

Vartiainen, Juhana; Pekkarinen, Tuomas

Working Paper

Gender Differences in Job Assignment and Promotion on a Complexity Ladder of Jobs

IZA Discussion Papers, No. 1184

Provided in Cooperation with:

IZA – Institute of Labor Economics

Suggested Citation: Vartiainen, Juhana; Pekkarinen, Tuomas (2004) : Gender Differences in Job Assignment and Promotion on a Complexity Ladder of Jobs, IZA Discussion Papers, No. 1184, Institute for the Study of Labor (IZA), Bonn

This Version is available at:

<https://hdl.handle.net/10419/20424>

Standard-Nutzungsbedingungen:

Die Dokumente auf EconStor dürfen zu eigenen wissenschaftlichen Zwecken und zum Privatgebrauch gespeichert und kopiert werden.

Sie dürfen die Dokumente nicht für öffentliche oder kommerzielle Zwecke vervielfältigen, öffentlich ausstellen, öffentlich zugänglich machen, vertreiben oder anderweitig nutzen.

Sofern die Verfasser die Dokumente unter Open-Content-Lizenzen (insbesondere CC-Lizenzen) zur Verfügung gestellt haben sollten, gelten abweichend von diesen Nutzungsbedingungen die in der dort genannten Lizenz gewährten Nutzungsrechte.

Terms of use:

Documents in EconStor may be saved and copied for your personal and scholarly purposes.

You are not to copy documents for public or commercial purposes, to exhibit the documents publicly, to make them publicly available on the internet, or to distribute or otherwise use the documents in public.

If the documents have been made available under an Open Content Licence (especially Creative Commons Licences), you may exercise further usage rights as specified in the indicated licence.

IZA DP No. 1184

Gender Differences in Job Assignment and Promotion on a Complexity Ladder of Jobs

Tuomas Pekkarinen
Juhana Vartiainen

June 2004

Gender Differences in Job Assignment and Promotion on a Complexity Ladder of Jobs

Tuomas Pekkarinen

*Nuffield College, Oxford
and IZA Bonn*

Juhana Vartiainen

FIEF, Stockholm

Discussion Paper No. 1184
June 2004

IZA

P.O. Box 7240
53072 Bonn
Germany

Phone: +49-228-3894-0
Fax: +49-228-3894-180
Email: iza@iza.org

Any opinions expressed here are those of the author(s) and not those of the institute. Research disseminated by IZA may include views on policy, but the institute itself takes no institutional policy positions.

The Institute for the Study of Labor (IZA) in Bonn is a local and virtual international research center and a place of communication between science, politics and business. IZA is an independent nonprofit company supported by Deutsche Post World Net. The center is associated with the University of Bonn and offers a stimulating research environment through its research networks, research support, and visitors and doctoral programs. IZA engages in (i) original and internationally competitive research in all fields of labor economics, (ii) development of policy concepts, and (iii) dissemination of research results and concepts to the interested public.

IZA Discussion Papers often represent preliminary work and are circulated to encourage discussion. Citation of such a paper should account for its provisional character. A revised version may be available on the IZA website (www.iza.org) or directly from the author.

ABSTRACT

Gender Differences in Job Assignment and Promotion on a Complexity Ladder of Jobs*

This paper studies gender differences in the allocation of workers across jobs of different complexity using panel data on Finnish metalworkers. These data provide a measure for the complexity of the workers' tasks that can be used to construct a complexity ladder of jobs. We study whether women have to meet higher productivity requirements than men in order to be assigned to more complex tasks. Gender differences in the promotion rates are examined. We use productivity measures that are based on the supervisors' performance evaluations and examine gender differences in the productivity of promoted and non-promoted workers. It is found that women start their careers in less complex tasks than men and that they are also less likely to get promoted than men who start in similar tasks. When we compare the productivity of men and women, both at the initial assignment and when some of these individuals have been promoted, we find that there is no gender-related productivity differential at the time of the initial assignment, but women become on average more productive than men afterwards, both in promoted and non-promoted subsets. The most plausible interpretation of these results is that women face a higher promotion threshold than men.

JEL Classification: J0, J7

Keywords: promotions, gender wage gap, discrimination

Corresponding author:

Tuomas Pekkarinen
Nuffield College
Oxford, OX1 1NF
United Kingdom
Email: tuomas.pekkarinen@nuffield.oxford.ac.uk

* We would like to thank David Autor, Andrea Ichino, Ian Jewitt, Markus Jäntti, Frank Vella and Rudolf Winter-Ebmer for helpful comments. The usual disclaimer applies. We are grateful to the Confederation of Finnish Industry and Employers (TT) for the permission to use their wage records. The data are archived at the Labour Institute for Economic Research, Helsinki, and the permission to use the data are controlled by the Confederation of Finnish Industry and Employers. Pekkarinen also thanks the Academy of Finland, Finnish Work Environment Fund and Yrjö Jahansson Foundation for financial support.

1 Introduction

The gender wage gap is a persistent phenomenon. Despite the fact that in many countries the female participation rates have been high for decades and the experience on anti-discriminatory legislation is long, women tend to earn lower wages than men. An important part of this wage gap is explained by occupational segregation. It is common to find, that when sufficiently narrow occupational categories are controlled for, the gender wage gap is considerably reduced.¹ Hence, one of the key elements in understanding the gender wage gap is the asymmetric allocation of men and women across occupations.

This paper uses panel data on Finnish metalworkers to study gender differences in the allocation of workers across jobs of different complexity. In the Finnish metal industry, as in many other industries, women typically work on less complex jobs than men. Our aim is to find out, whether this happens because women need to meet higher productivity requirements than men to be assigned to more complex jobs. We address this question by examining gender differences in the initial job assignments and in the probability of promotions from the initial jobs. Furthermore, we use a measure of individual productivity that is based on supervisors' performance evaluations to study how promoted and non-promoted women perform with respect to their male counterparts. We then use this information to infer whether the male and female productivity thresholds of assignment are different.

Very common explanation for gender differences in the assignment thresholds is taste-based discrimination. But there are also theoretical arguments that do not rely on discriminatory behavior on the part of the employers. Most notably, in the model by Lazear and Rosen (1990), the comparative advantage of women in non-market activities makes them more likely to quit than men. Consequently, women have to be more productive than men to be assigned to complex jobs.

Previous empirical literature on gender differences in job allocation has almost exclusively concentrated on estimating gender differences in the probability of promotion.² There are studies that use data from large surveys such as Winter-Ebmer and Zweimüller (1997) using Austrian census data, McCue (1996) using the PSID, and Booth et al (2001) using the BHPS. Other studies have used data from a single industry or firm: Granqvist and Persson (2002) analyze gender differences in the career mobility of workers in the Swedish retail trade industry, Hersch and Viscusi (1996) focus on one US public utility, and Jones and Makepeace (1996) look at workers in a British financial institution. A special branch of the literature are the studies on the career advancement of academics like Ginther and Hayes (1999, 2003) and McDowell et al (1999). The common finding is that women are less likely to get promoted than men.

But gender differences in the assignment thresholds have implications that go beyond promotion probabilities. The promotion process will also improve the relative productivity of women within jobs. Since the promoted women have passed a higher threshold than promoted men, they will, on average, be more productive than men who are promoted to the new job. On the other hand, the men that remain in the less complex jobs have failed to meet a lower productivity threshold and are therefore less productive than women who remain in the same job.

Here, we will directly address the question of whether women have to be more productive than men to be assigned to complex jobs. We focus on the early careers of workers who are initially assigned to jobs of similar complexity. The idea is that the productivity differences between men and women who are initially assigned to the same jobs should be

¹See for example Blau and Ferber (1987) or Groshen (1991).

