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How are work-related characteristics linked to sickness absence and presenteeism? Theory and data

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– Theory and Data –

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– Theory and Data –

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Abstract

This paper investigates how changes in work-related factors affect workers’ absence and presenteeism behavior. Previous studies (implicitly) assume that there is a substitutive relationship, i.e. a change in a work-related factor decreases the level of absence and simultaneously increases presenteeism (or vice versa). We set up a theoretical model in which work-related characteristics not only affect a worker’s absence decision but also the individual-specific sickness definition. Since work-related factors affect presenteeism through these two channels, non-substitutive relationships between absence and presenteeism are also conceivable. Using European cross-sectional data, we find only few substitutive and complementary relationships, while the bulk of the work-related characteristics is related only to one of the two sickness states.

Keywords: Sickness absence, presenteeism, annual duration, work-related characteristics, health at work

JEL Classification: J22, J28, I1, M50

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1 Introduction

1.1 Work-related Characteristics, Sickness Presenteeism and Absence

It is well established that sickness absence and presenteeism, i.e. going to work while sick, have negative economic effects through reduced or less productive labor supply (for absence see Pauly et al., 2002; for presenteeism see Pauly et al., 2008). Motivated by this stylized fact, a large number of papers investigate the determinants of absence and presenteeism behavior. Since most of the studies in this field look only at the determinants of one of the two sickness states, the possibility that one factor might influence absence and presenteeism behavior at the same time is neglected.\footnote{Only three studies (Böckerman and Laukkanen, 2009, 2010; Johansson and Lundberg, 2004) investigate both sickness states at once, and these will be discussed in more detail below. In contrast, the bulk of the literature investigates either sickness absence (for an early survey article, see Brown and Sessions, 1996; for literature using European cross-country data, see Frick and Malo, 2008; Livanos and Zangelidis, 2013; Lusinyan and Bonato, 2007) or presenteeism behavior (see Arnold, forthcoming; Aronsson et al., 2000; Aronsson and Gustafsson, 2005; Böckerman and Laukkanen, 2009; Hansen and Andersen, 2008, 2009; Leineweber et al., 2011; Preisendörfer, 2010).} Despite this lack of empirical evidence, some studies assume that a determinant which reduces absence leads to an increase in presenteeism (and vice versa) and insinuate thus a substitutive relationship. This assumption, however, is motivated rather implicitly by describing both sickness states as the result of the same decision process (Aronsson and Gustafsson, 2005) or by deducing hypotheses for determinants of presenteeism negatively from the literature on absence (Bierla et al., 2013).

In this paper, we contribute to the existing literature by explicitly and comprehensively analyzing how factors influence both sickness absence and presenteeism. Gaining a deeper insight of these interrelations is highly relevant for (personnel) managers and policy makers because it clearly makes a difference whether a factor aimed at reducing absence days is associated with more, unchanged or even fewer presenteeism days. While a decline in absence, c.p., is an economic improvement (for the manager, but of course also for the society), reducing absence at the cost of more presenteeism could reduce overall productivity, depending on the specific productivity effects of presenteeism.\footnote{See Schultz and Edington, 2007 for a survey on the productivity effects of presenteeism.} On the contrary, economic improvement clearly survives in
cases of unchanged or even lower presenteeism.

It is thus important to determine how different factors simultaneously affect both sickness states. This is in particularly true for factors that managers and policy makers can directly influence. For that reason, we focus on the influence of *work-related characteristics* on both sickness states. In our investigation, work-related characteristics comprise not only general features of the employer-employee relationship such as firm size, contract type, workload, work autonomy and others but also specific human resources management (HRM) practices such as piece rate, formalized performance evaluation or production targets. To highlight mechanisms through which both sickness states can be affected by work-related characteristics at the same time, we build a simple theoretical model. Subsequently, we use a rich data set in which indicators for sickness absence and presenteeism are compiled in one survey such that we are able to *simultaneously* analyze how work-related factors are correlated with both sickness absence and presenteeism.

In our theoretical model, the employee is hit by a non-binary health shock which entails disutility from work (see Brown and Sessions, 2004 for a similar approach). Comparing utility from being attendant at the workplace with that of being absent, the individual optimally makes her/his absence/attendance decision. This decision depends on work-related characteristics because they affect costs and returns of attendance. Presenteeism is given if an individual decides to be attendant at the workplace although s/he considers her/his health shock as sickness. We assume that an individual regards herself/himself as sick if disutility from work is relatively high and/or if productivity is relatively low. Because both determinants are affected by work-related factors, the individual-specific sickness definition and thus presenteeism can be influenced by those factors, too.

Our theoretical analysis shows that work-related characteristics can but not necessarily have to imply a substitutive relationship between absence and presenteeism, as is commonly assumed in the literature. Indeed, it can be possible that a change in one work-related factor leads to an increase (decrease) of both sickness states such that its interrelationship is complementary. Moreover, a change in one work-related factor might only influence one sickness state, leaving the other unaffected. The central mechanism behind these results is that presenteeism is affected through two channels. Suppose that a change in one work-related factor makes it more likely that
the individual is attendant at work, i.e. absence is reduced. Holding everything else constant, this would also increase the probability of presenteeism and hence imply a substitutive relationship. However, the same work-related factor could also change the individual’s sickness definition. If s/he evaluates fewer health shocks as sick, presenteeism could remain constant or could even decline as a response to the changing work-related factor. Furthermore, our theoretical model allows us to define illegitimate absence, i.e. absenteeism. In this case, employees are absent although they do not consider themselves sick.

In our empirical investigation, we estimate separately the relationship between work-related characteristics and the number of sickness absence and presenteeism days. For that purpose, we use the European Working Conditions Survey (EWCS), a cross-sectional survey which covers 34 European countries. This allows us to relate in OLS regressions 17 different work-related characteristics of more than 18,000 employees to their sickness absence and presenteeism behavior. Since there is no panel data on presenteeism available, we cannot deliver causal analysis of the interrelation between absence and presenteeism behavior, but our empirical investigation offers several improvements in other dimensions compared to the existing literature. We cover comprehensively work-related characteristics and use the annual number of days of absence and presenteeism, which is a better measure to investigate the relationship between both sickness states. In addition, we provide better external validity by using a large data set that is representative for each European country and Europe as a whole.

The main results of the empirical analysis are as follows: First, we find that only few work-related factors (namely supervisory status, over time and flexible working time) lead to a substitutive relationship between absence and presenteeism. This finding casts doubt on the predominant view in the literature that both sickness states are interlinked in a substitutive manner. Second, there are only two work-related factors (namely working conditions and tenure) which lead to a complementary relationship between absence and presenteeism.

3Specifically, we look at general work-related characteristics such as supervisory and blue collar status, temporary contracts, tenure categories, whether working part or over time, firm size, private sector employment, work autonomy, adverse working conditions, support by coworkers and the management, flexible working time and time pressure. Furthermore, we include HRM measures such as piece rate, formal performance evaluation and production targets.
and presenteeism. Specifically, averse working conditions and longer tenure are positively correlated with both sickness states. Third, the bulk of the considered work-related characteristics is only related to one of the two sickness states while being uncorrelated with the other. From a managerial and policy perspective, this suggests that it might be possible to reduce absence (by regulating or setting the respective work-related factors) without negative side-effects on presenteeism or to reduce presenteeism without the threat of higher absence. Finally, we can identify three factors (namely temporary contracts, employment in the private sector and firm size) that seem to affect illegitimate absence since they are only highly significantly related to absence while being unrelated to presenteeism. Our findings are robust against count data models and in differently defined sub-samples.

1.2 Related Literature

There are only few studies that look empirically at both sickness states and even fewer focusing on the interrelation between the two. Two Finnish studies investigate work-related determinants of both sickness states, but they do not focus on their interrelation and use binary measures for both sickness states (Böckerman and Laukkanen, 2009, 2010). In their first study, they find that only few determinants are related to both sickness states, be it complementary as shift work or substitutive as regular overtime. Only one of the two variables of interest is related to both sickness states in their follow-up study, while the other is only correlated with presenteeism. The match between desired and actual working hours is associated with a lower probability of sickness absence and presenteeism, whereas a strong emphasis of efficiency in the work place increases the likelihood of presenteeism. The data set, however, is not representative for the Finish workforce since it comprises only a small sample of Finnish trade union members. Furthermore, with employees deciding each day whether to come to work or stay at home, the relation of work-related factors to absence and presenteeism should be better measured in days than in incidence measures.

