing on workers' job stability.

The remainder of this paper is organized as follows: Section 2 states how job retention rates are measured. Section 3 describes the data and regression models in our study. In section 4 we present and discuss our regression results. Section 5 concludes our findings.

2. Measuring Job Stability and International Outsourcing

Hall (1982), Ureta (1992), and Diebold et al. (1997) evaluated workers' job stability using the *Current Population Survey* (CPS) data to compute job retention rates. According to Hall (1982), to obtain a clear picture of the duration of workers' jobs, it is necessary to know the projected likely additional time that a worker expects to spend in his current job. Job retention rates for a particular period provide us with an expected probability that a worker will retain his or her current job for a specific number of years. In Hall (1982), a job retention rate or probability is measured by a survival rate from one age-tenure category to another higher age-tenure category.

There are two ways to compute the job retention fraction. The first method uses cross-sectional data. If the number of people working in a calendar year t is N_t , the number of workers in a calendar year s with at least τ years seniority is $N_s(\tau)$. Then, if the focus is on workers with the *i*th skill level in the *j*th industry, the basic s-year estimated job retention rate for workers with τ years of tenure is computed as the ratio of the number of workers with at least τ years seniority in the tenure supplement s+ τ years, $N_{s+\tau,ij}(\tau)$, to the total number of workers in the tenure supplement s+ τ years, $N_{s+\tau,ij}$. Formally,

$$\hat{R}_{s,\ ij}^{c}(\tau) = \frac{N_{s+\tau,\ ij}(\tau)}{N_{s+\tau,\ ij}},\tag{1}$$

where $\hat{R}_{s,\ ij}^{c}(\tau)$ represents the cross-sectional τ -year estimated job retention rate with s years as basic years for workers with the *i*th skill level in the *j*th industry. Job retention probabilities can also be obtained by employing multiple historical CPS tenure supplements. Following the notation used earlier, the historical calculation of the job retention rate for workers with the *i*th skill level in the *j*th industry, $\hat{R}_{s,\ ij}^{h}(\tau)$, is

¹⁰ This argument is made based on the data collected and estimated by Barry T. Hirsch and David A. Macpherson. See unionstats.com for details.

$$\hat{R}^{h}_{s,\ ij}(\tau) = \frac{N_{s+\tau,\ ij}(\tau)}{N_{s,\ ij}}.$$
(2)

As discussed in Ureta (1992) and Diebold et al. (1997), the cross-sectional calculation requires that the number of beginners to be constant or at least remain similar across years. The historical calculation requires a stable survival function.¹¹ In this paper, because of the limitations of the data, we employ the cross-sectional approach after assuming that the participation rates between arrivals in different cohorts, i.e., the gender and race that Ureta (1992) argued about, are no different across industries.¹² The regression results for historical job retention rates are presented in the Appendix for the purpose of the robustness check.

This paper estimates international outsourcing by following the method in Feenstra and Hanson (1996, 1999). Outsourcing proportions are measured as the share of imported intermediate inputs in the total purchases of non-energy intermediates. There are two measures of imported intermediate inputs introduced in Feenstra and Hanson (1999). The first one considers all intermediate inputs that each manufacturing industry purchased from every other standard industrial classification (SIC) manufacturing industry.¹³ Some intermediate input purchases, however, may not involve outsourcing. For instance, the computer industry purchases plastic boxes. Hence, the second measure of international outsourcing only takes into account those intermediate inputs that are purchased from the same two-digit SIC manufacturing industry as the producing industry. Feenstra and Hanson (1999) referred to the first measure of international outsourcing as a broad measure of foreign outsourcing ratio could fall between (or on either of) the broad and narrow measures, this study includes both of them.

3. Data and Regression Equation

The Current Population Survey (CPS) data are employed to calculate the job retention rate to proxy job stability. The CPS data also contain information regarding the union membership ratio, the union coverage ratio and individual characteristics. During the 80s, four tenure supplements were available. This study chooses CPS tenure supplements for 1983, 1987 and 1991 to compute four-year retention rates in each industry type for each of the two skill levels (i.e., blue-collar and white-collar workers). There are two reasons for doing that. First, the tenure questionnaire in 1981 was different from that in 1983, 1987 and 1991. Second, outsourcing data has been computed from the *Economic Census* published in 1982 and 1987. Since the job tenure supplements are included in the January CPS, the 1983 CPS is more reliable.

Although the historical job retention rate can be measured under an unstable survival function and nonconstant arrival rate, the use of multiple CPS supplements carries the risk of inaccuracy because of the variety of sampling sizes.¹⁴ This

¹¹ Diebold et al. (1997) avoid the requirement of a stable function by adopting longer sequence CPS tenure supplements. See Diebold et al. (1997) for the details.

¹² Section 4 provides the detailed reasons for the data restrictions in this study.

¹³ See equation (9) in Feenstra and Hanson (1999).

¹⁴ Some adjustments were made to this issue. The Annual Survey of Manufactures (ASM), which could give us the number of employed workers, and the CPS data, which can indicate the total number

drawback is especially considerable in an industrial-level study. In order to coordinate the two sets of data on international outsourcing, this study organizes data by manufacturing industries, and some industries have only a limited number of responses. After workers are grouped by skill level, the influence of different sample sizes becomes even more crucial. Therefore, cross-sectional job retention rates (hereafter: job retention rates) are employed in this study, based on the assumption that survival functions are unstable over time, although those determinants of survival functions besides the independent variables in our regression are similar across all manufacturing industries. The same assumption is also applied to overall arrival rates, which were not found to be constant, but were similar across all manufacturing industries.

Following Feenstra and Hanson (1999), there are two variables related to outsourcing in the regression equations. The first is the narrow measure of foreign outsourcing and the second is the difference between the broad measure and the narrow measure of international outsourcing. Data on international outsourcing and trade are obtained by the same sources in Feenstra and Hanson (1999).¹⁵ Based on the discussion in Section 2, Munch (2010), Egger et al. (2007), and Geishecker (2008) found a negative and significant effect on blue-collar workers' job stability, but if international outsourcing (horizontal outsourcing) was caused by MNEs engaging in horizontal FDI mainly in the 1980s, a positive impact as in the results of Becker and Muendler (2008) is expected. The effect on white-collar workers' job stability could be either positive or have no significant impact. It depends on whether foreign outsourcing caused serious job-to-job turnover in skilled workers. The overall effects are the sum of the effect of outsourcing on blue-collar workers and white-collar workers. It can be negative or have no significant effect.

