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### 경제학석사학위논문

## Returns to Tenure and Labor Market Mobility in Korea

한국의 임금 연공성과 노동시장 이동성

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# Returns to Tenure and Labor Market Mobility in Korea

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in Korea

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Abstract

In this paper, I estimate the returns to tenure of the Korean labor market and investigate the relationship between returns to tenure and labor market mobility. I begin with the two methods introduced by Altonji and Shakotko (1987) and Topel (1991) to estimate returns to tenure in Korea and the US using panel data sets and confirm that both returns to tenure and job mobility are higher in Korea. Next, the industry- and occupation-wise returns to tenure are estimated for Korea and are found to be widely variable across different divisions. Finally, the correlation coefficients between returns to tenure and job mobility among industries and occupations are estimated and show almost zero or slightly positive correlations. These patterns contradict the conventional wisdom, which predicts a negative correlation, and suggest the possibility that returns to tenure work as an incentive device to retain workers in Korea.

Keywords: Returns to Tenure, Job Mobility, Korean Labor Market.

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#### 1 Introduction

Wage returns to job tenure and job mobility are intimately related in a labor market. Workers leaving or choosing jobs for better wages and firms offering higher wage contracts to attract or incentivize workers (Lazear, 1979; Burdett and Coles, 2003) are widely accepted concepts in labor economics. However, there has been little research conducted on the correlation between returns to tenure and labor market mobility, which would be essential to understanding the mechanism behind wage structures and labor dynamics in equilibrium.

In this paper, I investigate the relationship between returns to employer tenure and job mobility in a labor market, using panel data sets from Korea and the US. Conventional wisdom says that job-to-job mobility may decrease as returns to tenure increase. Returns to tenure are interpreted as the firm-specific component of wages that would be lost should a worker leave the firm (Topel, 1991; Deelen, 2012), making workers less mobile as the returns increase. This argument stems from the classic interpretation of wage growth during one's job tenure being the result of the accumulation of firm-specific human capital, which was first suggested by Becker (1964). Under this assumption, it is reasonable to believe that a negative correlation between returns to tenure and job mobility may appear in a labor market.

At the same time, there has been a wide variety of theories proposed—other than firm-specific human capital—to explain the positive returns to tenure that have been empirically confirmed in the US labor market (Topel, 1991; Altonji and Williams, 2005; Buchinsky et al., 2010). Most such theoretical explanations focus on the role of returns to tenure as an incentive device

(Lazear, 1979; Burdett and Coles, 2003): firms might adopt relatively high returns to tenure to give workers incentives to exert more effort or simply to stay. The consequent tendency toward wage-backloading has also been repeatedly depicted in the literature on labor market contracts (Burdett and Coles, 2003; Shi, 2009; Balke and Lamadon, 2020), but if some industries or occupations experience relatively high mobility of workers, zero or positive correlations may appear between returns to tenure and mobility across firms in equilibrium.

By investigating the actual relationship between wage structures and job mobility, I examine the validity of conventional wisdom. For this, I first estimate the returns to tenure in Korea and the US. In Korea, relatively high returns to tenure, which have been addressed in previous research, are suspected of causing a labor market mismatch that hampers the labor productivity growth of the nation. However, unlike the research on the US labor market, most papers investigating the Korean labor market have failed to consider unobserved heterogeneity using appropriate estimation methods and sufficient data (Hwang et al., 2005; Kim et al., 2015; Park, 2018).

In the current paper, the estimation methods from Altonji and Shakotko (1987) and Topel (1991), which control for unobserved heterogeneity and have been used widely throughout the literature, are employed to overcome these limitations. Furthermore, panel data sets from each country—the Korean Labor & Income Panel Study (KLIPS) and the US Panel Study of Income Dynamics (PSID)—are used, details of which will be described later. From the estimations, I determine that Korea has a relatively high return to tenure,

while the US is less mobile, which contradicts the conventional wisdom. This may be interpreted as evidence of firms paying a premium for seniority in order to retain workers, as demonstrated in Beffy et al. (2006).

To further investigate the returns to tenure in Korea, industry- and occupation-wise returns to tenure are estimated, and it is confirmed that the returns are widely variable across different industries and occupations. Using these estimates, I calculate the correlations between mobility and returns to tenure of different industries and occupations and find that they are slightly positive or almost zero. The findings show that the conventional wisdom of decreasing mobility with increasing returns to tenure may not appear in an equilibrium, possibly because of the role of wages as an incentive device for retaining workers.

The paper proceeds as follows. Section 2 provide a brief review of the literature, Section 3 explains the two main empirical methods for estimation and an overview of the data used, Section 4 presents and analyzes the results, and Section 5 concludes.

#### 2 Literature Review

In this section, I review the related literature, starting with the theoretical explanations that will help in understanding positive returns to employer tenure in a labor market. The most prominent approach is from the human capital theory of Becker (1964), which distinguishes general from specific human capital. Unlike general human capital, which is equally valuable in any firm, specific human capital can only be utilized and accumulated in specific firms

and thus contribute to greater firm-specific earnings.

Most other theories focus on the role of positive returns to tenure as an incentive device. For example, in Lazear (1979), it is shown that firms defer payment by paying young workers less and older workers more to solve the agency problem, while contract theory has offered persuasive explanations in several papers (Burdett and Coles, 2003; Shi, 2009; Balke and Lamadon, 2020)—according to the literature, in equilibrium, firms may use wage-tenure contracts that imply wage increases with employer tenure to incentivize a worker to stay.