Johansson and Lundberg (2004) is the only study that explicitly investigates the substitution between sickness absence and attendance, which they refer to as ‘illness flexibility’. Contrary to their expectations, presenteeism and absence have only a substitutive relationship with regards to attendance requirements, but not with regards to adjustment latitude (the possibility to
adjust work effort when ill). The latter is positively related to the frequency of sickness absence for females, while not affecting presenteeism. There are several differences in regards to our study. First, they use a very selective sample that is taken from Stockholm county and additionally exclude all respondents that report neither absence nor presence behavior. This sample selection could lead to biased estimates, if the excluded observations are systematically related to the explanatory variables, which is quite likely. Second, their dependent variable is measured in four vaguely defined ordinal categories (never, once, a few times, many times). Finally, controlling only for age, health, financial situation and family demands, the authors do not sufficiently address potential omitted variable bias.

There are three theoretical studies analyzing sickness absence and presenteeism behavior. Brown and Sessions (2004) enhance the Barmby et al. (1994) model of absenteeism by including sickness presenteeism. While our approach is inspired by their model, we depart in two ways. We do not focus on shirking and detection technology but on the interrelation between sickness absence and presenteeism. More importantly, we expand their model by defining presenteeism through an endogenously determined definition of sickness.

Chatterji and Tilley (2002) build a principal-agent model in which the employer offers supra-mandatory sick pay to prevent unproductive presenteeism. Since they are interested in optimal contractual design, they define presenteeism from the employer’s perspective as a situation in which the employee is too unproductive. While we use a similar framework in which bad health negatively affects the firms’ profits, we use a wider definition of presenteeism that is closer to our observational data. Specifically, we focus on the employee’s perspective that does not only take productivity into consideration but also disutility from work. Furthermore, the critical level that defines sickness is a function of the work-related characteristics in our model and not constant across individuals and firms.

Hirsch et al. (2015) present a theoretical model to deduce hypotheses for presenteeism behavior which they empirically investigate with German cross-sectional data. In contrast to the literature and our study, they offer
a very specific theoretical definition of absenteeism, i.e. a situation in which attendance at the workplace of healthy employees is lower than under optimal incentives. This does neither match with sickness absence, as in our case, nor with illegitimate absence.

The remainder of our paper is structured as follows. In Section 2, we present our theoretical model and derive conditions under which work-related factors imply a substitutive, complementary or no relationship between absence and sickness presenteeism. The empirical analysis is conducted in Section 3. Section 4 concludes.

2 Theoretical Model

In this section, we build a model that formalizes the absence/attendance decision of individuals and shows under which conditions presenteeism is conceivable. Furthermore, the model formalizes the influence of work-related factors on the probability of absence and the probability of presenteeism.

2.1 Absence/Attendance

2.1.1 Utility

We consider an individual $i$ who is employed at one particular firm. By assumption, there is a contract between both which specifies the wage rate $w_i > 0$ and the working hours per day $h_i > 0$. The individual is confronted with a health shock $\delta_i$ which recurs on a daily basis. The random variable $\delta_i$ is distributed over the interval $[0, 1]$ with the density $f(\delta_i)$ and determines the individual’s privately known health. The higher $\delta_i$, the worse the health shock is (see Brown and Sessions, 2004, for a similar approach).

The individual $i$’s utility has two possible realizations. If s/he is attendant at the workplace, utility is given by:

$$U^h_i = u^h (w_i, T - h_i, X_i) - \pi(\delta_i, X_i),$$

(1)

while in case of the individual’s absence from the workplace, we have:

$$U^a_i = u^a (s_i, T).$$

(2)

The variable $T$ stands for the individual’s endowment in time and $s_i$ ($0 \leq$
\[ s_i \leq w_i \) denotes the exogenously given firm-financed sick pay which the individual receives in the case of absence. The sub-utility functions \( u^b \) and \( u^a \) are increasing in consumption \((w_i \text{ and } s_i)\) and leisure \((T - h_i \text{ and } T)\). We assume that work-related characteristics \( X_i \) affect \( u^b \) and hence utility from being attendant at work. In order to keep our framework as general as possible, we use \( X_i \) as a vector that subsumes all relevant work-related factors. Whether one particular work-related factor raises \( u^b \) or reduces it, is a priori unclear. For example, an improvement of the job security could increase \( u^b \), whereas an impairment of working conditions could decrease it.

The function \( \pi \) represents the disutility from work conditional on the individual’s health shock, with \( \pi(\delta_i = 0) = 0 \) and \( \partial \pi / \partial \delta_i > 0 \). Moreover, we presume that \( \pi \) depends on work-related characteristics. For example, disutility at any given health shock could decrease if working conditions are improved. Notably, we focus on the disutility that is conditional on attendance and neglect the general disutility of different health shocks that is independent of the absence/attendance decision (see Chatterji and Tilley, 2002 for a modeling approach in this respect).

2.1.2 The Individual’s Decision

Under which conditions does individual \( i \) decide to be absent (attendant) from (at) the workplace? To find the answer to this question, we define a threshold health level, denoted by \( \theta_i \), at which the individual \( i \) is indifferent between attendance and absence. Formally, this requires \( U^h_i(\delta_i = \theta_i) = U^a_i \). Using (1) and (2), we obtain:

\[ B \equiv u^h_i(w_i, T - h_i, X_i) - u^a_i(s_i, T) - \pi(\theta_i, X_i) = 0, \tag{3} \]

which implicitly determines \( \theta_i \).

\( ^5 \) After the health shock \( \delta_i \) of the individual \( i \) is revealed, s/he chooses to be attendant on this day if \( \delta_i \leq \theta_i \iff B \geq 0 \) holds; otherwise \( \delta_i > \theta_i \iff B < 0 \) s/he chooses to be absent.

In line with the theoretical literature on absence (see, for instance, Brown and Sessions, 1996), work-related factors affect the absence decision because they influence the utility costs and returns of being attendant (see (3)). For

\( ^5 \) Notably, we assume that \( 0 < u^a_i(s_i, T) < u^b_i(w_i, T - h_i, X_i) \) holds to ensure an interior solution.
example, better working conditions could reduce the disutility from work
and hence make absence less likely. This leads to the following Lemma:

**Lemma 1** The threshold health level above which the individual chooses to
be absent from the workplace is (among others) a function of \( X_i \) and hence
influenced by work-related factors: \( \theta_i = \theta_i(X_i) \).

Because the individual \( i \) knows \( \theta_i \) before the health shock is revealed,
it is possible to calculate the probability that the individual will be absent
from the workplace on a given day. Formally, the probability of absence is
given by \( A_i = \Pr(\theta_i < \delta_i < 1) = F(\theta_i < \delta_i < 1) \), where \( F(\delta_i) \) denotes the
distribution function of \( \delta_i \). Using the simplifying assumption of a uniform
distribution \( F(\delta_i) = \delta_i \), we obtain:

\[
A_i(X_i) = 1 - \theta_i(X_i).
\]

The probability of being attendant on a given day is given by \( H_i = 1 - A_i \).

**2.2 Presenteeism**

As standard practice in the literature, we define presenteeism as a situ-
ation where the individual is attendant at the workplace despite s/he is
sick (Arnold, forthcoming; Böckerman and Laukkanen, 2010; Hansen and
Andersen, 2008). In the previous subsection, we have shown under which
conditions individual \( i \) chooses to be attendant at the workplace but we
have not introduced a formal binary definition of sickness. We thus have
to determine under which conditions the individual considers her/his health
shock as sickness.

**2.2.1 Definition of Sickness**

In contrast to most of the existing literature, we assume that the definition
of sickness is endogenous and individual-specific. While the individual per-
ception of sickness might be shaped by several things such as social norms
and others, we focus here on two aspects that are related to work-related
characteristics: (i) the individual \( i \)'s evaluation of the health shock \( \delta_i \) in
terms of utility, denoted by \( I \), and (ii) the firm's evaluation of the individ-
ual's health shock in terms of productivity, denoted by \( F \). The individual
\( i \) combines both factors to define a critical level of health \( \phi_i \), above which
s/he considers her/his health shock $\delta_i$ as sickness. The threshold level $\phi_i$ can hence be interpreted as the individual’s sickness definition. Formally, this is given by:

$$
\phi_i = \phi_i (I, F).
$$

(5)

Let us have a closer look to the determinants of the sickness definition, starting with $I$. We assume that the individual’s own evaluation of her/his health shock is driven by the associated disutility from work. If disutility exceeds an exogenously given threshold $V_i$, the individual $i$ evaluates a given health shock as bad. At the margin, we can define a cutoff health shock $\delta_{ind}$ such that $u(\delta_i = \delta_{ind}, X_i) = V_i$ holds. Hence, any health shock that exceeds $\delta_{ind}$ is evaluated as bad by the individual. Because disutility from work is affected by work-related characteristics, the individual’s evaluation of the health shock is also influenced by $X_i$. If, for instance, working conditions get worse, disutility increases and $\delta_{ind}$ declines such that the individual evaluates lower levels of $\delta_i$ as bad health shocks.

