

**Effective Working Hours and Wages:  
The Case of Downward Adjustment via Paid  
Absenteeism**

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# **Effective Working Hours and Wages: The Case of Downward Adjustment via Paid Absenteeism**

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## **Abstract**

This paper compares contractual with effective working hours and wages, respectively. Effective working hours are defined as contractual working hours minus absent working hours. This approach takes into account workers' downward adjustment of working time via paid absenteeism if working time constraints are present, which induce workers to accept contracts with larger than their optimal choice of working hours. A German personnel data set, which contains precise information on wages as well as working and absence hours, is used to assess the impact of such downward adjustment on wage inequality and wage differentials (gender, schooling, age).

**JEL Classification:** J22, J31

**Keywords:** absenteeism, earnings, inequality, wage differentials, working hours

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## 1. Introduction

One of the most frequently used application in econometrics is the estimation of earnings functions to assess wage differentials. The dependent variable is usually the log of earnings and measured on the basis of year, month, week, day, or hour. Most economic models rely on the hourly wage rate. In empirical practice, however, several hourly wage measures can be computed from the data and the question is which one is preferable. Contractual hourly wages (total income divided by contractual working hours) are normally used since they can be easily computed in most data sets. This might be problematic in an economic interpretation because a worker's utility does not depend on contractual working hours but on effective working hours and his perceived wage rate is not the contractual but the effective hourly wage (total income divided by effective working hours). Moreover, firms are interested in effective wages and not in contractual wages when making employment decisions. Thus, it is of central importance to define effective working hours and to assess the empirical importance of different hourly wage measures.

Some studies take into account overtime work when defining effective working hours and computing hourly wages (e.g., Bell and Hart, 1999; Bell et al., 2000; Hübler, 2002; Wolf, 2002; Anger, 2007). Overtime is however only one form of working time adjustment, which is upward orientated. In the presence of fixed contractual working hours, a worker might choose overtime work if his utility maximizing working hours are larger than contractual working hours. The important case of downward adjustment of working time via paid absenteeism is on the other hand largely ignored.<sup>1</sup> If working

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<sup>1</sup> Absenteeism refers to reported sickness related work absence of employees. Such sickness reports need not to be true. Even if an employee is really sick, he can choose to some degree how long he stays from

time constraints are present that induce a worker to accept a contract with larger than his optimal working hours, a worker can use absence to approach his optimal level of working hours (Allen, 1981; Brown and Sessions, 1996; Dunn and Youngblood, 1986).

This paper uses a personnel data set of a German company that is perfectly suitable to study the above issue because it comprises exact information about contractual, absent, and effective working hours as well as about wages. Moreover, the German case is of special interest for a first study due to its very generous institutional sickness benefits (Osterkamp and Röhn, 2007; Frick and Malo, 2008). Sick pay in Germany is regulated in the Act on Continued Payment of Remuneration ("*Lohnfortzahlungsgesetz*"). An employee who is sick for more than three days has to present a medical certificate of sickness from his physician. Sick employees have a legitimate claim of 100 percent wage replacement paid by the firm from the first absent day and for a period up to six weeks. In case of longer sickness absence due to the same disease, the wage replacement rate decreases to 70 percent and is paid by the health insurance up to 78 weeks. The issue of effective wages is, however, not only relevant for Germany but to some degree for every institutional setting in which workers receive sickness benefits when absent from work.

The next section illustrates the basic theoretical context, which is based on the static labour supply and demand models, and the relevance of effective hourly wages. Section 3 informs about the data set, basic descriptive statistics for hours and earnings, and

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work to recover from a disease or injury. Worker absenteeism is hence often used as a proxy for work effort, which is justified because absenteeism can be interpreted as shirking behaviour as well as a signal for work attachment (Barmby et al. 1994; Brown and Sessions, 1996; Ichino and Moretti 2009).

different inequality measures for contractual and effective wages. The regression results for hours and wages are presented in Section 4. The paper concludes with a summary and discussion of the results.

## 2. Theoretical framework

A worker's decision to be absent from work can be modelled in the framework of the static neo-classical labour supply model (Allen, 1981; Brown and Sessions, 1996). Let a worker's utility  $U$  in equation (1) depend positively on total consumption and total leisure. Leisure  $L$  is a fixed amount of time  $T$  minus the time spend at work  $H$  as depicted in equation (2). Consumption is generated from total income, which is for simplicity only labour income. Total labour income  $Y$  in equation (3) is constrained by the product of hourly wages  $w$  and the total number of working hours  $H$ .

$$(1) \quad U = U(Y, L) \quad \text{with} \quad \frac{\partial U}{\partial Y} > 0, \frac{\partial U}{\partial L} > 0$$

$$(2) \quad L = T - H$$

$$(3) \quad Y = wH$$

The worker's problem is to maximize his utility in equation (1) under the time constraint in equation (2) and the budget constraint in equation (3). The standard textbook solution is that a worker chooses to work as many hours until his marginal rate of substitution between leisure and consumption equals the hourly wage rate ( $\frac{\partial U / \partial L}{\partial U / \partial Y} = w$ ). In the graphical solution in Figure 1, the worker's optimal working hours choice  $H^*$  is the

tangential point of the indifference curve ( $U^*$ ) and the budget constraint with the slope  $-w = -w_C$ .

