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JEL Codes: D63, J28, J31 Keywords: Fair income, subjective well-being, quits, SOEP



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# Unfairness at Work: Well-Being and Quits\*

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#### Abstract

We here consider the effect of unfair income on both subjective well-being and objective future job quitting. In five waves of German Socio-Economic Panel data, those who perceive their labour income to be unfair have significantly lower subjective well-being, both in terms of cognitive evaluations (life and job satisfaction) and affect (the frequency of feeling happy, sad, angry and worried). Perceived unfairness also translates into objective labour-market behaviour, with current unfair income predicting future job quits.

Keywords: Fair income, subjective well-being, quits, SOEP.

#### JEL Classification Codes: D63, J28, J31.

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#### 1. Introduction

Not all aspects of our life are fair. Unfairness is perhaps particularly salient in the labour market, with its great variety of different job types and rewards, many of which are visible to others. Full-time workers in the OECD devote 62% of their day, or close to 15 hours, to personal care (eating, sleeping, and so on) and leisure (socialising with friends and family, hobbies, games, computer and television use, etc.).<sup>1</sup> Individuals' perceptions of the labour market are thus directly salient for almost 40% of the day, and may well colour individuals' feelings even when they are not at work.<sup>2</sup>

Unfairness can manifest itself in a variety of aspects of jobs, with unfair income likely being one of the most obvious. Income is of course only one aspect of a job, as underlined in the job-quality literature (see Clark, 2015, for a recent contribution), but is undeniably a key element of a good job and certainly quantifiable (as opposed to effort, say) and potentially visible. A worker who perceives their pay to be unfair may in return feel less committed to the job and take actions ranging from putting less effort into the job to the more extreme decision of quitting for an alternative job. Receiving less than what is deemed to be fair may generate a sense of relative deprivation with respect to some reference group, and thus reduce both job and life satisfaction. Given the reasonable degree of correlation between different measures of subjective well-being (Clark, 2016), we may expect the negative effect of unfairness to also be felt in workers' stated measures of affective well-being. We will test all of these correlations in the current paper.

The seminal work of Akerlof (1982) and Akerlof and Yellen (1990) introduced the fair wage-effort hypothesis and the consequences of unfairness perceptions on economic behaviour. Akerlof (1982) suggests that the wage-effort relationship is similar to a gift exchange that is regulated by norms of behaviour. Workers develop sentiments for their firm and co-workers, and provide effort in excess of the minimum work standard as a gift in exchange for a wage that is above the market-clearing level. Workers' conception of fair wages are based, among other things, on comparisons to a reference group composed of similar workers, or the wages that the individuals received themselves in the past. Perceiving labour income as unfair generates sentiments of anger and dissatisfaction, which lead workers to reduce their effort below the level they would offer if fully satisfied.

<sup>&</sup>lt;sup>1</sup> See http://www.oecdbetterlifeindex.org/topics/work-life-balance/.

 $<sup>^{2}</sup>$  Clark and Senik (2010) find that 60% of the individuals who say that they compare their income do so to work colleagues.

There is first a very large literature showing that individuals do compare. They evaluate outcomes not only in absolute terms but also relative to some reference level, which can be external (social comparisons) or internal (past or expected future outcomes). With respect to social comparisons, individuals compare their situation to that of others such as people working in the same firm or industry, neighbours, or friends (Clark, 2003, Clark and Oswald, 1996, Ferrer-i-Carbonell, 2005, and Luttmer, 2005). For the internal reference, individuals evaluate their actual situation relative to their own situation in the past (Frederick and Loewenstein, 1999), their aspirations (Stutzer, 2004) or expectations (McBride, 2010). These comparisons have been evoked to explain the Easterlin paradox (Easterlin, 1974, 2001), whereby in developed countries overall satisfaction has broadly not increased over time with economic growth, even though the cross-section relationship between satisfaction and income is positive.<sup>3</sup>

Other work has explicitly appealed to the notion of fairness. Experimental work has provided evidence for the fair wage-effort hypothesis, with individuals adjusting their effort according to fairness considerations (Mathewson, 1969, Cohn *et al.*, 2014, Blinder and Choi, 1990, Bewley, 1999, and Fehr *et al.*, 1993; Fehr and Gächter, 2000, provide a review of this literature).

Survey evidence on the wage-effort relationship is rarer. Clark *et al.* (2010) use 1997 ISSP data to show that individuals who have lower income relative to a comparison group (defined by country, sex, education and aged) are less likely to report working harder than they have to help their firm. The physiological responses to unfair treatment are explored in recent work by Falk *et al.* (2016), who look at the effect of earning an unfair wage on workers' health using both experimental and survey data (the survey data is the same as that used here). They find that workers who perceive their wage as unfair are more likely to suffer from stress-related diseases such as cardiovascular health problems.<sup>4</sup>

In contrast to work on the consequences of unfair wages on effort, the analogous empirical evidence on worker well-being is far more scant. Experimental evidence of the effect of fairness on emotions is provided by Pillutla and Murnighan (1996) in the context of the ultimatum game, and by Bosman and Van Winden (2002) and Ben-Shakhar *et al.* (2007) in the context of two-player power-to-take game. These show that participants reveal negative emotions when treated unfairly, and react by rejecting ultimatum offers and punishing unfair

<sup>&</sup>lt;sup>3</sup> There is of course a large literature on attitudes to inequality, but much of this does not explicitly refer to fairness. Clark and D'Ambrosio (2015) is a recent survey.