Turning to the determinant $F$, we assume that the health shock is considered bad from the firm’s perspective if the profit in case of the worker’s attendance, $\Pi_i^h$, falls short of the profit in case of her/his absence, $\Pi_i^a$ (see Chatterji and Tilley, 2002 for a similar approach). As for the individual’s evaluation, we can define a cutoff health shock $\delta_{firm}$ at which $\Pi_i^h(\delta_i = \delta_{firm}) = \Pi_i^a$ holds. To calculate this cutoff, we have to specify the firm’s profit situation. We define total profits of the firm as:

$$
\Pi = \pi_i + Y_{-i} \quad \text{with} \quad Y_{-i} = \sum_{n=1}^{N_{-i}} \pi_n,
$$

(6)

where $\pi_i$ denotes the profit produced by individual $i$ and $Y_{-i}$ represents aggregate profits generated by all other $N_{-i}$ workers. Taking the behavior of the $N_{-i}$ workers as given, we obtain:

$$
\Pi_i^h = r(h_i, X_i) - \tau(\delta_i) - w_i + Y_{-i}, \quad \Pi_i^a = -s_i + Y_{-i}.
$$

(7)

(8)

Revenue per worker is denoted by $r$ and positively depends on the hours worked per day. In addition, it is also influenced by work-related characteristics (summarized in the vector $X_i$), but as for the utility functions, the sign
(and magnitude) of this effect depends on the specific work-related factor considered. The function \( \tau(\delta_i) \) measures the impact of the worker’s health shock on the firm’s revenue. For simplicity, we assume that revenues decrease if the health shock becomes worse due to reduced productivity and other related costs, with \( \tau(0) = 0 \) and \( \partial \tau / \partial \delta_i > 0 \). In case of absence, revenue of the individual is zero and hence not plagued by its health shock. Furthermore, the firm has to pay the sick pay \( s_i \) instead of the wage \( w_i \).

Inserting (7) and (8) into the indifference condition above yields:

\[
C = r(h_i, X_i) - r(\delta_{firm}) - (w_i - s_i) \geq 0.
\]

If \( \delta_i \leq \delta_{firm} \iff C \geq 0 \) holds, the firm evaluates the individual’s health shock as good or at least acceptable; otherwise \( \delta_i > \delta_{firm} \iff C < 0 \), the firm considers the health shock as bad. Note that the firm cannot observe \( \delta_i \) due to our assumption that this is the individual’s private information.

Because profits are affected by work-related characteristics, the firm’s evaluation also depends on \( X_i \). For example, better working conditions could raise profits in the case of the worker’s attendance, thereby implying that even high levels of \( \delta_i \) are not considered as bad health shocks from the firm’s perspective. This yields to the following Lemma:

**Lemma 2** The definition of sickness is individual-specific and (among others) a function of \( X_i \) such that it is influenced by work-related factors:

\[
\phi_i = \phi_i[I[\delta_{ind}(X_i)], F[\delta_{firm}(X_i)]].
\]

### 2.2.2 The Individual’s Decision

With the threshold health level \( \theta_i \) that determines the absence/attendance decision and the (individual-specific) definition of sickness \( \phi_i \) at hands, we are able to show under which conditions presenteeism is conceivable, given the optimal behavior of individual \( i \). Suppose that \( \theta_i > \phi_i \) holds and that the realized health shock of the individual lies in the interval \( \phi_i < \delta_i < \theta_i \).

---

6In general, bad health shocks reduce not only the worker’s productivity but have further negative effects on the firm’s profit. For example, one worker could infect other employees (see Barmby and Larguem, 2009) or could lead to distortions in situations with production interdependencies (team production) as suggested by Pauly et al. (2008), which both reduce profits. To keep our analysis as simple as possible, we do not model these influences on the productivity of co-workers.
As a consequence, s/he chooses to be attendant at the workplace because \( \delta_i \) is smaller than her/his threshold level, although s/he considers her/his health shock as sickness. The individual’s decision – attendance despite sickness – hence leads to a situation of presenteeism. Recall that there is a daily health shock, implying that we measure presenteeism on a daily basis.

Similar to the absence/attendance decision, we can also compute the probability of presenteeism on a given day. In general, this is given by

\[
P_i = \Pr(\phi_i < \delta_i < \theta_i) = F(\phi_i < \delta_i < \theta_i).
\]

Using \( F(\delta_i) = \delta_i \), we obtain:

\[
P_i(X_i) = \theta_i(X_i) - \phi_i(X_i).
\]

(10)

Suppose that instead \( \theta_i < \phi_i \) holds. Then, a health shock realization of \( \theta_i < \delta_i < \phi_i \) implies that the individual chooses to be absent although s/he does not consider her/his health shock as sickness. We define this situation as absenteeism of the individual. Note, however, that in a situation where absenteeism is possible, there is no presenteeism by construction.

2.3 The Impact of Work-related Factors

2.3.1 Theoretical Effects

Our model shows that the probabilities of absence and of presenteeism depend on the utility based threshold level \( \theta_i \) and on the individual-specific definition of sickness \( \phi_i \) [see (4) and (10)]. In turn, both are affected by variations in work-related factors [see Lemma 1 and Lemma 2]. Hence, we can use our model to shed light on the following question: How does a variation in a work-related factor – holding everything else constant – influence both the probability of absence and the probability of presenteeism per day? For notational convenience, we suppress the index \( i \) in the following.

Suppose that one particular work-related factor included in the vector \( X \) changes and denote this factor as \( x \in X \). In general, we can distinguish three cases. First, the variation of \( x \) implies a decrease (increase) in the absence probability, while the probability of presenteeism increases (decreases). Then, a change in \( x \) leads to a substitutive relationship between absence and presenteeism. Second, the change in \( x \) leads to an increase (or decrease) in both the absence and the presenteeism probability. Then, the change in \( x \) entails a complementary relationship between absence and presenteeism. Third, the variation in \( x \) is associated with a change (no
change) in the probability of absence, while the probability of presenteeism remains constant (changes). Then, $x$ leads neither to a substitutive nor a complementary relationship between presenteeism and absence.

To determine under which conditions a change in work-related factor $x$ leads to a substitutive, a complementary or no relationship between absence and presenteeism, recall that variations of $x$ influence $\theta$ and $\phi$. Using (4), we can show that the probability of absence increases (decreases) when $\theta$ decreases (increases):

$$dA = \frac{\partial A}{\partial \theta} \theta > (\leq) 0 \Leftrightarrow d\theta < (\geq) 0. \tag{11}$$

Regarding the probability of presenteeism, (10) indicates that changes in $\theta$ and $\phi$ influence $P$. If $d\theta < 0$ (and thus $dA > 0$) holds, we get:

$$dP = \frac{\partial P}{\partial \theta} \theta < 0 + \frac{\partial P}{\partial \phi} \phi < (\geq) 0 \Leftrightarrow d\theta > 0 \text{ or } d\phi < 0 \text{ or } d\phi \leq d\theta < 0. \tag{12}$$

If $d\theta \geq 0$ (and thus $dA \leq 0$) holds, we find:

$$dP = \frac{\partial P}{\partial \theta} \theta > 0 + \frac{\partial P}{\partial \phi} \phi > (\leq) 0 \Leftrightarrow d\phi \leq 0 \text{ or } 0 < d\phi < d\theta \text{ or } 0 < d\theta \leq d\phi. \tag{13}$$

With these conditions at hand, we obtain the following Propositions.\(^7\)

**Proposition 1** Presenteeism and absence are substitutes with respect to a work-related factor $x$ (i) if $\theta$ changes while $\phi$ is unaffected, (ii) if the variations in $\theta$ and $\phi$ are oppositional or (iii) if the changes of $\theta$ and $\phi$ have the same sign but the (absolute) change in $\phi$ is sufficiently weak.

**Proof.** A substitutive relationship requires $dA > (<) 0$ and $dP < (> ) 0$. From (11), we obtain $dA > (<) 0 \Leftrightarrow d\theta < (>) 0$. If $d\phi = 0$ holds, Eqs. (12) and (13) imply that $dP < (> ) 0$, which proves part (i) of the Proposition. For $d\phi > (<) 0$, we also find $dP < (> ) 0$ as stated in part (ii) of the Proposition.