***- insert Figure 1 about here***

In the next step, the distinction between contractual and effective working hours and wages is made and illustrated in Figure 1. If the worker has to decide about accepting a job, firms offer him a fixed contractual numbers of working hours  $H_C$  and fixed contractual hourly wages  $w_C$  (e.g., due to collective contracts or inflexible work and pay schedules). The worker might have to accept a contract with larger than his optimal working hours ( $H_C > H^*$  and  $U_C < U^*$ ) because of such working hours constraints and a lack of better job opportunities. As a worker cares only about the working hours actually being at work, which are effective working hours  $H_E$ , he might deviate from contractual working hours. The difference between  $H_C$  and  $H_E$  can in principle be negative, i.e., the worker makes an upward adjustment of working time via overtime hours (Bell and Hart, 1999) which is not subject of this analysis, or positive, i.e., the worker makes a downward adjustment of working time via (fully) paid absent working hours  $H_A$ , which are valued by the worker as leisure hours.<sup>2</sup> The latter is possible as workers in Germany receive a 100 percent wage replacement rate in case of - sickness related - absenteeism. As total income  $Y_C$  is, at least if long-term aspects are not taken into account (Brown, 1994), independent of effective working hours, the worker perceives his effective hourly wage  $w_E$  as different from the contractual wage  $w_C$ . The

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<sup>2</sup> The assumption that all reported sickness absence is leisure increasing shirking behaviour of workers is quite strong. But it is useful to illustrate the issues of interested.

new time and budget constraints are as in equations (4) and (5) and the effective hourly wage is then calculated following equation (6).

$$(4) \quad L = T - H_C + H_A = T - H_E$$

$$(5) \quad Y_C = w_C H_C = w_E (H_C - H_A) = w_E H_E$$

$$(6) \quad w_E = \frac{Y_C}{(H_C - H_A)} = \frac{Y_C}{H_E}$$

The new budget constraint is a horizontal line at  $Y_C = w_C H_C$  due to the 100 percent wage replacement (see equation (5)). The worker has therefore an incentive to be as much absent as possible. In the extreme case, the worker would choose not to appear at work at all ( $H_E=0$  and  $L=T$ ). Such an behaviour is however unlikely to be tolerated by firms and likely to have negative career consequences (e.g., layoff, training, wage growth, promotion). Fairness and work group norms can also restrict such an extreme behaviour (e.g., unfair towards colleagues as they have to work more) (Bradley et al., 2007). Thus, a worker might choose reference points to determine his amount of absent and effective working hours (Munro and Sugden, 2003). He could for example choose his original optimal working hours ( $H_E=H^*$ ) or his original optimal utility level ( $U^*=U(Y_C, H_C > H_E > H^*)$ ) as reference points. The former example with  $H_E=H^*$  is used to briefly illustrate the effect of absenteeism on effective hourly wages and utility. The perceived effective hourly wage is total contractual income divided by effective working hours (see equation (6)) which results into a steeper hypothetical budget constraint because  $w_E > w_C$ . Further, utility in case of absenteeism is larger than to adhere contractual working hours ( $U_E(Y_C, H_E = H^*) > U_C(Y_C, H_C)$ ) because  $H_E < H_C$ .

Differences between workers in absence behaviour and consequently in effective wages can arise from heterogeneous preferences for leisure and consumption. Workers who defer in their optimal working hours are also likely to have different reference levels for effective working hours. A worker with lower optimal working hours is then likely to have more absent working hours and a higher effective wage compared to a worker whose optimal working hours do not deviate much from contractual working hours. Moreover, workers might be offered different contractual wages which also leads to different optimal working hours and reference levels. If workers with low contractual wages are more absent than workers with high contractual wages, the differences in effective wages between both groups would be lower than the differences in contractual wages.

Effective wages are furthermore crucial in the determination of labour demand, which is illustrated again in the static neo-classical model. A competitive firm maximizes its profits ( $\Pi$ ) in equation (7) if the difference between the value of total output ( $pQ$ ) and costs is maximised. We assume market output prices ( $p$ ), a fixed production technology ( $Q$ ), constant capital ( $K$ ), market capital prices ( $r$ ), and market contractual wages ( $w_C$ ). Moreover, total labour input consists of the number of workers ( $N$ ) times effective working hours ( $H_E$ ), which are contractual working hours ( $H_C$ ) minus absent working hours ( $H_A \leq H_C$ ). As the firm has to pay the contractual wage regardless of absent working hours (100 percent wage replacement), total labour costs are independent of absent working hours. Following the above computation of effective wages in equations (5) and (6), contractual labour costs can be reformulated into effective labour costs. Because workers are homogenous, all workers provide the same number of working hours. Furthermore, absent working hours are exogenously chosen by workers. Thus,



the firm's only choice variable is total labour input. The standard first order condition in (8) yields that a firm hires workers up to the point in which effective wages equal the value of marginal product.

$$(7) \quad \begin{aligned} \max_{H_E N} \quad \Pi &= pQ((H_C - H_A)N, K) - w_C H_C N - rK \\ &= pQ(H_E N, K) - w_E H_E N - rK \end{aligned}$$

$$\text{with } \frac{\partial Q}{\partial (H_E N)} > 0 \quad \text{and} \quad \frac{\partial^2 Q}{\partial (H_E N)^2} < 0$$

$$(8) \quad w_E = p \frac{\partial Q}{\partial (H_E N)}$$

The discrepancy between contractual and effective wages arises quite obviously, because effective wages are larger than contractual wages in case of workers' downward adjustment of working time via paid absenteeism. Thus, labour demand would be too high if falsely contractual instead of effective wages are taken into account, which would result into a loss of profits. Furthermore, a firm might statistical discriminate against worker groups with higher absenteeism, which would result into lower employment chances for these workers (Aigner and Cain, 1977; Pfeifer and Sohr, 2008).