²Exceptions are Booth et al (2001) and Hersch and Viscusi (1996) who also estimate gender differences in the wage growth upon promotion. Both find that women gain less from promotions than men.

so small that we can draw conclusions about the assignment thresholds by examining the gender differences in the probability of promotions and in productivity among promoted and non-promoted workers. If women are at the same time less likely to be promoted than men who start in similar tasks but also more productive than men among promoted and non-promoted workers, we conclude that the promotion threshold must be higher for women than for men. Finally, we use the estimated male and female quit rates to assess whether the gender differences in reservation wages could plausibly explain the asymmetry in the job assignment.

We believe that the Finnish metal industry data are suitable for this kind of analysis. First of all, they provide a measure for the complexity of the jobs. This complexity measure is valid for both within- and between-firm comparisons. In this industry, all the jobs are evaluated according to their complexity, and on the basis of this evaluation a minimum wage is attached to each job. We use these minimum wages to construct a complexity ladder of jobs. Second, the panel nature of the data allows us to distinguish between the initial job assignments and subsequent promotions. We can therefore compare the careers of men and women who start in jobs of similar complexity. Finally, the data provide information on bonuses that are based on individual performance evaluations, so that we do not have to rely on final wages when measuring individual productivity.

Using these features of the data, we find that women are initially assigned to less complex jobs than men and that they are also less likely to get promoted than men who start in the same jobs. When we measure individual productivity with bonuses, we find that there are no clear gender differences in productivity at the initial job assignment. However, gender differences in productivity become clear as some of the workers are selected for promotion. Women are consistently more productive in the groups of promoted and non-promoted workers. We interpret these results as supportive of the asymmetric threshold hypothesis.

In this industry, women are also more likely to quit than men. These differences in quit rates are especially significant among younger workers, who are the most likely candidates for promotion. However, as the differences in quit rates may also be a result of the different promotion prospects, we can only say that the quit behaviour of men and women in this industry is consistent with, but not necessarily exclusively supportive of, the reservation wage explanation of gender differences in assignment thresholds.

The rest of the paper is organized as follows. In the following section, we summarize the familiar Lazear and Rosen (1990) model and discuss the implications regarding the job assignments and gender productivity differences within jobs. We also point out what different assumptions about the underlying male and female productivity distributions imply for the inference on the assignment thresholds. In the third section, we discuss the data and the fourth section explains the construction of the complexity ladder of jobs. We then study the movement of workers on the complexity ladder. In the fifth section, we examine the allocation of workers across levels of complexity at the initial job assignment and replicate the standard analysis of gender differences in the probability of promotion. The most original analysis is reported in section 6, in which we use our productivity proxies to evaluate the relative performance of a given group of newly recruited men and women, both before and after the first promotion decisions have taken place. In section 7, we briefly examine the quit behavior of the workers and discuss whether the gender differences are large enough to explain the differences in the job allocation. Section 8 concludes.

2 Theoretical background

The model by Lazear and Rosen (1990) is a popular framework in which to think about gender differences in job assignment. In this model, the workers who are assigned to more demanding jobs undergo costly training to learn the tasks involved. Once the training is finished, the workers cannot commit to staying with the employer but leave the firm, if the value of their reservation wage exceeds the wage paid by the firm.

Women are assumed to have higher reservation wages than men because of their comparative advantage in non-market activities. Hence, women are more likely to leave the firm and the risk of loosing the training investment is higher when a woman is assigned to the demanding job. Thus, the productivity threshold that determines whether the worker is assigned to the demanding job is higher for women than for men.

2.1 Job assignments

Consider workers who enter the firm with some level of innate ability and who gradually acquire seniority that increases their effective ability. Denote the innate ability of worker i with δ_i and his/her seniority prior to period t with x_{it} . Following Gibbons and Waldman (1999), assume that the effective ability of the worker, η_{it} , is a function of the innate ability and seniority: $\eta_{it} = \delta_i f(x_{it})$, where $f' > 0, f'' \leq 0$.

There are two jobs, the more demanding job A and the less demanding job B . Consider the employer's job assignment decisions at time t . If the worker is assigned to job A , he/she will undergo costly training during the fraction τ of the period t and reaches the full productive capacity for the fraction $(1 - \tau)$ of the period. One way to interpret τ is to view it as the time spent in job specific monitoring, as in Lazear (1986). There is no training in task B .

The output per worker in job B during the whole period t is equal to his/her effective ability $\eta_{it} = \delta_i f(x_{it})$. If the worker is assigned to job A , his/her output is reduced to $\gamma_1 \eta_{it} = \gamma_1 \delta_i f(x_{it})$ during τ , where $0 < \gamma_1 < 1$. During $(1 - \tau)$, the output of the worker in job A reaches $\gamma_2 \eta_{it} = \gamma_2 \delta_i f(x_{it})$, where $\gamma_2 > 1$.

Assume that the workers are certain to stay in the firm during τ but that they cannot commit to staying during $(1 - \tau)$. Workers will leave the firm if the value of their reservation wage is higher than the wage paid by the firm. Lazear and Rosen assume that the worker's reservation wage is a random variable ω and that at the beginning of period t only the the cdf, $F(\omega)$, of this variable is known.

An efficient assignment rule should induce the workers to stay in the firm, if the market value of their output exceeds their reservation wage. Furthermore, the rule should assign only those workers to job A , whose social output is higher in job A than in job B .

Social output of the workers who are assigned to job A is equal to:

$$\tau \gamma_1 \eta_{it} + (1 - \tau) \left(\gamma_2 \eta_{it} \int_0^{\gamma_2 \eta_{it}} dF + \int_{\gamma_2 \eta_{it}}^{\infty} \omega dF \right) \quad (1)$$

Correspondingly, the social output of the workers in job B is:

$$\tau \eta_{it} + (1 - \tau) \left(\eta_{it} \int_0^{\eta_{it}} dF + \int_{\eta_{it}}^{\infty} \omega dF \right) \quad (2)$$

The difference between (1) and (2) can be written as:

$$D = -\tau \eta_{it} (1 - \gamma_1) + (1 - \tau) \int_{\eta_{it}}^{\gamma_2 \eta_{it}} F(\omega) d\omega \quad (3)$$

Workers are assigned to task A if $D > 0$ and to task B if $D < 0$. Hence, there is an assignment threshold $D(\eta^*) = 0$.

The employers do not observe the innate abilities, when making the initial assignment decisions. The effective ability of the worker is revealed after these assignments and the subsequent promotion decisions are based on the observed effective ability. Thus, for each level of seniority there is an innate ability threshold of promotion, δ^* , and for each level of innate ability there is a seniority threshold of promotion, x_t^*

2.2 Gender differences in reservation wages

Lazear and Rosen assume that men and women have different outside options. More specifically, female distribution of reservation wages, $F_f(\omega)$, first-order stochastically dominates that of men, $F_m(\omega)$. To introduce this assumption in the setting above, write the distribution of reservation wages as $F(\omega; \alpha)$, where α is a shifter, such that $\partial F / \partial \alpha > 0$. Define $F_f(\omega) \equiv F(\omega; \alpha_f)$ and $F_m(\omega) \equiv F(\omega; \alpha_m)$ so that $\alpha_m > \alpha_f$, which implies that $F_f(\omega) < F_m(\omega)$.

Differentiating $D(\eta^*) = 0$ with respect to α yields:

$$\frac{d\eta^*}{d\alpha} = \frac{\int_{\eta^*}^{\gamma_2 \eta^*} \frac{\partial F(\omega; \alpha)}{\partial \alpha} d\omega}{-\partial D(\eta^*) / \partial \eta_{it}}$$

which is negative since $\partial D(\eta^*) / \partial \eta_{it} > 0$. Hence, α decreases the threshold value of effective ability η^* and because $\alpha_m > \alpha_f$ we have that $\eta_m^* < \eta_f^*$. This means that in order to be assigned to the complex job women need to have a higher innate ability than men with the same level of seniority. Similarly, women need to acquire more seniority than men of equal innate ability to get promoted to the complex job.