\(^7\)As shown by (11), (12) and (13), the assumption of the uniform distribution implies that the partial derivatives equal 1 respectively –1. This simplifies the derived Propositions because changes in $\theta$ and $\phi$ are weighted by 1 (in absolute terms). In Appendix A.1, we prove that our qualitative findings are robust if the distribution of $\delta$ is not specified.
Finally, Eqs. (12) and (13) indicate that \( dP < (>)0 \) holds if the absolute change in \( \phi \) is lower than the absolute change in \( \theta \): \( d\theta < d\phi < 0 \) \((0 < d\phi < d\theta)\). This proves part (iii) of the Proposition. ■

**Proposition 2** Presenteeism and absence are complements with respect to a work-related factor \( x \) if the changes in \( \theta \) and \( \phi \) have the same sign and the (absolute) change in \( \phi \) is sufficiently strong.

**Proof.** A complementary relationship requires \( dA > (0) \) and \( dP > (0) \). Eq. (11) implies that \( dA > (0) \Leftrightarrow d\theta > (0) \). Observing (12) and (13), we find that \( dP > (0) \) holds if the absolute change in \( \phi \) is higher than the absolute change in \( \theta \): \( d\phi > d\theta > 0 \) \((0 < d\theta < d\phi)\). ■

**Proposition 3** Presenteeism and absence are neither substitutes nor complements with respect to a work-related factor \( x \) (i) if \( \theta \) remains constant while \( \phi \) changes or (ii) if the changes in \( \theta \) and \( \phi \) are identical.

**Proof.** There is no relationship between absence and presenteeism if \( dA = (\neq)0 \) and \( dP \neq (0) \) holds. From (11), we obtain \( dA = (\neq)0 \Leftrightarrow d\theta = (\neq)0 \). Given \( d\theta = 0 \), Eqs. (12) and (13) imply that \( dP \neq 0 \Leftrightarrow d\phi \neq 0 \), which proves part (i) of the Proposition. If \( d\theta \neq 0 \) holds, we see from (12) and (13) that \( d\theta = d\phi \) must hold in order to ensure \( dP = 0 \), which proves part (ii) of the Proposition. ■

These findings are based on the assumption \( \theta > \phi \). However, it can be the case that the reverse relation is true: \( \theta < \phi \). As discussed in section 2.2, there is then no presenteeism by construction, and we normalize its probability to zero: \( P \equiv 0 \Leftrightarrow \theta < \phi \). Note that in this scenario, the probability of absenteeism, i.e. illegitimate absence, is positive.

**Proposition 4** Presenteeism and absence are neither substitutes nor complements with respect to a work-related factor \( x \) if \( \theta < \phi \) holds.

### 2.3.2 Empirical Examination

Theoretically, the effects of work-related factors on absence and presenteeism are ambiguous, as shown by Propositions 1-4. Therefore, it is necessary to investigate these effects empirically.\(^8\) In the following, we use data from a

\(^8\)Note that, given the general form of the utility functions and the definition of sickness, our theory cannot be used to find clear-cut predictions whether a particular work-related factor leads to a substitutive, complementary or no relationship between absence and presenteeism.
cross-sectional survey where individuals self-report their absences and presenteeism days during the last 12 months. In addition, we make use of a rich set of work-related factors to analyze how they are – on average – related to absence and presenteeism days.

With our theory at hand, we are able to disentangle the channels through which work-related factors affect absence and presenteeism days in the data. In this regard, three remarks have to be made. First, we interpret the annual number of days in the data as the aggregated realization of daily absence decisions in the model. Second, we do not model inter-temporal feedback effects from presenteeism today to the health shock tomorrow to make our theoretical model as simple as possible. In the annual data, of course, such effects could not, at the first glance, be ruled out. To mitigate this contradiction between theory and data, we make use of the particularly rich health information available in the data set. An additional reason to accommodate for the heterogeneity in health in the data as well as we can is that our theoretical model is based on changes in the critical levels $\phi$ and $\theta$ for a given health status $\delta$. Third, we are not able to empirically investigate intra-individual changes as described in the theoretical model because we use cross-sectional data. Instead, we analyze differences between individuals with regards to their absence and presenteeism behavior.

Suppose that a change in one work-related factor leads to more absence days and less presenteeism days on average and implies hence a substitutive relationship between both. The increase in absence can be explained by a decline in the utility based threshold level $\theta$. Holding everything else constant, more absence implies that presenteeism is automatically reduced. This effect on presenteeism survives if the sickness definition $\phi$ is unchanged and is even amplified if the individual considers fewer health shocks as sick, i.e. $\phi$ increases.

It can, however, also be possible that a change in one work-related factor implies more absence and more presenteeism days on average. If we would find this complementary relationship in our data, the above mentioned change in absence has to be complemented by a sufficiently strong decrease in the definition of sickness $\phi$. Then, individuals are more likely
absent (due to lower $\theta$) which, c.p., reduces presenteeism, but also consider more health shocks as sick (due to lower $\phi$) which, c.p., increase presenteeism. If the latter effect is strong enough, a changing work-related factor could induce a simultaneous increase in absence and presenteeism days.

The endogenous definition of sickness also helps us to explain why a change in one work-related factor could lead to more presenteeism days (absence days), while leaving absence days (presenteeism days) unaffected. This would be the case if the utility based threshold level $\theta$ does not vary with the work-related factor while the definition of sickness $\phi$ changes (absence remains constant) or if both are affected in the same way (presenteeism remains constant). In addition, the constancy of presenteeism could also be explained if the work-related factor has an impact on absence and presenteeism. This would be the case if the utility based health threshold level falls short of the threshold level that defines sickness, implying that individuals choose to stay at home although they do not consider themselves as sick.

3 Empirical Analysis

3.1 Data and Empirical Strategy

To analyze the impact of work-related characteristics on the relationship between absence and presenteeism empirically, we use the fifth wave of the EWCS, a repeated cross-sectional survey on working conditions in Europe. The EWCS is conducted every five years by an agency of the European Union and profits from a single questionnaire guaranteeing consistent data across countries. In 2010, the EWCS covered for the first and only time an item on sickness presenteeism and is hence the first large-scale survey integrating information about sickness absence and presenteeism behavior. It comprises the population aged 15 and above living in 34 European countries. In our investigation, we consider employees aged 18-65 years who have been employed during the last 12 months prior to the interview and who have been working at least 10 hours per week, excluding the self-employed, students, apprentices, employees without work contracts and those working two jobs.\textsuperscript{10}
As the dependent variables, we use the annual duration of absence and presenteeism measured in days. The sickness absence item reads as follows: “Over the past 12 months how many days in total were you absent from work for reasons of health problems?” The sickness presenteeism item asks: “Over the past 12 months did you work when you were sick? a) Yes b) No. If yes, how many working days?”

We prefer the annual duration of absence and presenteeism over incidence or frequency measures for several reasons. First, the productivity effects of sickness depend among others on how long the employee’s productivity is impaired by the sickness (see Pauly et al., 2008). Hence, the annual duration of absence and presenteeism is a better proxy for productivity effects than their frequency or incidence. Second, since employees decide each day whether to come to work or stay at home, the relation of work-related factors to absence and presenteeism is better measured in days than in frequency or incidence measures. Furthermore, it fits well with our theoretical model describing daily absence decisions. Finally, asking for the number of sickness presenteeism and absence days in an open question is better than offering predefined frequency categories (as done by Johansson and Lundberg, 2004) because this avoids the establishment of reference points. Note that we disregard outliers, i.e. those with either more than 50 sickness presenteeism or 100 absence days within 12 months, resulting in a loss of around 150 observations. However, the central results do not depend on this sample selection (see section 3.3). We also exclude those 30 employees with a very bad health status. In total, the number of observations amounts to 18,804.