<sup>&</sup>lt;sup>4</sup> See also Pfeifer (2015), who finds that the perception of unfair pay is correlated with less sleep and more sleep disorders.

behaviours. The feeling of anger produced by unfairness is also correlated with physiological measures of skin conductance (see Ben-Shakhar *et al.*, 2007). Pfeifer (2016) uses the same survey data as we do here to show that unfairness perceptions increase the frequency of feeling anger.

We contribute to the literature on unfair wage perceptions using five waves of a nationally-representative survey, the German Socio-Economic Panel Survey (SOEP), to estimate the effect of earning an unfair income on a variety of measures of subjective wellbeing. We consider both cognitive measures of subjective well-being (job and life satisfaction) as well as measures of positive and negative affect. Controlling for the level of income, perceiving one's own income as unfair is associated with significantly lower levels of both job and life satisfaction. We also find lower levels of worker affect, here the frequency of feeling happy (positive affect), and sad, angry and worried (negative affects): the largest effect size of unfair income is with respect to anger. We also confirm the validity of these results by showing that unfair income is associated with not only subjective evaluations but also objective behaviour: the probability of quitting the job within the next year. We are not aware of any existing work that has shown that unfair income leads to quits. In SOEP data, controlling for income, perceived unfairness does indeed lead to a significant increase in the probability of quitting the job.

The remainder of the paper is organized as follows. Section 2 describes the data, and Section 3 contains the empirical strategy and results. The robustness checks are discussed in Section 4. Last, Section 5 concludes.

#### 2. Data

We use data from the SOEP, a longitudinal survey that has been conducted yearly since 1984 and that covers about 13,000 households and 24,000 individuals. Starting in 2005, SOEP respondents are asked if they think that the income they earn in their current job is fair and, if not, what the fair net monthly amount would be.<sup>5</sup> The question appears every second year: we thus here consider five SOEP waves (2005, 2007, 2009, 2011 and 2013). We restrict the sample to employed individuals aged 25 to 65 (as many Germans are still in education at younger ages: the results using those aged 18-65 turn out to be very similar).

More than one-third (37.4 percent) of respondents think that the income they earn is not fair, with very similar figures for men and women (see Table 1). Lower-educated individuals

<sup>&</sup>lt;sup>5</sup> Respondents are also whether they think their pay is fair in HILDA (see Long, 2005), but not about what the level of fair income would be.

are more likely to report their income as unfair (41 percent) than the highly educated (32 percent). The proportion of individuals reporting unfair income is also related to age, with the highest figures being found in the youngest cohort (aged up to 35: 41 percent). Half of the respondents from East Germany consider their income as unfair (50 percent), which is a much higher figure than that in West Germany (34 percent). The percentage of respondents perceiving their income as unfair was lowest in 2005 (at 33 percent) and highest in 2007 (at 42 percent). This time pattern is similar for men and women of all ages, and those living in either West or East Germany.

	Income is Unfair (%)	Mean Fair Income	Gap from Actual Income (%)
Gender			
Female	37.2	1703.7	529.6 (58.3)
Male	37.6	2537.1	753.4 (49.5)
Education			
Less than high school	40.8	1607.6	485.6 (64.4)
High school	39.3	1933.3	561.5 (51.9)
More than high school	31.7	2926.5	945.8 (54.1)
Age			
<=35	40.5	1865.0	534.9 (53.2)
36-45	35.7	2152.9	628.6 (51.6)
46-55	37.5	2248.3	707.2 (54.3)
>55	34.6	2298.1	725.2 (57.5)
West/East Germany			
West	33.5	2239.6	655.3 (48.8)
East	50.4	1903.7	623.9 (64.7)
Year			
2005	32.9	2092.5	566.7 (45.4)
2007	42.3	2193.2	645.0 (50.2)
2009	36.8	2041.7	608.1 (56.5)
2011	36.8	2144.1	683.9 (57.2)
2013	37.8	2191.1	711.7 (58.5)
Total sample	37.4	2135.5	645.5 (53.7)

Table 1 - Percentage of respondents considering their income unfair, average amount of fair income and income gap (fair income – income) by population subgroup

In terms of the income considered to be fair and the income gap, respondents who consider their income unfair earn on average a net income level per month that is about 645 Euros lower than what they consider to be fair, which corresponds to a gap of about 54% relative to their actual income (see column 3). The level of income perceived as fair is much higher for men than for women in absolute terms, and rises with education and age. However, the gap to the fair income is larger for women than for men. The same pattern of a higher fair

income figure but a smaller percentage gap is seen for the higher-educated relative to the loweducated, and in West compared East Germany. With the exception of 2009, the size of the income gap has been rising over time. Compared to the first survey wave (2005), the gap to fair income in 2013 is 13 percentage points higher.