Thus, past research suggests a relationship between mobility and wage returns. Other conditions being equal, it is reasonable for a worker to be less mobile as returns to tenure increase, considering firm-specific human capital. However, if we take into account firm heterogeneity and the incentive motives of firms, such that some may have to deal with more mobile workers and have methods to retain them, then an insignificant or positive correlation between returns to tenure and job mobility would also be plausible. In this paper, I explore this possibility.

Empirically, the existence of positive returns to tenure has been repeatedly supported in the literature. Since Altonji and Shakotko (1987) and Topel (1991) started the discussion on estimating returns to tenure by properly controlling unobserved heterogeneity, there have been numerous attempts to decompose wage growth with variations on the estimation methods from the two papers (Altonji and Williams, 2005; Kambourov and Manovskii, 2009; Deelen, 2012). Following the literature, I mainly use the two original methods, and the

details are explained in the next section.

Although some papers have explicitly considered job mobility in the estimation of wage returns (Buchinsky et al., 2010), little research has directly investigated the correlation between returns to tenure and job mobility. Of the papers that have, Beffy et al. (2006) estimates the returns to tenure in France with the empirical method from Buchinsky et al. (2010) and finds that both returns to tenure and job mobility are lower in France than in the US. Additionally, using an equilibrium search model with wage-tenure contracts from Burdett and Coles (2003), the paper explains that the relationship appears because high returns to tenure have a clear incentive effect in high-mobility countries, which is in line with the results of the current research. A second paper investigates the relationship between wage structure and job changes in the Netherlands and determines that steep wage-tenure profiles are correlated with low mobility (Deelen, 2012). This does not contradict the results of the current paper because the disparity would be due to the differences between two distinct economies. Rather, the results can be interpreted as another useful resource for later comparative studies on international labor markets to help understand the relationship between wage structures and labor market mobility.

In the Korean labor market, there have been repeated attempts to estimate returns to tenure (Ryoo, 2002; Hwang et al., 2005; Lee et al., 2008; Kim et al., 2015; Kang et al., 2016; Park, 2018), and much of the research has found them to be relatively high. However, most such papers fail to control for unobserved heterogeneity with appropriate methods or lack sufficient data for a thorough

investigation. In the current paper, I adopt two estimation methods and use panel data sets of sufficient volume to overcome these limitations.

#### 3 Data and Methodology

#### 3.1 Method

In order to estimate the returns to tenure, I use two different estimation methods, from Altonji and Shakotko (1987) and Topel (1991). Assume that the wage equation to be estimated is

$$w_{ijt} = X_{ijt}\beta_1 + T_{ijt}\beta_2 + \epsilon_{ijt}.$$

where  $w_{ijt}$  is the log real hourly wage of worker i in job j at period t, and  $X_{ijt}$  and  $T_{ijt}$  denote overall labor market experience and tenure with the employer, respectively. The model also includes dummies for demographic characteristics and a term for the square of labor market experience.

 $\beta_1$  and  $\beta_2$  of the wage equation represent returns to experience and tenure. In estimating the returns, unobservable individual- and match-specific components may generate biases; to analyze this issue and to introduce two methods of controlling unobservable components, one can decompose the error term as

$$\epsilon_{ijt} = \phi_{ijt} + \mu_i + v_{ijt},$$

following Topel (1991).  $\epsilon_{ijt}$  is the error term and consists of a match-specific effect  $\phi_{ijt}$ , an individual-specific effect  $\mu_i$ , and  $\nu_{ijt}$  which accounts for random

shocks and measurement error. For example, workers may have different qualities of match with their employers, resulting in different wages, or workers with the same observable characteristics may have different learning abilities. These factors cannot be observed in ordinary survey data, yet it would be reasonable to consider these components to be correlated with tenure or experience. It is well-known that a productive match is less likely to end, and one might therefore argue that the match-specific component  $\phi_{ijt}$  is positively correlated with tenure. Similarly, as a more productive worker would receive higher wages and stay longer in a job, the individual-specific fixed effect might also be positively correlated with tenure.

I will estimate the wage equation first with the OLS. However, the OLS will likely be biased due to the unobserved effects described above. To tackle this problem, I will introduce two estimators from the literature: an instrumental variable (IV) estimator from Altonji and Shakotko (1987) and two-step first-difference (2SFD) estimator form Topel (1991). First, the IV procedure proposed by Altonji and Shakotko (1987) considers the match-specific error to be the fixed "job effects" for each match ( $\phi_{ijt} = \phi_{ij}$ ). Under this assumption, the method uses the deviations of the tenure variables around their means for the sample observations of a given match to be instrumental variables, since they are uncorrelated with either match-specific or individual-specific error components. Specifically, let  $\overline{T}_{ij}$  be the mean of tenure for individual i in job j, and let the instrumental variable be  $\widetilde{T}_{ijt} = T_{ijt} - \overline{T}_{ij}$ . Then, by construction,  $\widetilde{T}_{ijt}$  is orthogonal to  $\phi_{ij}$  and  $\mu_i$ .