Unionization can be measured by computing how many workers are union members (membership) and what percentage of workers are union members (coverage), as described in Freeman and Medoff (1979). This study takes both membership and coverage into consideration. An interaction term for international outsourcing and unionization can reveal whether labor unions can moderate the impact of foreign outsourcing on workers. A significant positive coefficient of the interaction term means that an increase in unionization can release the negative impact of foreign outsourcing on the workers' job stability. We also construct a dummy variable that represents comparative labor union power. One industry's unionism dummy variable equals 1 if the industry's union coverage ratio is greater than the average ratio of all manufacturing industries.¹⁶ The coefficient of the interaction term of the measure of international outsourcing and the dummy variable denotes the impact of foreign outsourcing on the workers' job stability in more unionized industries. On the contrary, the coefficient of the interaction term of international outsourcing and one minus the dummy variable denotes the impact of foreign outsourcing on the workers' job stability in less unionized industries.

In addition to international outsourcing and unions, determinants such as the

of respondents in the sample, are employed in the adjustment. The after-adjustment job retention rates, however, are still unsatisfactory; some of them are still greater than one.

¹⁵ The authors would like to thank Dr. Hanson for providing outsourcing data.

¹⁶ Since some industries do not have data for the union membership of white-collar workers in some years, this study uses union coverage only in analyzing skilled workers.

real output, number of employees, research and development (R&D) expenditure share, skilled to less-skilled workers ratio, average age, and average income per person in each industry are also included in the regressions. Real output and the number of employees are indicators of the industries' economic activities. A straightforward example is that a decrease in output and employment means a decrease in labor demand which should lead to a deterioration in the workers' job stability. However, because the hiring and laying off policy is part of the negotiation and bargaining process between the employer and the labor union, the real effect could be complicated.

An increase in the share of R&D expenditure, which shows the willingness of employers to improve their technology, should be of benefit to skilled workers but not to less-skilled workers. Munch (2010) found that R&D intensity reduces the probability of job separation, as well as the job change hazard rate and unemployment hazard rate.¹⁷ However, the results of Geishecker (2008) showed that the share of R&D expenditure in total output increase the employment hazard rate, especially for less-skilled workers. Thus, the effects of the R&D expenditure share should be negative to blue-collar workers' job stability but positive to white-collar workers' job stability.

The skilled to less-skilled workers ratio may be viewed as an indicator of the average educational level or technology level of an industry. If the results reveal a positive relationship between the ratio and the workers' job stability, the employers who prefer to hire higher quality workers are also willing to retain their workers. A stable working environment provides workers with a chance to accumulate their job seniority. We expect there to be a positive relationship between the workers' average age and their job retention rate. The average income per person denotes how much the employers are willing to pay to retain their workers. We also expect there to be a positive relationship between income and job stability.

Since the dependent variable is the 4-year retention rate, all explanatory variables take the form of the changes over a period of 4 years. However, only considering the effect of the changes in independent variables on job stability might be problematic. For instance, the percentage of union membership might be stable in an industry with a higher percentage of union membership, but its unions will play a more important role in wage bargaining and job retention compared to an industry with a small percentage of union workers. Larger industries and smaller industries might also be affected differently by international outsourcing. Thus, the 4-year average of real output, the number of employees, the skilled to unskilled labor ratio, and the percentage of union membership or coverage are also included in the regression equations.

Since the job retention rate \hat{R}_{ij} is bounded between zero and one, a general logit specification is used.¹⁸ Variables preceded by a delta are the change in the variable during the 4-year span. The rest of the variables represent the averages for those variables in the 4-year period. For each category of workers, namely, *All*

¹⁷ The explanation in Munch (2010) is that since the industries with higher R&D intensity provide more training in firm-specific skills, they also wish to retain their workers.

¹⁸ Instead of using this log-adds ratio, one can use quasi-likelihood estimation method. See Papke and Wooldridge (1996) for details.

workers, *Blue-collar workers*, and *White-collar workers*, the regression equation is as follows:

$$\log(\frac{R_{ijt}}{1 - \hat{R}_{ijt}}) = c_t + \beta_1 \Delta S_{ijt} + \beta_2 S_{ijt} + \beta_3 \Delta U n_{ijt} + \beta_4 U n_{ijt} + \beta_5 R \& D_{jt} + \beta_6 W / B_{jt} + \beta_7 \Delta I p_{ijt} + \beta_8 \Delta A g e_{ijt} + \beta_9 A g e_{ijt} + \beta_{10} \Delta^D S o_{jt} + \beta_{11} \Delta^N S o_{jt} + \beta_{12} \Delta S o * U n_{ijt} + e_{ijt}, \quad (3)$$

where *i* represents the type of workers that belong to one of *All workers*, Blue-collar workers, and White-collar workers; j stands for each of the 77 manufacturing industries;¹⁹ t indicates the year that belongs to one of 1983-1987 and 1987-1991; \hat{R}_{ii} are the estimated job retention rates of each type of worker *i* in industry j; S_{ij} is the measurement of the scale of industry j, which can be the logarithm of the number of type *i* workers employed in industry *j*, E_{ij} or the logarithm of real output of industry j, Y_j ; Un_{ij} stands for the labor union power of type i workers in industry j and is represented by the percentage of either union membership (Un_{ij}^{mem}) or union coverage (Un_{ij}^{cov}) ; $R\&D_j$ is the average share of R&Dexpenditure to the total value of shipments in industry j; W/B_j is the white-collar to blue-collar ratio of industry j;²⁰ Ip_{ij} is the income per person of type *i* workers in industry j; Age_{ij} is the average age of the *i*th workers employed in industry j; So_i is the proportion of the production line outsourced, which includes the narrow definition of outsourcing $(^{N}So_{i})$ and is the difference between the broad definition and the narrow definition $({}^{D}So_{i})$; $\Delta So * Un_{ii}$ is the interaction term between international outsourcing and the unionization variable.