#### 3. Empirical strategy and results

We estimate the effect of perceiving one's own income as unfair on subjective wellbeing, measured by job and life satisfaction, and emotional well-being. Job and life satisfaction are both measured on 11-point scales ranging from 0 (completely dissatisfied) to 10 (completely satisfied). Positive and negative affect are measured by the frequency of feeling happy (positive), and sad, angry and worried (negative) in the past week on 5-point scales (1=very rarely; 5=very often). In order to control for unobservable factors such as personality traits, we exploit the panel nature of the dataset and estimate linear models with fixed-effects. All the models include the following controls: individual net monthly income, age and education dummies, marital status, number of children, hours worked, health status, an East Germany dummy, the regional unemployment rate, firm size, and industry, occupation and wave dummies. Given that subjective well-being is often considered to be concave in income, income is introduced in logarithm form. The summary statistics of all the variables used in the analyses appear in Appendix 1A. The standard errors in our regressions are clustered at the region-year level, as this is the aggregation level of regional unemployment.<sup>6</sup>

Our key fairness variable here is the gap between the level of income considered to be fair and the actual income received. This gap is set equal to zero for those who consider their income to be fair, and is positive for those whose income is lower than what they consider to be the fair amount. The income gap is entered in log form.

Table 2 displays the results for job and life satisfaction. For both dependent variables, we first estimate a baseline specification with income and the basic controls, and then add the income gap in a second specification.

The estimated coefficients on income are significant for both job and life satisfaction in the baseline specification (columns 1 and 3), with (perhaps unsurprisingly) a stronger effect on job satisfaction. The results with respect to the other control variables are very standard in the literature, and the full table of results is relegated to Appendix 2A. Compared to the

<sup>&</sup>lt;sup>6</sup> Clustering at the individual level does not change the nature of any of our results.

married, those who are separated or divorced are less satisfied with their life, while there is no difference regarding job satisfaction. Education has no significant impact on life satisfaction; however, we do find that higher-educated individuals are less satisfied with their jobs, which may well reflect occupational aspirations and job-related stress.<sup>7</sup> General health, as self-reported by respondents, is strongly positively correlated with both measures of satisfaction.<sup>8</sup> Higher annual hours are associated with lower job satisfaction. Individuals in East Germany are significantly less satisfied with their jobs. Firm size has a positive effect on both job and life satisfaction. Last, the regional unemployment rate is estimated to significantly reduce life satisfaction, which is a common finding in the literature.

The fair-income gap is significantly correlated with both job and life satisfaction. Those whose pay is unfair are significantly worse off in terms of both job and life satisfaction compared to individuals who perceive their income as fair. The estimated coefficient on the income gap is larger than that on income for both subjective well-being measures here. As such, relative income (here given by fairness considerations) matters more for well-being than does the absolute level of income. Unfair income has much larger effect on job than on life satisfaction: doubling the income gap is estimated to reduce life satisfaction by 0.1 of a point on the 0-10 scale, but to reduce job satisfaction by 0.5 of a point.

	Life satisfaction		Job sati	isfaction
	(1)	(2)	(3)	(4)
Net individual income (ln)	0.132***	0.106**	$0.477^{***}$	0.366***
	(0.040)	(0.042)	(0.051)	(0.053)
Income gap (ln)		-0.163***		$-0.702^{***}$
		(0.037)		(0.065)
Controls	Yes	Yes	Yes	Yes
Wave dummies	Yes	Yes	Yes	Yes
Individual fixed-effects	Yes	Yes	Yes	Yes
Ν	36746	36746	36746	36746
R-squared overall	0.18	0.18	0.05	0.06

Table 2 – The fair income gap and life and job satisfaction

<u>Notes</u>: These are linear models with individual fixed effects. \*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Income gap = ln(fair income) - ln(income). Additional controls: age dummies, marital status, education, number of children, health status, hours worked, regional unemployment rate, firm size, industry dummies, occupation dummies, and individual fixed-effects.

Our second type of well-being measure refers to affect. We show results separately for positive affect (the frequency of feeling happy), and for negative affect (the average frequency of feeling sad, angry and worried). The results for each separate negative affect can be found in the appendix. These measures are not available in 2005, so we have one wave less

<sup>&</sup>lt;sup>7</sup> It should be remembered that these are fixed-effect regressions. Education does not vary that much within individual over time, making the interpretation of the estimated coefficients a little more difficult.

<sup>&</sup>lt;sup>8</sup> Excluding self-reported health produces similar results.

than for the satisfaction analyses. Table 3 shows the estimates on emotional well-being (the full list of estimated coefficients appears in Appendix 3A).

Income is now insignificant for all affects,<sup>9</sup> while the gap between fair and actual income is significantly correlated with negative affect, increasing the average frequency of feeling sad, angry and worried.

