



## Does high involvement management improve worker wellbeing?☆

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### ABSTRACT

Employees exposed to high involvement management (HIM) practices have higher subjective wellbeing, fewer accidents but more short absence spells than “like” employees not exposed to HIM. These results are robust to extensive work, wage and sickness absence history controls. We highlight the possibility of higher short-term absence in the presence of HIM because it is more demanding than standard production and because multi-skilled HIM workers cover for one another’s short absences thus reducing the cost of replacement labour faced by the employer. We find direct empirical support for this. In accordance with the theoretical framework we find also that long-term absences are independent of exposure to HIM, which is consistent with long-term absences entailing replacement labour costs and with short absences having a negative effect on longer absences.

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## 1. Introduction

What people do affects how they feel at the time and how they subsequently evaluate themselves and their life more generally (Kahneman and Krueger, 2006). What happens at work matters partly because working individuals spend so much of their time at work, but also because it is salient in the way they think about themselves and the value they attach to their lives. This is borne out in empirical research. For instance, studies focusing on reflexive wellbeing indicate that job satisfaction is strongly positively associated with life satisfaction, even after controlling for satisfaction with other aspects of one’s life (Rice et al., 1980). Job satisfaction is also strongly associated with better mental health measured in a variety of ways (Warr, 2007; Llana-Nozal, 2009). However, recent research paints a more nuanced picture. Day reconstruction

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method (DRM) studies show that time spent with one's supervisor is often among the most stressful and least enjoyable parts of the day (Kahneman et al., 2004). So paid employment can be both good and bad for wellbeing. The type of work one undertakes also appears to be important. Thus, although moving into employment from non-employment is usually associated with improvements in mental health, the gains to entering non-standard employment contracts are often much lower (Llena-Nozal, 2009).

Standard models assume that employers make adjustments to the production process to maximise profits, rather than employee wellbeing. Consistent with this, there is empirical evidence that management practices will be adopted if their productivity benefits exceed the costs of introducing and maintaining them (Bloom and Van Reenen, 2007) and that firms will switch management practices – even if they are productivity enhancing – if the costs outweigh the benefits (Freeman and Kleiner, 2005). However, the way jobs are designed can also have a profound impact on workers' mental and physical wellbeing (Wood, 2008; Pouliakas and Theodoropoulos, in press). There is also evidence that happier workers are more productive at work (Oswald et al., 2009; Böckerman and Ilmakunnas, 2012). It does not follow, however, that employers will invest to maximise the wellbeing of their workers since such investments are themselves costly.

In recent decades many employers have introduced practices designed to maximise employees' sense of involvement with their work, and their commitment to the wider organisation, in the expectation that this will improve their organisation's performance. Although there is a good deal of debate as to the precise set of practices that are deemed "high involvement practices", core components include teams, problem-solving groups, information sharing, incentive pay, and supportive practices such as training and associated recruitment methods (Wood and Bryson, 2009). Collectively they constitute "high involvement management" (HIM). Rarely do analysts believe single practices constitute the presence of HIM. Rather, it is "bundles" of practices – often incorporating greater autonomy or control and greater performance-based pay – which analysts believe can help transform the working environment (Ichniowski et al., 1997; Ichniowski and Shaw, 2009; Shaw, 2009). A sizeable literature explores the links between these practices and firm performance (for a review see Bloom and Van Reenen, 2011), but far less is known about the effects of HIM on employees' health and other measures of wellbeing. The investigation of links between HIM and worker wellbeing is timely because HIM has become increasingly common in developed industrialised economies (Wood and Bryson, 2009) while, at the same time and perhaps coincidentally, there are indications of a decline in worker wellbeing (Oswald, 2010; Green, 2006, 2009). *A priori*, it is uncertain what impact HIM is likely to have on employee wellbeing. On the one hand, if HIM enriches employees' working lives by offering them greater job autonomy, more mental stimulation, team-based social interaction, and a heightened sense of achievement, this may improve worker wellbeing. On the other hand, if HIM is simply a means of intensifying worker effort, this may lead to a higher incidence of illness, injury, absence and stress.

In this paper, we explore the impact of HIM practices on worker wellbeing using an innovative combination of survey and register data. The use of linked data is a methodological advance over the existing studies. The key problem in previous research is that workers are not randomly assigned into HIM. This may bias the estimates of HIM on employee wellbeing considerably. If workers with 'good' work histories are more likely to be found in HIM jobs, the estimates of HIM on employee wellbeing are upwardly biased.<sup>1</sup> The size of this bias is not known. We tackle the problem caused by sorting of employees into HIM status by controlling for a particularly rich set of employees' work and sickness absence histories. This provides us with a better identification strategy than the ones that have been used previously.

Using linked data we contribute to the literature in five ways. Firstly, we establish whether healthier workers sort into jobs that involve using high involvement practices, as one might expect if HIM jobs demand more of workers than non-HIM jobs. We do so by linking register data on Finnish workers' absence histories to a nationally representative survey in which employees identify which, if any, high involvement practices they are exposed to in their jobs. Secondly, we estimate the impact of HIM practices on employee wellbeing having controlled for worker sorting into HIM jobs by conditioning on sickness absence histories and work and wage histories. Data limitations mean this has not been possible in the literature until now. Thirdly, we present theoretical arguments on why higher short-term absences in the presence of HIM are consistent with no association between HIM and long-term absences. Fourthly, unlike most of the literature that tends to focus on specific aspects of worker wellbeing we explore HIM effects across a broad range of wellbeing measures. Specifically, we estimate the effects of HIM on three types of wellbeing measure, namely sickness absences, both short-term and long-term; subjective wellbeing (job satisfaction, work capacity, the state of one's health, and feelings of tiredness); and physical discomfort at work, as measured by the experience of pain in four different parts of the body (lumbar, legs, arms and neck). Finally, we estimate the empirical models for a complete set of different "bundles" of HIM practices.

The remainder of the paper is structured as follows. Section 2 reviews the theoretical and empirical literatures linking HIM to employees' wellbeing. Section 3 introduces the data. Section 4 reports our results and Section 5 concludes.

## 2. Theoretical and empirical literatures

Since the early 1980s management theorists and practitioners have advocated innovations in job design expressly intended to elicit greater labour productivity via greater employee involvement (Beer et al., 1984, 1985; Walton, 1985).

<sup>1</sup> We use the adjective "good" to refer to work histories exhibiting high and/or rising wages and stable employment with few unemployment and sickness absence spells.

Scholars in the Harvard Business School tradition identify human resources as a key asset in value production and maintain that firms can gain a hard-to-replicate competitive advantage over rivals through investment in management practices which devolve responsibilities to employees in the organisation of work (Walton, 1987; Pfeffer, 1998). The shift towards job autonomy is often perceived as a move away from the deskilling imperatives associated with Taylorist principles of hierarchical work organisation towards job enrichment and “high commitment”. In return, employers might expect improved labour productivity through increased worker effort or through “smarter” working arising from employees’ increased opportunities to utilize tacit knowledge about efficient working which would not have been sought in a more hierarchically structured organisation.

### 2.1. Theory

One might assume that if HIM entails job enrichment it might improve worker wellbeing by increasing worker control over job tasks, increasing mental stimulation, providing greater opportunities for social interaction via team-working, and via a greater sense of achievement at work. However, demanding more of workers through the introduction of high involvement management practices may also have negative effects on employees’ subjective wellbeing. According to Karasek (1979) workers’ mental and emotional wellbeing is negatively related to job demands and positively related to job control. Increases in both demands and control are implied by a shift to HIM.

Even if HIM enhances job control, the *process* of HIM introduction can generate uncertainty leading to increased anxiety among workers, in much the same way as other processes of change. These effects on employee subjective wellbeing are unlikely to persist since those worst affected will choose to leave the organisation while the remainder are liable to adapt over time (Kahneman et al., 1999). Whether HIM innovations will lead to deterioration in employee wellbeing depends, in part, on what Payne (1979) and Karasek and Theorell (1990) term “social supports”. These supports, which might include union representation and consultative management, have the capacity to buffer individuals against the worst effects of workplace innovation.

High involvement management may also affect employee physical wellbeing either positively or negatively for a number of reasons. Since changes in physical health often accompany changes in mental and emotional wellbeing, HIM effects on subjective wellbeing may feed through to changes in physical wellbeing. Where workers have job autonomy they can instigate innovations in work practices which can reduce workers’ exposure to risks of injury and disease. Management can use the review of job tasks and work organisation accompanying the introduction of HIM to “build in” better working conditions for workers resulting in improved physical wellbeing, irrespective of the degree of job autonomy those HIM practices offer workers. Also the training that is integral to so many HIM innovations can raise worker competence thus reducing risks of accidents and injury. On the other hand, if HIM is used as a form of labour intensification it may lead to an increased risk of accidents, job-related pain or injury.

HIM effects on employees’ subjective and physical wellbeing may also affect their absence rates. HIM-induced increases (or reductions) in injury and illness should have a direct bearing on the amount of sickness absence employees take relative to what they would have taken in the absence of HIM. There are other less clear-cut scenarios in which whether a worker chooses to be absent from work is a marginal cost–benefit decision (Allen, 1981; Treble and Barmby, 2011). This choice will turn, in part, on whether HIM is viewed by the employee as an amenity or disamenity. If it is viewed as a disamenity which is not compensated with increased financial rewards – either through base pay or incentive pay – HIM may increase absence taking. However, certain HIM practices can be expected to reduce absenteeism. In the case of incentive pay, loss aversion will encourage workers to attend because absent workers forgo incentive payments (Merriman and Deckop, 2007). Where worker inputs are complementary, as in the case of team-working, workers may come under co-worker pressures to minimise absence, particularly if performance is judged on team outputs (Drago and Wooden, 1992; Kandel and Lazear, 1992; Knez and Simester, 2001; Heywood and Jirjahn, 2004).