R&D data are collected from the National Science Foundation (NSF).²¹ Industry-level data such as those for total output and employment can be found in the NBER Manufacturing Productivity Database (Bartelsman and Gray, 1996). Combining information from different datasets is an issue in this study. The NBER Manufacturing Productivity Database and the data on international outsourcing are obtained from the Annual Survey of Manufactures (ASM), whose industrial classification system is SIC for 1972 and 1987, but CPS data employs the census industrial classification system (CIC). By using the conversion bridge between 1972

¹⁹ There are 83 manufacturing industries in the CPS industrial code. Six of them do not have corresponding SIC codes and have thus had to be omitted. They are: not specified food industries (122); not specified metal industries (301); not specified machinery (332); not specified electrical machinery, equipment, and supplies (350); not specified professional equipment (382); and not specified manufacturing industries (392).

²⁰ In the results that are not shown, we also included the change in the R&D variable and the change in the white-collar to blue-collar ratio in our regressions, but the results are not significant for all kinds of workers.

workers.²¹ Total industrial R&D expenditure is recorded using two-digit SIC data. A part of the data in some years is being withheld to avoid the disclosure of information. We use the predicted values form simple regression estimation to fill out those blanks.

SIC and 1980 CIC data, this study converts all data classified by SIC into 1980 CIC data.

Table 1a summarizes the mean, standard deviation, and minimum and maximum values of the change in all variables during 1983-1987 and 1987-1991, and Table 1b the average value from 1982 to 1987 (or 1983 to 1987) and 1987 to 1991 (or 1987 to 1992). The corresponding industrial codes are presented in the Appendix.²² Job retention rates for all workers from 1983 to 1987 were about 54%. *Tobacco manufactures* had the highest job retention rate during 1983-1987. White-collar workers had a higher job retention rate than blue-collar workers during that period. The job retention rate for all workers was lower after 4 years. From 1987 to 1991, the job retention rate during 1987-1991.

Generally speaking, blue-collar workers had higher ratios of both labor union membership and union coverage than white-collar workers. *Household appliances* had the highest labor union membership and coverage ratio in 1983-1987, while if you only consider blue-collar workers during the same time, *Watches, clocks, and clockwork operated devices* had the highest union membership ratio and *Engines and turbines* had the highest union coverage ratio. Labor union membership and coverage ratios in *Dyeing & finishing textiles, except wool & knit goods* became the highest for both all workers and blue-collar workers during 1987-1991. From 1983-1991, both the ratio of labor union membership and the union coverage ratio fell. During 1983-1987, the average for workers in manufacturing industries with labor union membership was 5.2% and the ratio declined to 4.7% in 1987-1991. *Structural clay products* suffered the most both in terms of union membership and the coverage ratio in 1983-1987. *Watches, clocks, and clockwork operated devices* lost the most labor union membership and coverage for both all workers and blue-collar workers of union membership and the coverage ratio in 1983-1987. *Watches, clocks, and clockwork operated devices* lost the most labor union membership and coverage for both all workers and blue-collar workers in 1987-1991.

The average share of R&D expenditure in total output decreased slightly in the 1980s. *Guided missiles, space vehicles, and parts* and *photographic equipment and supplies* were characterized by the highest investment in R&D compared to their output value share in 1982-1987 and 1987-1992, respectively. For most of the 1980s, the white-collar to blue-collar ratio was increasing for workers as a whole. *Newspaper publishing and printing* had the highest white-collar to blue-collar ratios in both 1982-1987 and 1987-1992, while *Yarn, thread, and fabric mills* had the lowest white-collar to blue-collar ratios during the same periods of time. *Electronic computing equipment* and *Guided missiles, space vehicles, and parts* saw their white-collar to blue-collar ratios increase the most in 1982-1987 and 1987-1992, respectively.

Although the average increment of the international outsourcing share decreased slightly from 1982-1987 to 1992-1987, the change in the portion of outsourcing (based on both the difference between the broad and narrow definitions and the narrow definition) was still increasing from 1982-1987 to 1987-1992 in the industry that had the fastest rate of increase. *Watches, clocks, and clockwork-operated devices* saw their proportion of outsourcing (narrowly measured) increase the most in both 1982-1987 and 1987-1992. The growth rate of real output

 ²² See Table A-1 in the Appendix for a detailed description of industrial codes.
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went down over the two consecutive periods. The real output of *Motor vehicles and motor vehicle equipment* had the biggest increase in both 1982-1987 and 1987-1992.

Variable	Period	Mean	Std Dev	Min	Cod e	Max	Code
	83-87	54.27	10.99	25.93	230	82.93	130
$\hat{R}_{All Workers}$ (%)	87-91	49.51	10.95	18.18	220	78.05	210
<u>^</u>	83-87	52.31	11.91	22.22	230	83.33	130
$R_{Blue-collar}$ (%)	87-91	47.67	12.36	10.00	220	78.57	210
_	83-87	58.30	15.61	0.00	220	100.00	261, 361
$\hat{R}_{White-collar}$ (%)	87-91	54.78	13.86	16.67	361	100.00	381, 220
	83-87	-1.07	3.79	-11.54	252	14.29	140
$\Delta Un_{All Wor \ker s}^{mem}$ (%)	87-91	0.03	3.64	-10.53	381	11.62	190
	83-87	-0.69	6.49	-16.15	321	25.64	381
$\Delta Un^{mem}_{Blue-collar}$ (%)	87-91	-1.67	6.52	-33.33	381	23.02	190
	83-87	-1.19	3.82	-11.54	252	14.29	140
$\Delta U n_{All \ Wor \ker s}^{cov}$ (%)	87-91	0.07	3.55	-10.53	381	11.62	190
	83-87	-0.68	6.50	-16.15	321	25.64	381
$\Delta Un^{cov}_{Blue-collar}$ (%)	87-91	0.03	6.59	-33.33	381	23.02	190
	83-87	-1.60	3.80	-14.29	381	10.00	361
$\Delta Un_{White-collar}^{cov}$ (%)	87-91	0.03	2.77	-10.00	361	6.67	310
A 4	83-87	0.77	2.10	-5.16	381	5.69	140
$\Delta Age_{All Workers}$	87-91	0.44	0.17	-17.93	220	20.34	381
A 4	83-87	0.69	2.67	-12.82	381	7.27	252
$\Delta Age_{Blue-collar}$	87-91	0.06	3.67	-9.64	201	20.67	381
	83-87	0.62	3.82	-7.00	150	14.6	361
$\Delta Age_{White-collar}$	87-91	0.19	3.72	-12.1	140	7.56	152
	82-87	0.04	0.10	-0.28	391	0.43	151
$\Delta^{D}So$ (%)	87-92	0.04	0.11	-0.19	130	0.56	391
N~	82-87	0.05	0.16	-0.41	112	0.58	381
$\Delta^{N}So$ (%)	87-92	0.03	0.14	-0.41	162	0.77	381
A (1:11:	82-87	1925	5907	-3813	312	35410	351
Δy (billions in 1987)	87-92	827	2926	-881	261	18133	351
A E (in 1000-)	82-87	121	2303	-5845	312	14316	341
$\Delta E_{All \ Wor ker s}$ (in 1000s)	87-92	-327	1102	-7552	341	1631	212
$\Delta E_{Blue-collar}$ (in 1000s)	82-87	17	1234	-3416	312	5920	212
Blue-collar (=========)	87-92	-225	646	-3814	341	1118	212
	82-87	104	1256	-2428	312	9337	341
$\Delta E_{White-collar}$ (in 1000s)	87-92	-102	517	-3738	341	825	372
$\Delta Ip_{All Workers}$ (in \$1000)	82-87	5.11	1.82	0.74	100	10.76	130
	87-92	4.94	1.77	2.01	100	10.29	200
$\Delta Ip_{Rlva_{-collar}}$ (in \$1000)	82-87	4.31	1.78	-0.02	100	10.25	130
$\Delta Ip_{Blue-collar}$ (in \$1000)	87-92	3.73	1.56	-0.22	321	8.99	200
ΔIp_{White_collar} (in \$1000)	82-87	7.03	1.56	3.50	111	10.67	130
$\Delta Ip_{White-collar}$ (in \$1000)	87-92	7.39	2.02	3.21	152	13.01	181