In Topel (1991)'s 2SFD, the estimation proceeds in two steps. In the first

step, I estimate the combined effect of the experience and tenure terms by applying OLS to the first differences of the wage equation for those who stay, using the fact that fact that  $\Delta X = \Delta T = 1$ :

$$w_{iit} - w_{iit-1} = B + \epsilon_{iit} - \epsilon_{iit-1}$$

where  $B = \beta_1 + \beta_2$ . Here,  $\epsilon_{ijt} - \epsilon_{ijt-1}$  is assumed to have a mean of zero; then, the estimate of average within-job wage growth,  $\widehat{B}$ , is a consistent estimate. Using this, I can estimate

$$w - T\widehat{B} = X_0 \beta_1 + e$$

as the second step, where  $e = \epsilon + T(B - \widehat{B})$ . Finally,  $\beta_2$  is estimated to be  $\widehat{B} - \widehat{\beta}_1$ .

These are the two most-used methods for estimating returns to tenure in the literature (Altonji and Williams, 2005; Kambourov and Manovskii, 2009; Williams, 2009),<sup>1</sup> and I will estimate the country-, industry-, and occupation-wise returns to tenure using them. As the purpose of the current paper is to compare one representative measure of returns to tenure across different groups, I essentially use only linear terms for employer tenure and education, and I use linear and squared terms for work experience, as in Abraham and Farber (1987). However, considering that it is more common in the literature

<sup>&</sup>lt;sup>1</sup>Estimates from the IV and 2SFD methods can still be biased, as discussed in Altonji and Williams (2005) and Buchinsky et al. (2010). Both methods are known to produce an upward bias in the estimate for  $\beta_1$  and a downward bias for  $\beta_2$ , but for Topel's 2SFD especially, because the estimate for  $\beta_1 + \beta_2$  is unbiased, the author argues that the estimate for  $\beta_2$  in his method provides a lower bound for returns to tenure.

to include higher terms for tenure and experience, I also perform regressions with squared terms for tenure and education and a cubed term for experience, as in Altonji and Shakotko (1987) and Kambourov and Manovskii (2009), for comparison, although these generate very similar results.

In addition, a marital status dummy, a union dummy, a head-of-household dummy, and one-digit occupation and industry dummies are included in each regression. Details of the industry and occupation specifications are in the next subsection. Finally, for a labor mobility measure of different industries and occupations, I mainly use job-to-job mobility, which I estimate as the proportion of individuals in a group who changed jobs in the subsequent period.

#### 3.2 Data

I mainly use KLIPS for estimating returns to tenure and job mobility in Korea, which is a longitudinal survey of the Korean labor market and the income activities of households and individuals residing in urban areas. It was started in 1998 and is the longest collection of panel data on Korean individuals, covering a variety of social and economic aspects, from job mobility and unemployment experiences to schooling and health. For the purpose of this research, the data provides rich information about the past jobs each individual has had, making it possible to estimate not only country-level, but also industry- and occupation-wise returns to tenure.

The sample used in this study is from 13 waves of the survey, covering the period from 2004 to 2016. The reason for setting the starting year to be 2004 is because the data provides information on after-tax income only since that year. The sample is restricted to employed male Koreans aged 18–60. Individuals who were working in the agriculture, fishery, forestry, or mining industries or in related occupations are excluded, as are those who worked less than 250 hours a year or earned less than minimum wage. The wage data used for the study is the log hourly real wage. To deflate the wages, I use the yearly CPI estimates by Statistics Korea.

For industry- and occupation-level comparisons, I use the Korean Standard Industrial Classification (8th) and Korean Standard Classification of Occupations (5th) designed by Statistics Korea and based on the International Standard Industrial Classification of the United Nations and the International Standard Classification of Occupations of the International Labour Organization. Throughout this paper, industry or occupation "sections" denote one-digit classifications and "divisions" denote two-digit classifications. I estimate returns to tenure for both section-level and division-level classifications and use only division-level industries and occupations for investigating the correlations between returns to tenure and labor mobility.

To compare the returns to tenure in the US with those in Korea, I use survey years 2005 to 2017 of the PSID, which cover the same period as the Korean sample. The US sample is also restricted in a similar way to the Korean sample. For calculating tax, I use TAXSIM32 of NBER (Feenberg and Coutts, 1993), and the wages are deflated with the CPI estimated by the US Bureau of Labor Statistics.

As noted by Topel (1991) and Buchinsky et al. (2010), the tenure variables in the PSID data are not reliable. To overcome possible inconsistencies, I

reconstruct the variable using the procedure detailed in both papers. The same procedure is also conducted for the Korean sample. Table 1 shows the summary statistics with sample means and standard deviations for the Korean and US samples.

Table 1: Summary Statistics; KLIPS and PSID; 2004-2016

|                    | Ko      | orea     | Ţ       | JS        |  |
|--------------------|---------|----------|---------|-----------|--|
| Wage               | 0.1583  | (0.4980) | 2.3401  | (0.5380)  |  |
| $\Delta Wage$      | 0.0446  | (0.3137) | 0.0612  | (0.3058)  |  |
| Tenure             | 6.2972  | (7.4420) | 7.3249  | (8.1704)  |  |
| Experience         | 11.8172 | (8.8218) | 16.7077 | (10.9932) |  |
| Education          | 13.3807 | (2.9112) | 13.8575 | (2.9301)  |  |
| Head               | 0.7991  | (0.4007) | 0.9992  | (0.0278)  |  |
| Married            | 0.7220  | (0.4481) | 0.8309  | (0.3749)  |  |
| Union              | 0.1241  | (0.3297) | 0.1457  | (0.3528)  |  |
| Individuals        | 4,      | 545      | 3,      | 3,278     |  |
| Wage obs. 26,      |         | 775      | 11      | ,654      |  |
| $\Delta$ Wage obs. | 19,     | 036      | 6,607   |           |  |

Notes: KLIPS and PSID samples of employed males aged 18-64. Workers in the agriculture, fishery, forestry, and mining industries and associated occupations and those who worked less than 250 hours a year or earned less than minimum wage are excluded. Wages are log real hourly wages deflated using the CPIs of their respective countries. Wage changes are only for those who stay. Tenure, experience, and education are all measured in years. Standard deviations in parentheses.