### 3. Data set and descriptive statistics

The data set was extracted directly from computerized personnel records of a large German limited liability company that produces innovative products for the world market. The company has a works council and is subject to an industry wide collective

contract. The personnel records contain information on all employees in the company's headquarter on a monthly basis from January 1999 to December 2005. The subsequent empirical analysis includes all blue-collar and white-collar workers, who are neither apprentices nor trainees, who are not in early retirement schemes, and who are not absent on a permanent basis (e.g., parental leave, sabbaticals). Moreover, monthly observations are aggregated on the basis of calendar years because individual absenteeism is very volatile over the year. Therefore, workers who are not observed for all twelve months in a calendar year are excluded from the sample. In sum, the sample contains 9633 yearly observations of 1790 workers in an unbalanced panel design.

Table 1 presents descriptive statistics about working hours and earnings. Workers have on average 1815.5 contractual working hours per year. Because workers also have on average 58.4 absent working hours per year, which are officially sickness related and fully paid, effective working hours are only 1757.1 and hence significantly lower.<sup>3</sup> Workers are on average 3.25 percent of their contractual working time absent. The average probability that a worker reports absence in a year is larger than 70 percent. Nominal yearly gross income is on average 36727.6 Euros. The contractual hourly wage is computed by dividing yearly income by contractual working hours, whereas the effective hourly wage takes into account effective working hours in the denominator. The mean contractual wage is 20.00 Euros and the mean effective wage is significantly larger with 20.63 Euros, which are about three percent higher hourly wages.

***- insert Table 1 about here***

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<sup>3</sup> Unfortunately the personnel data set does not contain reliable information about unpaid overtime, which would be the opposite of downward adjustment via paid absenteeism.

Figure 2 depicts the development of working hours and wages from 1999 to 2005. It can be seen that differences between contractual and effective working hours are quite stable, although working hours vary over time. Wages are largely increasing due to their nominal character; but the differences between contractual and effective wages are again stable. Overall, the descriptive statistics indicate that downward adjustment of working time via paid absenteeism is a non negligible factor which has a significant impact on hourly wages.

***- insert Figure 2 about here***

Figure 3 depicts the distribution (kernel density estimations using Epanechnikov kernel function) of log contractual and effective hourly wages. It is not surprising that the distribution of effective wages is on the right hand side of the distribution of contractual wages. More interesting is, however, that especially workers at the lower tail of the wage distribution benefit if effective working hours are incorporated instead of contractual hours. The first picture, thus, suggests that inequality is lower for effective than for contractual wages because low wage workers might more frequently adjust their working time downwards via absenteeism to increase effective wages. This impression is supported by different inequality measures, which are summarized in Table 2. Although the standard deviation of contractual wages is slightly lower than of effective wages, which is due to the higher mean of the latter, all other inequality measures show a reduction in inequality when taking into account effective working time and wages, respectively.

***- insert Figure 3 about here***

***- insert Table 2 about here***

#### 4. Regression analyses

In this section, I estimate several regressions for hours and earnings using random effects GLS (generalized least squares) to exploit the panel nature of the data set. The Breusch and Pagan (1980) Lagrange multiplier test shows that the random effects model is more appropriate than cross sectional OLS (ordinary least squares), because the null hypothesis that the variance of the random effects equals zero is rejected at high significance levels in all regressions. Since most of our variables of interest (gender, schooling) are time invariant, fixed effects models are not very useful for the aim of this paper, which is to compare coefficients between different groups, namely by gender, highest schooling degree, and age categories.<sup>4</sup> In addition to these group indicating variables, all regressions include the observation years to control for aggregated influences and to deal with the nominal character of earnings.

Two specifications are estimated for every outcome variable. The first specification does not control for job levels obtained from wage groups in the collective contract, because these levels are highly correlated with earnings as well as hours and can be interpreted as an outcome of other covariates.<sup>5</sup> Nevertheless, it might be interesting to

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<sup>4</sup> For a detailed discussion of random and fixed effects as well as their application to wage regressions with monthly data of the analyzed company see Pfeifer (2009). A comparison of the two models shows very little differences between the coefficients, which are however significant in a Hausman (1978) specification test due to the large sample size and very small standard errors.

<sup>5</sup> See Pfeifer (2008) for a discussion of the impact of tenure, age, schooling, and job levels on wages in the analyzed company. A description of the job levels obtained from the collective contract, which is binding to the company, can also be found in Pfeifer (2008). In the present paper, job levels fulfil only the purpose of an additional control variable.

see if effects can still be found within job level, so that a second specification is estimated which includes twelve job level dummies. As the dependent variables have already been discussed at length in the previous section, Table 3 contains only a list of the independent variables and their descriptive statistics.

*- insert Table 3 about here*

Table 4 informs about the results for hours regressions. Five different outcome variables are used to get an impression of differences between gender, schooling degrees, and age categories: (1) number of contractual working hours, (2) number of absent working hours, (3) absence rate  $((2)/(1))$ , (4) absence probability  $((2)>0)$ , (5) number of effective working hours  $((1)-(2))$ . Women have on average 137 contractual working hours less, are 14 hours more absent, have a higher absence rate of one percentage point, have a higher absence probability of nine percentage points, and work in total 150 effective working hours less than men. It can also be seen that higher schooling degrees are associated with more contractual working hours, fewer absent working hours, a lower absence rate and absence probability, and consequently with more effective working hours. When additionally controlling for job levels, the differences between the gender and schooling groups are reduced but remain largely significant. Overall the results on working hours and absenteeism are in line with previous research (e.g., Allen, 1981; Brown and Sessions, 1996; Dunn and Youngblood, 1986; Bradley et al., 2007; Ichino and Moretti, 2009).