Looking at the three negative affects separately, we can see that this result is driven by the effect of unfairness on anger (see Appendix, Table 4A). The effect of fairness on anger confirms what has been suggested in the literature on fairness and reciprocity that the punishment of unfair treatment, in the form for example of less effort or negative reciprocity, is due to feelings of dissatisfaction and anger. Fairness is insignificant for the other affects, while the frequency of feeling worried turns out to be significantly correlated with the absolute level of income.

	Positive affects <sup>(a)</sup>		Negative affects <sup>(b)</sup>	
	(1)	(2)	(3)	(4)
Net individual income (ln)	0.007	0.007	-0.018	-0.011
	(0.020)	(0.021)	(0.016)	(0.016)
Income gap (ln)		-0.000		$0.039^{*}$
		(0.019)		(0.020)
Controls	Yes	Yes	Yes	Yes
Wave dummies	Yes	Yes	Yes	Yes
Individual fixed-effects	Yes	Yes	Yes	Yes
Ν	30596	30596	30596	30596
R-squared overall	0.06	0.06	0.07	0.07

#### Table 3 – Effect of fair income on emotional well-being

<u>Notes</u>: <sup>(a)</sup> Frequency of feeling happy; <sup>(b)</sup> Average of frequency of feeling sad, angry and worried. Income gap = ln(fair income) - ln(income). These are linear models with fixed effects. \*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Additional controls: age, marital status, education level, number of children, health status, hours worked, regional unemployment rate, firm size, industry dummies, occupation dummies, and individual fixed-effects.

#### 4. Unfair income and the probability of quitting the job

We now turn from subjective to objective outcomes, and estimate the relationship between unfair income and the probability of quitting the job. We relate the information on unfair income to the probability of quitting the job over the next year. We define a job quit as a job change that was caused by the worker leaving the job intentionally (i.e. resigning). About seven percent of employees in the 2005, 2007, 2009, 2011 and 2013 waves separate

<sup>&</sup>lt;sup>9</sup> Kahneman and Deaton (2010) use data on US respondents to the Gallup-Healthways Well-Being Index to show that income is more strongly correlated with a cognitive/evaluative measure of subjective well-being (the Cantril Ladder) than with positive and negative affect. In the cross-country regressions of Gallup World Poll data in Layard *et al.* (2012), GDP per capita is correlated with neither positive nor negative affect once controls are introduced (but is correlated with the Cantril Ladder (see their Table 3.1, page 65).

from their job over the next year; 42% of these separations come about from the worker resigning (see Table 4). We create a dummy variable for the respondent quitting their job and use this as the dependent variable to estimate the relationship between fairness and job quitting.

	Freq.	Percent
Company Shut Down	200	7.0
Own Resignation	1,210	42.3
Dismissal	465	16.2
Mutual Agreement	324	11.3
Temporary Contract Expired	428	15.0
Reached Retirement Age, Pension	76	2.7
Leave of Absence, Sabbatical	54	1.9
Business Closed Down (Self-Employed)	105	3.7
Total	2,862	100

 Table 4 – Reason for job termination

The results appear in Table 5 (the full table of coefficients is in Appendix 5A). We estimate linear probability models (the results using probit models are similar and are available upon request). Column 1 contains the results including only income and the other controls; column 2 then adds the gap between fair and actual income; and column 3 is the same model as in column 2 but also controls for the Big 5 personality traits.<sup>10</sup> As almost no-one quits more than once over these five waves, we cannot introduce individual fixed effects; our use of personality traits helps to control for unobserved differences between individuals.

As expected, income reduces job quits. In column 2, we find a significant effect of unfairness on the probability of quitting. Income remains significant here, but with a smaller estimated coefficient. The effect of the income gap is significantly higher than that of absolute income, and the results in column 2 are robust to the inclusion of personality traits in column 3. Given the small proportion of people quitting their job from one year to another (about 3% on average in the overall period), these effects are quite large: doubling the income gap increases the quit probability by just under two percentage points.<sup>11</sup>

<sup>&</sup>lt;sup>10</sup> These are extraversion, agreeableness, openness, neuroticism, and conscientiousness. They are available in the SOEP dataset in three of the five waves we use in the paper, namely 2005, 2009 and 2013. We assume that personality traits are stable over a two-year period and impute the values for the 2007 and 2011 waves from the observations in the preceding waves.

<sup>&</sup>lt;sup>11</sup> In column 3,  $0.024*\ln(2) = 0.017$ .

	(1)	(2)	(3)
Net individual income (ln)	-0.016****	-0.011***	-0.011***
	(0.003)	(0.003)	(0.003)
Income gap (ln)		$0.026^{***}$	$0.024^{***}$
		(0.006)	(0.006)
Controls	Yes	Yes	Yes
Wave dummies	Yes	Yes	Yes
Big 5 Personality traits	No	No	Yes
N	30960	30960	28655
R-squared overall	0.02	0.02	0.02

#### Table 5 – Probability of quitting job by t+1

<u>Notes</u>: Income gap = ln(fair income) - ln(income). These are cross-section linear probability models. Standard errors in parentheses are clustered at the region\*year level. Controls: gender, age dummies, educational level, marital status, hours work, East, regional unemployment rate, number of children, health, firm size, industry dummies, occupation dummies, and year dummies. \*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10.