HIM may also affect absences through its impact on the firms’ optimising behaviour. One can think of firms choosing an optimal rate of absence. Increasing worker wellbeing is likely to benefit firms, but at a decreasing rate. On the other hand, the marginal cost of decreasing absenteeism can be increasing. Equality of marginal benefits and costs determines the absence rate that is optimal from the point of view of the firm. Since firms differ in terms of production processes, the optimal rate varies across firms. In particular, HIM practices may have a bearing on the optimal absence rate. What Coles et al. (2007) and Coles and Treble (1996) term the “shadow price of absenteeism” may differ in HIM firms and non-HIM firms. In the sort of multitasking environment which predominates in many HIM firms, workers can substitute for one another in the short term without the firm having to bring in additional labour. Therefore, it may be worthwhile paying the additional short run cost of absences if it means that the firm can meet production schedules. Additional tiredness associated with the intensity of HIM production may require short absences to recuperate in order to avoid longer term absences. These arguments suggest likely differences in the impacts of HIM on the length of absences, with HIM employees taking more short absences.

Coles and Treble (1993, 1996) have formed models to explain firms’ demand for absences. In their models this rises from the fact that workers may have true sickness absences but may also be absent for personal reasons. Worker absences cause disruptions in the production process when there is an assembly line technology where workers are complementary and at least a minimum number of workers are required to be present to obtain positive output. Therefore, the firms will use wage policies that reduce absences. However, the Coles and Treble model does not fully fit our purposes. Firstly, the institutional setting in our empirical application is such that the workers get full pay for a relatively long time during sickness absences

(see Appendix 1). Thus, there is little room for absence-reducing wage policies and we can assume that the same wage is paid whether the worker is at work or absent. Secondly, we want to emphasize such work organisation where absences can partly be covered by co-workers and therefore absences do not cause zero production. Rather, we assume a more standard technology where absences lead to a proportional drop in the labour input and the effective labour input has decreasing returns. Therefore, a firm with absences would have lower marginal productivity than a firm without absences. Thirdly, we want to make a distinction between short-term absences, which are easier to cover by the other employees and long-term absences, which are more difficult to cover and which may lead to a need for replacement hiring and are therefore more costly to the firm. We interpret long-term absences as work accidents.

The impact of the HIM practices can be described with their relationship to absence probability and production loss.<sup>2</sup> The purpose of the practices is to increase productivity. However, high work intensity can also increase the probability of short absences  $p_S$ . With multi-tasking team work it may be easier for the other team members to replace the absent workers in the short run, so the proportional production loss from absences  $\alpha_S$  ( $0 \leq \alpha_S \leq 1$ ) would be lower than without team work. The expected share of lost production is  $p_S \alpha_S$ . The firm could then optimize work intensity (and thereby the absence rate): the marginal benefit is the increased production through higher work intensity and the marginal cost is the expected production loss through absences. Correspondingly, the marginal benefit or worker welfare (lower absences) is the reduced production loss and the marginal cost of welfare is the lower production with less intensive work.

The short absences may also be a mechanism for reducing accidents. The probability of accidents is  $p_A$ , the corresponding proportional production loss is  $\alpha_A$  ( $0 \leq \alpha_A \leq 1$ ), and the expected share of lost production is  $p_A \alpha_A$ . Since long-term absences are more difficult to cover by the other workers even with team work,  $\alpha_A$  is likely to be higher than the loss from short absences as the firm may have to hire new workers with less firm-specific skills. It is therefore optimal for HIM and non-HIM firms alike to avoid long-term absences. Although higher work intensity in HIM work may have a direct accident-increasing effect, it is counteracted if the workers can use short absences to recuperate. The accident probability can then be treated as a function  $p_A = p_A(p_S)$ , with  $dp_A/dp_S < 0$ . This would decrease the marginal cost of work intensity, leading to higher work intensity and therefore a higher short-term absence rate in HIM work than in non-HIM work, but not necessarily a higher rate of long-term absences.

## 2.2. Evidence

The evidence on the link between subjective wellbeing and job control and job demands tends to support Karasek's theory. Using linked employer–employee data for Britain Wood (2008) confirms that worker wellbeing is negatively related to job demands and positively related to job control, and that high job controls reduce the negative association between job demands and wellbeing. Studies which examine the effects of specific HIM practices indicate that they are often associated with high levels of work intensity and worker stress (Barker, 1993; Godard, 2001), even when they are also associated with higher work commitment (Ramsay et al., 2000) or higher job control (Gallie, 2005).<sup>3</sup>

The process of innovating can also generate anxiety. In a case study Bordia et al. (2004) link organisational change to psychological stress through perceived loss of control. Pollard (2001) shows that workplace reorganisation caused significant increases in distress and in systolic blood pressure and that uncertainty was a key factor. However, as predicted by theory, social supports can help workers cope with workplace innovation. Using the same linked employer–employee survey as Wood (2008), Bryson et al. (forthcoming) find supportive evidence for the buffering effect of unionisation in ameliorating the negative impacts of workplace innovation on job anxiety.

Using data for the late 1990s Green (2006) shows that task discretion has been declining in most European countries. Green and Tsitsianis (2005) show that in Britain there has also been a decline in job satisfaction which is accounted for by declining task discretion and the intensification of work effort. Rather than being a force for job enrichment, it appears that HIM was introduced over the period as part of a lean production system geared to cost reductions and just-in-time production. There is direct evidence that this is the case in Britain (Wood and Bryson, 2009). Just-in-time production is associated with poorer sick pay provision (Lanfranchi and Treble, 2010), as predicted under Coles and Treble's (1996) model. Taken together, findings from these studies suggest HIM may well be associated with injuries, accidents and higher levels of absenteeism. However, other British studies suggest HIM increases satisfaction at work. Green and Heywood (2008) document that performance pay increases job satisfaction while Jones et al. (2009) report that satisfaction with employer-provided training reduces absenteeism. Also, Poulidakas and Theodoropoulos (forthcoming) find that the British private sector establishments that link their pay with individual performance have significantly lower absence rates.

The evidence for continental Europe is also ambiguous. Askenazy and Caroli (2010) report that in France innovative workplace practices are positively associated with mental strain and with worker perceptions of occupational risks, but not with occupational injury. Heywood and Jirjahn (2004) find absence rates are lower in German manufacturing in the presence of team-working. However, Frick and Simmons's (2010) case study of a large German steel plant supports the contention that HIM increases accidents and absenteeism via labour intensification. In their study the introduction of production bonuses for

<sup>2</sup> The theoretical model is fully presented in the working paper version (<http://www.niesr.ac.uk/pdf/dp380.pdf>).

<sup>3</sup> In his review of the literature Godard (2004) suggests the evidence is more mixed. For instance, there are some studies such as Appelbaum et al. (2000), that find no adverse effects.

teams leads to an increase in both absence rates and the number and severity of accidents. The steel workers face a capped incentive structure allowing them to achieve their maximum bonus without fully utilising labour, thus enabling them to share out leisure time in the form of coordinated absences. Furthermore, incentive payments in the absence of teams result in an increased accident rate which they suggest is evidence of “excessive incentivisation . . . workers work too hard and cause accidents through carelessness and/or fatigue” (2010, p. 14). In the presence of teams, incentive pay is not associated with increased accidents, a finding the authors say is consistent with team members taking care of one another to ensure they make the team bonus. Finally, [Bender et al. \(2012\)](#) relate piece rates to increased workplace injuries with European data.

Empirical evidence for North America is equally ambiguous ([Handel and Levine, 2004](#)). Establishment-level studies for the United States have identified a positive link between managerial innovations and workplace injuries ([Askenazy, 2001](#); [Fairris and Brenner, 2001](#)) and cumulative trauma disorders ([Brenner et al., 2004](#)).<sup>4</sup> On the other hand, using linked employer–employee data for Canada, [Mohr and Zoghi \(2008\)](#) find a robust positive association between HIM and job satisfaction and no association with work-related stress. Using the longitudinal component in their data they find higher job satisfaction predicts increased participation in HIM whereas participation does not predict future satisfaction, a result which raises questions about a causal linkage between HIM and improved worker wellbeing. Furthermore, using similar survey instruments in Canada and England in 1998 and 2003/2004, [Godard \(2010\)](#) reports different relationships between workplace practices and worker subjective wellbeing over time and place, leading him to suggest that the associations “may be historically and institutionally contingent and thus should be interpreted using a historical/institutional perspective” (2010, p. 466).

Our study utilises nationally representative data from Finland, a country with very high rates of unionisation (~70 percent) and a Scandinavian social model which places a much greater emphasis on social dialogue in the workplace than the European and North American countries which account for most of the empirical studies. One might expect Finnish employees to have a greater say in the process of workplace innovation, offering them opportunities to influence the nature of HIM and the way it is introduced and implemented in a manner which may be less common in other settings. In fact, [Green \(2006, p. 103\)](#) notes that, whilst job discretion has been on the decline in many countries, it has been rising in Finland. The Finnish Quality of Working Life Surveys (QWLSs) provide consistent data over time to map changes in work organisation in Finland, and they paint a more nuanced picture. For instance, while employees’ ability to influence the way their own work is organised has increased in most dimensions, perceptions of work intensity have also increased ([Lehto and Sutela, 2009](#)). Furthermore, Finland has the highest sickness absence rate in the European Union ([Gimeno et al., 2004](#)). Finland is thus of particular interest when analysing the effects of work practices on worker wellbeing.

Using the QWLS 2003 – the same survey we use in this paper – [Kalmi and Kauhanen \(2008\)](#) find HIM is negatively correlated with worker stress and positively correlated with both job satisfaction and job security. These associations strengthen with the number of HIM practices to which the employee is exposed. [Böckerman et al. \(2012\)](#) examine the relationship between HIM and sickness absence and accidents using the QWLS 2008. Their results are not so clear cut. Using single equation models, they find that innovative work practices increase short-term sickness absence for blue-collar and lower white-collar employees. In contrast, in two-equation models that treat innovative workplace practices as endogenous variables and control for unobserved correlations between HIM and the wellbeing outcomes they do not find significant relationships between innovative work practices and sickness absence or accidents at work. However, [Böckerman et al. \(2012\)](#) neither condition on absence and work histories, nor do they consider the effects of different “bundles” of HIM.