2.3 Implications of the gender differences in reservation wages

Gender differences in the assignment thresholds have implications for the job assignments and productivity differences within jobs. The direction in which these impacts work depends on the underlying ability distributions of men and women.

Denote the male ability distribution with $G_m(\eta)$ and the female distribution with $G_f(\eta)$. Women will be less likely to be assigned to complex job, if:

$$\Pr(\eta \geq \eta_m^*) = 1 - G_m(\eta_m^*) \geq 1 - G_f(\eta_f^*) = \Pr(\eta \geq \eta_f^*) \quad (4)$$

Furthermore, women will be on average more productive within jobs if:

$$E[\eta \mid \eta \geq \eta_m^*] = \int_{\eta_m^*}^{\infty} \eta dG_m(\eta \mid \eta \geq \eta_m^*) \leq \int_{\eta_f^*}^{\infty} \eta dG_f(\eta \mid \eta \geq \eta_f^*) = E[\eta \mid \eta \geq \eta_f^*] \quad (5)$$

Below, we will use gender differences in promotion rates and productivity within groups of promoted and non-promoted workers to infer whether the assignment thresholds are different for men and women. In order to do this, we need to know under which assumptions the fact that (4) and (5) hold simultaneously necessarily implies $\eta_m^* \leq \eta_f^*$. It is easy to see that if $G_m(\eta)$ and $G_f(\eta)$ are identical, both (4) and (5) can only hold if $\eta_m^* \leq \eta_f^*$. When the male ability distribution first-order stochastically dominates the female distribution, $1 - G_m(\eta_m) \geq 1 - G_f(\eta_f)$, the condition (4) can also hold for some values $\eta_m^* > \eta_f^*$ but the condition (5) holds only when $\eta_m^* \leq \eta_f^*$. On the other hand, if the female distribution dominates, $1 - G_m(\eta_m) \leq 1 - G_f(\eta_f)$, the condition (4) can only hold if $\eta_m^* \leq \eta_f^*$, whereas the condition (5) also holds for some values such that $\eta_m^* > \eta_f^*$. Thus, in these cases women can be *simultaneously* less likely to be assigned to

the complex job and more productive within jobs only if the female assignment threshold is higher than the male assignment threshold.

Naturally, one cannot draw such conclusions if male and female ability distributions intersect. It might appear to be a very strong assumption to rule out the intersecting ability distributions. After all, it is often argued that the variance of male ability is higher than that of the female ability. However, in our case we are not comparing the "global" male and female ability distributions. Instead, we focus on "ability brackets" of men and women who have been initially assigned to similar tasks by their employers. It seems very unlikely that the male and female ability distributions would systematically intersect within all of these ability subsets. Indeed, in our empirical analysis we will show that the male and female distributions of measured productivity seem to be remarkably similar within the initial jobs. Hence, in these circumstances, observing that women are both less likely to be promoted and more productive than promoted and non-promoted men implies that the female assignment threshold is higher than the male one.

3 The data

The data used in this paper come from the wage records of the Confederation of Finnish Industry and Employers (*Teollisuus ja työnantajat*). The wage records contain detailed information on the wages and working hours of all the workers who are affiliated with the confederation. In the Finnish metal industry, this covers practically 100% of the firms.

The wage records' data on wages and working hours can be considered as exceptionally reliable since the information comes directly from the firms' wage accounts. However, the information on the individual characteristics is rather scarce. Basically only age, gender, and seniority can be identified from the raw data. For the objectives of this paper, the most disturbing piece of missing information are the variables concerning marital status of the worker and the number of dependent children.

In this paper, we use all the workers who start their careers in the Finnish metal industry between 1990-1995 and whom we can follow for at least five years up to year 2000 when our data end. We chose to restrict the sample like this because for our purposes it is essential to observe the workers at their initial job assignments and follow them for a reasonable number of years. Furthermore, restricting the analysis to workers who stay for more than four years selects the workers with a strong labour market attachment.

This panel of newcomers to the metal industry has 83 474 employee/year observations on 11 661 workers of whom 2 705 (23 %) are women. We have 64 198 episodes where the both current and next year's jobs are observed and the worker stays within the same firm.

In table 1, we present the descriptive statistics on this sample at the first year of seniority and compare them with the full cross-section of workers in the metal industry in 1990. It is clear that this is a strongly male-dominated industry. This fact should be taken into account, when interpreting the results reported below.

3.1 Wage determination in the Finnish metal industry

The sample was restricted to include only workers from the metal industry because the peculiar wage determination process in this industry provides particularly interesting information on the complexity of the jobs. According to the industry's collective agreement, wages should be determined by the complexity of the job, the individual performance of the worker, and by various individual and firm-specific factors.

The complexity of the job specifies a minimum wage for each job. This minimum wage is called the *occupation-related wage*. Worker's individual performance on the job affects

the wage outcome through a *personal bonus* of 2 to 17% of the occupation-related wage.³ In this paper, we will use the occupation-related wages as a measure of the complexity of the job and personal bonuses as a proxy for the productivity of the individual.

3.2 Job complexity

The complexity of the job is evaluated with a grading system that is similar to the ones used in some large establishments in the US. The evaluation is carried out by a group of experts, that considers various aspects of the jobs and assigns them points according to their complexity. The complexity level is based on three criteria: 1) how long does it take to learn the tasks, 2) the degree of responsibility, and 3) the working conditions. The evaluation should be independent of the workers performing the job and it should make jobs comparable both between and within firms.

Based on the evaluation of jobs, an occupation-related wage is determined for each job in the collective agreement. The more demanding the job, that is the more complexity points it gets, the higher is the corresponding occupation-related wage. Basically, there should be a one-to-one mapping from the occupation-related wages to the complexity points. The occupation-related wages can therefore be interpreted as a continuous variable that measures the complexity of the job. There are typically around 50 different levels of occupation-related wages per year.

3.3 Individual performance

The individual performance of the worker is evaluated by his/her immediate supervisor. The performance evaluation is based on three criteria: 1) how well the tasks are carried out, 2) the worker's output relative to what is considered normal in the job, and 3) how well the worker follows the instructions and regulations. The performance of the worker is always evaluated relative to what is considered normal on the job in question. Hence, if, for example, the output of all the workers in the firm increases, the performance evaluations of the individual workers should not be affected.

Based on the supervisor's evaluation each worker is paid a personal bonus that should amount to 2-17% of the worker's occupation-related wage. The collective agreement states, that the bonuses should be symmetrically distributed around the mean of 9.5% within each complexity tercile in the firm.

In order to check whether these principles of the collective agreement are also followed in practice, we examined the variation of bonuses across complexity levels for groups of workers with a given level of seniority. There was some indication that bonuses were increasing in the level of complexity. Furthermore, straightforward analysis of variance revealed that firm dummies explain approximately 5% of the variation in bonuses. In order to account for this variation across complexity levels and firms, we measure individual performance using the deviation of the worker's personal bonus from the firm - complexity level cell mean. In the following, this productivity measure will be called *personal bonus deviation*.

4 Complexity ladder of jobs

As was explained above, the occupation-related wages order the jobs according to their complexity. In this paper, we use this ordering of jobs as a job ladder where the within-

³It is important to note that the sum of the occupation-related wage and personal bonus is not the final wage outcome. The firms are free to set their final wages as long as they are above the minimum levels set at the collective agreement. Thus, there is considerable within and between firm variation in wages even among workers in similar tasks.

firm upwards movement on the ladder is interpreted as a promotion and downwards movement as a demotion.

The fact that the occupation-related wages are a component of the final wages makes their use as a complexity measure somewhat problematic. There is some year-to-year variation in the occupation-related wages that is not related to changes in the complexity of the jobs. For example, in certain years all the occupation-related wages are increased to account for the effects of inflation. This means that the scale with which the complexity of the jobs is measured is not constant in time.