*- insert Table 4 about here*

Age has a more complex impact on hours. First, older workers have more absent hours and a higher absence rate, even though the absence probability is not significantly

affected by age. These results are quite similar in both specifications with and without job levels. Working hours, however, give not a conclusive picture, because statistical significance is low and some age coefficients even change their signs if job levels are incorporated. Nevertheless, it seems as older workers provide on average fewer effective hours since they are on average more hours absent.

The results of earnings regressions are presented in Table 5. The dependent variables are (1) log yearly income, (2) log contractual hourly wage, and (3) log effective hourly wage. All regressions reveal the usual findings that women earn significantly less than men, that more schooling is associated with significant higher earnings, that age-earnings profiles are upward sloping concave with slightly negative wage growth for the oldest workers, and that coefficients are smaller in the specifications with job levels as these absorb parts of the effects (Pfeifer, 2008). The aim of this paper is, however, to compare wage differential estimates between the different earnings measures with a special focus on differences between contractual and effective wages.<sup>6</sup> Hausman (1978) specification tests between the regressions reject the null hypothesis of no systematic differences between the models at high significance levels. Moreover, a fourth specification is estimated in which the dependent variable is the difference between log contractual wages and log effective wages. The significance of the estimated coefficients indicates that the estimated wage differentials differ significantly between contractual and effective wages.

***- insert Table 5 about here***

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<sup>6</sup> Bell and Hart (1999) perform a quite similar exercise for unpaid overtime.

On the one hand, wage differentials between men and women as well as between schooling degrees are smaller in effective wage regressions than in contractual wage regressions. The gender gap in hourly wages is reduced by more than one percentage point if effective instead of contractual working hours are taken into account. The wage premiums of workers with high school and college degrees are about two percentage points smaller for effective than for contractual hourly wages. On the other hand, the effects of age on wages are larger in effective wage regressions than in contractual wage regressions. Figure 4 illustrates the predicted age-earnings profiles of average workers for contractual and effective hourly wages from the specifications without job levels and Figure 5 illustrates the same from the specifications in which job levels are incorporated. It can be seen that age-earnings profiles are steeper and that wages of older workers do not decrease as much for effective hourly wages compared to contractual hourly wages.

*- insert Figure 4 about here*

*- insert Figure 5 about here*

## **5. Discussion and concluding remarks**

The main result of this paper is that wage differential estimates are systematically biased if the compared groups differ in work absence. The gender gap in effective hourly wages is more than one percentage point smaller than the gender gap in contractual wages, because women are on average more absent than men. Moreover, workers with lower schooling are more absent, which leads to an upward bias in

estimates for rates of return to schooling when contractual instead of effective wages are used. Older workers are also more absent so that contractual age-earnings profiles are significantly flatter than effective age-earnings profiles. Since workers at the lower tail of the wage distribution (e.g., women, low schooling) have higher absenteeism rates, the concept of effective wages reduces wage inequality compared to contractual wages. Bell and Hart (1999) report quite similar findings for unpaid overtime. Worker groups with more unpaid overtime have lower effective wages and hence wage differences are smaller than for contractual wages (e.g., men, higher schooling). A complete assessment of effective wages has therefore to account for paid and unpaid as well as for upward and downward adjustments of working hours. The general implication of my results is hence that researchers should be aware of potential biases in wage differential estimates and hourly wage inequality measures, if they are interested in the effective wage rate, which is the core of most economic models, and if absenteeism and other working time adjustments are not observed in the data, which is unfortunately the case for most data sets.

As firms are primarily interested in effective wages paid for effective labour supply, differences in effective working hours and wages are also likely to affect firms' employment decisions. If two groups differ on average in their effective wages, firms might statistical discriminate against workers from a group with on average higher absenteeism (women, low skilled, older workers) and hence prefer to recruit workers from a group with higher effective working hours (men, high skilled, younger workers). In a dynamic context, less employment chances for high absent workers might lead to a reduction of reservation wages and consequently to contractual wage differentials between low and high absent workers.



Some general implications on wage inequality and institutional sickness benefit systems can also be drawn. Sickness benefits are likely to reduce effective wage inequality because low wage workers are on average more absent than high wage workers. Therefore, sickness benefits, which are often financed by taxes or insurance premiums, lead to redistribution from high wage (low absent) workers to low wage (high absent) workers. If taxes and premiums are progressive, as in most countries, this redistribution effect is even larger. It might be worthwhile for future research to test relationships between the design of sickness benefit and health insurance systems on the one hand and inequality measures on the other hand. For this purpose, between and within country comparisons as well as natural experiments in micro data sets are possible empirical strategies. Such results would also help to improve our understanding of the political economy of sickness benefits and health insurances.