### 2.3. Sorting of employees

The difficulty in interpreting the results from the studies reviewed above is establishing whether the relationship between HIM and wellbeing outcomes is causal. If HIM jobs are more demanding than other jobs, it is plausible that only healthier employees, or those who are mentally and physically more resilient, will put themselves forward for HIM jobs, or be offered them by HIM employers. Failure to account for selection of healthier workers into HIM jobs will upwardly bias any estimated effect of HIM on worker wellbeing since the wellbeing of HIM workers would have been higher than their non-HIM counterparts even in the absence of HIM. Market frictions mean workers cannot simply choose to shift easily between the HIM and non-HIM sectors so that the sector they work in will not necessarily reflect preferences but it remains a source of potential estimation bias. We address this concern by conditioning on employees’ prior sickness absence. To our knowledge, the only other author to do this is [Llena-Nozal \(2009\)](#) in her study of the effect of labour market transitions on mental health. She finds that failure to account for previous health histories leads to an upward bias in the mental health returns to entering employment.

A further threat to causal interpretation of the link between HIM and employee wellbeing arises from the fact that HIM and non-HIM workers may differ in dimensions other than their health histories which are unobservable to the analyst but which may nevertheless influence their propensity to take HIM jobs and their current state of wellbeing. For example, we do not observe risk preferences, yet those with high risk preferences may be more prepared to take the demanding and responsible work in an HIM job and be more prepared to engage in risky behaviour which adversely affects health. If so,

<sup>4</sup> [Batt \(2004\)](#) argues that the effects on wellbeing may also differ by worker type.

this would induce a negative bias in the effects of HIM on employees' wellbeing. To help overcome this problem we also condition on employees' work and earnings histories which are plausibly highly correlated with unobserved worker traits, thus reducing the potential for omitted variables bias.

Omitted variables bias may also arise due to unobserved differences between HIM and non-HIM jobs. For instance, HIM jobs may simply be 'better' jobs than non-HIM jobs in terms of pay or working conditions, in which case they may generate higher worker wellbeing for reasons that are not strictly due to the amount of employee involvement they entail. For this reason, we test the sensitivity of our results to a full set of job controls including a range of highly detailed job disamenities.

### 3. Data

Our data are the Quality of Working Life Survey (QWLS) 2003 of Statistics Finland (SF). The initial sample for QWLS is derived from a monthly Labour Force Survey (LFS), where a random sample of the working age population is selected for a telephone interview. The QWLS 2003 was based on LFS respondents in October and November who were 15–64-year-old wage and salary earners with a normal weekly working time of at least 5 h. 5270 LFS participants satisfied these conditions and were invited to participate in a personal face-to-face interview for the QWLS. Eventually 4104 persons participated (Lehto and Sutela, 2005) in the interviews (a 77.9 percent response rate), which took place mostly in October–December 2003, with some taking place in the beginning of January 2004.<sup>5</sup> Owing to missing information on some variables for some workers, the sample size used in this study is 3755 observations.

In addition to the HIM practices the worker is exposed to in her employment, the QWLS contains information on the type of job the employee does and the nature of the employer, together with employees' personal characteristics and work experience. SF supplements QWLS with information from the LFS on, for example, working time and exact labour market status, and information on annual earnings from tax registers and on education (level and field) from the register of completed degrees. Supplementary information on the industry and location of the employer is gathered from various other registers maintained by SF.

The QWLS is a cross-section data set that includes only limited self-reported information on past labour market experience. However, we match the QWLS data to comprehensive longitudinal register data. These are the Finnish Longitudinal Employer–Employee Data (FLEED). FLEED is constructed from a number of different registers on individuals and firms that are maintained by Statistics Finland. In particular, FLEED contains information from Employment Statistics, which records each employee's employer during the last week of each year. We match QWLS and FLEED using unique personal identifiers (i.e. ID codes for persons). We can follow the employees backwards over the period 1990–2003. In each year, we can link information on the firm and establishment to each person.

The dependent variables describe different aspects of worker wellbeing. Firstly, we consider sickness absence. The QWLS has information on the number and length of absences during the last 12 months. The questions relating to absences are the following: "How many times have you been absent 1–3 days?"; "How many times have you been absent 4–9 days?"; "How many times have you been absent at least 10 days?"; and regarding the longest absences, "How long were you absent from work? (Add up several absences of over 10 days.)". With this information we can form variables for the total number of absence spells. In addition, we can approximate the total days of absence by using 2 days as the length of the short 1–3 day absences, 6.5 as the length of the 4–9 day absences, and the actual number of days for the long absences.<sup>6</sup> An alternative measure is based on information from the Social Insurance Institution (KELA). This is the number of days for which the worker has obtained sickness allowance from the sickness insurance system. (The details of the Finnish sickness insurance system are described in [Appendix 1](#).) Since there is a waiting period of 10 days until eligibility to the sickness allowance, this measure only includes long absences. On the other hand, the information on the allowance days is available for the whole period 1995–2006, so we can use it both as a control variable for past absence history and as an outcome variable.<sup>7</sup> The likelihood of systematic reporting error is minor, because the KELA data are based on comprehensive registers. Related to the absence variables is an indicator for accidents. QWLS has a question on whether the person has had an accident at work that has resulted in absence from work in the last 12 months.

<sup>5</sup> The response rate of QWLS is not low compared to other surveys. For example, the response rate of 77.9 percent compares favourably to virtually any UK based survey. There has been a general trend that the response rates of the surveys are falling in Finland and elsewhere for various reasons. QWLS 2003 has still a very high response rate for a complex and burdensome face-to-face survey. Lehto and Sutela (2005) provide a detailed analysis of response v. non-response. Their overall assessment is that non-response does not seriously undermine the representativeness of the QWLS data. We have also learned from personal communication with Anna-Maija Lehto (Statistics Finland) that researchers at Statistics Finland have made experiments by calibrating weights to the QWLS to account for non-response in the survey. The use of weights to account for non-response proved to have only a minor effect. This supports the argument that non-response does not seriously undermine the representativeness of the data.

<sup>6</sup> Although the question in the survey refers to days of absence from work, it is not clear whether the respondents think of these days as actual working days, e.g. Monday to Friday, 'official' working days which also include Saturdays but not Sundays or holidays, or as calendar days which may include the whole weekends.

<sup>7</sup> According to the rules of the sickness insurance and labour contracts, the workers get paid during the waiting period, after which there is an earnings-related allowance. The replacement rate declines with earnings. According to many labour contracts the workers are actually paid for a considerably longer period. However, we do not have information on the contracts that the workers in our data belong to, so we cannot tell whether they have received full pay or the allowance during their illness. Therefore, we cannot use the replacement rate as an explanatory variable.

The other wellbeing variables are from the QWLS. The second set captures subjective measures of employee wellbeing. There is a question on job satisfaction measured on a four-point Likert scale from “Very dissatisfied” (coded 1) to “Very satisfied” (coded 4). There is also a question on working capacity: “Assuming that your top working capacity would score 10 points while your total inability to work would score zero, how many points would you give to your working capacity at the moment?”. The state of self-assessed health is measured in the survey with answers on a 5-point scale from “Poor” (coded 1) to “Good” (coded 5). We also have a measure of tiredness from answers to the question: “How often do you feel reluctant or mentally tired on leaving for work?”. The answers range on a 6-point scale from “Daily or almost daily” (coded as 1) to “Never” (coded as 6).

The third set of dependent variables capture pain felt at work. We use answers to the questions on specific conditions to identify whether the person suffers from recurrent aches or pains in (a) neck, cervical spine or shoulders; (b) hands or arms; (c) lumbar region; or (d) legs, including hips. We code these answers so that 1 indicates no pain and 0 indicates pain.

The explanatory variable of interest is HIM. Following Kalmi and Kauhanen (2008) we capture four different aspects of HIM using dummy variables for them. These indicators are *incentive pay* for those who are personally subject to performance-related pay; *training* for employees who have participated in employer-provided training during the past 12 months; *self-managed teams* for individuals who work in a team that selects its own foreman and decides on the internal division of responsibilities; and *information sharing* for employees who are informed about the changes at work at the planning stage rather than shortly before the change or at its implementation.<sup>8</sup> Other empirical studies (e.g. Frick and Simmons, 2010) suggest that the particular combination of HIM practices may determine their effects on worker wellbeing. We therefore construct a categorical variable “any HIM” for being exposed to any (or any combination of several) of the HIM practices, as well as separate indicators for identifying all possible combinations of the four HIM practices to fully establish the effects of different “bundles”.

As control variables, we use indicators for gender, age, marital status, educational level, plant size, multi-plant firms, foreign ownership, public sector employer and a set of 14 single digit industry dummies. All of these variables are based on the data on individuals in QWLS. Furthermore, we have several work and earnings history variables for the period 1990–2001. These include the number of past job switches (defined as a change of establishment), unemployment episodes (both number of episodes and their length in months), past employment months, an indicator for having worked in a large firm (firm with more than 300 employees), past average earnings (1990–2001) and past earnings growth (average over periods 1999–2000 and 2000–2001). The past earnings data are introduced as the log of annual earnings. Earnings include the base wage, overtime pay, bonuses, and wage supplements. All of the above work history variables are from the longitudinal register data (FLEED). In addition, we use information in the QWLS to measure the length of tenure with current employer and to form an indicator for persons who have had more than three different professions over their working life. We also control for employees’ past sickness absence history by using the total number of sickness absence days over the period 1995–2001, as recorded by KELA.

Finally, we have three job disamenity variables, based on the QWLS. For perceived harms, there is a five-point scale in which the highest category corresponds to the perception by a worker that a certain feature of working conditions is ‘very much’ an adverse factor at the workplace. Harms include heat, cold and dust, among other things. For perceived hazards, the highest category among three possibilities is the one in which the respondent considers a certain feature at the workplace as ‘a distinct hazard’. Hazards include accident risk, risk of strain injuries and risk of grave work exhaustion, among other things. For insecurities, the respondents answer whether certain aspects are insecurity factors or not. These aspects include the threat of temporary dismissal and the threat of unemployment. Responses to the questions about adverse working conditions are aggregated by forming a dummy variable that equals one if there is at least one clearly adverse factor (Harm), a dummy that equals one if there is at least one distinct hazard (Hazard), and a dummy if there is at least one insecurity factor (Uncertainty).<sup>9</sup>

Descriptive statistics for all the variables used in the analysis are presented in Appendix 2 (Table A.1). In Table A.2 we document the incidence of HIM bundles along with ‘any HIM’. Discussion of the estimation results is focused on the bundles that are common enough to support robust conclusions. We report the estimates for all the HIM bundles for the sake of completeness. Table A.3 reports the associations for the HIM bundles along with ‘any HIM’. We report Kendall’s tau-*b* correlation coefficients because all the HIM variables are 0/1-variables. These associations are for the most part statistically significant, but small. HIM bundles are positively correlated with ‘any HIM’ but negatively correlated with each other, as expected.