#### 5. Subgroup analysis

In this last section we consider heterogeneity in the effect of fairness considerations on well-being (life satisfaction, job satisfaction and anger) and quits with respect to two individual characteristics: gender and income. We consider only the affect of anger as this was the only that was significantly influenced by fairness considerations.

A large literature has documented a gender gap in income. We might therefore expect women to perceive their income as less fair than do men, although our summary statistics actually reveal no difference in the percentage of males and females perceiving their income as unfair. To see whether unfairness affects men and women differently, we introduce an interaction between gender and the income gap into our regressions: Table 6 displays the results. This interaction term attracts a large and significant coefficient in the life-satisfaction regression. The coefficient is negative in sign: men's well-being is more affected by unfairness than is that of women. This may reflect gender differences in preferences over competition, with women perhaps reacting less to competitive environments than do men (see Gneezy *et al.*, 2003, and Croson and Gneezy, 2009, for example). There is no significant gender difference for the other well-being variables, nor for job quits (either with an interaction term in column 4, or in separate regressions in columns 5 and 6).

The second characteristic is income. The emotional and behavioural consequences of fairness may well differ along the income distribution. Shaw and Gupta (2001) show, for example, that individuals in financial need, defined as those with children, married, and without alternative sources of income, are less likely to quit their jobs because they are economically dependent, but they suffer from greater dissatisfaction and depression. We here ask whether those in the bottom half of the income distribution have less economic

independence than those in the upper half.<sup>12</sup> The results including an interaction term between the income gap and a dummy for being poor are shown in Table 7.

	Life satisfaction	Job satisfaction	Anger	Job quit	Job quit	Job quit
			-	-	Men	Women
Net individual income (ln)	$0.105^{**}$	0.365***	0.009	-0.011***	-0.017***	-0.011***
	(0.042)	(0.053)	(0.025)	(0.003)	(0.005)	(0.004)
Income gap (ln)	-0.077	-0.652***	$0.102^{***}$	0.021***	$0.029^{***}$	$0.021^{***}$
	(0.049)	(0.090)	(0.035)	(0.007)	(0.007)	(0.007)
Income gap*Male	-0.184**	-0.107	0.033	0.010		
	(0.071)	(0.121)	(0.062)	(0.008)		
Controls	Yes	Yes	Yes	Yes	Yes	Yes
Wave dummies	Yes	Yes	Yes	Yes	Yes	Yes
Individual fixed-effects	Yes	Yes	Yes	No	No	No
Ν	36746	36746	30596	30960	15921	15039
R-squared overall	0.19	0.06	0.03	0.02	0.02	0.02

Table 6 – Fair income and gender

<u>Notes</u>: \*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Controls: age dummies, educational level, marital status, hours work, East Germany, regional unemployment rate, number of children, health, and firm size.

We find a significant effect of the interaction term for well-being. In contrast with Shaw and Gupta (2001), both the life satisfaction and anger of those in the bottom half of the income distribution are less affected by income unfairness.<sup>13</sup> We find no significant difference in the effect of unfairness by income for job satisfaction and job quits.

Net individual income (ln) Income gap (ln)	0.077 <sup>*</sup> (0.042) -0.310 <sup>***</sup>	0.340 <sup>***</sup> (0.059)	0.006	-0.008**	Poor -0.009*	Non Poor
Income gap (ln)		(0.059)			-0.009	-0.006
Income gan (ln)	-0 310 <sup>***</sup>	(0.00))	(0.026)	(0.004)	(0.005)	(0.005)
meome sup (m)	-0.510	-0.733****	0.213***	0.017**	$0.028^{***}$	0.018**
	(0.061)	(0.107)	(0.054)	(0.008)	(0.007)	(0.008)
Income gap*poor	0.223***	0.049	-0.137**	0.011		
	(0.065)	(0.112)	(0.057)	(0.010)		
Poor (1 if Income <median)< td=""><td>-0.099***</td><td>-0.072</td><td>0.010</td><td><math>0.006^{**}</math></td><td></td><td></td></median)<>	-0.099***	-0.072	0.010	$0.006^{**}$		
	(0.031)	(0.049)	(0.023)	(0.003)		
Controls	Yes	Yes	Yes	Yes	Yes	Yes
Wave dummies	Yes	Yes	Yes	Yes	Yes	Yes
Individual fixed-effect	Yes	Yes	Yes	No	No	No
N	36746	36746	30596	30960	14909	16051
R-squared overall	0.18	0.06	0.03	0.02	0.02	0.02

<u>Notes</u>: \*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Controls: gender, age dummies, educational level, marital status, hours work, East Germany, regional unemployment rate, number of children, health, and firm size.

<sup>&</sup>lt;sup>12</sup> Using an index of financial need based on single-parent household, level of savings and financial worries did not produce significant results.