The descriptive statistics provided in Table 1 show that sickness absence and presenteeism is a widespread and quantitatively relevant phenomenon in Europe. In our sample, the average number of sickness presenteeism and absence days amounts to 2.3 and 5.2, respectively. Conditioning on those who go to work while sick at all leads to an average of almost seven presenteeism days. For absence, this number amounts to more than ten absence days. There is a large cross-sectional variation with standard deviations of more than five presenteeism days and over ten absence days.\textsuperscript{11}

Since we are interested in how work-related characteristics are related to
\textsuperscript{11}The distribution of the conditional sickness presenteeism and absence days is shown in Figure A.1 in the Appendix A.2.
Table 1: Descriptive Statistics

<table>
<thead>
<tr>
<th>Variable</th>
<th>Obs.</th>
<th>Mean</th>
<th>Std. Dev.</th>
<th>Min.</th>
<th>Max.</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sickness presenteeism days</td>
<td>18,804</td>
<td>2.25</td>
<td>4.8</td>
<td>0</td>
<td>50</td>
</tr>
<tr>
<td>Sickness absence days</td>
<td>18,804</td>
<td>5.23</td>
<td>10.6</td>
<td>0</td>
<td>100</td>
</tr>
<tr>
<td><strong>Work-related characteristics</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Tenure (&lt;1 years)</td>
<td>18,804</td>
<td>0.07</td>
<td>0.26</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Tenure (1-2 years)</td>
<td>18,804</td>
<td>0.17</td>
<td>0.37</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Tenure (3-14 years)</td>
<td>18,804</td>
<td>0.50</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Tenure (≥ 15 years)</td>
<td>18,804</td>
<td>0.26</td>
<td>0.44</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Flexible schedule</td>
<td>18,804</td>
<td>0.04</td>
<td>0.20</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Work autonomy index</td>
<td>18,804</td>
<td>1.92</td>
<td>1.19</td>
<td>0</td>
<td>3</td>
</tr>
<tr>
<td>Supervisor</td>
<td>18,804</td>
<td>0.16</td>
<td>0.37</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Part time (10-30h)</td>
<td>18,804</td>
<td>0.12</td>
<td>0.33</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Over time (≥ 45h)</td>
<td>18,804</td>
<td>0.12</td>
<td>0.32</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Lack of time to get work done</td>
<td>18,804</td>
<td>2.11</td>
<td>1.0</td>
<td>1</td>
<td>5</td>
</tr>
<tr>
<td>Adverse working conditions</td>
<td>18,804</td>
<td>2.81</td>
<td>2.60</td>
<td>0</td>
<td>15</td>
</tr>
<tr>
<td>Social support</td>
<td>18,804</td>
<td>5.60</td>
<td>1.93</td>
<td>0</td>
<td>8</td>
</tr>
<tr>
<td>Blue collar worker</td>
<td>18,804</td>
<td>0.33</td>
<td>0.47</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Temporary contract</td>
<td>18,804</td>
<td>0.12</td>
<td>0.33</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Size (&lt;10 employees)</td>
<td>18,804</td>
<td>0.27</td>
<td>0.44</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Size (10-49 employees)</td>
<td>18,804</td>
<td>0.34</td>
<td>0.47</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Size (50-99 employees)</td>
<td>18,804</td>
<td>0.13</td>
<td>0.33</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Size (100-249 employees)</td>
<td>18,804</td>
<td>0.11</td>
<td>0.31</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Size (≥250 employees)</td>
<td>18,804</td>
<td>0.15</td>
<td>0.37</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Private sector</td>
<td>18,804</td>
<td>0.67</td>
<td>0.47</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Piece rate</td>
<td>18,804</td>
<td>0.12</td>
<td>0.33</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Production target</td>
<td>18,804</td>
<td>0.43</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Performance assessment</td>
<td>18,804</td>
<td>0.45</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
</tr>
</tbody>
</table>

Source: 2010-EWCS. Own calculations, survey weights used.

absence and presenteeism behavior, we have to select specific work-related factors and cover them empirically. In our cross-sectional model, this selection of explanatory variables is key. Therefore, we guide our choice by the literature on sickness absence behavior (Frick and Malo, 2008; Livanos and Zangelidis, 2013; Puhani and Sonderhof, 2010; Ziebarth and Karlsson, 2010) and by the literature on sickness presenteeism (Arnold, forthcoming; Aronsson et al., 2000; Aronsson and Gustafsson, 2005; Böckerman and Laukkanen, 2009, 2010; Hansen and Andersen, 2008; Leineweber et al., 2011; Preisendörfer, 2010). Furthermore, we are interested to see whether HRM measures reduce absence at the cost of presenteeism. We present a parsimonious model as preferred specification excluding some insignificant explanatory variables, which does not change our results (see section 3.3).

In our main specification, we include rather formal job characteristics
such as supervisory and blue collar status, temporary contracts, tenure categories (1-2 years, 3-14 years, ≥ 15 years), part time (10-30h per week), over time (≥ 45h per week), flexible working time, firm size, industry (modified NACE-17 classification) and sector information (private sector). To measure the working conditions, we construct an index measuring adverse working conditions. The index is a sum of dummy variables measuring whether the individual is exposed half of the time or more to a number of adverse working conditions. These comprise: exposure to vibrations from hard tools or machinery; to noise; high temperatures; low temperatures; breathing in smoke, fumes, powder or dust; breathing in vapours; handling or being in skin contact with chemical products or substances; breathing tobacco smoke from other people; handling or being in direct contact with materials which can be infectious; having a job that involves tiring or painful positions; lifting or moving people; carrying or moving heavy loads; standing; performing repetitive hand or arm movements; and handling angry clients or patients. Additionally, we take subjective properties of a job into account such as work autonomy, support by co-workers and the management, and time pressure (lack of time to get work done). Finally, we include three HRM measures, namely piece- or productivity pay, production targets that determine work speed and individual formalized performance assessment. The corresponding descriptive statistics of the two dependent variables and the work-related characteristics are provided in Table 1.

Additionally to the work-related characteristics, we control for sociodemographic variables and health status (for descriptives of the controls see Table A.1 in the Appendix A.2). As sociodemographic variables, we include sex (female=1), having children, living with a partner, age categories (aged 18-24, 25-34, 35-44, 45-54, and 55-65 years), educational status (primary, secondary and higher education status) and having financial problems as a proxy for the household income.

As argued in section 2.3.2, it is important to filter out individual heterogeneity in health as well as possible. Fortunately, the EWCS data offers a

\footnote{Work autonomy is captured by an index measuring the number of autonomy dimensions in which the employee has control – specifically, work order, methods and speed. The other subjective variables are measured on different Likert scales (see Table 1).}

\footnote{The EWCS includes also a crude net income measure that exhibits many non-responses. Including this variable does not change our findings. The respective estimations are available upon request to the authors.}
rich set on health information that goes far beyond those included in other labor market surveys. Hence, we do not only control for self-rated general health (very good, good, fair and bad), which alone is already a very good predictor of general health and mortality (Idler and Benyamini, 1997), but we also include a set of dummy variables that measures specific diseases or health issues from which the respondents have suffered within the last 12 months. The dummies on specific diseases and symptoms include: hearing problems, skin problems, backache, muscular pain in shoulders, neck and/or upper limbs, muscular pain in lower limbs, headaches and eyestrain, stomach ache, respiratory difficulties, cardiovascular diseases, injuries, depression or anxiety, overall fatigue, insomnia or general sleeping problems and other health problems. Since absence and presenteeism behavior might not only be affected by the respondent’s physical health, we also include a mental health index. This index is a sum of five Likert scales measuring for the last two weeks whether the respondent has ‘felt cheerful and in good spirits’, ‘felt calm and relaxed’, ‘felt active and vigorous’, ‘woken up feeling fresh and rested’, and whether her or his ‘daily life has been filled with things that interest [her or him]’. The index runs from zero to 25 with higher values indicating better mental health.

Finally, we include country dummies in order to account for average aggregated country differences in absence and presenteeism behavior that might be due to country specific effects such as labor market institutions, social norms, or health care institutions. In doing so, we indirectly control also for the generosity of sick pay entitlements that strongly vary across countries and affect absence behavior (Frick and Malo, 2008; Puhani and Sonderhof, 2010; Ziebarth and Karlsson, 2010).14

To find out whether a change of a work-related factor implies a substitutive, complementary or no relationship between sickness presenteeism and absence, we investigate separately how they are related to the number of presenteeism and absence days and classify them accordingly. A substitutive relationship between presenteeism and absence days is given if a work-related factor leads to an opposite change in these two sickness states. For a complementary relationship between presenteeism and absence days,

---

14 We prefer country dummies in cross-sectional data since they are a general way to control for cross country differences. Furthermore including both country dummies and sick pay generosity is not feasible due to multi-collinearity.
a work-related factor affects both sickness states at the same time either positively or negatively. Finally, if a work related factor is either significantly related to sickness presenteeism or absence, its relationship is neither substitutive nor complementary.

We investigate the relationship between the work-related characteristics and the number of presenteeism and absence days by estimating OLS regression models with cluster adjusted standard errors at the country level. Additionally, we present the results from seemingly unrelated regression models (SUR) which allow for the correlation of the error terms between both equations and hence take advantage of the fact that we observe absence and presenteeism behavior of the same individuals (for SUR or system OLS, see Wooldridge, 2002, pp. 143ff.). Since there is no panel data on presenteeism available at the moment, we can offer only cross-sectional correlations, which should be kept in mind when interpreting the ensuing results.

Despite having a count data structure, we prefer OLS to count data models since they are less contingent on distributional assumptions and easier to interpret (for count data models, see Cameron and Trivedi, 1998, pp. 59ff.). Particularly, assuming an average linear relationship between independent and dependent variable, OLS models make it easier to draw a general picture with regards to a substitutive, a complementary or no relationship than count data models where there might be different effects at different points on the distribution. To be sure that our results do not depend on this simplifying assumption, we present a Poisson regression model as a robustness check in section 3.3.