<sup>8</sup> Kauhanen (2009) provides a detailed descriptive account and discussion on the distribution of innovative workplace practices among different types of workers in Finland by using the 2003 QWLS. It is possible that employees exposed to HIM in 2003 had already earlier been in HIM jobs. However, there has been a rapid expansion of HIM practices in Finland during the past 15 years, so the probability of being in a HIM job in 2003 is much higher than in the earlier years.

<sup>9</sup> A full description of these variables is available in Böckerman and Ilmakunnas (2008).

## 4. Estimation results

### 4.1. Basic results

As discussed earlier, those taking HIM jobs may be healthier than other workers if HIM jobs are more demanding than non-HIM jobs and, recognising this, employers select and allocate workers accordingly. To examine this proposition we first established whether absence histories are related to current self-assessed working capacity. It was indeed the case that absence history had a significant negative relationship with working capacity.<sup>10</sup> Next we examined the relationship between previous absences and current HIM status. The results from probit estimations revealed that sickness absence history over the period 1995–2002 was not related to current exposure to any of the four HIM practices in our data.<sup>11</sup>

Next we turn to the relationship between HIM practices and sickness absences. Table 1 (Panel A) shows the average marginal effects from a probit model for having any absences, where the dependent variable is based on absences reported in QWLS 2003. The columns refer to different control sets, starting from the baseline model that includes only an indicator for being exposed to any HIM practice (v. none), and then successively adding sickness absence history, employment history, personal and firm characteristics, and finally the three measures of job disamenities.<sup>12</sup> This allows us to test for the significance of sickness absence histories as we load in more information to the models. It is particularly useful to explore the sensitivity of the results to the inclusion of job disamenities in the final column because it is possible that these variables are picking up stressful outcomes that we measure via sickness absence, at least to some degree. The tables only report the average marginal effects for the HIM variables.

HIM practices are associated with a 5 percent increase in the probability of having a sickness absence spell (Table 1, Panel A, columns 1–3). However, when demographic and employer characteristics are included the estimate drops and loses significance. This remains the case when we add job disamenities in Column 5 that constitutes our preferred estimate.

In Panel B, we use the total number of absence spells as the dependent variable. Since this is a count, but has a very high concentration of zeros, we use zero inflated negative binomial models in the estimation. Overdispersion tests clearly showed that the overdispersion parameter differed statistically significantly from zero and Vuong tests revealed that we have to use zero inflated models.<sup>13</sup> The coefficient for any HIM practice is positive and statistically significant across all model specifications. However, Panels C and D reveal that the effect is driven by the total number of short absence spells (spells that lasted less than 4 days). HIM is positively correlated with the number of short absence spells in all specifications, but it remains negative and non-significant for long absence spells (spells that lasted 4 days or more).<sup>14</sup> This is consistent with the possible effects of HIM practices discussed in Section 2 above.<sup>15</sup>

We also estimated models for the number of days of absence reported by KELA, a probit for having any absences, and a zero inflated negative binomial model for the number of absence days in 2003. The probit showed positive, but non-significant marginal effects for HIM, while in the negative binomial model the marginal effects were negative and remained statistically non-significant throughout as we loaded in more controls (results not reported). Since all KELA absences are relatively long

<sup>10</sup> We estimated this relationship using an ordered probit model containing the controls for demographic characteristics used throughout the paper. The marginal effect for the probability of the top category of working capacity was  $-0.0011$  with a robust standard error of  $0.00016$ . This implies that one standard deviation change in absence history reduces the probability of being in the top category by 4 percent. The economic significance of this effect is therefore notable.

<sup>11</sup> Sickness absence histories were statistically non-significant in probits explaining any HIM practice v. none, any profit related pay (i.e. in any combination with other practices) v. no HIM, any training v. no HIM, any self-managed teams v. no HIM, and any information sharing v. no HIM. In the probit specification in which we explained the exposure to 'any HIM v none' the variable for sickness absence history obtained a marginal effect  $-0.000309$  with robust standard error  $0.000161$ . Thus, the point estimate was very close to zero and the non-significance was clearly not being driven by a particularly large standard error. This was also the case with the other HIM variables. We also investigated the impact of very recent sickness absence history (over the period 2000–2002) before QWLS 2003, because it is arguably more easily observable to the current employer and thus could have a larger impact on the allocation of workers into various tasks. The recent absence history was not statistically significant in any of the models. For the recent sickness absence history (2000–2002) the marginal effect was  $-0.0001615$  with robust standard error  $0.0003367$  using the main specification of 'any HIM v none'. Furthermore, we excluded the employment history variables from the set of controls, because employment and sickness absence history may be closely related. This did not have any significant effect on the estimates. However, work history is an important determinant of current HIM status. Those employees with 'good' work histories are much more likely to be found in HIM jobs. These effects are fully explored in our related research.

<sup>12</sup> The individuals exposed to HIM and those not exposed are not strikingly dissimilar in terms of their observable characteristics. We used propensity score matching and found that the number of observations out of common support is very low. In the baseline specifications that use 'any HIM v none' between 0.2 and 1.0 percent of employees are lost through the enforcement of common support. This implies that the estimation of specifications conditional on common support produces very similar results.

<sup>13</sup> Under overdispersion the estimated parameters from a Poisson model are consistent, but the standard errors are not. Using the zero inflated Poisson model with sandwich standard errors gave results that were fairly close to those from the zero inflated negative binomial model.

<sup>14</sup> Table A.4 documents the full estimation results for the specification in Column 5 of Panel B of Table 1. Absence history over the period 1995–2001 is a strong predictor of the probability of absence in 2003. The results for the standard control variables are very similar compared to the ones in a study using the QWLS 1997 (Böckerman and Ilmakunnas, 2008). There is evidence that females have more absences and older employees have fewer absence spells although they have longer absences measured by absence days (see also Ilmakunnas et al., 2010, for age effects on absenteeism in the QWLS data). We also find that the role of working conditions is pronounced.

<sup>15</sup> We obtained evidence to support this also by estimating a model for the total number of absence days, conditionally on a positive number of days. One can think of the probit model and this model for positive days as a two-part model. To guarantee that the predictions are positive, the model was estimated using the logarithm of the number of days as the dependent variable. The HIM coefficient was negative, which is consistent with HIM increasing short absences.



**Table 1**  
Innovative workplace practices as determinants of sickness absence (QWLS).

	Baseline	Sickness absence history	Sickness absence history + work history	Sickness absence history + work history + controls	Sickness absence history + work history + controls + job disamenities
<b>Panel A: Any absence, probit</b>					
Any HIM v none	0.0455** (0.0191)	0.0464** (0.0191)	0.0466** (0.0193)	0.0251 (0.0195)	0.0316 (0.0195)
<b>Panel B: Total number of absence spells, zero infl. negative binomial regression</b>					
Any HIM v none	0.1375* (0.0694)	0.1443* (0.0691)	0.1581** (0.0680)	0.1238* (0.0687)	0.1483** (0.0668)
<b>Panel C: Total number of long absence spells, zero infl. negative binomial regression</b>					
Any HIM v none	-0.0451 (0.0370)	-0.0357 (0.0371)	-0.0407 (0.0363)	-0.0024 (0.0314)	0.0174 (0.0264)
<b>Panel D: Total number of short absence spells, zero infl. negative binomial regression</b>					
Any HIM v none	0.1931*** (0.0533)	0.1930*** (0.0533)	0.2048*** (0.0533)	0.1388** (0.0556)	0.1460*** (0.0550)

Notes: In Panel A the dependent variable is the indicator for a positive number of absence in QWLS 2003. In Panel B the dependent variable is the total number of sickness absence spells in QWLS 2003. In Panel C long absences are those that have lasted for 4 days or more and in Panel D short absences are those that have lasted 1–3 days. Sickness absence history refers to the total number of sickness absence days over the period 1995–2001, as recorded by KELA. Work history refers to variables that describe past labour market experiences. Controls that are included in Columns 4 and 5 consist of the individual and employer characteristics described in Table A.1 along with a full set of industry indicators. Job disamenities refer to Harm, Hazard and Uncertainty, as explained in the text. The specifications in Column 1 contain the HIM indicator only. Average marginal effects and robust standard errors reported.  $N = 3755$ .

\* Statistical significance:  $p < 0.1$ .

\*\* Statistical significance:  $p < 0.05$ .

\*\*\* Statistical significance:  $p < 0.01$ .

spells, the non-significant effects are fully consistent with the results for longer spells (4 days or more) using the QWLS data in Panel C of Table 1.

In Table 2, we rerun the short absences model from Table 1 but this time we consider the effects of a full set of different “bundles” of the four HIM practices on the number of short absence spells. All combinations of the four HIM practices are included in the models. The reference group in all of the specifications is no HIM practices. The most robust results across all model specifications are the associations of absences with PRP and training in isolation and with the combination of PRP and employer-provided training. Those aspects of HIM are strongly related to a higher number of short absences. In the preferred model (column 5) that includes all the controls the bundles that consists of “PRP and self-managed teams” and “PRP, self-managed teams and information sharing” are also statistically significant. However, these are very rare bundles, as shown in Table A.2.

Some of these effects are quantitatively large, as indicated by the average marginal effects reported in Table 2. PRP alone, for example, increases the number of short absence spells by 40–50 percent, depending on the model specification.<sup>16</sup> The link between PRP and more short absences is intuitive since PRP can be thought to involve work intensification. However, the positive link to training is somewhat surprising. A plausible explanation is that on-the-job training is almost always accompanied by changes or adjustments in work roles and it is these, rather than training *per se*, that generates increased short absences. Alternatively, if training reduces the amount of “down-time” at work, it could be linked to labour intensification. We examined the link between HIM bundles and work intensification by running a model equivalent to that in column 4 in Table 2 where the dependent variable was agreement with the statement that “time pressure increases sickness absence” (Table A.5, column 1) on a five-point Likert scale. Those exposed to PRP, training or a combination of the two were most likely to agree to the statement, further supporting the proposition that, at least for a subset of HIM practices, attendant work intensification was significantly associated with a higher probability of absence.