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*Figures and Tables Included in Text*

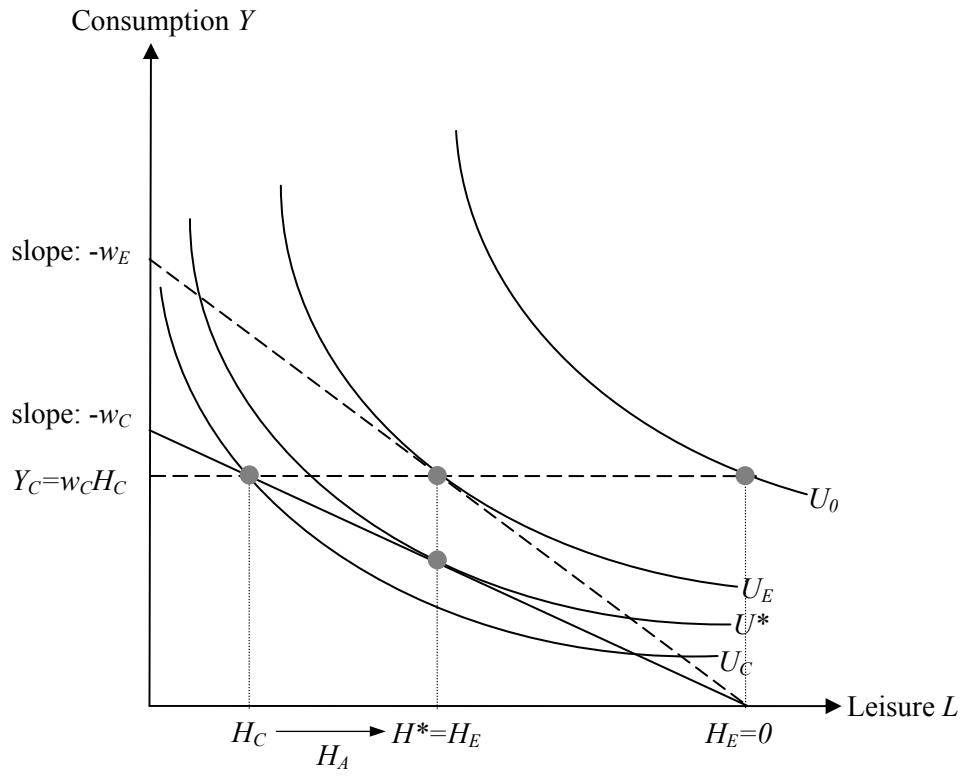


Figure 1: Effective working hours and wages in the labour supply model

Table 1: Descriptive statistics of hours and earnings

	Mean	Std. Dev.	Min.	Max.
<i>Hours</i>				
Contractual working hours	1815.5000	181.1042	207.8400	2088.6000
Absent working hours	58.3980	77.1738	0	658.0000
Absence rate	0.0325	0.0428	0	0.3602
Absence probability	0.7121	0.4528	0	1
Effective working hours	1757.1020	197.3925	207.8400	2088.6000
<i>Earnings</i>				
Yearly gross income in Euros	36727.58	14796.60	3855.14	157478.30
Log yearly income	10.4447	0.3538	8.2572	11.9670
Contractual hourly wage in Euros	19.9996	6.7531	9.5514	75.4206
Log contractual wage	2.9474	0.3018	2.2567	4.3231
Effective hourly wage in Euros	20.6326	6.7553	9.5514	76.8937
Log effective wage	2.9814	0.2928	2.2567	4.3424

Note: Number of yearly observations is 9633 of 1730 workers.