The econometric model reads as follows:

\[
\text{absence/presenteeism days}_i = \alpha_0 + \text{workcharacteristics}_i' \alpha_1 + \text{sociodemographics}_i' \alpha_2 + \text{healthstatus}_i' \alpha_3 + \text{country}_i' \alpha_4 + \epsilon_i.
\]

Presenteeism/absence days indicate the number of days either spent at work while being sick or absent during 12 months for individual \(i\). Work characteristics, health status, and sociodemographics represent the different vectors of independent variables. We include country dummies and \(\epsilon_i\) is the error term.
3.2 Results

The regression outcomes are depicted in Table 2, in which we present the determinants of absence and presenteeism days in columns (1) and (2), respectively. Since the coefficients in the OLS models are by construction the same as in the SUR models, we present additionally the z-statistics of the SUR models in squared brackets, which deliver virtually the same qualitative results as the OLS models with one exception.

As our first result, we find that there are only few work-related variables that lead to a substitutive relationship. This is a very remarkable finding because it is at odds with the common view in the literature that employees’ choice between absence and presenteeism is a zero-sum game. Furthermore, it suggests that the individual-specific definition of sickness – the main innovation of our theoretical model – plays an important role because non-substitutive relationships between both sickness states can only be explained by sufficiently large changes in the sickness definition, as explained in section 2.3.2.

More specifically, being supervisor, working more than 45 hours per week and being autonomous about the working time are associated with less absence days at the expense of more presenteeism days. This group could be described as career-oriented employees because they work long hours and take responsibility for their firm as supervisor, but have also autonomy over their working time. Hence, one could argue that these employees have incentives to come to work at a more severe health shock but without adjusting their sickness definition. Consequently, absence days decline while presenteeism days increase. Note that the flexible time variable becomes insignificant in the presenteeism regression in the SUR model.
## Table 2: Regression Results

<table>
<thead>
<tr>
<th></th>
<th>(1) Absence (OLS)</th>
<th>(1) Absence [SUR]</th>
<th>(2) Presenteeism (OLS)</th>
<th>(2) Presenteeism [SUR]</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Substitutes</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Supervisor</td>
<td>-0.503***</td>
<td>(-2.19) [-2.12]</td>
<td>0.314***</td>
<td>(3.12) [2.91]</td>
</tr>
<tr>
<td>Over time (≥ 45 h)</td>
<td>-0.831***</td>
<td>(-3.30) [-3.19]</td>
<td>0.485***</td>
<td>(4.43) [4.09]</td>
</tr>
<tr>
<td>Flexibel time arr.</td>
<td>-1.225***</td>
<td>(-3.90) [-2.79]</td>
<td>0.223**</td>
<td>(2.11) [1.12]</td>
</tr>
<tr>
<td><strong>Complements</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adverse working condit.</td>
<td>0.233***</td>
<td>(4.63) [6.04]</td>
<td>0.058**</td>
<td>(2.45) [3.30]</td>
</tr>
<tr>
<td>Tenure (&lt;1 years) (base)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Tenure (1-2 years) (base)</td>
<td>1.085***</td>
<td>(3.54) [2.98]</td>
<td>0.597***</td>
<td>(3.28) [3.61]</td>
</tr>
<tr>
<td>Tenure (3-14 years) (base)</td>
<td>1.728***</td>
<td>(4.29) [4.96]</td>
<td>0.702***</td>
<td>(3.52) [4.41]</td>
</tr>
<tr>
<td>Tenure (≥ 15 years)</td>
<td>1.671***</td>
<td>(3.46) [4.26]</td>
<td>1.047***</td>
<td>(4.66) [5.88]</td>
</tr>
<tr>
<td><strong>Only absence</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Private sector</td>
<td>-0.928***</td>
<td>(-4.99) [-4.19]</td>
<td>0.014</td>
<td>(0.18) [0.14]</td>
</tr>
<tr>
<td>Temporary contract</td>
<td>-0.960***</td>
<td>(-3.61) [-3.46]</td>
<td>-0.069</td>
<td>(-0.57) [-0.54]</td>
</tr>
<tr>
<td>Part time (10-30h)</td>
<td>-0.685*</td>
<td>(-1.80) [-2.44]</td>
<td>-0.146</td>
<td>(-0.78) [-1.15]</td>
</tr>
<tr>
<td>Size (&lt;10 employees) (base)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Size (10-49 employees) (base)</td>
<td>0.719***</td>
<td>(3.57) [3.43]</td>
<td>0.084</td>
<td>(0.76) [0.88]</td>
</tr>
<tr>
<td>Size (50-99 employees)</td>
<td>0.655**</td>
<td>(2.39) [2.35]</td>
<td>0.007</td>
<td>(0.06) [0.06]</td>
</tr>
<tr>
<td>Size (100-249 employees)</td>
<td>1.045***</td>
<td>(3.20) [3.51]</td>
<td>0.287*</td>
<td>(1.95) [2.12]</td>
</tr>
<tr>
<td>Size (≥250 employees)</td>
<td>1.418***</td>
<td>(4.19) [4.95]</td>
<td>0.135</td>
<td>(1.14) [1.03]</td>
</tr>
<tr>
<td><strong>Only presenteeism</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Work autonomy</td>
<td>-0.083</td>
<td>(-0.99) [-1.11]</td>
<td>0.156***</td>
<td>(3.16) [4.62]</td>
</tr>
<tr>
<td>Lack of time</td>
<td>-0.106</td>
<td>(-0.89) [-1.20]</td>
<td>0.157***</td>
<td>(3.04) [3.92]</td>
</tr>
<tr>
<td>Production targets</td>
<td>-0.013</td>
<td>(-0.07) [-0.07]</td>
<td>0.198***</td>
<td>(2.90) [2.43]</td>
</tr>
<tr>
<td>Performance assessment</td>
<td>0.065</td>
<td>(0.28) [0.38]</td>
<td>0.218**</td>
<td>(2.25) [2.79]</td>
</tr>
<tr>
<td>Social support</td>
<td>-0.025</td>
<td>(-0.39) [-0.53]</td>
<td>-0.085***</td>
<td>(-3.69) [-4.03]</td>
</tr>
<tr>
<td>Blue collar</td>
<td>0.224</td>
<td>(0.92) [0.95]</td>
<td>-0.186*</td>
<td>(-1.92) [-1.75]</td>
</tr>
<tr>
<td><strong>Insignificant</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Piece or product. rate</td>
<td>0.116</td>
<td>(0.46) [0.45]</td>
<td>-0.079</td>
<td>(-0.75) [-0.68]</td>
</tr>
<tr>
<td><strong>Control variables</strong></td>
<td>Yes</td>
<td>Yes</td>
<td></td>
<td></td>
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<tr>
<td><strong>Health controls</strong></td>
<td>Yes</td>
<td>Yes</td>
<td></td>
<td></td>
</tr>
</tbody>
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<table>
<thead>
<tr>
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<th>(1)</th>
<th>(2)</th>
</tr>
</thead>
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<tr>
<td>N</td>
<td>18804</td>
<td>18804</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.12</td>
<td>0.14</td>
</tr>
</tbody>
</table>

Source: 2010-EWCS, own calculations. Notes: Coefficient estimates are from OLS regressions. The dependent variables are the number of sickness absence days in column (1) and the number of sickness presenteeism days in column (2), including those with zero days. All variables shown in the table except adverse working conditions, lack of time, work autonomy and social support are dummies (see Table 1 for descriptive statistics). Sociodemographic variables as well as country and industry dummies are included as control variables but are not shown. The sociodemographic controls comprise sex, having children, partnership status, age categories, financial problems and educational status. The health control variables include self rated health status (3 dummies), 14 different symptoms and diseases from which the respondent suffered during the last 12 months as well as an index measuring mental health. T-statistics based on standard errors clustered on the country level are in parentheses, z-statistics from SUR regression in squared brackets. The correlation coefficient of the residuals in the SUR amounts to 0.0977. * $p < 0.1$, ** $p < 0.05$, *** $p < 0.01$. 
The second result of our empirical investigation is that only a few complementary relationships exist between absence and presenteeism days with respect to work-related factors.\textsuperscript{15} To be more specific, adverse working conditions and tenure are positive and significantly related to both sickness categories. Accordingly, longer tenure as well as worse working conditions give employees an incentive to stay at home even for less severe health shocks which raises absence days. At the same time, the employees’ definition of sickness is widened by both factors such that more health shocks are considered as sick which raises presenteeism days.