We corrected for these changes by descaling the occupation-related wages in the following way. We first grouped the workers according to their occupation-related wages within each year and examined the within group distributions of changes in the occupation-related wages. This analysis revealed that for most of the workers in these groups the year-to-year changes in occupation-related wages were identical. We interpreted the group mode changes of the occupation-related wages as increases that were not related to changes in the complexity of the jobs. The occupation-related wages were then corrected by subtracting the annual mode changes from occupation-related wages. After this correction, the occupation-related wages form a consistent ladder for all the years 1990-2000.

Table 2 is a transition matrix of movements between complexity levels in 1990-1991. It shows all the within-firms movement between complexity levels, including entries, exits and stays as percentages of the number of workers on each level. The table can be interpreted in the same way as the job-to-job transition matrices in Baker et al (1994) or Treble et al (2001). For the purposes of this table, the complexity levels were aggregated into integers.

If complexity levels really are a true job ladder, the movements on this ladder should resemble the stylized facts about promotions and demotions. Basically, there should be more movement upwards than downwards. In table 2, there are 62 304 workers who stayed with the same employer between 1990-1991. Approximately 9% of these workers worked in a more complex and about 5% in a less complex job in 1991 than in 1990. Shaded areas in the table indicate the levels that were the most frequent destinations of movement between complexity levels. There is a considerable amount of movement, most of which is upwards and rarely leaping over many levels. In our view, the information in table 2 is in line with the stylized facts about promotions and demotions. It seems appropriate to interpret the complexity axis as a job ladder.

5 Gender differences in job assignments

How are men and women then allocated on this ladder? Figure 1 plots the percentages of male and female workers across the same complexity levels as in table 2 in the 1990 cross-section. The figure reveals that women are concentrated at the low end of the complexity axis while most men work on more complex jobs. This pattern is repeated throughout the years 1990-2000.

Why are women concentrated at the low-end of the complexity axis as in figure 1? Perhaps the most common approach to studying gender differences in job allocation, is to examine movement of workers between jobs. If the assignment thresholds are identical for men and women, we shouldn't see any differences in the movement of men and women once the productive characteristics are controlled for. In this section, we study the gender differences in the allocation of workers across jobs of different complexity at the initial job assignment and in the promotion process.

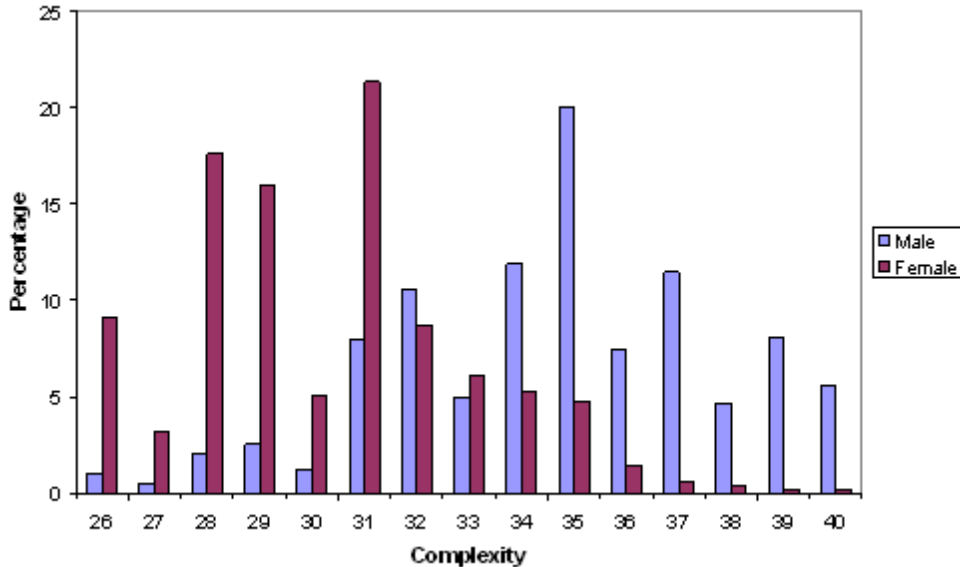


Figure 1: Percentages of male and female workers across tasks of different complexity, 1990 cross-section

5.1 Initial job assignments

In our sample, we observe the jobs where the workers started their careers and their subsequent promotions and demotions. In figure 2, we have plotted the distributions of the first job assignments of the men and women in our sample.

The asymmetry in figure 2 is striking. Women seem to start their careers in clearly less complex jobs than men. The Duncan and Duncan index of dissimilarity for the initial job assignments gives a value 49.3. That is, nearly half of the women in our sample should change their initial job assignment to arrive to the male allocation of jobs.

However, figure 2 should not be interpreted as evidence on gender differences in the assignment thresholds. After all, it is possible that the asymmetry in figure 2 only reflects differences in the ability distributions of men and women who choose to work in the metal industry. If the underlying productivity of women in this industry is lower than that of men, it is not surprising that women start their careers in less complex jobs than men. Still, whatever the reason for the asymmetry in figure 2, it is clear that the initial job assignment has to be taken into account in the analysis of gender differences of promotion and productivity.

5.2 Promotions

Are women then less likely to move upwards on the complexity ladder than men? In table 3, we report the gender differences in the change of complexity. We take advantage of the fact that the occupation-related wage is a continuous variable. Thus, the changes of occupation-related wages conveniently summarize both the extent and the direction of the change in the complexity of the jobs that workers are performing.

The first row of table 3 reports the sample means of changes in log occupation-related wages for men and women in our sample. The male and female sample mean of changes in the complexity of the jobs are virtually identical. If anything, the female mean is

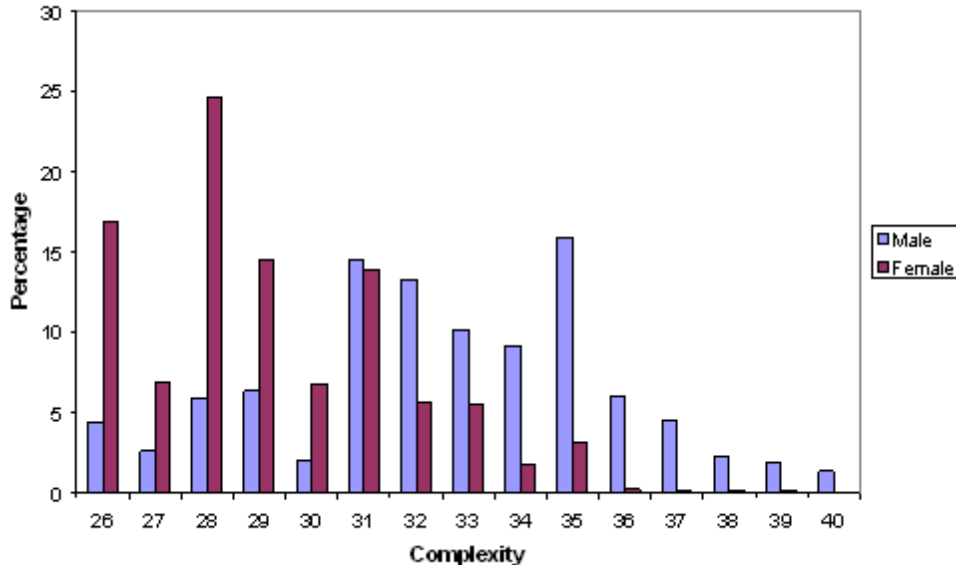


Figure 2: Percentages of male and female workers across jobs of different complexity at the initial job assignment

slightly higher than the male mean, although the difference is not statistically significant. However, once we split the workers according to the complexity of their initial job assignments, the gender differences become clear. In the following rows of table 3, we report the male and female means of changes in log occupation-related wages in the complexity groups of initial job assignment. In most groups, the mean change in log complexity is clearly lower for women than for men.