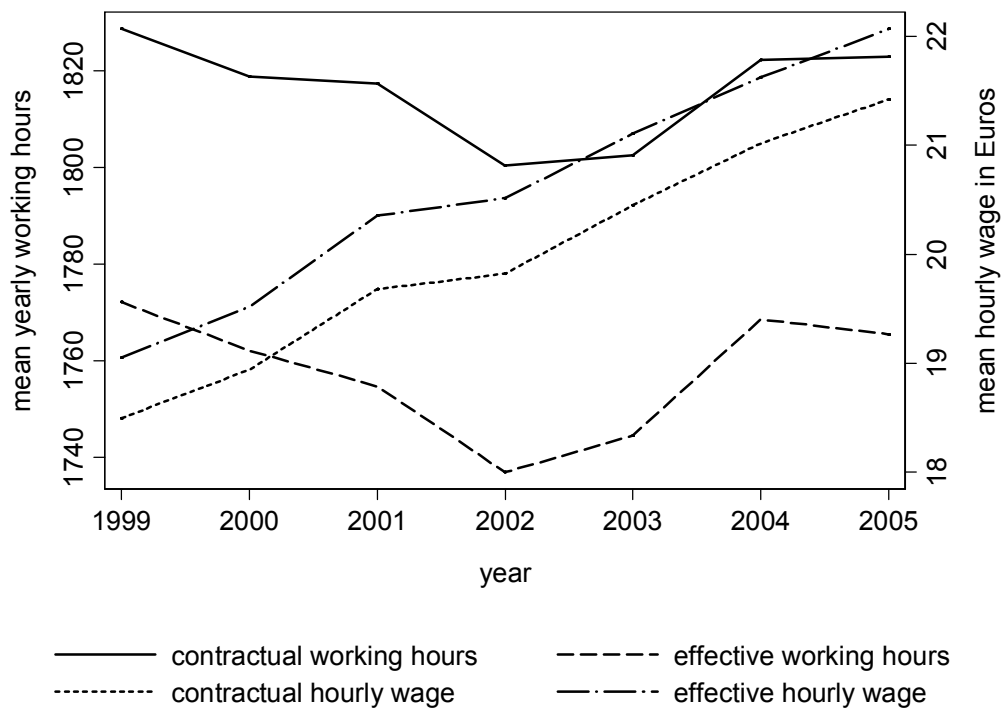


Figure 2: Time trends of hours and wages

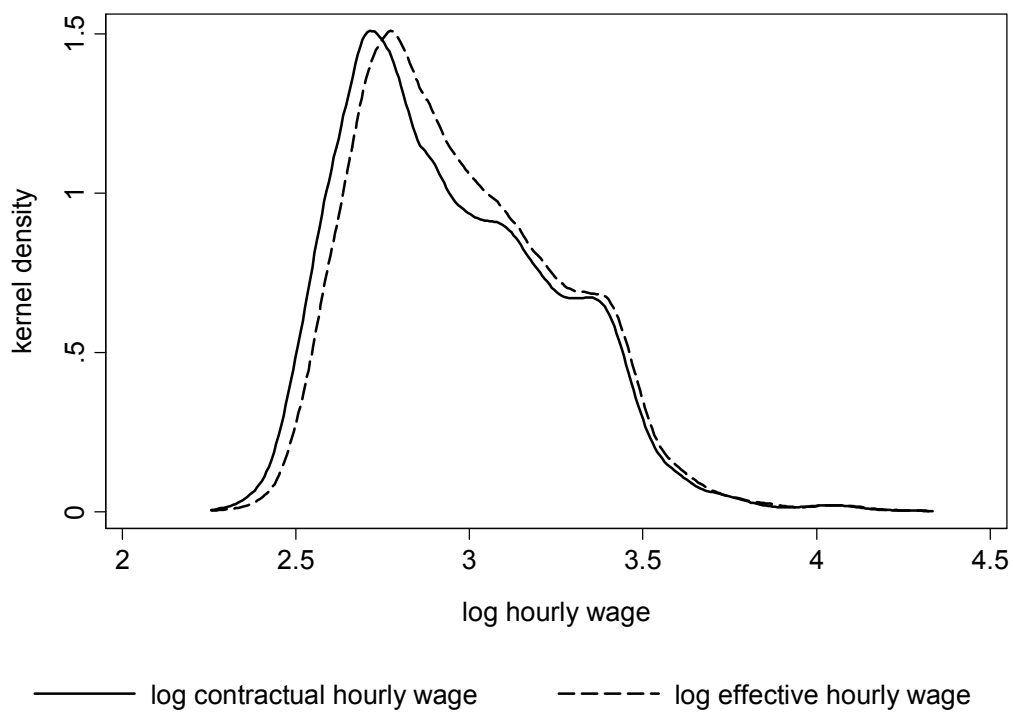


Figure 3: Distribution of contractual and effective wages

Table 2: Inequality of earnings

	(1) Yearly income	(2) Contractual wage	(3) Effective wage	(2) - (3)
Mean	36727.58	19.9996	20.6326	-0.6330
Standard deviation	14796.60	6.7531	6.7553	-0.0022
Standard deviation of logs	0.3538	0.3018	0.2928	0.0090
Relative mean deviation	0.1500	0.1308	0.1259	0.0049
Coefficient of variation	0.4029	0.3377	0.3274	0.0103
Gini coefficient	0.2048	0.1766	0.1710	0.0056
Mehran measure	0.2775	0.2424	0.2351	0.0073
Piesch measure	0.1685	0.1436	0.1390	0.0047
Kakwani measure	0.0391	0.0289	0.0272	0.0017
Theil entropy measure	0.0708	0.0512	0.0482	0.0030
Theil mean log deviation measure	0.0665	0.0484	0.0455	0.0029
Atkinson measure	0.0644	0.0472	0.0445	0.0027



Table 3: Descriptive statistics of control variables

	Mean	Std. Dev.
Female (dummy)	0.2363	0.4248
Less than high school degree (dummy, reference)	0.6966	0.4598
High school degree (dummy)	0.0969	0.2958
College degree (dummy)	0.2066	0.4049
Age <25 years (dummy, reference)	0.0225	0.1484
Age 25-35 years (dummy)	0.1916	0.3936
Age 35-45 years (dummy)	0.3920	0.4882
Age 45-55 years (dummy)	0.3081	0.4617
Age 55-65 years (dummy)	0.0857	0.2800
Year 1999 (dummy, reference)	0.1285	0.3347
Year 2000 (dummy)	0.1361	0.3429
Year 2001 (dummy)	0.1445	0.3516
Year 2002 (dummy)	0.1566	0.3635
Year 2003 (dummy)	0.1522	0.3592
Year 2004 (dummy)	0.1452	0.3524
Year 2005 (dummy)	0.1368	0.3437
Blue-collar level 1 (dummy, reference)	0.0339	0.1811
Blue-collar level 2 (dummy)	0.0270	0.1621
Blue-collar level 3 (dummy)	0.0396	0.1949
Blue-collar level 4 (dummy)	0.1638	0.3701
Blue-collar level 5 (dummy)	0.0799	0.2712
Blue-collar level 6 (dummy)	0.0470	0.2117
Blue-collar level 7 (dummy)	0.0195	0.1383
White-collar level 1 (dummy)	0.0523	0.2227
White-collar level 2 (dummy)	0.1396	0.3466
White-collar level 3 (dummy)	0.1478	0.3549
White-collar level 4 (dummy)	0.0910	0.2877
White-collar level 5 (dummy)	0.0694	0.2542
White-collar level 6 (dummy)	0.0890	0.2847

Note: Number of yearly observations is 9633 of 1730 workers.