Table 1a. Summaries of the Changes in All Variables

Variable	Year	Mean	Std Dev	Min	Code	Max	Code
$Un_{All Wor \ker s}^{mem}$ (%)	83-87	5.21	3.13	0	220,38 0	13.27	340
All workers	87-91	4.69	3.29	0	a	15.48	140
T T MOM	83-87	7.29	4.37	0	b	20.51	381
$Un_{Blue-Collar}^{mem}$ (%)	87-91	6.86	4.66	0	с	22.65	140
	83-87	5.63	3.29	0	380,22 0	14.22	340
$Un_{All\ Wor\ ker\ s}^{cov}$ (%)	87-91	5.07	3.33	0	201,22 0	15.48	140
	83-87	7.62	4.49	0	b	21.26	310
$Un^{\rm cov}_{Blue-Collars}$ (%)	87-91	7.29	4.60	0	201,22 0	22.65	140
$\mathbf{L}_{\rm L}$ COV (0()	83-87	2.33	2.29	0	d	8.33	292
$Un_{White-collar}^{cov}$ (%)	87-91	1.55	1.73	0	e	6.62	111
100	83-87	38.70	1.84	31.38	171	42.43	291
$Age_{All \ Wor \ker s}$	87-91	39.17	1.85	33.99	100	43.38	381
100	83-87	38.31	1.89	33.14	370	41.86	291
$Age_{Blue-collar}$	87-91	38.69	2.15	31.94	370	43.27	140
100	83-87	39.83	2.62	29.81	171	45.50	150
$Age_{White-collar}$	87-91	40.24	2.37	33.46	171	46.47	221
$D g_{r} D (0/)$	82-87	29.65	4.12	0.14	142	14.45	362
<i>R</i> & <i>D</i> (%)	87-92	26.80	3.15	0.16	132	9.15	380
$W \mid P(0/)$	82-87	47.96	34.03	12.59	142	182.74	171
W / B (%)	87-92	49.47	37.38	12.37	142	200.46	171
γ (billions in 1987)	83-87	680	1234	6	261	9505	351
<i>y</i> (01110113 111 1907)	87-91	817	1554	5	261	11436	351
$E_{All Workers}$ (in 1000s)	83-87	5189	7747	68	220	42057	342
$E_{All Wor \ker s}$ (in 1000s)	87-91	5086	7777	62	220	39757	342
E (in 1000s)	83-87	3448	5146	57	220	29554	351
$E_{Blue-collar}$ (in 1000s)	87-91	3343	5087	51	220	29542	351
$E_{White coller}$ (in 1000s)	83-87	1741	2925	12	220	14716	341
$E_{White-collar}$ (in 1000s)	87-91	1742	3097	12	220	17515	341

Table 1b. Summaries of the Average of Variables in 1982, 1983, and 1987

a: 141, 201, 220, 232, 370. b: 370,220, 380. c: 220, 232, 141, 370, 201. d:110, 130, 132, 140, 141, 150, 181, 190, 191, 201, 220, 221, 222, 230, 231, 232, 241, 252, 261, 281, 380, 390.

e:101, 102, 121, 130, 132, 140, 141, 150, 152, 161, 180, 201, 211, 220, 221, 222, 230, 231, 232, 241, 252, 261, 270, 280, 281, 290, 320, 370, 381, 391.

In 1987-1992, the number of employed workers, both blue-collar and white-collar, decreased. The *Radio, TV, and communication equipment* industries saw their numbers of both blue-collar and white-collar employees decrease the most in 1987-1992. The number of employed workers in *Construction and material handling machines*, both blue collar and white collar, decreased the most in 1982-1987. The demand for blue-collar labor in *Miscellaneous plastics products* increased the most in both 1982-1987 and 1987-1992. Although average income per person also decreased from 1982-1987 to 1987-1992, the income per person of white-collar workers was greater in 1987-1992 than in 1982-1987. The *Drugs* industry was the industry whose income per person for white-collar workers increased the most in 1987-1992. *Tobacco manufactures* saw both its blue-collar and white-collar workers' income per person increase the most in 1982-1987.