The numbers in table 3 are just cell means which do not control for anything else than the complexity of the initial job assignment. In order to control for observable productive characteristics, we ran a following regression on the pooled data:

$$\Delta c_{it} = \beta F_i + X'_{it}\gamma + \delta I_{it} + \varepsilon_{it} \quad (6)$$

where $\Delta c_{it} = c_{i,t+1} - c_{it}$ is the difference in log occupation related wages, F_i is the female dummy, and X_{it} is a set of productive characteristics including age and seniority and their squares. In I_{it} we include a full (15) set of initial job complexity dummies to control for the initial job assignment.

The regression results are reported in table 4. The coefficient on the female dummy is again negative, -0.008 , and clearly significant. Thus, the results in tables 3 and 4 clearly indicate that women take smaller steps on the complexity ladder than men who start in jobs of similar complexity.

Another way of looking at gender differences in the promotion patterns is to examine duration to promotion. In order to do this, we defined a promotion indicator that takes value one if the individual experienced a positive change in the occupation-related wage within the same firm and zero otherwise. In table 5, we report the Kaplan-Meier estimates of the survivor function for men and women. The first column shows the estimates for the whole sample. Although the female survivor function is consistently above the male one, the differences are relatively small.

However, the gender differences become clear once we focus on workers who start their careers in jobs of more or less similar complexity. This is done in columns 2-4 of table 5, where we report the estimates of the survivor function within subsamples of workers. The groups were formed by dividing the workers into "low-complexity", "medium-complexity", and "high-complexity" groups according to the complexity of the initial job assignment. Female survivor functions are clearly above the male ones in the low and medium groups of initial job complexity. A straightforward log-rank test for the equality of the Kaplan-Meier estimates, results of which are reported in the last row of table 5, rejects the null in all the cases expect the high complexity jobs.

In order to control for observable productive characteristics in the duration analysis, we ran a discrete-time proportional hazards model of promotion, using the log-likelihood suggested by Jenkins (1995):

$$\log \mathcal{L} = \sum_{i=1}^N \sum_{t=1}^T y_{it} \left[\frac{h_{it}}{(1-h_{it})} \right] + \sum_{i=1}^N \sum_{t=1}^T \log(1-h_{i,t}) \quad (7)$$

where N is the number of individuals in our sample, T is the number of years, and y_{it} is a dummy that takes the value 1 for individual i at t if he/she is promoted at that period and zero otherwise. We use a complementary loglog specification for the hazard rate:

$$h_{it} = 1 - \exp \{ - \exp [\theta(t) + F_i + X'_{it}\gamma + \delta I_{it}] \} \quad (8)$$

where the baseline hazard $\theta(t)$ is left unspecified and the rest of the variables are as in (6). The results are presented in table 6. The estimated coefficient of the female dummy is negative (-0.4) and clearly significant.

We interpret the results in tables 3-6 as implying that women are initially assigned to jobs that are less complex but where the promotions are also more frequent. However, compared to the male workers who start their careers on those same jobs, women are much less likely to move to more complex tasks. Thus, women tend to "get stuck" in the initial jobs of low complexity whereas men are more successful in advancing to more complex jobs.

6 Gender differences in productivity

The regressions in the previous section controlled for productive characteristics in a very crude way. An alternative approach to study gender differences in the assignment thresholds is to compare the relative productivity of men and women before and after the promotion decisions have taken place. If the threshold of promotion is higher for women than for men, the promoted and non-promoted women should be, on average, more productive than their male counterparts.

We now use personal bonus deviations to measure individual productivity and examine the productivity of the workers in our newcomer sample. We first measure the productivity of each worker at the initial job assignment and then partition the workers into two sets of workers: those who have got promoted up to some specific year (the "promoted" group) and those who have stayed at the initial assignment until that year (the "stagnant" group).

In table 7, we report the median personal bonus deviations for workers in each of these groups. In the first row, we display the personal bonus deviations of the new entrants during their first year in the industry. The next row depicts the personal bonus deviations of men and women as measured in the second year of the career, separately for those who were promoted after the initial year and those who were not. The following row depicts

the median personal bonus deviations two years after the initial assignment, similarly differentiated between those who had been promoted up to their third year of the career and those who were not; and analogously for the fourth and the fifth row.

The medians reported in table 7 indicate that there are hardly any gender differences in productivity at the initial task assignments. However, once the workers are split into promoted and non-promoted groups, the gender differences become clear. Women tend to dominate in both groups. Among the non-promoted workers the gender difference reaches 0.009 and among the promoted workers 0.011 in favour of women.

Table 7 only reports the differences in medians. In order to have a better view of what happens to the productivity distributions, we have plotted the kernel estimators of the cumulative distributions of personal bonus deviations for men and women separately for stagnant and promoted workers in the first four years of seniority in figure 3.⁴ The male and female distributions of personal bonus deviations are almost indistinguishable at the initial job assignment. After the first year, the workers are split into stagnant and promoted groups. In both groups, the female distributions clearly dominate the male distributions and the differences seem to become clearer with seniority.

Whether the results presented above, together with the results on gender differences in promotion probabilities, imply that the female assignment thresholds are higher than the male ones depends on the underlying cumulative productivity distributions of men and women who choose to work in the metal industry. This implication only follows if the underlying cumulative productivity distributions of men and women do not intersect. As we argued in section 2.3, it seems plausible to assume that this is the case here. First of all, we focus on men and women who start their careers within jobs of similar complexity. The employers have already selected the workers that they judge to be appropriate for given tasks. Any differences in the underlying distributions of productivity of men and women within these groups should be small. Furthermore, it seems very unlikely that distributions would intersect within all these groups. Finally, the results in table 7 and in figure 3 indicate that there are no visible gender differences in the personal bonus deviations within initial jobs. Hence, in our view, it is justified to conclude that these results as indicating that women need to be more productive than men to be assigned to complex jobs.

Finally, personal bonus deviation comparisons also tell us something about the behaviour of the employers. First of all, it is clear that the workers that the employers choose to promote tend to be more productive already in their initial jobs. In terms of Lazear and Rosen story, this implies that training costs do play a role in these firms. After all, if the assignment to more complex jobs was costless, the employer would assign all the workers to more complex tasks. However, it seems that women need to be more productive at their initial job than men to get promoted. For promoted women, the average personal bonus deviation in the initial job is $-.021$ while for promoted men it is only $-.025$. Thus, promoted women are approximately 18% more productive at the initial job than promoted men. Since the Lazear and Rosen model does not assume any differences in the training costs of men and women, the only way to explain these differences in terms of their model are gender differences in the quit behaviour.

7 Gender differences in quit behavior

According to the Lazear and Rosen model, gender differences in reservation wages give rise to differences in the assignment thresholds. We will now examine whether the observed

⁴The estimation was done using the Epanechnikov kernel with a bandwidth of .005. Different bandwidths were tried out without relevant changes in the qualitative results.