Table 4: Hours regressions

	(1) Contractual working hours	(2) Absent working hours	(3) Absence rate	(4) Absence probability	(5) Effective working hours	(6) Contractual working hours	(7) Absent working hours	(8) Absence rate	(9) Absence probability	(10) Effective working hours
Female (dummy)	-136.7559*** (8.1680)	13.6377*** (3.1904)	0.0099*** (0.0018)	0.0910*** (0.0170)	-150.2349*** (8.5955)	-124.4736*** (8.3162)	6.3345* (3.5604)	0.0056*** (0.0020)	0.0663*** (0.0198)	-134.0164*** (8.8675)
High school degree (dummy)	-18.5385 (11.7859)	-38.9132*** (4.6502)	-0.0217*** (0.0026)	-0.1056*** (0.0248)	20.8897* (12.4295)	-19.2151* (11.2615)	-17.8699*** (4.7073)	-0.0102*** (0.0026)	-0.0524** (0.0260)	1.9734 (11.9217)
College degree (dummy)	103.5522*** (8.3267)	-40.2240*** (3.2693)	-0.0232*** (0.0018)	-0.1673*** (0.0174)	144.6246*** (8.7699)	37.5580*** (9.4073)	-13.9547*** (4.0838)	-0.0080*** (0.0023)	-0.0772*** (0.0227)	52.5169*** (10.0737)
Age 25-35 years (dummy)	2.8233 (8.8006)	9.5235* (5.6237)	0.0052* (0.0031)	0.0346 (0.0343)	-7.2795 (10.4785)	-5.3358 (8.8408)	11.0182* (5.7020)	0.0064** (0.0032)	0.0385 (0.0352)	-17.2298 (10.5232)
Age 35-45 years (dummy)	-7.4289 (9.5165)	8.1959 (5.8050)	0.0048 (0.0032)	-0.0131 (0.0348)	-16.1024 (11.2200)	-25.9359*** (9.5633)	11.7586** (5.9376)	0.0073** (0.0033)	0.0016 (0.0362)	-40.0274*** (11.2827)
Age 45-55 years (dummy)	5.8924 (10.1823)	10.4499* (5.9971)	0.0058* (0.0033)	-0.0209 (0.0357)	-0.7777 (11.9055)	-18.7376* (10.1926)	17.0068*** (6.1528)	0.0100*** (0.0034)	0.0065 (0.0373)	-34.5452*** (11.9368)
Age 55-65 years (dummy)	16.5921 (11.5945)	14.3830** (6.6653)	0.0075** (0.0037)	-0.0576 (0.0394)	7.0986 (13.4819)	-13.5386 (11.5292)	23.4457*** (6.7940)	0.0132*** (0.0038)	-0.0186 (0.0409)	-34.9853*** (13.4231)
Year (6 dummies)	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Job level (12 dummies)	No	No	No	No	No	Yes	Yes	Yes	Yes	Yes
Mean dependent variable	1815.50	58.3980	0.0325	0.7121	1757.10	1815.50	58.3980	0.0325	0.7121	1757.10
R squared (overall)	0.2128	0.0580	0.0645	0.0319	0.2282	0.3226	0.1134	0.1190	0.0496	0.3277
Breusch Pagan test ( $\chi^2$ )	14672.84***	3787.49***	3757.59***	2215.78***	12021.37***	12840.16***	3023.07***	3004.51***	2020.66***	10211.46***

Note: Number of yearly observations is 9633 of 1730 workers. Reference groups are men, have less than high school degree, and are younger than 25 years. Random effects GLS. Standard errors in brackets. Coefficients and test values are significant at \*\*\* 1%, \*\* 5%, and \* 10%.

Table 5: Earnings regressions

	(1) Log yearly income	(2) Log contractual wage	(3) Log effective wage	(4) (Log $w_C$ ) - (Log $w_E$ )	(5) Log yearly income	(6) Log contractual wage	(7) Log effective wage	(8) (Log $w_C$ ) - (Log $w_E$ )
Female (dummy)	-0.2411*** (0.0143)	-0.1516*** (0.0125)	-0.1402*** (0.0121)	-0.0105*** (0.0019)	-0.1363*** (0.0082)	-0.0588*** (0.0057)	-0.0425*** (0.0061)	-0.0059*** (0.0021)
High school degree (dummy)	0.1297*** (0.0204)	0.1416*** (0.0178)	0.1223*** (0.0173)	0.0232*** (0.0028)	-0.0214* (0.0111)	0.0045 (0.0079)	-0.0179** (0.0083)	0.0109*** (0.0028)
College degree (dummy)	0.4644*** (0.0145)	0.4084*** (0.0127)	0.3838*** (0.0123)	0.0247*** (0.0020)	0.0640*** (0.0092)	0.0742*** (0.0064)	0.0326*** (0.0069)	0.0085*** (0.0025)
Age 25-35 years (dummy)	0.0743*** (0.0089)	0.0611*** (0.0049)	0.0700*** (0.0066)	-0.0057* (0.0034)	0.0423*** (0.0081)	0.0448*** (0.0043)	0.0505*** (0.0059)	-0.0069** (0.0034)
Age 35-45 years (dummy)	0.1061*** (0.0099)	0.0923*** (0.0055)	0.1077*** (0.0073)	-0.0053 (0.0035)	0.0602*** (0.0088)	0.0722*** (0.0048)	0.0831*** (0.0064)	-0.0080** (0.0036)
Age 45-55 years (dummy)	0.1172*** (0.0109)	0.0862*** (0.0062)	0.1060*** (0.0081)	-0.0063* (0.0036)	0.0738*** (0.0095)	0.0746*** (0.0052)	0.0903*** (0.0069)	-0.0109*** (0.0037)
Age 55-65 years (dummy)	0.1119*** (0.0126)	0.0659*** (0.0072)	0.0932*** (0.0094)	-0.0082** (0.0040)	0.0748*** (0.0107)	0.0675*** (0.0060)	0.0902*** (0.0079)	-0.0144*** (0.0041)
Year (6 dummies)	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Job level (12 dummies)	No	No	No	No	Yes	Yes	Yes	Yes
Mean dependent variable	10.4447	2.9474	2.9814	-0.0340	10.4447	2.9474	2.9814	-0.0340
R squared (overall)	0.4274	0.3944	0.3854	0.0625	0.8375	0.8951	0.8769	0.1158
Breusch Pagan test ( $\chi^2$ )	19975.58***	22346.56***	20892.66***	3670.24***	12409.31***	13606.85***	10772.14***	2939.20***
Hausman test ( $\chi^2$ )		45.58***			620.85***			

Note: Number of yearly observations is 9633 of 1730 workers. Reference groups are men, have less than high school degree, and are younger than 25 years. Random effects GLS. Hausman test refers to regressions for contractual and effective wages. Standard errors in brackets. Coefficients and test values are significant at \*\*\* 1%, \*\* 5%, and \* 10%.