#### 4 Empirical Results

The results are organized as follows: Section 4.1 compares the estimated returns to tenure and job-to-job mobility in Korea and the US, Section 4.2 presents the returns to tenure of each industry and occupation, Section 4.3 shows the correlations between the industry- or occupation-wise returns to tenure with job mobility and discusses the implications, and Section 4.4 investigates possible selection bias that might affect those correlations. The main conclusion is that Korea has higher returns to tenure than the US and, contrary to conventional wisdom, returns to tenure and job mobility in Korea show a slightly positive correlation.

#### 4.1 Returns to Tenure and Mobility in Korea and the US

The estimates from OLS, IV, and 2SFD are presented in Tables 2 and 3. Columns 1 and 3 of Table 2 and column 1 of Table 3 show the results for the Korean sample; columns 5 and 7 of Table 2 and column 3 of Table 3 show the results for the US sample. As explained in Section 3.1, the 2SFD procedure occurs in two steps, and Table 3 presents those two parts: the first six rows show the results from the first-difference estimation and the next three rows are for the second-step estimation. The last row presents the returns to tenure calculated by the two-step procedure. In all three estimations, Korea shows higher returns to tenure than the US, which is consistent with the previous results in the Korean literature, although this is effectively the first attempt to estimate the returns while considering unobserved heterogeneity and using panel data of a sufficient volume.

Table 2: Wage Equation Estimates; Korea and US; OLS and IV

|                               |                        | Korea                  |                        |                        |                        | US                     |                        |                        |  |  |
|-------------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|------------------------|--|--|
|                               | O.                     | LS                     | Ι                      | V                      | OLS                    |                        | IV                     |                        |  |  |
|                               | (1)                    | (2)                    | (3)                    | (4)                    | (5)                    | (6)                    | (7)                    | (8)                    |  |  |
| Tenure                        | 0.0164***<br>(0.0004)  | 0.0222***<br>(0.0010)  | 0.0218***<br>(0.0010)  | 0.0230***<br>(0.0019)  | 0.0076***<br>(0.0006)  | 0.0138***<br>(0.0015)  | 0.0094***<br>(0.0009)  | 0.0172***<br>(0.0021)  |  |  |
| $\mathrm{Ten.^2} \times 10^2$ |                        | -0.0230***<br>(0.0040) |                        | 0.0056 $(0.0064)$      |                        | -0.0243***<br>(0.0050) |                        | -0.0307***<br>(0.0064) |  |  |
| Experience                    | 0.0238***<br>(0.0010)  | 0.0233***<br>(0.0020)  | 0.0211***<br>(0.0011)  | 0.0240***<br>(0.0022)  | 0.0313***<br>(0.0014)  | 0.0418***<br>(0.0030)  | 0.0306***<br>(0.0014)  | 0.0403***<br>(0.0031)  |  |  |
| $\mathrm{Exp.}^2 \times 10^2$ | -0.0533***<br>(0.0028) | -0.0604***<br>(0.0131) | -0.0535***<br>(0.0029) | -0.0766***<br>(0.0141) | -0.0634***<br>(0.0033) | -0.1300***<br>(0.0169) | -0.0637***<br>(0.0033) | -0.1280***<br>(0.0168) |  |  |
| $\rm Exp.^3\times 10^3$       |                        | 0.0029 $(0.0025)$      |                        | 0.0046*<br>(0.0026)    |                        | 0.0120***<br>(0.0027)  |                        | 0.0118***<br>(0.0027)  |  |  |
| $R^2$                         | 0.454                  | 0.456                  | 0.451                  | 0.452                  | 0.390                  | 0.434                  | 0.390                  | 0.433                  |  |  |
| Observations                  |                        | 26,                    | 775                    |                        |                        | 11,                    | 654                    |                        |  |  |

Notes: The dependent variables is the log real hourly wage. Other variables include education variables, a head dummy, a marital status dummy, a union dummy, and section-level industry and occupation dummies. Robust standard errors are in parentheses. \*: p < 0.1, \*\*: p < 0.05, \*\*\*: p < 0.01.

To further investigate the returns, I also conduct the same exercises with additional squared tenure, cubed experience, and squared education variables as in Abraham and Farber (1987) and Kambourov and Manovskii (2009), which results in similar estimates in terms of scale. These are presented in the even-numbered columns in Table 2 and in columns 2 and 4 of Table 3. The wage-tenure profiles are plotted in Figure 1 using the linear and squared tenure terms for each estimation method and country. Based on the profiles, it is obvious not only that the linear returns to tenure in Korea are higher than in the US, but also that the curvature of Korea's profile is smaller, meaning that Korean workers experience steadier growth of returns to tenure in the long term than US workers.