We repeated the same kind of analysis as in Table 2 also for the number of long absence spells. Overall, the HIM bundles seemed to have no significant connection to long spells (not reported). When all controls were included, “PRP only” was the only HIM variable with a significant coefficient. Its relationship to spells was positive, so that PRP seems to increase both short and long spells, but when used in combination with other HIM practices, the effect is confined to short spells.

Under the theoretical arguments presented in Section 2 this positive association between HIM and short-term absence taking is related to HIM employers' ability to call on the multi-tasking skills of HIM workers to avoid replacement labour costs, something that is not available to non-HIM employers. We can test for this relationship directly with our data. Employees are asked whether they agree with the statement: “Replacements are not hired to cover temporary absences”. The probit model results presented in Table A.5 (column 2) reveal very clearly that replacements are least likely to be hired where

<sup>16</sup> For example, with all the controls individuals with PRP only have  $\exp(0.3596) = 1.4328$  times the number of absences of workers with no HIM.

**Table 2**  
Innovative workplace practices as determinants of short absence spells: specific bundles.

	Baseline	Sickness absence history	Sickness absence history + work history	Sickness absence history + work history + controls	Sickness absence history + work history + controls + job disamenities
<b>Total number of short absence spells, zero infl. negative binomial regression</b>					
PRP only	0.3944** (0.1867)	0.3824** (0.1718)	0.4800*** (0.1826)	0.3780*** (0.1473)	0.3596*** (0.1404)
PRP and training	1.5947** (0.7872)	0.3343*** (0.0961)	0.3397*** (0.0989)	0.2047** (0.0987)	0.2067** (0.0970)
PRP, training and self-managed teams	1.6266* (0.8999)	2.5285*** (0.9440)	1.8097*** (0.6951)	0.9710 (0.6012)	1.0294 (0.7148)
All four HIM practices	0.3635 (0.3968)	0.3579 (0.2256)	0.3965 (0.2491)	0.2598 (0.2622)	0.2927 (0.2501)
PRP, self-managed teams and information sharing	0.7459* (0.4224)	0.7266 (0.4428)	0.7539 (0.4724)	0.6155 (0.4554)	0.7193* (0.4133)
PRP and information sharing	0.0710 (0.1414)	0.0864 (0.1400)	0.1500 (0.1464)	0.1250 (0.1469)	0.1583 (0.1484)
PRP, training and information sharing	0.9200* (0.4982)	1.7834*** (0.6385)	1.7420*** (0.6635)	1.0096* (0.6050)	0.8919 (0.5564)
PRP and self-managed teams	1.2752** (0.5674)	1.1945** (0.4970)	1.3944*** (0.4989)	1.1094*** (0.3987)	0.8753** (0.3687)
Training only	0.2948*** (0.0915)	0.2890*** (0.0854)	0.3222*** (0.0880)	0.2518*** (0.0792)	0.2131*** (0.0762)
Training and self-managed teams	1.5752* (0.8503)	2.4591*** (0.9022)	1.7442*** (0.6621)	1.0132* (0.5765)	1.0632 (0.6872)
Training, self-managed teams and information sharing	0.0860 (0.1528)	1.3245** (0.5233)	0.8559** (0.3574)	0.5770 (0.3537)	0.6561 (0.4104)
Training and information sharing	0.0953 (0.3516)	0.0748 (0.1163)	0.1184 (0.2679)	0.0642 (0.1574)	0.1066 (0.2067)
Self-managed teams only	-0.0168 (2.3194)	-0.0807 (0.1683)	0.0106 (0.1773)	0.0504 (0.1921)	0.1144 (0.2044)
Self-managed teams and information sharing	-0.2292 (0.2947)	-0.2413 (0.2437)	0.3172 (0.3663)	0.2460 (0.3478)	0.2825 (0.3705)
Information sharing only	-0.2045 (0.1355)	0.4287 (0.9026)	0.3968 (0.2671)	0.2124 (0.2504)	0.2929 (0.2999)

Notes: The dependent variable is the total number of short absences, defined as those that have lasted 1–3 days. The reference category is no HIM. See also notes to Table 1. Average marginal effects and robust standard errors reported.  $N = 3755$ .

\* Statistical significance:  $p < 0.1$ .

\*\* Statistical significance:  $p < 0.05$ .

\*\*\* Statistical significance:  $p < 0.01$ .

PRP and/or training are present, thus lending support to the argument that the shadow price of short-term absence to the employer is somewhat lower in these circumstances.

Table 3 repeats the analyses in Table 2 but for having had work accidents leading to absence in the last 12 months. In many ways the results are the mirror image of those in Table 2: those “bundles” associated with a *higher* incidence of short-term absences are also associated with *fewer* accidents. PRP and training are the practices most likely to be associated with fewer accidents, the most robust result being the negative association between accidents and the “bundle” of PRP plus training. With all controls, “PRP only” and the bundle “PRP and training” were the only significant ones: both HIM regimes were negatively associated with the probability of accidents. The effects are sizeable given that slightly fewer than 5 percent of the sample have had accidents (Table A.1). As noted earlier, HIM practices lead to work intensification, but also to increased control over the working environment. The former may increase accident rates, but the latter should have an opposite effect. Our finding is therefore consistent with increased control dominating intensification.

To explore the relationships between absences, accidents and HIM a little further we ran bivariate probit models where the two dependent variables were dummies for having any spell of short-term absence and for having any accident leading to sickness absence. The motivation for this specification is that there may be some common shocks that affect both accidents and absence. This may pose a problem for inference in the single equation models. The controls and model specifications were identical to those presented in column 5 of Tables 2 and 3. There was a highly statistically significant rho identifying a positive unobserved correlation between short absences and accidents (not reported). The positive estimate of rho was also

**Table 3**  
Innovative workplace practices as determinants of accidents: specific bundles.

	Baseline	Sickness absence history	Sickness absence history + work history	Sickness absence history + work history + controls	Sickness absence history + work history + controls + job disamenities
<b>Person has had an accident, probit</b>					
PRP only	-0.0110 (0.0133)	-0.0110 (0.0133)	-0.0126 (0.0135)	-0.0237* (0.0135)	-0.0260** (0.0133)
PRP and training	-0.0554*** (0.0151)	-0.0554*** (0.0151)	-0.0548*** (0.0155)	-0.0501*** (0.0154)	-0.0511*** (0.0152)
All four HIM practices	-0.0607 (0.0424)	-0.0606 (0.0424)	-0.0498 (0.0428)	-0.0363 (0.0419)	-0.0369 (0.0407)
PRP, self-managed teams and information sharing	-0.0128 (0.0500)	-0.0128 (0.0500)	-0.0204 (0.0458)	-0.0259 (0.0429)	-0.0185 (0.0419)
PRP and information sharing	-0.0114 (0.0198)	-0.0115 (0.0198)	-0.0132 (0.0193)	-0.0103 (0.0196)	-0.0061 (0.0196)
PRP, training and information sharing	-0.0445*** (0.0170)	-0.0445*** (0.0170)	-0.0409** (0.0171)	-0.0282 (0.0177)	-0.0232 (0.0177)
PRP and self-managed teams	0.0221 (0.0404)	0.0217 (0.0406)	0.0240 (0.0407)	0.0217 (0.0408)	0.0086 (0.0394)
Training only	-0.0320*** (0.0106)	-0.0320*** (0.0106)	-0.0255** (0.0106)	-0.0123 (0.0106)	-0.0158 (0.0105)
Training and self-managed teams	-0.0112 (0.0213)	-0.0111 (0.0213)	-0.0020 (0.0211)	-0.0126 (0.0211)	0.0095 (0.0205)
Training, self-managed teams and information sharing	-0.0899** (0.0387)	-0.0899** (0.0387)	-0.0781** (0.0390)	-0.0463 (0.0401)	-0.0398 (0.0400)
Training and information sharing	-0.0478*** (0.0134)	-0.0477*** (0.0134)	-0.0382*** (0.0135)	-0.0113 (0.0139)	-0.0076 (0.0137)
Self-managed teams only	-0.0104 (0.0258)	-0.0104 (0.0258)	-0.0095 (0.0261)	-0.0074 (0.0262)	-0.0055 (0.0264)
Self-managed teams and information sharing	-0.0337 (0.0331)	-0.0336 (0.0331)	-0.0300 (0.0329)	-0.0154 (0.0319)	-0.0114 (0.0308)
Information sharing only	-0.0278** (0.0141)	-0.0278** (0.0141)	-0.0256* (0.0138)	-0.0178 (0.0135)	-0.0125 (0.0134)

Notes: The dependent variable is the indicator whether a person has had an accident at work which has resulted in absence from work in the last 12 months. The combination "PRP, training and self-managed teams" is not included among the explanatory variables, because there is no variation in the outcome. See also notes to Table 1. Average marginal effects and robust standard errors reported.  $N = 3755$ .

\* Statistical significance:  $p < 0.1$ .

\*\* Statistical significance:  $p < 0.05$ .

\*\*\* Statistical significance:  $p < 0.01$ .

robust to model specification. This is something we would fully expect since employees suffering an accident at work will often need to take a short absence to recuperate. Having accounted for this correlation the results presented in Tables 2 and 3 were confirmed. The links between HIM "bundles" containing PRP or training and a higher probability of short-term absence were very well-determined, as were the links between these practices and a reduced likelihood of an accident.

The negative relationship between PRP and training bundles and accidents, on the one hand, and their positive association with short absences is consistent with a safer working environment in which accidents are less likely and employees are encouraged not to practice "presenteeism" whereby they turn up for work even when they are not fully fit. This interpretation is also consistent with the fact that PRP has a much stronger effect on short absence spells than on long ones. It is plausible that accidents lead more often to long rather than short absences from work.