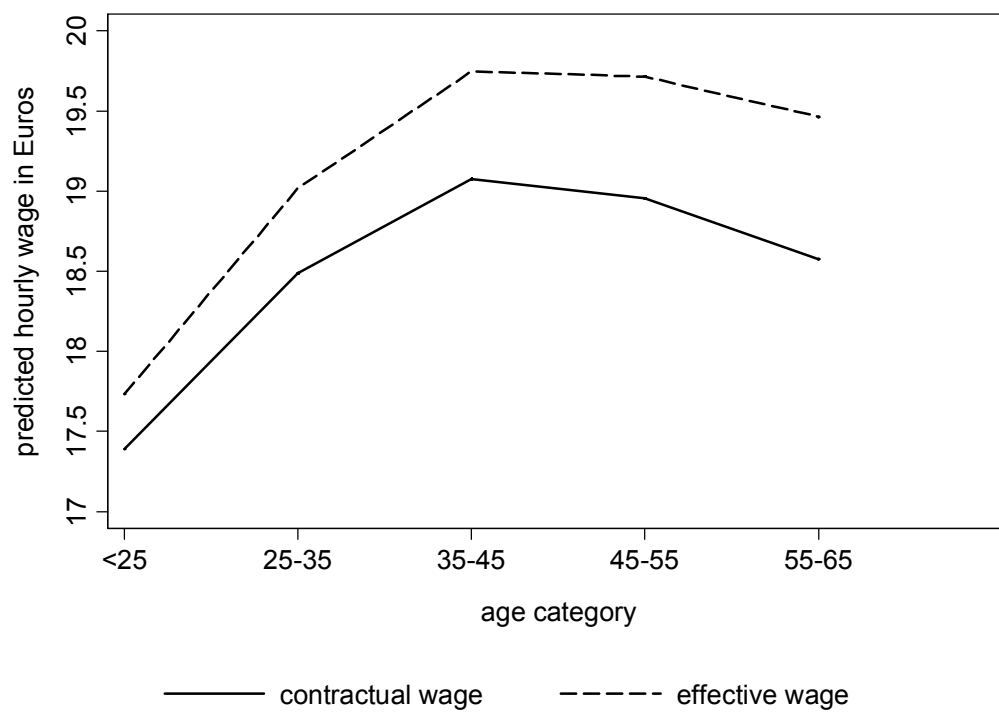


Figure 4: Age-earnings profiles for contractual and effective wages (without job levels)

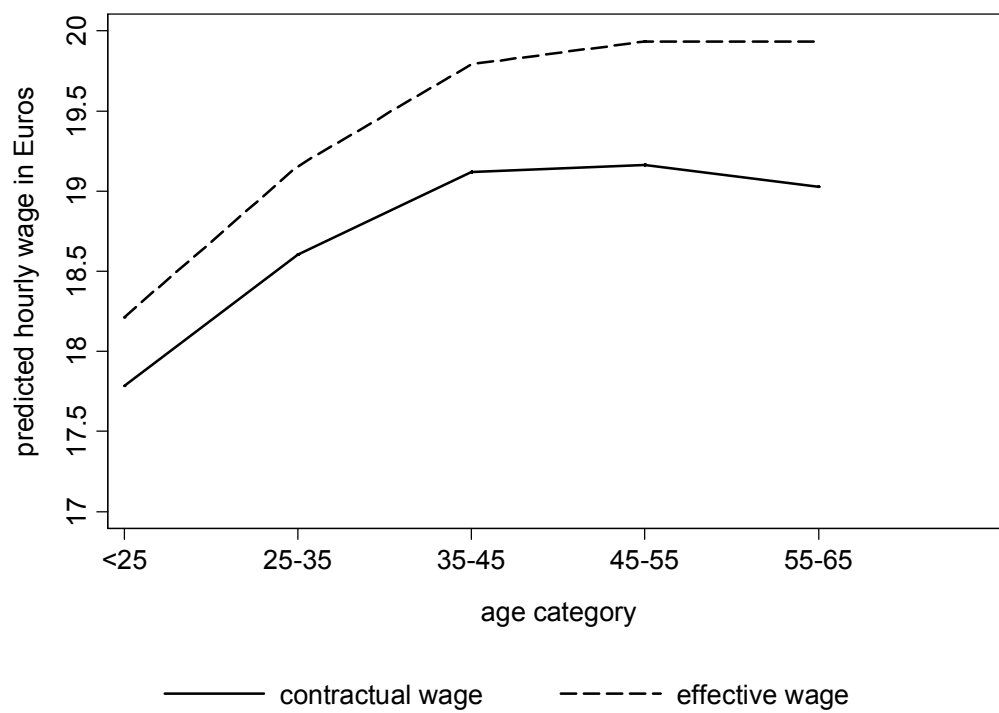


Figure 5: Age-earnings profiles for contractual and effective wages (with job levels)

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