4. Empirical Results

The regression estimation results are organized and analyzed according to the workers' skill levels. *All workers* are considered first to see the overall effects, and then *Blue-collar workers* and *White-collar workers* are considered separately. The regression settings for *All workers* and *Blue-collar workers* are the same. The correlation coefficients between all independent variables show that the number of employed workers is highly correlated with real output, and the labor union membership ratio is correlated with the union coverage ratio.²³ Thus, as shown in Tables 2 and 3, only regression 2 uses real output, and the rest of the regressions use the number of employed workers. Similarly, only regressions 3 and 5 have the union membership ratio as an explanatory variable. The union coverage ratio will be discussed with the rest of the regressions. Regression equations 4 and 5 have the interaction term for international outsourcing and unionization.

The difference in the regression setting of *White-collar* to *All workers* and *Blue-collar workers* is that there is no union membership ratio in the regressions. As shown in Table 4, there are three regressions instead of five. Again, the real output is still highly correlated with the number of white-collar workers employed. Thus, regression 2 takes real output into consideration and employment will be discussed in the rest of the regressions, i.e., regressions 1, and 3. Regression equation 3 has the interaction term for international outsourcing and unionization. R & D is included in the regressions and has important implications behind the results.²⁴

4.1 All Workers

The results for all workers are presented in Table 2. Both the changes in and value of real output and employment have no significant effect on workers' job retention rates. Although the change in labor union membership and the coverage ratio had no significant effect on the job retention rate for all workers in the 1980s, this study still found a support to the claim that labor unions enhance the workers'

²³ The correlation coefficients for all workers, blue-collar workers, and white-collar workers are available from the authors upon request.

²⁴ In unreported results, we found that the coefficients of R&D are statistically insignificant when data of *All workers* and *Blue-collar workers* are employed. Regression results of *Blue-collar workers and All workers*' equation contains R&D as a independent variable are available from the authors upon request.

job stability since the average level of labor union membership and the coverage ratio had a significant positive impact on labor's job retention rate.

Independent Variable	1	2	3	4	5
	-0.168		-0.176	-0.476**	-0.474**
$\Delta E_{All \ Wor \ ker \ s}$	(1.50)		(1.59)	(3.09)	(3.06)
_	0.000		0.000	0.027	0.026
E _{All Workers}	(0.01)		(0.00)	(0.96)	(0.93)
		0.025			
Δy		(0.29)			
у		0.025			
<i>y</i>		(1.29)			
A T T COV	1.347*	1.425*		0.629	
$\Delta Un^{ m cov}_{All\ Wor\ m ker\ s}$	(1.85)	(1.94)		(0.50)	
TT COV	3.536**	3.154**			
$Un_{All Wor \ker s}^{cov}$	(3.89)	(3.29)			
ATT MPM			1.242		0.537
$\Delta Un^{mem}_{All\ Wor\ ker\ s}$			(1.59)		(0.44)
тт тет			3.592**		
$Un_{All Wor \ker s}^{mem}$			(3.99)		
W / B	0.216**	0.212**	0.229**		
W / D	(3.66)	(4.13)	(3.99)		
A T	1.414**	1.015**	1.362**	1.963**	1.988**
$\Delta Ip_{All Workers}$	(2.48)	(9.76)	(2.40)	(2.33)	(2.35)
	1.121**	1.258**	1.059**		
$\Delta Age_{All Workers}$	(2.33)	(2.71)	(2.17)		
	4.290**	4.538**	4.299**		
$Age_{All Workers}$	(8.83)	(9.69)	(9.01)		
	-0.712**	-0.636**	-0.710**	-2.194**	-2.227**
$\Delta ^{D}So \times 100$	(3.12)	(2.69)	(3.10)	(2.71)	(2.80)
NG 100	-0.255	-0.135	-0.274	-0.331	-0.334
$\Delta^{N}So \times 100$	(1.35)	(0.80)	(1.45)	(1.20)	(1.21)
$\Delta^D So * Un_{All Wor \ker s}^{cov}$				26.253	
				(1.40)	
$\Delta^D So * Un_{All Wor \ker s}^{mem}$. /	28.902
					(1.47)
	-16.270***	-17.423**	-16.289**	-0.607**	-0.604**
Constant	(9.60)	(10.41)	(9.76)	(2.19)	(2.18)
R-Squared	0.650	0.649	0.649	0.279	0.281

Table 2. Regression Results of All Workers' Job Retention Rates

Note: N=154 in each specification and standard errors in all regressions are robust to heteroskedasticity. All regressions include a time dummy variable and are weighted by the average employment share of all manufacturing industries. The figures in parentheses are absolute values of *t*-statistics. The superscripts over the coefficient values indicate the levels of statistical significance as follows: **, at 5% or higher; *, 10%.

The main focus of this paper is international outsourcing. Outsourcing (narrowly measured) was found to have no significant effect on job retention rates, but outsourcing (difference), which is the difference between the broad and narrow

measures of outsourcing, had a significantly negative impact on the workers' job retention rates. Although the coefficient of the interaction term for international outsourcing and labor unions is insignificant,²⁵ it could be a mixed result from both

Labor in the industries that have higher white-collar to blue-collar ratios had higher job retention rates in the 1980s, which implies that workers have better job security if they work for the industries that adopt higher technology and have higher average education levels. As we expected, average income per person and average age had a positive and significant effect on the workers' job stability.

4.2 Blue-Collar Workers

blue-collar workers and white-collar workers.

After checking the correlation coefficients between all independent variables using blue-collar workers only, this study has the same regression settings for blue-collar workers as for all workers.

Table 3 presents the results of regressions employing data only for blue-collar workers. Similar to the results in Table 2, the change in real output had no effect on the workers' job stability. Surprisingly there was a significantly negative relationship between the change in employment and job retention rates for blue-collar workers. This result implies that when employers lay off their employees, the workers' stability will not be reduced. One explanation for this is that because unions follow the "last-in, first-out" or "last hired, first fired" rule, the workers who have a higher likelihood of being laid off are those who are new hires.

International outsourcing (difference) decreased the number of blue-collar workers' jobs in the 1980s. Labor unions, however, increased blue-collar workers' job retention rates and reduced the negative impact of international outsourcing on blue-collar workers' job stability. Blue-collar workers in industries with higher ratios of white-collar to blue-collar workers and higher incomes had higher job retention rates in the 1980s.