In Table 4, we turn to subjective indicators of worker wellbeing and pain. The entries in the table are again the average marginal effects of the HIM variables on various wellbeing measures from separate specifications which use the full set of controls used in the penultimate columns of Tables 2 and 3, i.e., they exclude job disamenities.<sup>17</sup> Each column corresponds to a different dependent variable. In columns 1–3 and 8, the dependent variables are ordered categories, so we use ordered probit models and present average marginal effects on the probability of the top category. In columns 4–7, the dependent

<sup>17</sup> The results reported in Table 4 are not sensitive to the inclusion of job disamenities.

**Table 4**  
Innovative workplace practices as determinants of subjective wellbeing: specific bundles.

	Job satisfaction	Working capacity	The state of health	No pain in neck	No pain in arms	No pain in lumbar	No pain in legs	Not being tired
PRP only	0.0115 (0.0310)	0.0127 (0.0260)	-0.0503 (0.0340)	-0.0119 (0.0346)	0.0014 (0.0289)	0.0355 (0.0324)	0.0261 (0.0293)	0.0361 (0.0231)
PRP and training	0.0753** (0.0302)	0.0124 (0.0221)	0.0304 (0.0290)	0.0349 (0.0313)	0.0416 (0.0268)	0.0362 (0.0291)	0.0281 (0.0267)	0.0261 (0.0202)
PRP, training and self-managed teams	0.0902 (0.0652)	-0.0031 (0.0493)	-0.0394 (0.0804)	-0.0181 (0.0861)	0.0154 (0.0719)	0.1739* (0.0955)	0.0121 (0.0730)	0.0553 (0.0653)
All four HIM practices	0.3297*** (0.0767)	0.0283 (0.0514)	-0.0639 (0.0804)	0.0719 (0.0777)	0.0969 (0.0731)	-0.0051 (0.0719)	0.0809 (0.0738)	0.1067* (0.0548)
PRP, self-managed teams and information sharing	0.2570* (0.1317)	0.1754* (0.0949)	0.1442 (0.0942)	0.1097 (0.1253)	0.0890 (0.1089)	0.0559 (0.1133)	0.1634 (0.1218)	0.0204 (0.0822)
PRP and information sharing	0.2176*** (0.0493)	0.0538 (0.0382)	0.0104 (0.0493)	-0.0057 (0.0498)	0.0343 (0.0434)	0.0567 (0.0469)	-0.0001 (0.0411)	0.0954*** (0.0365)
PRP, training and information sharing	0.2374*** (0.0355)	0.0784*** (0.0272)	0.1023*** (0.0337)	0.0464 (0.0372)	0.0601* (0.0328)	0.0911*** (0.0356)	0.0769** (0.0330)	0.1365*** (0.0293)
PRP and self-managed teams	0.1191 (0.1152)	-0.1813*** (0.0271)	-0.2546** (0.1029)	0.1040 (0.1259)	-0.0292 (0.0931)	-0.0585 (0.1126)	-0.0530 (0.0995)	-0.0769* (0.0440)
Training only	0.0082 (0.0226)	-0.0113 (0.0174)	-0.0456* (0.0244)	0.0113 (0.0253)	-0.0363* (0.0208)	-0.0220 (0.0232)	-0.0175 (0.0209)	0.0104 (0.0163)
Training and self-managed teams	0.1928*** (0.0505)	0.0151 (0.0393)	-0.0790 (0.0554)	0.0412 (0.0550)	0.0377 (0.0463)	0.0824 (0.0521)	0.0878* (0.0466)	0.1146*** (0.0402)
Training, self-managed teams and information sharing	0.2627*** (0.0512)	0.0540 (0.0401)	0.1204** (0.0537)	0.1216** (0.0551)	0.1183** (0.0528)	0.1448*** (0.0561)	0.1328*** (0.0507)	0.1891*** (0.0421)
Training and information sharing	0.2261*** (0.0285)	0.0754*** (0.0223)	0.0535* (0.0281)	0.1009*** (0.0296)	0.0429* (0.0253)	0.0513* (0.0278)	0.0723*** (0.0254)	0.1168*** (0.0219)
Self-managed teams only	0.1055* (0.0585)	-0.0319 (0.0336)	0.0443 (0.0555)	0.0564 (0.0638)	-0.0007 (0.0517)	0.0085 (0.0569)	0.0017 (0.0520)	0.1677*** (0.0530)
Self-managed teams and information sharing	0.2549*** (0.0717)	0.0553 (0.0552)	0.0509 (0.0711)	0.0108 (0.0715)	0.1911** (0.0792)	0.1303* (0.0698)	0.0233 (0.0604)	0.1134** (0.0562)
Information sharing only	0.1046*** (0.0304)	0.0425* (0.0241)	0.0452 (0.0304)	0.0753** (0.0327)	0.0290 (0.0274)	-0.0042 (0.0297)	0.0471* (0.0269)	0.0845*** (0.0244)

Notes: All specifications include individual and employer characteristics and work history variables as controls along with a full set of industry indicators. All models also control for sickness absence history. The specifications in columns 1–3 and column 8 are estimated by using ordered probit and the specifications in columns 4–7 are estimated by using probit. Average marginal effects and robust standard errors reported. For ordered probit models the marginal effects are reported for the probability of the top category.  $N = 3755$ .

\* Statistical significance:  $p < 0.1$ .

\*\* Statistical significance:  $p < 0.05$ .

\*\*\* Statistical significance:  $p < 0.01$ .

pain measure variables are binary, so we use probit models and the average marginal effects from these models give the impact on the probability of not having the negative symptoms.

Eleven of the fifteen HIM regimes have a positive, statistically significant association with job satisfaction, and none were negatively associated with job satisfaction (column 1). It is clear, therefore, that HIM is positively associated with employee positive affect. Column 8 provides overwhelming evidence that HIM is also associated with a lower likelihood of being tired: nine of the fifteen HIM regimes were positively associated with not feeling tired at work. On the other hand, only three HIM regimes were consistently associated with a lower likelihood of feeling pain on at least three of the four types of pain recorded in the survey. All three of these “bundles” included training and information sharing. Positive associations of HIM with work capacity and self-assessed state of one’s health were less evident. The HIM “bundle” most consistently associated with employees’ wellbeing is the “bundle” containing PRP, training and information sharing. It has a positive and statistically significant relationship to all wellbeing outcomes in Table 4 with the exception of no neck pain.

In general there is little evidence of HIM being associated with poorer employee wellbeing. But there is one exception: the combination of PRP and team working is associated with having a lower self-assessed working capacity, having a lower assessment of one’s own state of health, and feeling more tired.<sup>18</sup> Although we are not able to conclude that there is a causal relationship between exposure to team working allied to PRP and poorer employee wellbeing, this correlation is independent

<sup>18</sup> Note that this is not a very common bundle.

of a particularly rich set of controls, including employee demographic characteristics, the nature of the workplace, and the employees' own absence, work and earnings histories. It is also consistent with the proposition that PRP coupled with team working can incentivise workers who respond by working more intensively. In particular, team work combined with PRP may lead to increased peer pressure and thereby to higher effort levels (Kander and Lazear, 1992; Barron and Paulson Gjerde, 1997). The fact that this negative relationship does not show up in other bundles including these two HIM practices suggests that the effects are ameliorated when combined with training and information sharing.

#### 4.2. Additional aspects

To evaluate the earlier results, we have estimated alternative specifications. We briefly discuss these results without presenting them in tables. Another line of the literature – the share capitalism strand – emphasises the distinction between individual performance pay, on the one hand, and group, team or firm-based performance pay on the other (Kruse et al., 2010). It is argued that it is really the latter that inculcates cohesion and a team-working orientation. We constructed indicators for both group (team) and organisation based PRP using information on what kind of bonuses the person receives. Organisation based PRP is clearly the most important component of PRP in Finland. Twice as many employees are paid organisation based bonuses compared to team based PRP. The estimations with the new indicators showed that the results for PRP in the tables that use specific bundles are for the most part driven by organisation based PRP. We also tested whether the specifications that use 'any HIM v none' are sensitive to the definition of PRP. The key result of the paper according to which HIM status is positively correlated with the number of short absence spells, but remains non-significant for long absence spells, obtained support also when 'any HIM v none' was defined based on organisation based PRP.

The basic models do not include occupational controls, because it is not possible to identify all of them especially in the specifications that use the comprehensive set of HIM bundles. We conducted robustness tests that included 2-digit occupational indicators to control for potential heterogeneity among occupations. Their inclusion did not change the conclusions.

If work organisation affects claims for sickness allowance, it is possible that absence history is correlated with HIM and therefore may not be a good control variable. We checked whether the correlation between absences according to KELA and absences according to QWLS is independent from HIM status in 2003, regressing QWLS absences on KELA absences. Exclusion/inclusion of HIM status in 2003 as a variable in the regression had only a very small influence on the relationship between the two absence variables. This gives evidence for HIM not affecting the allowance claims. The conclusion was not sensitive to the use of individual and employer controls in the regression.

To downplay the potential impact of confounding factors, we have used an unusually extensive vector of control variables in the specifications, including employees' comprehensive work and absence histories that account for sorting of employees into HIM status. However, this is not ideal to fully establish causality. To shed further light on the causal effects, we used the lagged incidence of HIM in the same 2-digit industry cell in 1997 to instrument for exposure to HIM in 2003.<sup>19</sup> The idea is that HIM is a technology which diffuses across time and space according to certain structural features of firms and their peers, e.g. via networks, or as an experience good, or through herding mentality. This affects the propensity of specific firms to deploy HIM. However, having conditioned on the full set of detailed industry effects, there is no reason to suspect any effect of lagged industry HIM on current worker wellbeing. The first stage of these IV models worked well. The *F*-test statistics were well above 10 that is the threshold proposed by Staiger and Stock (1997) for a weak instrument. Thus, any HIM in the industry in 1997 is a strong predictor for a person being exposed to any HIM in 2003. We estimated linear probability models with the IV approach for specifications that corresponded to Panel A of Table 1. The original results remained intact. Thus, there was a positive and significant effect in models corresponding to columns 1–3 of Panel A, but the estimate turned non-significant in specifications corresponding to columns 4–5. The point estimates were larger, but the standard errors were also (much) larger, as expected. For the count models corresponding to Panels B–D of Table 1 we applied GMM. There were some differences in the results, but still the basic pattern clearly remained the same.