Finally, Table 4 presents the job-to-job mobility of different age groups

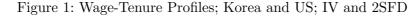
Table 3: Wage Equation Estimates; Korea and US; 2SFD

|                                      | Ko                     | rea                        | U                               | JS                     |
|--------------------------------------|------------------------|----------------------------|---------------------------------|------------------------|
|                                      | (1)                    | (2)                        | $\overline{\qquad \qquad } (3)$ | (4)                    |
| $\Delta$ Tenure                      | 0.0625***<br>(0.0043)  | 0.0759***<br>(0.0067)      | 0.0662***<br>(0.0043)           | 0.0997***<br>(0.0078)  |
| $\Delta \mathrm{Ten.^2} \times 10^2$ |                        | 0.0065 $(0.0188)$          |                                 | -0.0300**<br>(0.0132)  |
| $\Delta \text{Exp.}^2 \times 10^2$   | -0.0698***<br>(0.0136) | -0.1960***<br>(0.0502)     | $-0.0957^{***}$<br>(0.0095)     | -0.3040***<br>(0.0404) |
| $\Delta \text{Exp.}^2 \times 10^3$   |                        | $0.0252^{***}$<br>(0.0095) |                                 | 0.0362***<br>(0.0060)  |
| $R^2$                                | 0.021                  | 0.022                      | 0.055                           | 0.062                  |
| Observations                         | 19,                    | 036                        | 6,6                             | 607                    |
| Initial exp.                         | 0.0354***<br>(0.0004)  | 0.0507***<br>(0.0004)      | 0.0472***<br>(0.0005)           | 0.0787***<br>(0.0005)  |
| $R^2$ Observations                   | 0.398<br>26,           | 0.510<br>775               | 0.478<br>11,                    | 0.733<br>654           |
| Returns to tenure                    | 0.0272                 | 0.0251                     | 0.0190                          | 0.0209                 |

Notes: The first four rows show the first-step estimates of 2SFD. The dependent variable is the change in log wages. Rows 7-9 rows show the second-step estimates. The dependent variable is the log wage minus the estimated tenure and experience terms. Other variables include education variables, a head dummy, a marital status dummy, a union dummy, and section-level industry and occupation dummies. The last row is the estimated returns to tenure. Robust standard errors are in parentheses. \*: p < 0.1, \*\*: p < 0.05, \*\*\*: p < 0.01.

and the overall mobility for both countries. As briefly explained in Section 3.1, job-to-job mobility is measured simply as the proportion of workers who have changed their jobs the following year. In every age group, mobility is higher in Korea than in the US, and paired t-tests show that the differences are statistically significant.

In summary, Korea has higher returns to tenure than the US in both



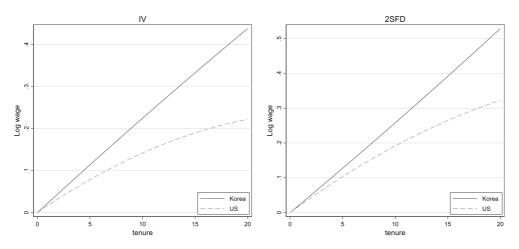


Table 4: Job-to-Job Mobility; Korea and US

| Age          | Korea  |      | US     |
|--------------|--------|------|--------|
| $18 \sim 29$ | 0.1480 | >*** | 0.1151 |
| $30 \sim 39$ | 0.0951 | >*** | 0.0751 |
| $40 \sim 49$ | 0.0621 | >**  | 0.0497 |
| $50 \sim$    | 0.0582 | >*** | 0.0319 |
| Total        | 0.0852 | >*** | 0.0701 |

Notes: Inequalities show the results from paired t-tests. \*\*\* : p < 0.01, \*\* : p < 0.05, \* : p < 0.1.

linear wage growth and long-term wage profile with curvature. At the same time, Korea's job-to-job mobility is higher than in the US for each age group and for the overall sample. Although this is only a two-country comparison using different panel data sets, this suggests a possible relationship between job mobility and returns to tenure.

As mentioned above, conventional wisdom has been that job mobility decreases as returns to tenure increase. However, Korea, which has a more mo-

bile labor market than the US, also shows higher returns to tenure, which is in line with the findings in Beffy et al. (2006), in which France was found to have lower returns to tenure and lower mobility than the US. Using a job search model with equilibrium wage-tenure contracts introduced by Burdett and Coles (2003), that paper shows that returns to tenure may increase as the mobility rate of workers increases and that this can explain the results of the comparison between France and the US.

The findings in this section may add additional evidence to Beffy et al. (2006). However, more thorough analysis that considers workers' mobility decisions and wage contracts would be needed to confirm that the difference in returns to tenure between Korea and the US is due to the difference in mobility.