<sup>&</sup>lt;sup>13</sup> In Schneider (2012), fairness perceptions are measured by comparing what respondents say they think individuals in a certain number of occupations earn to what the same respondents say that these individuals should earn. This produces an individual measure of the fairness of the income distribution. Schneider finds that more unfairness is associated with lower life satisfaction, and that this correlation is stronger for those with higher income.

#### 6. Conclusions

This paper has considered how income is related to labour-market outcomes in largescale panel survey data. Our results suggest that income on its own is not a sufficient statistic to predict well-being or behaviour. As has been emphasised in previous experimental work, there is in particular an independent role for fairness. The gap between what individuals receive and what they believe would be a fair income systematically predicts their job and life satisfaction in panel regressions; it also predicts measures of positive and negative affect (and particularly anger). In terms of effect size, fairness considerations matter more for subjective well-being than does absolute income.

Moving onto objective outcomes, workers are more likely to quit their job if they perceive their income as unfair, conditional on the level of income received. Workers are again at least as sensitive to fairness considerations as to absolute income levels in terms of their objective future behaviour on the labour market.

Fairness then drives both well-being and behaviour. The complete understanding of this phenomenon requires knowledge of where fairness evaluations come from. It is tempting to consider the latter as being at least partly influenced by general movements in income inequality or macroeconomic conditions. However, with comparisons being in the majority to work colleagues, some part of fairness concerns can likely directly be affected by the firm. In this context, it might be useful to think about wage secrecy: Are fairness concerns harmed by the provision of information on the actual structure of pay (as in Card *et al.*, 2012), or does this provision rather actually correct erroneous perceptions? The relationship between the actual distribution of income and what individuals believe it to be<sup>14</sup> will likely continue to be an area of continuing interest for academic research.

<sup>&</sup>lt;sup>14</sup> See the OECD's Compare Your Income tool (<u>http://www.compareyourincome.org/</u>) in this context.

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	Mean	Std. Dev.	Min	Max
Gender (male)	0.50	0.50	0	1
Age (Ref.: <= 35)				
36-45	0.29	0.46	0	1
46-55	0.32	0.47	0	1
> 55	0.17	0.38	0	1
Education level (Ref.: < High school)				
= High school	0.63	0.48	0	1
> High school	0.29	0.45	0	1
Marital status (Ref.: Married)				
Single	0.21	0.41	0	1
Widowed	0.02	0.13	0	1
Divorced/Separated	0.12	0.33	0	1
Individual Monthly income (ln)	7.24	0.7	1	11
Income gap (ln)	0.37	0.29	0	3
Income is Unfair	0.37	0.48	0	1
Health status	3.52	0.85	1	5
No. Children	0.60	0.90	0	9
Annual hours worked/100	19.43	7.22	0	67
East Germany	0.23	0.42	0	1
Firm size (Ref.: < 20)				
20-200	0.29	0.46	0	1
200-2000	0.22	0.42	0	1
> 2000	0.25	0.43	0	1
Regional unemployment rate	7.85	3.77	3	21
Job quit	0.03	0.16	0	1
Life satisfaction	7.16	1.60	0	10
Job satisfaction	6.99	1.98	0	10
Нарру	3.57	0.81	1	5
Sad	2.30	0.99	1	5
Angry	2.91	0.96	1	5
Worried	1.88	0.9	1	5
Big 5 Personality traits				
Extraversion	14.52	3.39	3	21
Agreeableness	16.06	2.88	3	21
Conscientiousnes	14.49	1.81	3	21
Neuroticism	11.34	3.60	3	21
Openness	13.44	3.48	3	21