We also estimated a set of recursive bivariate probit models in which the dependent variable of the first equation was any HIM in 2003 and the explanatory variables were any HIM in the same industry cell in 1997 and the controls. In the second equation the dependent variable was any absence and the explanatory variables were any HIM in 2003 and the controls. We estimated these models with different set of controls as in Table 1. The pattern of results remained exactly the same as in Table 1. Thus, any HIM in 2003 was positive and significant in specifications similar to columns 1–3, but non-significant in specifications similar to columns 4–5.<sup>20</sup> This suggests that unobservables are not driving the results in Table 1.

To explore the potential heterogeneity in the effects, we estimated separate specifications for females and males and for young (aged less than 45) and old (aged 45 or more) workers. It is useful to study the relationships in these groups, because the prevalence of sickness absence is at much higher level for females and old workers. Generally, there were no large differences in the results between different groups of workers. However, the effect of HIM practices on experiencing no pain in neck was not significant for the young workers. This finding is not particularly surprising, because the prevalence of neck pain is much higher for the older workers. We also estimated a set of specifications separately for blue-collar workers and white-collar workers, defined based on socio-economic status in 2000 from FLEED. The most interesting finding was that

<sup>19</sup> The QWLS data is available for both 2003 and 1997.

<sup>20</sup> The correlation between the error terms of the equations was negative in all specifications.

the white-collar workers are a very heterogeneous group in terms of accidents. The share of accidents is roughly 3 percent for the lower white-collar workers, but they are practically non-existent for the upper white-collar workers. For the lower white-collar workers the negative effect of HIM practices on accidents seemed to be somewhat larger than in the full sample (without adding all the controls to the model). Finally, we split the full sample by using the Harm and Hazard variables and experimented with different ways to define workers who are facing substantial risk at the workplace. We obtained a very clear-cut result that the negative effect of HIM practices on accidents was roughly twice the effect in the full sample (after adding all the controls) if workers who are “at risk” were defined as those who had experienced both harms and hazards at their workplaces. Thus, the protective effects of HIM practices are clearly more pronounced for those who are in the zone of risk.

## 5. Conclusions

Using nationally representative survey data for Finnish employees linked to register data on their work, wage and sickness absence histories we observe that high involvement management (HIM) practices are generally positively and significantly associated with various aspects of employee wellbeing. In particular, HIM is strongly associated with higher evaluations of subjective wellbeing including higher job satisfaction and non-tiredness. HIM is also associated with a lower probability of having a workplace accident. However, HIM exposure – especially performance-related pay and training – is also associated with having more short absence spells.

We tackle the problem that workers are not randomly assigned to HIM by using an innovative combination of survey and register data. In the empirical specifications work and sickness absence histories of the employees are used to control for individual effects that might be correlated with exposure to HIM. This extension of previous analyses brings two important insights. Firstly, we show that links between HIM exposure and worker wellbeing are not for the most part biased by the omission of work and sickness absence histories. Thus, in this sense the role played by these controls is relatively small. Secondly, it is nevertheless important to control for work and sickness absence histories since conditioning on them has a sizeable quantitative impact on the HIM coefficients in our models. The point estimates of HIM on employee wellbeing can be biased without these controls that account for sorting of employees into HIM status.

The positive association between HIM and the incidence of short absence spells is consistent with the view that for firms using HIM practices zero absences may not be optimal. Rather, the optimal absence rate may involve short spells if the firm can meet production schedules by intensifying work using multi-tasking workers. On the other hand, avoiding long absences is likely to be beneficial for all firms. We find no strong evidence that HIM reduces the number of long-term absences, but we do find clear evidence that HIM practices are associated with a lower probability of having accidents at work.

Our results are rather positive from the employee point of view whereas the previous literature presents more mixed findings. It is plausible that the co-operation between employees and employers which characterises employment relations in Finland, together with the strong role of trade unions in implementing work reorganisation, results in mutual gains for firms and workers (see also Kalmi and Kauhanen, 2008). That said the combination of PRP and team working, which is central to notions of devolved responsibilities underpinned by incentive structures, is also the HIM “bundle” most clearly associated with negative outcomes for employee wellbeing.

Short-term absences and workplace accidents have positively correlated unobservable components, but the former is positively and the latter negatively correlated with HIM. This result might be driven by unobservable features of the working environment, such as having a “good” employer capable of investing in HIM, keeping absences low and minimising accidents all by virtue of good management rather than HIM *per se*. Future research on this issue would benefit from taking into account employer unobserved heterogeneity which may simultaneously affect worker wellbeing and the propensity for HIM adoption.<sup>21</sup> Our data have only a few observations from many of the firms, thus preventing us from exploring this issue.

## Appendix 1. The Finnish sickness insurance system

When an employee becomes sick, he/she has to provide a doctor’s note on sick leave, usually after the first day of sickness. After the first day, there is a nine working days’ (including Saturdays, but not Sundays or public holidays) waiting period until the employee is eligible to sickness allowance from the Social Insurance Institution (KELA). During this period, the employee gets his/her pay from the employer, if the employment relationship has lasted at least a month. If it has lasted less than a month, the employee obtains 50 percent of the pay. This waiting period is waived if the same sickness has already caused absence in the previous 30 days.

After the waiting period the employee starts to get an earnings-related sickness allowance. However, most labour contracts stipulate that the firms actually continue to pay the employees much longer than the waiting period. The length of the pay usually depends on length of tenure. For example, according to the metal workers’ contract, workers with tenure over one month but less than 3 years are paid for 28 calendar days, workers with tenure of at least 3 years but less than 5 years are paid for 35 days, those with tenure of at least 5 but less than 10 years are paid for 42 days, and the workers with

<sup>21</sup> Dionne and Dostie (2007) is an example of a linked employer-employee data study where employer heterogeneity in absenteeism is accounted for.

at least 10 years' tenure get paid for 56 calendar days. The exact rules vary across contracts. When the employees get their pay even after the waiting period, the employer gets the sickness allowance.

After 60 work days (approximately 80 calendar days) on sickness allowance there is an assessment of rehabilitation needs. If the employee goes to rehabilitation, sickness allowance ends and he/she gets rehabilitation allowance. The maximum period for sickness allowance is 300 working days (approximately a full calendar year); all allowance days within the last 2 years are counted to this. After this maximum has been reached, there is an assessment of eligibility to disability pension. The person can get sickness allowance again only after having worked for at least a year.

The sickness allowance is based on past earnings (previous year or previous months if earnings have increased). Work-related expenses are deducted from earnings, and in addition a deduction is made to account for pension and unemployment insurance contributions. In 2003 (the year of the QWLS survey from which we have also the survey information on absences), the allowance was calculated in the following way. A 4.8 percent deduction of earnings was first used for the insurance contributions. There was no daily allowance if annual earnings were below €1004; in the range €1004–26,124 the allowance was  $0.7 \times \text{earnings}/300$ ; in the range €26,125–40,192 it was  $60.96 + 0.4 \times (\text{earnings} - 26,124)/300$ ; finally for annual earnings over €40,192 the daily allowance was  $79.71 + 0.25 \times (\text{earnings} - 40,192)/300$ . Therefore, the replacement rate falls with earnings. Those receiving no allowance or a very low one could get minimum allowance €11.45 after 55 days' sickness. Over time, there have been changes in the earnings limits and replacement rates.

Source: KELA (The Social Insurance Institution of Finland), A Guide to Benefits 2003.

## Appendix 2. Additional tables

**Table A.1**  
Descriptive statistics of the variables.

Variable	Average	Standard deviation	Source
<b>Outcomes</b>			
<i>Sickness absence</i>			
Any absence days, dummy	0.6028	0.4894	QWLS
Total number of absence days	8.6787	21.6096	QWLS
Total number of absence spells	1.3580	1.9036	QWLS
Any absence days, dummy	0.1211	0.3263	KELA
Total number of absence days	3.2351	15.5850	KELA
<i>Accidents</i>			
Has had an accident at work which has resulted in absence from work in the last 12 months	0.0495	0.2169	QWLS
<i>Subjective wellbeing</i>			
Job satisfaction	3.2464	0.5989	QWLS
Working capacity	8.5673	1.3348	QWLS
The state of health	4.4023	0.7724	QWLS
No pain in neck	0.5666	0.4956	QWLS
No pain in arms	0.7599	0.4272	QWLS
No pain in lumbar	0.6901	0.4625	QWLS
No pain in legs	0.7559	0.4296	QWLS
Not being tired	4.6231	1.1695	QWLS
<b>Sickness absence history</b>			
The total number of sickness absence days	14.695	40.755	KELA
<b>Work history</b>			
Number of job switches	1.7816	1.5464	FLEED
Number of employment months	102.6729	45.1923	FLEED
Number of unemployment months	8.6227	15.9072	FLEED
Ever worked in the manufacturing sector	0.2493	0.4327	BR
Ever worked in a firm with over 300 workers	0.2930	0.4552	BR
Number of layoff episodes	0.3041	0.9464	FLEED
Past average earnings	6.3748	1.5636	FLEED
Past average earnings change	0.1119	0.4972	FLEED
Worked over 10 years with the current employer	0.4027	0.4905	QWLS
Had over 3 professions over working life	0.1423	0.3494	QWLS

Table A.1 (Continued)

Variable	Average	Standard deviation	Source
<b>Controls</b>			
<i>Individual</i>			
Female	0.5230	0.4995	QWLS
Age <= 34	0.2811	0.4496	QWLS
Age 35–44 (ref.)	0.2612	0.4394	QWLS
Age 45–54	0.2959	0.4565	QWLS
Age 55–64	0.1616	0.3681	QWLS
Married	0.7506	0.4327	QWLS
Comprehensive education only (ref.)	0.1663	0.3724	QWLS
Secondary education	0.4381	0.4962	QWLS
Polytechnic education	0.2800	0.4491	QWLS
University education	0.1155	0.3197	QWLS
Union member	0.7911	0.4066	QWLS
Usual weekly hours	34.2205	7.1307	QWLS
<i>Employer</i>			
Plant size < 10 (ref.)	0.2290	0.4202	QWLS
Plant size 10–49	0.3725	0.4835	QWLS
Plant size ≥ 50	0.3985	0.4897	QWLS
Part of multi-plant firm	0.4217	0.4939	QWLS
Foreign firm	0.0945	0.2926	QWLS
Public sector	0.3535	0.4781	QWLS
<b>Job disamenities</b>			
Harm	0.2771	0.4476	QWLS
Hazard	0.3810	0.4857	QWLS
Uncertainty	0.6018	0.4896	QWLS

Notes: BR, business register; FLEED, Finnish Longitudinal Employer–Employee Data; KELA, Social Insurance Institution; QWLS, Quality of Work Life Survey. Controls also include 14 industry indicators (not shown).