4.3 White-Collar Workers

Table 4 shows the regression equations in regard to white-collar workers. White-collar workers' job retention rates, unlike blue-collar workers' job retention rates, were not affected by international outsourcing (both narrowly measured and different), income, and labor unions. We did not find any evidence to support the view that job-to-job changes or reemployment caused by international outsourcing among white-collar workers is important enough to affect white-collar workers' job stability. Our results of white-collar workers are in line with Munch (2010) but not Geishecker (2008). The average employment and output of industries has a surprisingly negative effect on white-collar workers' job retention rates. One possible explanation is that larger industries provide larger job markets which provide more chances of reemployment and job matching to white-collar workers.

²⁵ Since the average and not the change in labor union membership and the coverage ratio had a significant impact on labor's job retention rate, the interaction term between foreign outsourcing and unionization only employs average unionization ratios.

By combining this with the weak possible effect of income per person, we can say that it is likely that a white-collar worker in an industry, which has a greater than average size, will easily leave his or her job for better pay.

Independent Variable	1	2	3	4	5
	-0.315**		-0.329**	-0.657**	-0.659**
$\Delta E_{\it Blue-Collar}$	(2.54)		(2.63)	(4.63)	(4.62)
$E_{Blue-Collar}$	0.006		0.005	0.031	0.031
L Blue-Collar	(0.29)		(0.22)	(1.07)	(1.07)
A		-0.171*			
Δy		(1.88)			
у		0.026			
		(1.10)			
A T T COV	0.668	0.735		0.294	
$\Delta Un_{Blue-Collar}^{ m cov}$	(1.21)	(1.30)		(0.35)	
TT COV	3.668**	3.305**			
$Un_{Blue-Collar}^{cov}$	(4.17)	(3.50)			
A T T., mem			0.669		0.244
$\Delta Un^{mem}_{Blue-Collar}$			(1.21)		(0.29)
тт тет			3.790**		
$Un_{Blue-Collar}^{mem}$			(4.17)		
W / B	0.210**	0.229**	0.223**		
$H \neq D$	(3.04)	(3.30)	(3.16)		
$\Delta I p_{Blue-Collar}$	1.549**	1.542**	1.525**	2.545**	2.551**
P Blue-Collar	(2.89)	(5.52)	(2.82)	(3.74)	(3.72)
$\Delta Age_{Blue-collar}$	0.686	0.739	0.691		
Blue-collar	(1.26)	(1.37)	(1.27)		
	3.138**	3.436**	3.005**		
$Age_{Blue-collar}$	(5.34)	(5.55)	(4.92)		
$\Delta^{D}So \times 100$	-0.788**	-0.733**	-0.784**	-2.604**	-2.605**
Δ 50 ×100	(3.01)	(2.72)	(3.03)	(3.07)	(3.10)
NG 100	-0.348*	-0.241	-0.370*	-0.463	-0.471
$\Delta^{N}So \times 100$	(1.78)	(1.38)	(1.90)	(1.49)	(1.52)
$\Delta^D So * Un_{Blue-Collar}^{cov}$				29.654*	
				(1.94)	
$\Delta^{D}So * Un^{mem}_{Blue-Collar}$					31.069**
					(1.98)
Constant	-12.199**	-13.544**	-11.702**	-0.787**	-0.789**
Constant	(5.98)	(6.15)	(5.52)	(2.82)	(2.84)
R-Squared	0.606	0.596	0.606	0.292	0.293

 Table 3. Regression Results of Blue-Collar Workers' Job Retention Rates

Note: N=154 in each specification and standard errors in all regressions are robust to heteroskedasticity. All regressions include a time dummy variable and are weighted by the average employment share of all manufacturing industries. The figures in parentheses are absolute values of *t*-statistics. The superscripts over the coefficient values indicate the levels of statistical significance as follows: **, at 5% or higher; *, 10%.

Independent Variable	1	2	3
$\Delta E_{White-Collar}$	-0.154		-0.299
	(0.69)		(1.21)
F	-0.121**		-0.045
$E_{White-Collar}$	(2.66)		(1.08)
A		0.220	
Δy		(1.54)	
у		-0.106**	
<i>y</i>		(2.29)	
	1.759	2.079	-0.807
$\Delta Un_{White-Collar}^{ m cov}$	(0.57)	(0.76)	(0.32)
TT COV	2.105	4.795*	
$Un_{White-collar}^{cov}$	(0.81)	(1.81)	
R&D	4.969**	4.201**	
	(3.99)	(3.43)	
$W \mid B$	0.029	-0.087	
W / D	(0.19)	(0.77)	
$\Delta Ip_{_{White-Collar}}$	1.380	1.243	-1.074
P White-Collar	(0.86)	(0.78)	(0.88)
$\Delta Age_{White-collar}$	1.089	1.046	
White-collar	(1.15)	(1.03)	
	3.488**	3.407**	
$Age_{White-collar}$	(3.89)	(4.39)	
A P S = 100	-0.078	-0.193	-1.486
$\Delta ^{D}So \times 100$	(0.16)	(0.41)	(1.58)
$\Delta^{N}So \times 100$	0.299	0.673	0.367
Δ SO ×100	(0.46)	(1.02)	(0.50)
A D C = A I L COV			7.506
$\Delta^{D}So * Un_{White-collar}^{cov}$			(0.16)
Constant	-12.328**	-11.437**	0.733**
Constant	(3.82)	(4.25)	(2.07)
R-Squared	0.316	0.299	0.113

Table 4. Regression Results of White-collar Workers' Job Retention Rates

Note: Since there is no white-collar worker found in the industry code 220 in the year 1987, N =152 in each specification. Standard errors in all regressions are robust to heteroskedasticity and correlation in the errors within two-digit SIC industries. All regressions include a time dummy variable and are weighted by the average white-collar employment share of all manufacturing industries. The figures in parentheses are absolute values of *t*-statistics. The superscripts over the coefficient values indicate the levels of statistical significance as follows: **, at 5% or higher; *, 10%.

The share of R&D expenditure to total output, however, has a positive and significant impact on white-collar workers' job retention rates. It seems that the explanation of Munch (2010) can also be applied here. The industries that are willing to invest in R&D also provide more job training to their employees and are also willing to retain their more highly-skilled workers.

5. Concluding Remarks

The study of the impact of international outsourcing on job security or job stability has been one of the fastest growing areas in economic research in recent decades. For instance, Munch (2010) employed Danish manufacturing data, Egger et al. (2007) analyzed Austrian male workers over the period 1988-200, and Geishecker (2008) employed German data in the 1990s. Most of them found that international outsourcing more or less led to a deterioration in the job security of unskilled labor.