## Appendix 1A - Descriptive statistics

	Life satisfaction		Job sati	sfaction
-	(1)	(2)	(3)	(4)
Net individual income (ln)	0.132***	0.106**	0.477***	0.366***
	(0.040)	(0.042)	(0.051)	(0.053)
Income gap (ln)	()	-0.163***	()	-0.702****
		(0.037)		(0.065)
Age (Ref.: <= 35)	_	-	_	(0.000)
Age 36-45	-0.034	-0.034	$0.107^{**}$	$0.108^{**}$
1160 50 15	(0.037)	(0.037)	(0.047)	(0.047)
Age 46-55	-0.039	-0.037	0.160**	0.169**
Age 40-55	(0.046)	(0.046)	(0.065)	(0.064)
A co > 55	0.029	0.030	0.155*	0.159*
Age > 55				
	(0.055)	(0.055)	(0.084)	(0.083)
Education level (Ref.: <high school)<="" td=""><td>-</td><td>-</td><td>-</td><td>-</td></high>	-	-	-	-
=High school	0.003	0.003	-1.406***	-1.408***
	(0.321)	(0.319)	(0.321)	(0.311)
>High school	-0.122	-0.106	-1.350***	-1.285***
	(0.336)	(0.334)	(0.333)	(0.329)
Marital status (Ref.: Married)	-	-	-	-
Single	-0.025	-0.025	-0.033	-0.033
	(0.037)	(0.037)	(0.063)	(0.062)
Widowed	-0.498**	-0.508***	$0.314^{*}$	0.271
	(0.193)	(0.192)	(0.185)	(0.182)
Divorced/Separated	-0.112***	-0.110***	0.003	0.010
Ĩ		(0.043)	(0.059)	(0.060)
Health status	(0.043) 0.449 <sup>***</sup>	0.447***	0.398***	0.391***
	(0.012)	(0.012)	(0.017)	(0.017)
No. hours worked/100	-0.002	-0.002	-0.017***	-0.014***
ito. nouis worked, 100	(0.002)	(0.002)	(0.002)	(0.003)
East Germany	-0.157	-0.154	-0.427**	-0.415**
Last Germany	(0.140)	(0.140)	(0.169)	(0.165)
Decienci un ampleument rate	-0.037***	-0.038***		
Regional unemployment rate			-0.011	-0.012
Na ahildaan	(0.006)	(0.006)	(0.009)	(0.009)
No. children	0.011	0.011	-0.022	-0.020
	(0.015)	(0.015)	(0.021)	(0.020)
Firm size (Ref.: < 20)	-	-	-	-
20 - 200	$0.122^{***}$	0.124***	0.181***	$0.190^{***}$
	(0.031)	(0.031)	(0.050)	(0.050)
200 - 2000	0.134***	0.137***	0.151**	0.165***
	(0.040)	(0.040)	(0.060)	(0.060)
> 2000	0.157***	$0.158^{***}$	0.191***	0.198***
	(0.034)	(0.035)	(0.061)	(0.060)
Constant	4.531***	4.715***	$4.509^{***}$	5.298 <sup>***</sup>
	(0.462)	(0.464)	(0.601)	(0.586)
Wave dummies	Yes	Yes	Yes	Yes
Industry dummies	Yes	Yes	Yes	Yes
Occupation dummies	Yes	Yes	Yes	Yes
N	36746	36746	36746	36746
R-squared overall	0.18	0.18	0.05	0.06

## Table 2A – Effect of fair income on life and job satisfaction. Full table.

<u>Notes</u>: Linear models with fixed effects.\*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Source: SOEP, waves 2005, 2007, 2009, 2011 and 2013.

	Positiv	e affects	Negativ	e affects
	(1)	(2)	(3)	(4)
Net individual income (ln)	0.007	0.007	-0.018	-0.011
	(0.020)	(0.021)	(0.016)	(0.016)
Income gap (ln)	(0.0_0)	-0.000	(0.010)	0.039*
sure on e gap ()		(0.019)		(0.020)
Age (Ref.: <= 35)	-	-	_	-
Age 36-45	-0.033	-0.033	0.004	0.004
1160 00 10	(0.026)	(0.026)	(0.019)	(0.018)
Age 46-55	0.006	0.006	0.012	0.011
1150 10 55	(0.030)	(0.030)	(0.027)	(0.027)
Age > 55	0.008	0.008	0.033	0.033
Ngc > 55	(0.038)	(0.038)	(0.030)	(0.030)
Education level (Ref.: <high school)<="" td=""><td>(0.038)</td><td>(0.058)</td><td>(0.050)</td><td>(0.050)</td></high>	(0.038)	(0.058)	(0.050)	(0.050)
= High school	0.277	0.277	-0.037	-0.036
- 111511 501001	(0.278)	(0.278)	(0.153)	(0.152)
> High school	0.245	0.245	-0.095	-0.099
> mgn school				
Marital status (Ref.: Married)	(0.256)	(0.256)	(0.168)	(0.168)
	-0.011	-0.011	-0.039	-0.039
Single				
XX7' 1 1	(0.024) -0.411 <sup>***</sup>	(0.024) -0.411 <sup>***</sup>	(0.025) 0.439 <sup>***</sup>	(0.025)
Widowed				0.442***
$\mathbf{D}$ : $1/\mathbf{C}$ (1)	(0.110)	(0.110)	(0.077)	(0.077)
Divorced/Separated	0.015	0.015	0.064**	0.064**
TT 1.1	(0.034)	(0.034)	(0.029)	(0.029)
Health status	0.146***	0.146***	-0.176***	-0.176***
N. 1 1/100	(0.007)	(0.007)	(0.006)	(0.006)
No. hours worked/100	0.000	0.000	0.002*	0.002*
	(0.001)	(0.001)	(0.001)	(0.001)
East Germany	-0.039	-0.039	0.077	0.075
	(0.067)	(0.067)	(0.062)	(0.062)
Regional unemployment rate	0.001	0.001	0.016***	0.016***
	(0.004)	(0.004)	(0.003)	(0.003)
No. children	-0.004	-0.004	-0.031**	-0.031***
	(0.012)	(0.012)	(0.012)	(0.012)
Firm size (Ref.: < 20)	-	-	-	-
20 - 200	0.013	0.013	-0.003	-0.004
	(0.023)	(0.023)	(0.019)	(0.019)
200 - 2000	-0.000	-0.000	-0.025	-0.026
	(0.024)	(0.024)	(0.020)	(0.020)
> 2000	-0.037	-0.037	-0.012	-0.012
	(0.026)	(0.026)	(0.021)	(0.021)
Constant	2.655***	2.655***	3.154***	3.102***
	(0.302)	(0.303)	(0.260)	(0.255)
Wave dummies	Yes	Yes	Yes	Yes
Industry dummies	Yes	Yes	Yes	Yes
Occupation dummies	Yes	Yes	Yes	Yes
N	30596	30596	30596	30596
R-squared overall	0.06	0.06	0.07	0.07