#### 4.2 Returns to Tenure in Korean Industries and Occupations

I now focus on the Korean labor market. In this section, returns to tenure of industries and occupations in Korea are estimated and the section-level results presented for two purposes. Firstly, although the previous section confirmed that Korea has relatively high returns to tenure, if the industries and occupations have significantly different returns to tenure than each other, it would be more reasonable to look into how the returns vary across different groups to understand the country's overall returns to tenure. Secondly, to investigate the relationship between returns to tenure and job mobility, a comparison between only two countries is not sufficient. As matched employer—employee data to conduct firm-level analysis is not currently available, I address this issue in

Table 5: Returns to Tenure by Industry Section

|  | IV        |            |       | 2SFD                |            |          |          |
|--|-----------|------------|-------|---------------------|------------|----------|----------|
|  | Tenure    | Experience | obs.  | Tenure              | Experience | 1st obs. | 2nd obs. |
| Manufacturing                                | 0.0258*** | 0.0200***  | 6,450 | 0.0278§             | 0.0417***  | 4,883    | 6,450    |
| Electricity, gas, and water supply           | 0.0089    | 0.0467***  | 198   | 0.0149              | 0.0417***  | 161      | 198      |
| Construction                                 | 0.0119*** | 0.0185***  | 3,626 | $0.0226^{\S}$       | 0.0299***  | 2,646    | 3,626    |
| Wholesale and retail trade                   | 0.0229*** | 0.0256***  | 1,940 | $0.0225^{\S}$       | 0.0390***  | 1,406    | 1,940    |
| Accommodation and food service activities    | 0.0270*** | 0.0387***  | 499   | $0.0452^{\ddagger}$ | 0.0112***  | 314      | 499      |
| Transportation and storage                   | 0.0245*** | 0.0044     | 1,457 | $0.0317^{\S}$       | 0.0374***  | 1,116    | 1,457    |
| Information and communication                | 0.0217    | 0.0308***  | 225   | $0.0328^\dagger$    | 0.0289***  | 180      | 225      |
| Financial and insurance activities           | 0.0173*** | 0.0309***  | 673   | $0.0200^{\ddagger}$ | 0.0300***  | 537      | 673      |
| Real estate and renting activities           | 0.0297*** | 0.0156***  | 580   | $0.0531^{\ddagger}$ | 0.0220***  | 403      | 580      |
| Business support services                    | 0.0149*** | 0.0273***  | 1,921 | $0.0236^{\S}$       | 0.0236***  | 1,388    | 1,921    |
| Public administration and defense            | 0.0420*** | -0.0109    | 1,251 | $0.0350^{\S}$       | 0.0285***  | 1,050    | 1,251    |
| Education                                    | 0.0118*   | 0.0301***  | 984   | $0.0126^{\dagger}$  | 0.0214***  | 786      | 984      |
| Human health and social work activities      | 0.0233*** | 0.0218**   | 440   | $0.0168^{\S}$       | 0.0569***  | 322      | 440      |
| Arts, sports and recreation related services | 0.0083    | 0.0335**   | 212   | $0.0251^{\S}$       | 0.0576***  | 149      | 212      |
| Other services                               | 0.0165*** | 0.0307***  | 939   | $0.0266^{\S}$       | 0.0919***  | 680      | 939      |

Notes: Estimates for IV and 2SFD of 15 industry sections. First and second obs. of 2SFD are the observations for the first- and second-step estimations. \* : p < 0.1, \*\* : p < 0.05, \*\*\* : p < 0.01. For the returns to tenure estimates of 2SFD, the superscripts show the statistical significance of the  $\Delta$ tenure estimates in the first step. † : p < 0.1, † : p < 0.05, § : p < 0.01.

the next section using different industries and occupations, while the current section provides the necessary estimates and the full picture of the returns to tenure for the final step.

Table 5 and 6 present the industry and occupation section-wise estimation results for the IV and 2SFD methods. For both estimations, real estate and public administration have relatively high returns to tenure while electricity,

Table 6: Returns to Tenure by Occupation Section

|  | IV        |            |       | 2SFD          |            |          |          |
|--|-----------|------------|-------|---------------|------------|----------|----------|
|  | Tenure    | Experience | obs.  | Tenure        | Experience | 1st obs. | 2nd obs. |
| Legislators, senior officials and managers | 0.0255*** | 0.0065     | 450   | $0.0232^{\S}$ | 0.0674***  | 311      | 450      |
| Professionals                              | 0.0041    | 0.0391***  | 3,688 | $0.0119^{\S}$ | 0.0313***  | 2,697    | 3,688    |
| Technicians and associate professionals    | 0.0166*** | 0.0188***  | 2,529 | $0.0277^{\S}$ | 0.0382***  | 1,728    | 2,529    |
| Clerks                                     | 0.0213*** | 0.0321***  | 4,778 | $0.0253^{\S}$ | 0.0430***  | 3,560    | 4,778    |
| Services workers                           | 0.0377*** | 0.0133***  | 1,284 | $0.0399^{\S}$ | 0.0396***  | 865      | 1,284    |
| Sales workers                              | 0.0180**  | 0.0285***  | 1,060 | $0.0326^{\S}$ | 0.0256***  | 653      | 1,060    |
| Craft and related trades workers           | 0.0186*** | 0.0230***  | 5,057 | $0.0228^{\S}$ | 0.0461***  | 3,649    | 5,057    |
| Plant and machine operators and assemblers | 0.0246*** | 0.0121***  | 5,085 | $0.0301^{\S}$ | 0.0359***  | 3,721    | 5,085    |
| Elementary occupations                     | 0.0279*** | 0.0065**   | 2,844 | 0.0463§       | 0.0083***  | 1,852    | 2,844    |

Notes: Estimates for IV and 2SFD of 15 industry sections. Specifications are the same as for Table 5.

gas, and water supply and education have low returns. For occupations, services workers and elementary occupations are high, while professionals and craft workers are low.