Apart from the previous literature, this paper uses another measure, which is the worker's job retention rate, to evaluate the impact of international outsourcing on the worker's job stability. Diebold et al. (1997) found that the job stability of blue-collar workers in the U.S. tended to decrease in the 1980s. Combining the effects of international outsourcing on wage inequality, which are analyzed along with the results of Diebold et al. (1997), raises the question of whether international outsourcing should be responsible for the decrease in the blue-collar workers' job retention rate in the 1980s. In addition, assessing whether labor unions can moderate the effects of international outsourcing on workers' job stability is another issue in this study.

This study employs CPS data, the NBER Productivity Database, and the outsourcing data in Feenstra and Hanson (1999) to perform the analysis. The results show that the broad measure of international outsourcing decreases blue-collar workers' job retention rates. This part of our results is in line with the results in Munch (2010). Labor unions not only have a positive impact on the blue-collar workers' job stability, but moderate the negative impact of foreign outsourcing on the blue-collar workers' job stability. Workers in the industries with higher average income and age also are also characterized by higher job retention rates. There is no effect of international outsourcing on the white-collar workers' job stability but the industries with higher shares of expenditure on R&D in relation to their output are more willing to retain their white-collar workers.

The analyses in this study are based on industry-level data. Future studies could employ plant-level or establishment-level data to analyze this issue. At the firm level, the impact of outsourcing on employees can be analyzed from one decision to another. In particular, the negotiations that take place between employers and labor unions give rise to an interesting topic to study.

Acknowledgements

Financial support by Ministry of Science and Technology, grant number MOST NSC 102-2410-H-004 -015, is gratefully acknowledged. The usual disclaimer applies.

Appendix: Robustness Check

As introduced in the paper, the historical job retention rate can be used under an unstable survival function and nonconstant arrival rate. Because the sampling error is noticeable in this study, cross-sectional computation is employed instead. Here we still report the regression results for equation (3) with a historical job retention rate for All workers and Blue-collar workers.²⁶

Table A-2 reports the regression results for the *All workers'* historical job retention rate. The setting of Table A-2 is the same as the one in Table 2 for the purpose of comparison. The change in unionization has a positive and significant effect on *All workers'* job retention rates but the effect of the average unionization is positive but not significant. Average age and outsourcing (difference) still have a strong influence on *All workers'* job stability. Workers in industries with comparatively weak unionization have lower job security than those in industries with strong unionization. Thus, the main results in Table 2 still hold.

The regression results for the *Blue-collar workers*' historical job retention rate are presented in Table A-3. Similar to the results in Table 3, the average unionization has a positive and significant impact on the blue-collar workers' job retention rates. Slightly different from the results for the international outsourcing effect in Table 3, both the broad measure of outsourcing and the narrow measure of outsourcing have a negative impact on the blue-collar workers' job stability. Although unionization cannot moderate the negative impact of foreign outsourcing, workers in industries that have comparatively strong unionization have better job stability. Increasing personal income and working for an industry with a higher average age can significantly enhance the blue-collar workers' job stability. By contrast, the ratio of white-collar to blue-collar workers has no significant effect on the workers' job retention rate.

To sum up, the conclusions based on the results of using the historical job retention rate are in essence similar to those obtained from using cross-sectional job retention rates. International outsourcing when at least broadly measured has a negative impact on the blue-collar workers' job stability. The other determinants such as income per person and average age still give rise to similar results.

 $^{^{26}}$ Due to the incompleteness and inconsistency of the historical retention rate for white-collar workers, there is little reason to report the results here. Basically, the share of R&D expenditure in total output still has a positive effect on the white-collar workers' job stability, while the impact of international outsourcing is negative but is sensitive to the setting of the regressions.

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<i>Table A-1</i> . Indust	Ital Coues			
Meat products	Printing, publishing 100 & allied industries, except newspapers	¹⁷² Miscellaneous wood products	241 Metalworking machinery	320
Dairy products	101 Plastics, synthetics, and resins	180 Furniture and fixtures	242 Office and accounting machines	321
Canned and preserved fruits and vegetables	102 Drugs	181 Glass and glass products	250 Electronic computing equipment	322
Grain mill products	110 Soaps and cosmetic	products	251 Machinery, except electrical, n.e.c.	331
Bakery products	111 Paints, varnishes, ar related products	d 190 Structural clay products	252 Household appliances	340
Sugar and confectionery products	112 Agricultural chemicals	191 Pottery and related products	Radio, T.V., and 261 communication equipment	341
Beverage industries	Industrial and 120 miscellaneous chemicals	Miscellaneous 192 nonmetallic mineral & stone products	Electrical machinery, 262 equipment, and supplies, n.e.c.	342
Miscellaneous food preparations & kindred products	121 Petroleum refining	Blast furnaces, 200 steelworks, rolling & finishing mills	Motor vehicles and 270 motor vehicle equipment	351
Tobacco manufactures	Miscellaneous 130 petroleum and coal products	201 Iron and steel foundries	271 Aircraft and parts	352
Knitting mills	132 Tires and inner tube	s 210 Primary aluminum industries	272 Ship and boat building and repairing	360
Dyeing & finishing textiles, except wool & knit goods	Other rubber 140 products, and plastic footwear and belting	industries	280 Railroad locomotives and equipment	361
Floor coverings except hard surface	141 Miscellaneous plastics products	$212 \frac{\text{Cutlery, hand tools,}}{\text{and other hardware}}$	Guided missiles, 281 space vehicles, and parts	362
Yarn, thread, and fabric mills	142 Leather tanning and finishing	220 Fabricated structural metal products	Cycles and 282 miscellaneous transportation equipment	370
Miscellaneous textile mill products	150 Footwear, except rubber and plastic	221 Screw machine products	Scientific and 290 controlling instruments	371
Apparel and accessories, except knit	151 Leather products, except footwear	222 Metal forgings and stampings	291 Optical and health services supplies	372
Miscellaneous fabricated textile products	152 Logging	230 Ordnance	Photographic 292 equipment and supplies	380
Pulp, paper, and paperboard mills	160 Sawmills, planning, mills, and millwork	Miscellaneous 231 fabricated metal products	Watches, clocks, and 300 clockwork operated devices	381
Miscellaneous paper and pulp products	161 Wood buildings and mobile homes	232 Engines and turbines	310 Toys, amusement, and sporting goods	390
Paperboard containers and boxes	162	Farm machinery and equipment	Miscellaneous 311 manufacturing industries	391
Newspaper publishing and printing	171	Construction and material handling machines	312	

Appendix

Table A-1. Industrial Codes

Note: n.e.c. is an abbreviation for not elsewhere classified.