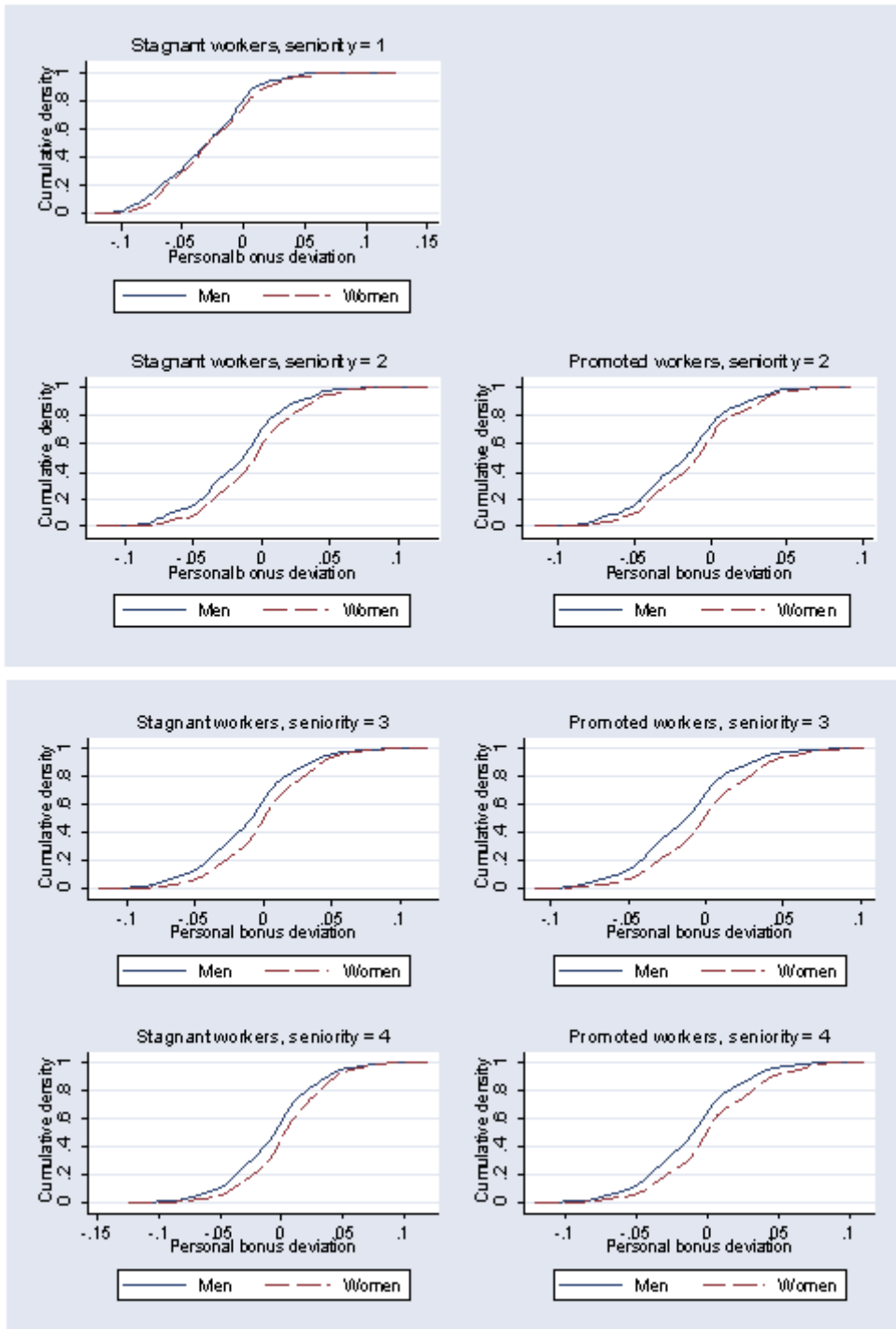


Figure 3: Kernel density estimates of the cumulative distributions of personal bonus deviations

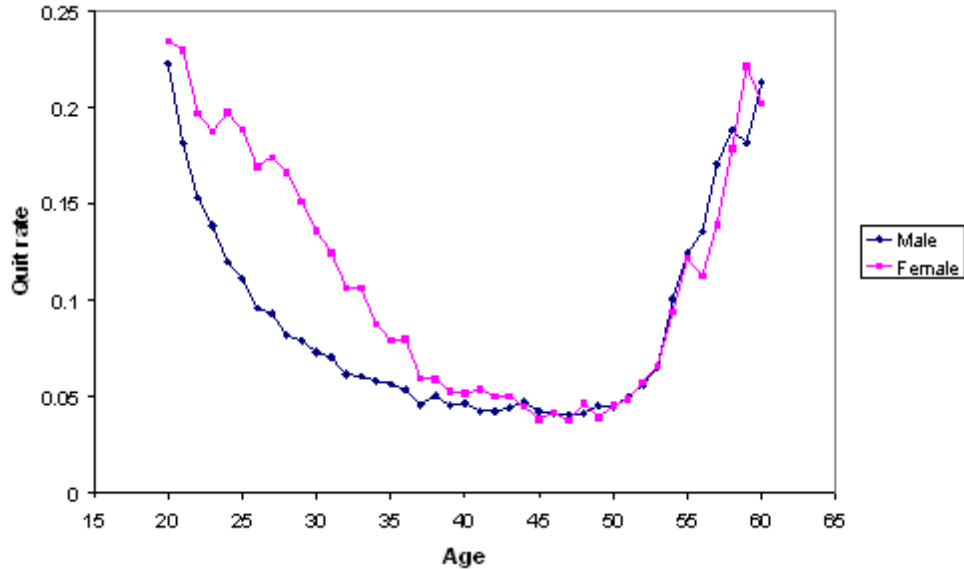


Figure 4: Average quit rates by age and gender in the whole metal industry population, 1990-2000

quit behaviour of men and women in the Finnish metal industry is consistent with the Lazear and Rosen argument. However, we emphasize that, in our view, these data do not really allow for any exclusive "testing" of the Lazear and Rosen argument against all alternative models such as taste-based discrimination in promotions. After all, if there are other forces that hamper women's promotion prospects, women's higher exit rates will also reflect these other mechanisms. That is, any gender differences in quit behaviour that we observe may be caused by gender differences in the assignment thresholds which themselves could be a result of taste-based discrimination. Hence, this exercise should be seen more as a simple check of whether gender differences in quit behaviour could be the underlying cause for the differences in the assignment thresholds in the most favourable case where the quit rates are exogenous.

We do not observe the actual reasons of job separations in our data. Hence, it is not possible to distinguish layoffs and voluntary quits directly. However, since voluntary quits are what we are interested in, we tried to tackle this problem by defining as a "quit" those separations, where more than 75% of the employees remain in the departure firm. This definition rules out the situations where a large proportion of the firm's workforce is laid off at the same time, and it is likely to catch more truly voluntary separations. We do not distinguish between job-to-job turnover and job-to-nonemployment turnover, since from the point of the view of the employer these are equivalent.

In the whole metalworker population, the average quit rate of men was 7.8% and that of women 9.6%. Figure 4 plots the average quit rates for men and women by age in the whole metalworker population. As is clear from the figure, the male and female quit rates differ significantly among younger workers, but tend to converge among older workers

Table 8 reports the marginal effects from a probit model of quits that uses the whole population of metalworkers. In this model, we allow for different intercepts and age profiles for men and women. The effects are calculated at the mean values of the continuous

independent variables. According to these results, women are five percentage points more likely to leave the firm.

However, from the point of view of the employer who is considering whether to promote one of the workers in our newcomer sample, the most relevant differences are the differences in the quit rates of the younger workers. In table 9, we report the predicted quit rates of men and women in different age groups that are calculated using the coefficients of the probit model in table 8. The predicted quit rates of men and women are substantially different among workers who are younger than 30 years old. In fact, in both groups 25 and below as well 25-30 years old, the female quit rate is approximately 40% higher than the male quit rate. In our newcomer sample, nearly 50% of the observations come from these two age groups.

Thus, we find a pattern of exit rates that is certainly consistent with the Lazear and Rosen story but, of course, not exclusively supportive of that model only. The women are more likely to quit the firm precisely in the group of workers where one would expect the Lazear and Rosen argument about gender differences in the reservation wages to apply, i.e. young workers. Interestingly, when the stochastic dominance analysis of figure 3 is replicated for the population of over 40 years old workers, one does not find a similar pattern of gender productivity differences among promoted and non-promoted workers as in figure 3.

8 Conclusions

The asymmetric allocation of men and women across jobs is one of the most common explanations for the persistence of gender wage gaps. It is often argued that women have to meet higher productivity requirements than men to be assigned to demanding jobs. Because of these differences in the assignment thresholds women tend to start their careers in less demanding jobs than men and find it more difficult to get promoted than men.