From the estimates, one can confirm that there are significant differences in returns to tenure across industries and occupations. The returns are also widely variable across industry and occupation divisions. For industry divisions with more than 200 observations, 2SFD returns are 0.0257 on average and the standard deviation is 0.0125. For occupations with more than 200 observations, the average is 0.0286 and the standard deviation is 0.0162. This implies that there is considerable heterogeneity in the wage structures across different industries and occupations, which may affect the employment and mobility decisions of heterogeneous workers, although this has been frequently ignored

Table 7: Correlations between Job Mobility and Returns to Tenure

|            |           | Job-to-job mobility of workers with                   |         |         |         |  |  |
|------------|-----------|---|---------|---------|---------|--|--|
|            | All moves | Ten. $\leq 1$ Exp. $\leq 6$ Exp. $\geq 10$ Age $\geq$ |         |         |         |  |  |
| Industry   |           |   |         |         |         |  |  |
| IV         | 0.1272    | 0.2566  | 0.0396  | 0.2174  | 0.1848  |  |  |
| 2SFD       | 0.0176    | 0.1298  | -0.0876 | 0.1325  | 0.1451  |  |  |
| Occupation |           |   |         |         |         |  |  |
| IV         | -0.0307   | 0.2743  | 0.1534  | -0.1669 | -0.0533 |  |  |
| 2SFD       | 0.2786    | 0.2966  | 0.3169  | -0.0243 | 0.0487  |  |  |

Notes: The numbers are correlation coefficients between returns to tenure and the job mobility of industry and occupation divisions. Divisions with fewer than 200 observations are excluded, leaving 36 industries and 23 occupations. The first column use ordinary job-to-job mobility; the second column uses the proportion of workers with tenure  $\leq 1$  who move the following year as mobility; and the mobility measures for the remaining columns are determined in a similar manner.

in previous research.

#### 4.3 Correlations between Returns to Tenure and Mobility

Using the returns to tenure of industry and occupation divisions estimated in the previous section, I now investigate the relationship between job mobility and returns to tenure in the Korean labor market. Table 7 presents the correlation coefficients between job mobility and returns to tenure of the industry and occupation divisions. For this, the first column uses the proportion of workers who move the following year as a mobility measure, while the remaining columns use the proportions of workers who move the following year among those who satisfy the given conditions. For example, the mobility measure used in the second column is the proportion of those with tenure less

than or equal to 1 who move. Divisions with fewer than 200 observations are excluded, leaving 36 industries and 23 occupations.

The table shows that the correlations are weakly positive or almost zero in most cases. For a robustness check, I also estimate the returns to tenure with the linear tenure term from the wage equations in addition to the squared tenure, cubed experience, and squared education terms, as in column 4 of Table 2 and column 2 of Table 3. Under this specification, the correlation coefficients are 0.2059 and 0.1352 for industry-wise IV and 2SFD estimates, respectively, and 0.0764 and 0.2279 for occupation-wise IV and 2SFD estimates, respectively. Considering the number 200 is somewhat arbitrary, I further estimate the original equations for the divisions using cut-offs of 150 and 100 observations, and the resulting correlations are still weakly positive or almost zero.

The results contradict the conventional wisdom of decreasing mobility with increasing returns to tenure. All other things being equal, it would be reasonable for a worker to move less as he or she receives more for staying in a firm. However, with heterogeneous productivity, it is possible that a firm might use returns to tenure as an incentive device to retain workers, which has proven theoretically possible in the previous literature under various conditions (Lazear, 1979; Burdett and Coles, 2003; Balke and Lamadon, 2020). Then, in equilibrium, there may be no or even a positive correlation between returns to tenure and mobility.

Table 8: Estimated Returns to Tenure for Long-Term Workers

|                 | Return to tenure of a division that is |                |           |            |  |  |  |
|-----------------|--|----------------|-----------|------------|--|--|--|
|                 | $< Q_2$                                | $\geq Q_2$     | $< Q_1$   | $\geq Q_3$ |  |  |  |
| $maxten \ge 10$ |  |                |           |            |  |  |  |
| Industry        |  |                |           |            |  |  |  |
| IV              | 0.0130***                              | $0.0450^{***}$ | 0.0074    | 0.0594***  |  |  |  |
| 2SFD            | 0.0123***                              | 0.0308***      | 0.0086*** | 0.0320***  |  |  |  |
| Occupation      |  |                |           |            |  |  |  |
| IV              | 0.0123***                              | $0.0400^{***}$ | -0.0039   | 0.0480***  |  |  |  |
| 2SFD            | 0.0108***                              | 0.0269***      | 0.0017*** | 0.0278***  |  |  |  |
| $maxten \ge 15$ |  |                |           |            |  |  |  |
| Industry        |  |                |           |            |  |  |  |
| IV              | 0.0563**                               | 0.0703***      | 0.0853    | 0.1160***  |  |  |  |
| 2SFD            | 0.0115**                               | 0.0318***      | 0.0092*** | 0.0322**   |  |  |  |
| Occupation      |  |                |           |            |  |  |  |
| IV              | $0.0466^{***}$                         | $0.0786^{***}$ | 0.0213    | 0.0900***  |  |  |  |
| 2SFD            | 0.0102*                                | 0.0284***      | 0.0028    | 0.0216     |  |  |  |

Notes:  $Q_1$ ,  $Q_2$ , and  $Q_3$  are the first, second, and third quartiles, respectively, The first column presents the estimates of long-term workers in the divisions with returns to tenure less than the median  $(Q_2)$ . The other columns are defined similarly. Two alternative definitions of long-term workers are used: those with maximum tenure greater than 10 year  $(maxten \geq 10)$  or 15 years  $(maxten \geq 15)$ . For 2SFD, the asterisks are estimates for  $\Delta$ tenure in the first-step estimations. The estimates for the initial variable in the second-step estimations are all significant at the 1% level. \*: p < 0.1, \*\*: p < 0.05, \*\*\*: p < 0.01.