Independent Variable	1	2	3	4	5
A E	0.088		0.081	-0.222	-0.219
$\Delta E_{All \ Wor \ ker \ s}$	(0.37)		(0.35)	(0.96)	(0.94)
Г	0.031		0.029	0.060	0.058
$E_{All \ Wor \ ker \ s}$	(0.98)		(0.92)	(1.44)	(1.39)
Δy		0.097			
Δy		(0.81)			
у		0.063*			
		(1.93)			
$\Delta Un_{All \ Wor ker s}^{cov}$	3.676**	3.706**		2.714	
$\Delta O n_{All}$ Workers	(2.25)	(2.32)		(1.61)	
$Un_{All \ Wor \ ker \ s}^{cov}$	3.485*	2.747			
All Workers	(1.78)	(1.51)			2 2 2 7
$\Delta Un^{mem}_{All\ Wor \ker s}$			3.226**		2.287
All Workers			(2.04)		(1.44)
$Un^{mem}_{All \ Wor \ ker \ s}$			3.503*		
	0.159	0.122	(1.70)		
W / B	0.158 (1.12)	(0.94)	0.180		
	(1.12) 2.459*	(0.94) 2.106**	(1.28) 2.373*	2.737**	2.756**
$\Delta I p_{All Wor \ker s}$	(1.91)	(4.06)	(1.82)	(2.58)	(2.60)
	-0.017	0.130	-0.131	(2.50)	(2.00)
$\Delta Age_{All \ Wor \ker s}$	(0.02)	(0.14)	(0.14)		
	4.366**	4.381**	4.388**		
$Age_{All Workers}$	(3.75)	(4.08)	(3.70)		
	-0.858**	-0.749**	-0.831**	-2.173**	-2.170**
$\Delta^{D}So \times 100$	(2.75)	(2.37)	(2.62)	(2.21)	(2.21)
Ng 100	-0.420	-0.427	-0.442	-0.551	-0.556
$\Delta ^{N}So \times 100$	(1.02)	(1.10)	(1.09)	(1.26)	(1.28)
				21.064	
$\Delta^D So * Un_{All Wor \ker s}^{cov}$				(0.89)	
DC. IT mem					22.928
$\Delta^D So * Un_{All Wor \ker s}^{mem}$					(0.89)
Constant	-16.978**	-17.513**	-17.028**	-1.050**	-1.038**
Constant	(4.12)	(4.59)	(4.06)	(2.52)	(2.48)
R-Squared	0.382	0.397	0.376	0.176	0.172

Table A-2. Regression Results of All Workers' Historical Job Retention Rates

Note: N=154 in each specification and standard errors in all regressions are robust to heteroskedasticity. All regressions include a time dummy variable and are weighted by the average employment share of all manufacturing industries. The figures in parentheses are absolute values of *t*-statistics. The superscripts over the coefficient values indicate the levels of statistical significance as follows: **, at 5% or higher; *, 10%.

Rates						
Independent Variable	1	2	3	4	5	
ΛF	-0.076		-0.093	-0.364*	-0.366*	
$\Delta E_{Blue-Collar}$	(0.37)		(0.46)	(1.82)	(1.83)	
F	0.036		0.034	0.065	0.064	
$E_{Blue-Collar}$	(0.96)		(0.90)	(1.35)	(1.34)	
		-0.080				
Δy		(0.55)				
у		0.062				
<i>.</i>		(1.51)				
A T T COV	1.091	1.115		0.932		
$\Delta Un_{Blue-Collar}^{cov}$	(1.04)	(1.06)		(0.87)		
TT COV	3.938**	3.286**				
$Un_{Blue-Collar}^{cov}$	(2.36)	(2.07)				
АТ Т тет			1.020		0.780	
$\Delta Un^{mem}_{Blue-Collar}$			(0.96)		(0.72)	
т т тет			4.102**			
Un ^{mem} _{Blue-Collar}			(2.35)			
W / B	0.141	0.110	0.141			
,, , <u>D</u>	(0.96)	(0.74)	(0.93)			
$\Delta Ip_{Blue-Collar}$	2.608**	2.585**	2.585**	3.212**	3.225**	
- r Blue-Collar	(2.61)	(2.56)	(2.56)	(3.67)	(3.67)	
$\Delta Age_{Blue-collar}$	-0.558	-0.515	-0.566			
S - Blue-collar	(0.56)	(0.52)	(0.56)			
100	2.436**	2.609**	2.280**			
$Age_{Blue-collar}$	(2.57)	(2.71)	(2.30)			
$\Delta^{D}So \times 100$	-0.967**	-0.885**	-0.956**	-2.159*	-2.236*	
A 50 ×100	(2.94)	(2.69)	(2.92)	(1.85)	(1.90)	
Δ ^N So ×100	-0.663*	-0.654**	-0.686*	-0.810*	-0.815*	
A 50 ×100	(1.85)	(1.96)	(1.93)	(1.93)	(1.95)	
$\Delta^{D}So * Un_{Blue-Collar}^{cov}$				17.730		
				(0.86)		
$\Delta^{D}So * Un^{mem}_{Blue-Collar}$					20.153	
					(0.92)	
Constant	-10.073**	-11.171**	-9.489**	-1.199**	-1.196**	
	(2.98)	(3.18)	(2.69)	(2.51)	(2.50)	
R-Squared	0.319	0.327	0.320	0.172	0.172	

Table A-3. Regression Results of Blue-Collar Workers' Historical Job Retention

Note: N=154 in each specification and standard errors in all regressions are robust to heteroskedasticity. All regressions include a time dummy variable and are weighted by the average employment share of all manufacturing industries. The figures in parentheses are absolute values of *t*-statistics. The superscripts over the coefficient values indicate the levels of statistical significance as follows: **, at 5% or higher; *, 10%.

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