In this paper, we examine gender differences in the job allocations using panel data on Finnish metalworkers. These data are exceptionally suitable for the analysis of the assignment thresholds. They provide a measure for the job complexity that can be used to construct a complexity ladder of jobs. Thus, we are able to compare men and women, who start their careers in similar jobs. Furthermore, the data contain information on personal bonuses that can be used to study the productivity implications of the gender differences in the assignment thresholds.

In the Finnish metal industry, women work on clearly less complex jobs than men. This gender difference in the job allocation is already visible at the initial job assignment. Nearly 50% of the women would have to change their initial jobs to arrive at the male allocation of initial jobs. Furthermore, women find it more difficult to move to more complex tasks than men: Women are clearly less likely to get promoted than men who start their careers in the same jobs, even after controlling for observable productive characteristics.

Using personal bonuses as a measure of individual productivity, we also find that there are no clear productivity differences between men and women within the jobs where the workers are initially assigned to. However, when the workers are split into groups of promoted and non-promoted workers, the gender differences in productivity become clear. Promoted and non-promoted women are consistently more productive than their male counterparts.

Whether one can interpret these results as evidence on gender differences in the assignment thresholds, depends on the underlying productivity distributions of male and female workers in this industry. We argue that since we do not observe any productivity differences within the initial jobs, we can interpret the combination of the results that

women are less likely to be promoted and that women are more productive among the promoted and non-promoted workers as evidence on higher female thresholds. Whereas the result that women are less likely to get promoted than men has been reported often, it hasn't been combined with the results on gender differences in productivity that would match the asymmetric assignment threshold hypothesis before.

Finally, the observed quit behaviour of men and women in this industry is broadly consistent with the Lazear and Rosen explanation of the gender differences in the promotion thresholds. Namely, the fact that women are more likely to quit the firm than men, may imply that women face higher thresholds because their promotion is more costly to the employer. However, since the quit behaviour may be a result of the differences in the promotion thresholds, it is not possible to draw definite conclusions on the causes of these threshold differences with these data.

References

- [1] Baker, G. P., M. Gibbs & B. Holmström, (1994): "The internal economics of the firm: Evidence from personnel data", *Quarterly Journal of Economics*, 109: 921-955.
- [2] Blau, F. D. & M. A. Ferber, (1987): "Discrimination: Empirical evidence from the United States", *American Economic Review*, 77, 316-20.
- [3] Booth, A. L., M. Francesconi & J. Frank, (2003): "A sticky floors model of promotion, pay, and gender", forthcoming *European Economic Review*, 47(2), 295-322..
- [4] Gibbons, R. & M. Waldman, (1999): "A theory of wage and promotion dynamics inside firms", *Quarterly Journal of Economics*, 114 (4), 1321-1358.
- [5] Ginther, D. K. & K. J. Hayes, (1999): "Gender differences in salary and promotion in the humanities", *American Economic Review, Papers and Proceedings*, 89, 397-402.
- [6] Ginther, D. K. and K. J. Hayes, (2003): "Gender differences in salary and promotion for faculty in the humanities 1977-1995", *Journal of Human Resources*, 38(1), 34-73.
- [7] Granqvist, L. & H. Persson, (2002): "Gender differences in career mobility - an example from the wholesale and retail trade", mimeo, Swedish Institute for Social Research, Stockholm University.
- [8] Hersch, J. & W.K. Viscusi, (1996): "Gender differences in promotions and wages", *Industrial Relations*, 35, 461-72.
- [9] Jenkins, S.P., (1995): "Easy estimation methods for discrete-time duration models", *Oxford Bulletin of Economics and Statistics*, 57, 129-138.
- [10] Jones, D. R. & G. H. Makepeace, (1996): "Equal worth, equal opportunities: Pay and promotion in an internal labour market", *Economic Journal*, 106, 401-409.
- [11] Lazear, E.P. & E. Rosen, (1990): "Male-female wage differentials in job ladders", *Journal of Labor Economics*, 13 (3), 426-471.
- [12] McCue, K. (1996): "Promotions and wage growth", *Journal of Labor Economics*, 14 (2), 175-209.
- [13] McDowell, J.M, L.D. Singell & J.P. Ziliak, (1999): "Cracks in the glass ceiling: Gender and promotion in the economics profession", *American Economic Review, Papers and Proceedings*, 89, 392-396.

- [14] Treble, J. & E. van Gasteren & S. Bridges & T. Barnby, (2001): "The internal economics of the firm: Further evidence from personnel data", *Labour Economics*, 8, 531-552.
- [15] Winter-Ebmer, R. & J. Zweimüller, (1997): "Unequal assignment and unequal promotion in job ladders", *Journal of Labor Economics*, 15:1, 43-71.

Table 1 Descriptive statistics

Variable	Men in our sample		Women in our sample		Men in 1990 cross-section		Women in 1990 cross-section	
	Mean	Standard deviation	Mean	Standard deviation	Mean	Standard deviation	Mean	Standard deviation
Age	28.43	8.35	31.43	9.22	37.23	10.40	40.19	10.37
Complexity	33.03	3.24	29.55	2.35	35.21	3.17	30.62	2.62
Bonus	.05	.04	.06	.04	.09	.04	.09	.04
Average hourly earnings	41.35	8.23	34.92	4.61	46.79	7.64	38.48	5.07
Individuals	8 956		2 705 (23%)		52377		14 607 (22%)	
Ind/year observations	83474							

Our sample consists of all the observations during 1990-2000 on workers who enters the metal industry during 1990-1995 and stays for at least five years. The descriptive statistics are 1990 means and standard deviations. 1990 cross-section consists of all the workers in the metal industry population in 1990. Complexity refers to the occupation-related wage in FIM 1990. Bonus is reported as a proportion of occupation-related wage. Average hourly earnings consist of worker's earnings in different hourly wage schemes in FIM 1990.

Table2 Transition matrix between jobs for newcomers in the metal industry, 1990-1991

	Exit	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40	Total	N
Entry		36.5	45.1	18.2	19.5	18.4	14.3	12.8	15.4	8.1	7.7	6.3	4.3	4.9	2.8	3.7		7405
26	38.9	44.3	0.1	9.4	2.9	.	1.1	1.3	0.1	1.6	0.3	.	.	0.1	.	.	100	1762
27	39.9	0.1	45.3	3.0	0.1	4.6	3.4	.	1.4	0.3	1.3	0.4	.	0.1	.	.	100	704
28	30.0	1.2	0.4	59.9	2.9	1.0	2.3	0.9	0.5	0.4	0.5	0.2	100	3358
29	28.6	0.6	.	2.0	60.1	0.1	5.5	2.2	.	0.7	0.1	.	0.1	.	.	.	100	3475
30	28.6	.	0.7	1.6	.	56.3	5.5	0.2	3.1	0.4	2.1	0.6	0.9	0.1	.	.	100	1344
31	24.9	0.1	.	0.8	1.0	0.4	62.0	4.8	2.2	1.3	1.4	0.7	0.1	0.1	0.1	.	100	6796
32	25.7	0.1	.	0.3	0.5	.	2.3	62.2	0.1	6.0	2.1	.	0.5	.	0.2	.	100	6457
33	22.6	.	0.1	0.3	.	0.5	1.5	.	64.6	0.2	5.5	3.5	.	0.9	.	0.3	100	3303
34	20.9	.	.	0.1	0.2	.	0.7	3.3	.	63.8	9.1	0.3	1.3	.	0.4	.	100	6529
35	16.6	.	.	0.1	.	0.1	0.3	0.7	0.9	2.5	68.9	2.9	5.7	0.5	0.6	0.2	100	10382
36	13.8	0.1	0.2	.	0.7	.	4.3	74.0	0.1	5.3	0.1	1.6	100	3754
37	12.5	.	.	.	0.1	.	0.2	0.4	.	1.2	6.5	.	74.3	0.1	4.8	.	100	5677
38	13.1	0.1	.	0.3	.	0.9	4.0	0.7	76.7	0.1	4.1	100	2199
39	11.8	0.3	.	0.4	1.5	.	4.3	0.1	81.6	0.2	100	3964
40	12.2	0.3	.	0.5	2.4	.	2.7	1.7	80.1		100	2600
Total	20.6	1.4	0.6	3.8	3.8	1.4	7.9	7.7	4.0	8.1	14.2	5.5	8.3	3.3	5.9	3.6	100	69709

Shows all transitions between complexity levels, including entry, exit, and stays from 1990 to 1991, as percentage of movements from a complexity level to another. Aggregating occupation-related wages into integers creates complexity levels. Shaded cells indicate the level that was the most frequent destination of the complexity level moves. Numbers in boxed cells indicate stays within a complexity level. Zeros denote nonempty cells that round up to zero and “.”s denote empty cells.