#### 4.4 Robustness Tests for Possible Selection Bias

It is possible that the correlations in the previous section are biased: if lowpaid workers leave firms more often in some divisions, there may be a selection bias that creates higher returns to tenure for those divisions. To check whether this is the case, I conduct two exercises. First, for the wage equations used, I add a new variable, "maximum tenure" (maxten), which is a worker–job pair-wise maximum value of tenure—or in other words, the length of a match—and estimate the equations using IV and 2SFD. If a selection bias exists, those leaving would be those who had been earning less, at least for the divisions with high returns to tenure, making  $\hat{\beta}_{maxten}$ —the coefficient for maxten—greater than zero.

Second, the same procedure of estimation in Section 4.1 is performed for long-term workers only, who are defined as those with  $maxten \geq 10$  or, in the alternative,  $maxten \geq 15$ . If the bias were the main source of difference between divisions of high and low returns to tenure, the estimated returns of the divisions with high returns to tenure would be similar to or smaller than those of the low-return divisions when estimated only using long-term workers.

Using the country-level sample, the value of  $\widehat{\beta}_{maxten}$  for the first exercise is -0.0025 in IV and -0.0113 in 2SFD, both statistically significant at the 1% level. Furthermore, most of the industries and occupations with high returns to tenure show  $\widehat{\beta}_{maxten}$  less than zero. Among the 20 industries and occupations with the highest returns to tenure and with more than 100 observations, only three industries and three occupations have  $\widehat{\beta}_{maxten}$  greater than zero. In conclusion, the people leaving early are those who are earning more on average, and this pattern is significant in the divisions with high returns to tenure.

The results for the estimations of long-term workers are presented in Table 8. To compare high-return and low-return divisions, I divide the industry or

occupation divisions by those with returns to tenure less than the median  $(Q_2)$  and those with returns greater than or equal to the median, and the estimates are given in the first and second columns. I also compare divisions of returns to tenure smaller than the first quartile  $(Q_1)$  with those greater than the third quartile  $(Q_3)$  in the third and fourth columns. For each high or low group of divisions, I aggregate the sample of workers with  $maxten \geq 10$  or 15 and estimate the returns to tenure with IV and 2SFD. In each case, divisions with higher returns to tenure for the entire sample also have higher returns to tenure for long-term workers.

In summary, there is insufficient evidence to support the notion that those who leave firms in industries or occupations with high returns to tenure earned less before leaving than those who stayed. Furthermore, those who stay long term show a similar pattern of returns to tenure as the entire sample. Therefore, a selection bias caused by those who leave firms is insignificant or implausible.

#### 5 Concluding Remarks

The current research estimates the returns to tenure in Korea and the US, the returns to tenure of industries and occupations in Korea, and the correlations between mobility and returns to tenure across Korean industries. The results show that, firstly, Korea has higher returns to tenure and mobility than the US; although some papers have estimated the returns to tenure of the Korean labor market, the current research is the first to consider unobserved heterogeneity and to use more than 10 years of panel data. Secondly, the returns

to tenure are widely variable across Korean industries and occupations. Finally, the correlations between mobility and returns to tenure are almost zero or even slightly positive, which contradicts the conventional wisdom in which job mobility is expected to decline as returns to tenure increase. Instead, the newly found pattern raises the possibility of returns to tenure being used as an incentive device by firms to retain workers.

Limitations such as a lack of consideration of endogenous decision-making and insufficient control over industry- and occupation-specific traits mean that the analysis in the current research is insufficient to understand the mechanisms behind wage structures and job mobility in an economy. The findings, however, should motivate further research to investigate labor market dynamics. More specifically, a structural model that includes both firm heterogeneity and the endogenous mobility decisions of workers would help with understanding the relationship between labor market mobility and the wage dynamics of an economy.

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## 국문초록

## 한국의 연공임금과 노동시장 이동성

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본 연구에서는 한국 노동시장의 임금 연공성을 추정하고 임금 연공성이 노동시장 이동성과 갖는 관계를 조사하였다. 이를 위해서 우선 Altonji and Shakotko (1987)와 Topel (1991)에서 사용된 방법을 통해 패널 데이터를 이용하여 한국과 미국의 임금 연공성을 추정하였다. 그 결과 한국에서 임금 연공성과 노동 이동성이 모두 높게 나타나는 것이 확인되었다. 다음으로 한국 노동시장에서 산업 및 직업별 임금 연공성을 추정하였고 서로 다른 산업 및 직업 간에 상당한 연공성 차이가 존재함을 확인하였다. 마지막으로 산업 및 직업 간 임금 연공성과 노동 이동성의 상관계수를 추정하였고 상관관계가 거의 없거나약한 양의 상관관계가 나타나는 것으로 확인되었다. 이러한 결과는 음의 상관관계를 예측하는 기존의 통념과 배치되며 한국에서 임금 연공성이 근로자의 이직을 줄이는 인센티브 장치로 작용할 가능성을 시사한다.

주요어 : 임금 연공성, 노동 이동성, 한국 노동시장

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