As shown in Table 2, the majority of the work-related factors is either related to sickness absence days or to presenteeism days while being uncorrelated with the other. Let us first look at the case in which absence behavior changes while presenteeism remains unaffected. Working under a temporary contract and in the private sector are associated with significantly less absence days (part time only at the ten per cent significance level), while working in a larger plant is positively related to the number of absence days.

In our theoretical model, this pattern can be explained if there is illegitimate absence instead of presenteeism. Individuals then decide to be absent from the workplace without considering the respective health shock as sick. Accordingly, temporary contracts, employment in the private sector and in smaller firms make particularly illegitimate absence more costly. This might be explained by less employment protection in these jobs which has been shown to affect absence behavior (cf. Riphahn, 2004; Ichino and Riphahn, 2005). Although there is a reduction of absence, these characteristics do not, in this case, come at the cost of increased presenteeism. With data collected in 2010, on the peak of the economic crisis in Europe with its increasing unemployment and general economic insecurity, which should increase effort in form of presenteeism particularly among those with low employment protection, it is a strong result that we find only differences in absence but none in presenteeism.

Finally, we consider the case where presenteeism changes while absence is unaffected. As shown in Table 2, subjective work load (lack of time),

\textsuperscript{15}This lack in substitutes and complements is not a specific result for the work-related characteristics but does hold also true for the sociodemographics. Here, only sex leads to a complementary relationship between absence and presenteeism. Self-rated health status and some of the specific illness dummies are significantly related to both sickness states, too.
work autonomy and two of the HRM measures (production target and formal performance assessment) are highly significantly associated with more sickness presenteeism days, while support by colleagues and management is highly significantly associated with less (blue collar only at the ten per cent level). As our theoretical model explains this pattern by differences in the individual sickness definition, these results suggest that pressure on the employee (work load and performance assessment) and support of the employee affect perceived sickness in an oppositional way. From the perspective of a manager, the existence of this category is interesting because it highlights possibilities to reduce presenteeism without the threat of higher absence.

### 3.3 Robustness Checks

As robustness checks, we use count data models as well as OLS models in differently defined (sub)samples and provide the respective estimation results in Appendix A.2. Estimating a Poisson regression model confirms our results from the OLS regressions (see columns 1 and 2 in Table A.2). Restricting our sample to EU member states which are characterized by a more homogeneous institutional setting (columns 3 and 4 in Table A.2) does not fundamentally alter our results, either. Only the part time variable becomes insignificant in this sample.

As a further robustness check, we restrict our sample to employees that have been sick during the last 12 months because one can only substitute between sickness states when being sick at all (columns 5 and 6 in Table A.2). In this sample, we find one additional substitutive relationship, namely work autonomy. Here, work autonomy is associated with less absence and more presenteeism days, while the qualitative findings for the other controls remain mostly stable. Being associated with responsibility and intrinsic motivation, work autonomy fits quite well into the group of the other substitutes.

Finally, we do some further sensitivity checks that are not presented in Table A.2. Including (some) of the outliers, i.e. those with up to 150 absence or presenteeism days in 12 months, does not fundamentally change our results. Specifically, flexible working time becomes insignificant in the presenteeism regression, while supervisor is only significant at the ten per

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16 The respective estimation results are available upon request from the authors.
cent level in both regressions. In contrast, blue collar status becomes highly statistically significant in the presenteeism regression. Hence, only over time and supervisor status – albeit less significantly – remain substitutes in these models. This could be seen as a sign that substitution plays rather a role in a sample of employees with less severe health issues which allows them to actually reduce absence at the cost of presenteeism.

To see whether our results are sensible to the exclusion of employees with particularly severe health issues, we rerun our models excluding employees having suffered during the last twelve months from a.) injuries (around eight per cent of the sample) or b.) other (non listed) health problems (around three per cent).\footnote{These are the two least prevalent illness categories that are particularly strongly related to both sickness states. Excluding those with cardiovascular or respiratory disease does not qualitatively change our results either.} Again, the results are virtually unchanged in these two sub-samples. Only flexible working time becomes insignificant in both presenteeism models. Finally, results are not qualitatively affected by the inclusion of further work-related characteristics that turned out to be insignificant at conventional levels and hence were excluded from our preferred specification. The additional controls comprise work interdependence (whether work speed depends on others), whether working during evenings or weekends, a categorical income measure (21 categories) and perceived job insecurity (5 point Likert scale).

## 4 Summary

How are work-related factors simultaneously related to sickness absence and presenteeism? To address this question, we explore a theoretical and empirical analysis. Using cross-sectional data from the EWCS, we find in OLS regressions that only three out of 17 work-related factors, namely supervisor status, over time and flexible working time, show a substitutive relationship between absence and presenteeism. This finding is at odds with the predominant view in the literature that presenteeism behavior is simply the residuum of sickness absence. In addition, there are also only two work-related factors, namely working conditions and tenure, for which a complementary relationship between both sickness states can be observed.

The bulk of work-related factors is only related to one of the sickness states while leaving the other unaffected. This finding shows that it might
be possible to reduce either absence or presenteeism without shifting the negative productivity effect of sickness to the other sickness state, hence possibly raising the overall number of unimpaired working days. Finally, we are able to identify situations which are associated with absenteeism and that would benefit from according measures. This is the case for employees in the public sector, in larger firms and with open-ended contracts who are associated with more absence but not fewer presenteeism days. Our results are remarkably robust against count data models and in differently defined sub-samples.

In our theoretical model, we show that non-substitutive relationships between absences and presenteeism (with respect to work-related factors) can be explained by shifts in the individual-specific definition of sickness. If, for instance, a change in one work-related factor implies more absence days, this would, c.p., be combined with reduction of presenteeism days. This effect interacts, however, with a change in the sickness definition. If the same work-related factor implies that individuals consider more health shocks as sick, presenteeism days would, c.p., increase. If the latter effect is strong enough, absence and presenteeism days could simultaneously rise with respect to the work-related factor. Presenteeism days remain unchanged if the aforementioned effects cancel out or if there is absenteeism. If, in turn, the sickness definition changes but the individuals’ absence decision does not vary, only presenteeism days will adjust. The fact that the non-substitutive relationship between absence and presenteeism is the dominant empirical finding thus indicates the importance of the individual-specific sickness definition.

A Appendix

A.1 Theoretical Model

As in the main text, we assume that $\delta$ is randomly distributed over the interval $[0, 1]$ with the density $f(\delta)$, where we suppress the index $i$ for notational simplification. The distribution function is denoted by $F(\delta)$, with $F(0) = 0$, $F(1) = 1$ and $\partial F/\partial \delta > 0$. The probability of absence reads:

$$A = P(\theta < \delta < 1) = 1 - F(\theta),$$
which implies:
\[
\frac{\partial A}{\partial \theta} = -\frac{\partial F}{\partial \theta} < 0. \tag{A.1}
\]

The probability of presenteeism is given by:
\[
P = F(\phi < \delta < \theta) = F(\theta) - F(\phi),
\]

which leads to:
\[
\frac{\partial P}{\partial \theta} = \frac{\partial F}{\partial \theta} > 0 \quad \text{and} \quad \frac{\partial P}{\partial \phi} = -\frac{\partial F}{\partial \phi} < 0. \tag{A.2}
\]

With (A.1) and (A.2) at hand, we can rewrite (11) as:
\[
dA = \frac{\partial A}{\partial \theta} d\theta > (\leq) 0 \Leftrightarrow d\theta < (\geq) 0. \tag{A.3}
\]

If \(d\theta < 0\) (and thus \(dA > 0\)) holds, we get from (12):
\[
dP = \left[\frac{\partial P}{\partial \theta} \frac{d\theta}{>0} < 0 + \frac{\partial P}{\partial \phi} \frac{d\phi}{<0} < 0\right] > 0 \Leftrightarrow \frac{d\phi}{<0} \leq 0 \Leftrightarrow |\Gamma_1| < |\Gamma_2|,
\]

with \(\Gamma_1 \equiv (\partial P/\partial \theta)d\theta\) and \(\Gamma_2 \equiv (\partial P/\partial \phi)d\phi\). If instead \(d\theta \geq 0\) (and thus \(dA \leq 0\)) holds, we find [see (13)]:
\[
dP = \left[\frac{\partial P}{\partial \theta} \frac{d\theta}{>0} \geq 0 + \frac{\partial P}{\partial \phi} \frac{d\phi}{<0} > 0\right] > 0 \Leftrightarrow \frac{d\phi}{<0} \geq 0 \Leftrightarrow |\Gamma_1| > |\Gamma_2|,
\]

Observing (A.4) and (A.6), we find that presenteeism and absence are substitutes (i) if the variations in \(\theta\) and \(\phi\) are oppositional or (ii) if the changes of \(\theta\) and \(\phi\) have the same sign but the weighted (absolute) change in \(\phi\) is sufficiently weak. This finding corresponds with Proposition 1. In addition, (A.5) and (A.7) imply that presenteeism and absence are com-
plements if the changes in $\theta$ and $\phi$ have the same sign and the weighted (absolute) change in $\phi$ is sufficiently strong. This confirms Proposition 2. Finally, (A.5) and (A.7) also show that presenteeism and absence are neither substitutes nor complements (i) if $\theta$ remains constant while $\phi$ changes or (ii) if the weighted changes in $\theta$ and $\phi$ are identical, which is equivalent to Proposition 3.