Table 3 Change in complexity – Gender differences

	Male	Female	Difference
Total	.010 (.000)	.011 (.000)	.000 (.000)
Complexity group 26	.026 (.001)	.018 (.001)	-.008** (.001)
Complexity group 27	.035 (.002)	.018 (.002)	-.017** (.002)
Complexity group 28	.026 (.001)	.013 (.001)	-.013** (.001)
Complexity group 29	.017 (.001)	.009 (.001)	-.008** (.001)
Complexity group 30	.022 (.001)	.013 (.001)	-.009** (.001)
Complexity group 31	.014 (.000)	.005 (.001)	-.008** (.001)
Complexity group 32	.010 (.000)	.003 (.001)	-.007** (.001)
Complexity group 33	.010 (.000)	.005 (.001)	-.005** (.001)
Complexity group 34	.006 (.000)	.004 (.002)	-.002 (.002)
Complexity group 35	.005 (.000)	.001 (.001)	-.005** (.001)
Complexity group 36	.005 (.001)	-.002 (.003)	-.007** (.001)
Complexity group 37	.000 (.001)	.004 (.011)	.004 (.003)
Complexity group 38	.001 (.001)	.002 (.003)	.001 (.001)

The numbers in the second and third columns are the sample mean changes of log occupation-related wages for men and women. The fourth column gives the female-male difference. The groups were constructed by aggregating the occupation-related wages of the workers' initial task assignments to integers. The numbers in parentheses are standard errors. ** denotes significance at 5%-level and * at 10-level.

Table 4 Change in complexity – regression results

Variable	(3)
Female	-.008** (.000)
Age/10	-.011** (.001)
(Age/10) ²	.001** (.000)
Seniority/10	-.086** (.003)
(Seniority/10) ²	.066** (.002)
Initial task complexity dummies	Yes (15)
R ²	.135
N	64 198

Dependent variable is the difference between log of the occupation-related wage in the next period and the log of the occupation-related wage at the current period. Seniority measured as number of years that individual has been present in the metal industry. Regression also includes a full set of initial task complexity group and firm dummies. Numbers in parenthesis are robust standard errors. ** denotes significance at 5%-level and * at 10-level.

Table 5 Kaplan-Meier estimates of the survivor function by gender

Seniority	(1) The whole sample		(2) Low complexity tasks		(3) Medium complexity tasks		(4) High complexity tasks	
	Men	Women	Men	Women	Men	Women	Men	Women
1	0.709	0.746	0.596	0.707	0.679	0.808	0.859	0.943
2	0.564	0.628	0.410	0.580	0.536	0.709	0.770	0.862
3	0.473	0.531	0.327	0.478	0.441	0.624	0.696	0.813
4	0.412	0.468	0.274	0.413	0.385	0.567	0.624	0.766
5	0.371	0.424	0.237	0.371	0.345	0.521	0.583	0.723
Log-rank test	X ² (1)=21.72 (p=0.000)		X ² (1)=104.68 (p=0.000)		X ² (1)=116.85 (p=0.000)		X ² (1)=2.59 (p=0.108)	

Table reports the Kaplan-Meier estimates of the survivor function, where hazard is the probability of promotion conditional on not being promoted up to the year of seniority. Low-complexity refers to tasks that have occupation-related wages less than or equal to 30. Medium complexity refers to tasks that have occupation-related wages larger than 30 but less than or equal to 35. High complexity refers to tasks that have occupation-related wages higher than 35. The last row reports the log-rank test statistic for the hypothesis that the male and female survivor functions are equal. ** denotes significance at 5%-level and * at 10-level.

Table 6 Duration to promotion – Discrete-time complementary log-log proportional hazards estimates.

Variable	
Female	-.357** (.026)
Age/10	-.191** (.087)
(Age/10) ²	-.000 (.012)
Initial task complexity dummies	Yes (15)
N	64 198

Estimates from a discrete-time proportional-hazards model of promotion. Complementary log-log specification was chosen for the hazard of promotion. In addition to the variables listed in the table, 11 duration dummies were included in the estimation Standard errors are reported in the parentheses. ** denotes significance at 5%-level and * at 10-level.

Table 7 Median personal bonus deviations in the groups of promoted and not-promoted workers

	Stagnant men	Stagnant women	Difference	Promoted men	Promoted women	Difference
First assignment	-.027 (.000) N=8 956	-.027 (.001) N=2 705	.000 (.001)			
After 1 year	-.010 (.000) N=6 345	-.001 (.001) N=2 017	.009** (.001)	-.014 (.001) N=2 611	-.003 (.001) N=688	.011** (.001)
After 2 years	-.005 (.001) N=5 057	.001 (.001) N=1 675	.007** (.001)	-.010 (.001) N=3 899	.000 (.001) N=1 030	.011** (.001)
After 3 years	-.002 (.001) N=4 229	.005 (.001) N=1 431	.007** (.001)	-.007 (.000) N=4 727	.001 (.001) N=1 274	.008** (.001)
After 4 years	.000 (.001) N=3 687	.007 (.001) N=1 255	.007** (.001)	-.005 (.001) N=5 269	.003 (.001) N=1 450	.008** (.001)

Numbers in cells are medians of bonuses as deviations from firm- and task-specific means. Stagnant refers to the group of workers who remain in their first assignment job. Promoted refers to the groups of workers who move to more complex jobs. Medians of bonus deviations are reported for each group. The numbers in parentheses are standard errors. N refers to the sample size of the cell. ** denotes significance at 5%-level and * at 10-level.

Table 8 Probit model of voluntary quits, marginal effects

Variable	Marginal effect	Standard error
Female dummy	.050**	(.011)
Age/10	-.216**	(.002)
Age squared /100	.029**	(.000)
(Female x age)/10	-.008*	(.004)
(Female x age squared)/100	-.001	(.001)
Seniority/10	-.044**	(.001)
Seniority squared /100	.005**	(.000)
Firm size/1000	.000**	(.000)
14 complexity dummies	Yes	
N	630 839	
% correctly predicted	91.8	
Log-likelihood	-165709.14	
Pseudo R ²	.072	

Marginal effects of the probit model of quit behaviour using the whole population of metalworkers 1990-2000. Marginal effects of the female dummy reports the effect of 0 to 1 change. Quit is defined as a separation where at least 75% of the employees remain in the firm. ** denotes significance at 5%-level and * at 10-level.

Table 9 Predicted probabilities of quitting in different age groups

Age group	Male	Female
25 and below	.173	.242
25-30	.086	.124
30-35	.054	.079
35-40	.041	.059
40-45	.041	.054
45-50	.052	.059
50-55	.081	.079
55 and above	.166	.141

Predicted probabilities of quitting by age groups. The probabilities are calculated using the coefficients of the probit model in table 8. Quit is defined as a separation where at least 75% of the employees remain in the firm. ** denotes significance at 5%-level and * at 10-level.