Table A.2

The incidence of different HIM variables.

	Mean
<b>Any HIM</b>	0.7742
<b>Specific bundles</b>	
PRP only	0.0736
PRP and training	0.1062
PRP, training and self-managed teams	0.0083
All four HIM practices	0.0120
PRP, self-managed teams and information sharing	0.0043
PRP and information sharing	0.0286
PRP, training and information sharing	0.0664
PRP and self-managed teams	0.0043
Training only	0.1873
Training and self-managed teams	0.0244
Training, self-managed teams and information sharing	0.0249
Training and information sharing	0.1250
Self-managed teams only	0.0158
Self-managed teams and information sharing	0.0128
Information sharing only	0.0803

Notes: Specific bundles are mutually exclusive components of 'any HIM'. The sum of the incidence of specific bundles constitutes the incidence of 'any HIM'.



**Table A.3**  
The association between different HIM variables.

	Any HIM	b1	b2	b3	b4	b5	b6	b7	b8	b9	b10	b11	b12	b13	b14	b15
Any HIM	1															
b1	0.1522 <sup>***</sup>	1														
b2	0.1862 <sup>***</sup>	-0.0972 <sup>***</sup>	1													
b3	0.0494 <sup>***</sup>	-0.0258	-0.0315 <sup>*</sup>	1												
b4	0.0596 <sup>***</sup>	-0.0311 <sup>*</sup>	-0.0381 <sup>**</sup>	-0.0101	1											
b5	0.0354 <sup>**</sup>	-0.0185	-0.0226	-0.0060	-0.0072	1										
b6	0.0927 <sup>***</sup>	-0.0484 <sup>***</sup>	-0.0592 <sup>***</sup>	-0.0157	-0.0190	-0.0113	1									
b7	0.1440 <sup>***</sup>	-0.0751 <sup>***</sup>	-0.0919 <sup>***</sup>	-0.0244	-0.0294 <sup>*</sup>	-0.0175	-0.0458 <sup>***</sup>	1								
b8	0.0354 <sup>**</sup>	-0.0185	-0.0226	-0.0660	-0.0072	-0.0043	-0.0113	-0.0175	1							
b9	0.2593 <sup>***</sup>	-0.1353 <sup>***</sup>	0.1655 <sup>***</sup>	-0.0439 <sup>***</sup>	-0.0530 <sup>***</sup>	-0.0315 <sup>*</sup>	-0.0824 <sup>***</sup>	-0.1280 <sup>***</sup>	-0.0315 <sup>*</sup>	1						
b10	0.0853 <sup>***</sup>	-0.0445 <sup>***</sup>	-0.0545 <sup>***</sup>	-0.0144	-0.0174	-0.0104	-0.0271 <sup>*</sup>	-0.0421 <sup>***</sup>	-0.0104	-0.0758 <sup>***</sup>	1					
b11	0.0863 <sup>***</sup>	-0.0450 <sup>***</sup>	-0.0551 <sup>***</sup>	-0.0146	-0.0176	-0.0105	-0.0274 <sup>*</sup>	-0.0426 <sup>***</sup>	-0.0105	-0.0767 <sup>***</sup>	-0.0252	1				
b12	0.2041 <sup>***</sup>	-0.1065 <sup>***</sup>	-0.1303 <sup>***</sup>	-0.0346 <sup>**</sup>	-0.0417 <sup>***</sup>	-0.0248	-0.0649 <sup>***</sup>	-0.1008 <sup>***</sup>	-0.0248	-0.1814 <sup>***</sup>	-0.0597 <sup>***</sup>	-0.0604 <sup>***</sup>	1			
b13	0.0684 <sup>***</sup>	-0.0357 <sup>*</sup>	-0.0437 <sup>***</sup>	-0.0116	-0.0140	-0.0083	-0.0217	-0.0338 <sup>**</sup>	-0.0083	-0.0608 <sup>***</sup>	-0.0200	-0.0202	-0.0479 <sup>***</sup>	1		
b14	0.0616 <sup>***</sup>	-0.0321 <sup>*</sup>	-0.0393 <sup>**</sup>	-0.0104	-0.0126	-0.0075	-0.0196	-0.0304 <sup>*</sup>	-0.0075	-0.0548 <sup>**</sup>	-0.0180	-0.0182	-0.0431 <sup>***</sup>	-0.0144	1	
b15	0.1596 <sup>***</sup>	-0.0833 <sup>***</sup>	-0.1019 <sup>***</sup>	-0.0270 <sup>*</sup>	-0.0326 <sup>***</sup>	-0.0194	-0.0507 <sup>***</sup>	-0.0788 <sup>***</sup>	-0.0194	-0.1418 <sup>***</sup>	-0.0467 <sup>***</sup>	-0.0472 <sup>***</sup>	-0.1116 <sup>***</sup>	-0.0374 <sup>**</sup>	-0.0337 <sup>**</sup>	1

Notes: Kendall's tau-b correlation coefficients reported.

Key to specific bundles: b1, PRP only; b2, PRP and training; b3, PRP, training and self-managed teams; b4, All four HIM practices; b5, PRP, self-managed teams and information sharing; b6, PRP and information sharing; b7, PRP, training and information sharing; b8, PRP and self-managed teams; b9, training only; b10, training and self-managed teams; b11, training, self-managed teams and information sharing; b12, training and information sharing; b13, self-managed teams only; b14, self-managed teams and information sharing; b15, information sharing only.

<sup>\*</sup> Statistical significance:  $p < 0.1$ .

<sup>\*\*</sup> Statistical significance:  $p < 0.05$ .

<sup>\*\*\*</sup> Statistical significance:  $p < 0.01$ .

**Table A.4**

The full estimation results.

<b>Total number of absence spells, zero infl. negative binomial regression</b>	
<i>HIM practices</i>	
Any HIM v none	0.1483** (0.0668)
<i>Sickness absence history</i>	
The total number of sickness absence days	0.0042*** (0.0007)
<i>Work history</i>	
Number of job switches	0.0601*** (0.0220)
Number of employment months	-0.0003 (0.0013)
Number of unemployment months	-0.0028 (0.0024)
Ever worked in the manufacturing sector	-0.1254 (0.0878)
Ever worked in a firm with over 300 workers	0.0582 (0.0710)
Number of layoff episodes	-0.0034 (0.0322)
Past average earnings	-0.0032 (0.0340)
Past average earnings change	0.1212* (0.0542)
Worked over 10 years with the current employer	0.0406 (0.0831)
Had over 3 professions over working life	0.1565** (0.0774)
<b>Controls</b>	
<i>Individual</i>	
Female	0.4034*** (0.0690)
Age ≤34	0.1744* (0.0892)
Age 45–54	-0.3489*** (0.0814)
Age 55–64	-0.5841*** (0.1100)
Married	0.0657 (0.0667)
Secondary education	0.0189 (0.0902)
Polytechnic education	-0.2580** (0.1013)
University education	-0.3180** (0.1336)
Union member	-0.0616 (0.0741)
Usual weekly hours	0.0081* (0.0045)
<i>Employer</i>	
Plant size 10–49	0.2392*** (0.0787)
Plant size ≥50	0.3821*** (0.0863)
Part of multi-plant firm	-0.0390 (0.0830)
Foreign firm	0.0832 (0.1002)
Public sector	0.1248 (0.1075)
<b>Job disamenities</b>	
Harm	0.3545*** (0.0683)

Table A.4 (Continued)

Hazard	0.3003*** (0.0635)
Uncertainty	0.2518*** (0.0598)

Notes: The table reports the estimates for all the included explanatory variables (excluding the indicators for 14 industries) from the specification of column 5 of Panel B in Table 1. Average marginal effects and robust standard errors reported.  $N = 3755$ .

\* Statistical significance:  $p < 0.1$ .

\*\* Statistical significance:  $p < 0.05$ .

\*\*\* Statistical significance:  $p < 0.01$ .

Table A.5

Innovative workplace practices as determinant of replacement policies and perceived work intensity.

	Time pressure increases sickness absence	Replacements not hired
<b>Probit</b>		
PRP only	0.0563** (0.0226)	0.0854*** (0.0325)
PRP and training	0.0403* (0.0208)	0.0860*** (0.0286)
PRP, training and self-managed teams	0.0417 (0.0566)	0.0060 (0.0746)
All four HIM practices	-0.0561 (0.0571)	0.0279 (0.0694)
PRP, self-managed teams and information sharing	0.1078 (0.0747)	0.1557 (0.1318)
PRP and information sharing	-0.0744* (0.0404)	-0.0050 (0.0440)
PRP, training and information sharing	-0.0605** (0.0286)	0.0608* (0.0346)
PRP and self-managed teams	0.1628** (0.0660)	0.0537 (0.0986)
Training only	0.0304* (0.0171)	0.0677*** (0.0230)
Training and self-managed teams	0.0138 (0.0354)	0.0597 (0.0479)
Training, self-managed teams and information sharing	-0.0349 (0.0411)	0.0199 (0.0477)
Training and information sharing	-0.0537** (0.0218)	0.0178 (0.0262)
Self-managed teams only	-0.0245 (0.0452)	0.0719 (0.0598)
Self-managed teams and information sharing	-0.0613 (0.0611)	-0.0117 (0.0627)
Information sharing only	-0.0429* (0.0250)	-0.0035 (0.0290)

Notes: Controls as per Column 4 of Table 2 consist of the individual and employer characteristics and industry indicators along with work and sickness absence histories. The reference category is no HIM. Average marginal effects and robust standard errors reported.  $N = 3584$ .

\* Statistical significance:  $p < 0.1$ .

\*\* Statistical significance:  $p < 0.05$ .

\*\*\* Statistical significance:  $p < 0.01$ .

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