## Table 3A – Effect of fair income on emotional well-being. Full table.

<u>Notes</u>: Linear models with fixed effects.\*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Source: SOEP, waves 2007, 2009, 2011 and 2013.

Table 4A – Effect of fair income on positive	and negative affect
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	Нарру	Sad	Angry	Worried
Net individual income (ln)	0.007	0.010	0.009	-0.052***
	(0.021)	(0.027)	(0.025)	(0.019)
Income gap (ln)	-0.000	0.003	$0.117^{***}$	-0.003
	(0.019)	(0.030)	(0.028)	(0.028)
Controls	Yes	Yes	Yes	Yes
Wave dummies	Yes	Yes	Yes	Yes
Individual fixed-effects	Yes	Yes	Yes	Yes
Ν	30596	30596	30596	30596
R-squared overall	0.06	0.05	0.03	0.04

<u>Notes</u>: Linear models with fixed effects.\*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Standard errors in parentheses are clustered at the region\*year level. Additional controls: age, marital status, education level, number of children, health status, hours worked, regional unemployment rate, firm size, industry dummies, occupation dummies, and individual fixed-effects. Source: SOEP, waves 2007, 2009, 2011 and 2013.

	(1)	(2)	(3)
Net individual income (ln)	-0.016***	-0.011***	-0.011***
	(0.003)	(0.003)	(0.0031)
Income gap (ln)		0.026***	$0.024^{***}$
		(0.006)	(0.006)
Male	$0.004^{*}$	0.003	$0.004^{*}$
	(0.002)	(0.002)	(0.002)
Age (Ref.: <= 35)			
Age 36-45	-0.021***	-0.021***	-0.019***
-	(0.003)	(0.003)	(0.004)
Age 46-55	-0.030****	-0.030****	-0.028***
	(0.003)	(0.003)	(0.004)
Age > 55	-0.039***	-0.039***	-0.038***
C	(0.003)	(0.003)	(0.004)
Education level (Ref.: < High school)	-	-	-
= High school	-0.000	0.000	0.000
	(0.003)	(0.003)	(0.004)
> High school	0.012***	0.011***	0.011***
	(0.004)	(0.004)	(0.004)
Marital status (Ref.: Married)	(0.001)	-	(0.001)
Single	0.003	0.003	0.003
Single	(0.003)	(0.003)	(0.004)
Widowed	-0.001	-0.001	0.001
Wide wed	(0.005)	(0.005)	(0.005)
Divorced/Separated	0.009***	0.009***	0.009***
Divorced/Separated	(0.003)	(0.003)	(0.003)
Health status	0.003**	0.003***	0.003**
icalui status	(0.001)	(0.001)	(0.001)
No. hours worked/100	0.001**	0.000	0.000
vo. nours worked/100	(0.000)	(0.000)	(0.000)
East Germany	-0.009**	-0.010***	-0.010***
	(0.003)	(0.003)	(0.004)
Regional unemployment rate	0.000	0.000	0.000
Regional unemployment fate	(0.000)		
No. children	0.002	(0.000) 0.001	(0.000) 0.001
No. cilitaren			
$\overline{C}$	(0.001)	(0.001)	(0.001)
Firm size (Ref.: < 20)	-0.007***	-	-
20 - 200		-0.008****	-0.008***
200 2000	(0.003)	(0.003)	(0.003)
200 - 2000	-0.010****	-0.010****	-0.009***
2000	(0.003)	(0.003)	(0.003)
> 2000	-0.009***	-0.009****	-0.009****
_	(0.003)	(0.003)	(0.003)
Constant	0.082***	0.049**	0.029
	(0.023)	(0.024)	(0.024)
Wave dummies	Yes	Yes	Yes
industry dummies	Yes	Yes	Yes
Occupation dummies	Yes	Yes	Yes
Big 5 Personality Traits	No	No	Yes
V	30960	30960	28655
R-squared overall	0.02	0.02	0.02

#### Table 5A - Probability of quitting job by t+1. Full Table

<u>Notes</u>: Linear probability models. Standard errors in parenthesis are clustered at the region\*year level. Controls: gender, age dummies, educational level, marital status, no. hours work, east, regional unemployment rate, no. children, health, firm size, industry dummies, occupation dummies, year effects. \*\*\*=p<0.01; \*\*=p<0.05; \*=p<0.10. Source: SOEP, waves 2005-2014.