A.2 Empirical Analysis

Figure A.1: Distribution of sickness absence and presenteeism days conditional on absence and presenteeism. Observations with zero sickness absence and presenteeism days not shown but included in analysis (53 and 62 % of the full sample, respectively). Source: 2010-EWCS. Own calculations.
<table>
<thead>
<tr>
<th>Variable</th>
<th>Obs.</th>
<th>Mean</th>
<th>Std. Dev.</th>
<th>Min.</th>
<th>Max.</th>
</tr>
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<tbody>
<tr>
<td>Socio-demographic controls</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sex (female=1)</td>
<td>18,804</td>
<td>0.45</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
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<tr>
<td>Children</td>
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<td>0.53</td>
<td>0.50</td>
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<tr>
<td>Partnership</td>
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<td>0.44</td>
<td>0</td>
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<tr>
<td>Age (18-24 years)</td>
<td>18,804</td>
<td>0.07</td>
<td>0.26</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Age (25-34 years)</td>
<td>18,804</td>
<td>0.25</td>
<td>0.44</td>
<td>0</td>
<td>1</td>
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<tr>
<td>Age (35-44 years)</td>
<td>18,804</td>
<td>0.29</td>
<td>0.45</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Age (45-54 years)</td>
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<td>0.27</td>
<td>0.44</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Age (55-65 years)</td>
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<td>Primary education</td>
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<td>0.48</td>
<td>0</td>
<td>1</td>
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<tr>
<td>Financial problems</td>
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<td>0.36</td>
<td>0.48</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>Health status</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Very good health</td>
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<td>0.44</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Good health</td>
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<td>0.50</td>
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<tr>
<td>Fair health</td>
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<td>0.38</td>
<td>0</td>
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<tr>
<td>Bad health</td>
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<td>0.01</td>
<td>0.12</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Mental health index</td>
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<td>16.7</td>
<td>4.9</td>
<td>0</td>
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</tr>
<tr>
<td>Hearing problems</td>
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<td>0.06</td>
<td>0.23</td>
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<tr>
<td>Skin problems</td>
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<td>0.08</td>
<td>0.27</td>
<td>0</td>
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<td>Back ache</td>
<td>18,804</td>
<td>0.45</td>
<td>0.50</td>
<td>0</td>
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<tr>
<td>Muscular pain in upper limbs</td>
<td>18,804</td>
<td>0.42</td>
<td>0.50</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Muscular pain in lower limbs</td>
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<td>0.28</td>
<td>0.45</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Headaches and eyestrain</td>
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<td>0.49</td>
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</tr>
<tr>
<td>Stomach ache</td>
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<td>0.34</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>Respiratory difficulties</td>
<td>18,804</td>
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<td>0.22</td>
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<td>1</td>
</tr>
<tr>
<td>Cardiovascular diseases</td>
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<td>0.04</td>
<td>0.19</td>
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<tr>
<td>Injuries</td>
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<td>0.27</td>
<td>0</td>
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</tr>
<tr>
<td>Depression or anxiety</td>
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<td>0.28</td>
<td>0</td>
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<tr>
<td>Overall fatigue</td>
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<td>0.48</td>
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</tr>
<tr>
<td>Insomnia or general sleeping problems</td>
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<td>0.48</td>
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<tr>
<td>Other health problems</td>
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<td>0.02</td>
<td>0.15</td>
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</tr>
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</table>

Source: 2010-EWCS. Own calculations, survey weights used.
<table>
<thead>
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<th>Col. 1</th>
<th>Col. 2</th>
<th>Col. 3</th>
<th>Col. 4</th>
<th>Col. 5</th>
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<tr>
<td>Absence</td>
<td>Poisson</td>
<td>EU-27</td>
<td>Absence</td>
<td>EU-27</td>
<td>Absence</td>
<td>EU-27</td>
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<td>Supervisor</td>
<td>-0.539*** (-2.19)</td>
<td>0.300*** (3.20)</td>
<td>-0.618** (-2.58)</td>
<td>0.267*** (2.39)</td>
<td>-0.677* (-1.91)</td>
<td>0.525*** (3.74)</td>
</tr>
<tr>
<td>Over time (≥ 45h)</td>
<td>-0.813*** (-2.91)</td>
<td>0.457*** (5.06)</td>
<td>-0.807*** (-2.76)</td>
<td>0.610*** (5.00)</td>
<td>-1.405*** (-3.53)</td>
<td>0.826*** (5.20)</td>
</tr>
<tr>
<td>Flexibl time arr.</td>
<td>-1.439*** (-3.19)</td>
<td>0.252*** (2.60)</td>
<td>-1.073*** (-4.32)</td>
<td>0.230** (2.13)</td>
<td>-1.467*** (-3.52)</td>
<td>0.417*** (2.97)</td>
</tr>
<tr>
<td>Adverse working condit.</td>
<td>0.198 (4.76)</td>
<td>0.055** (2.57)</td>
<td>0.255*** (4.72)</td>
<td>0.066** (2.59)</td>
<td>0.328*** (4.17)</td>
<td>0.080** (2.26)</td>
</tr>
<tr>
<td>Tenure (1-2 years)</td>
<td>1.505*** (3.03)</td>
<td>0.668*** (3.09)</td>
<td>1.280*** (3.79)</td>
<td>0.614*** (3.11)</td>
<td>0.754 (1.34)</td>
<td>0.444* (1.70)</td>
</tr>
<tr>
<td>Tenure (3-14 years)</td>
<td>2.271*** (3.66)</td>
<td>0.830*** (3.61)</td>
<td>2.006*** (4.84)</td>
<td>0.679*** (3.26)</td>
<td>1.482** (2.10)</td>
<td>0.490* (1.86)</td>
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<tr>
<td>Tenure (≥ 15 years)</td>
<td>2.152*** (3.04)</td>
<td>1.153*** (4.81)</td>
<td>1.996*** (4.00)</td>
<td>0.980*** (4.20)</td>
<td>0.991 (1.25)</td>
<td>0.916*** (2.80)</td>
</tr>
<tr>
<td>Private sector</td>
<td>-0.919*** (-4.96)</td>
<td>0.036 (0.49)</td>
<td>-0.903*** (-4.84)</td>
<td>0.058 (0.69)</td>
<td>-1.432*** (-6.30)</td>
<td>0.055 (0.51)</td>
</tr>
<tr>
<td>Temporary contract</td>
<td>-1.045*** (-2.95)</td>
<td>0.003 (0.02)</td>
<td>-0.771*** (-2.99)</td>
<td>-0.062 (-0.48)</td>
<td>-1.631*** (-4.06)</td>
<td>0.030 (0.15)</td>
</tr>
<tr>
<td>Part time (10-30h)</td>
<td>-0.908*** (-2.23)</td>
<td>-0.174 (-0.84)</td>
<td>-0.658 (-1.64)</td>
<td>-0.050 (-0.25)</td>
<td>-0.712 (-1.33)</td>
<td>0.025 (0.08)</td>
</tr>
<tr>
<td>Size (10-49 employees)</td>
<td>0.760*** (3.54)</td>
<td>0.102 (0.86)</td>
<td>0.813*** (3.68)</td>
<td>0.175 (1.46)</td>
<td>1.161*** (3.93)</td>
<td>0.124 (0.74)</td>
</tr>
<tr>
<td>Size (50-99 employees)</td>
<td>0.696** (2.37)</td>
<td>0.071 (0.54)</td>
<td>0.621** (2.22)</td>
<td>0.027 (0.22)</td>
<td>0.669* (1.70)</td>
<td>-0.059 (-0.33)</td>
</tr>
<tr>
<td>Size (100-249 employees)</td>
<td>1.058*** (3.29)</td>
<td>0.326*** (2.11)</td>
<td>1.146*** (3.19)</td>
<td>0.348*** (2.09)</td>
<td>1.347*** (3.03)</td>
<td>0.269 (1.27)</td>
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<tr>
<td>Size (&gt;250 employees)</td>
<td>1.332*** (3.98)</td>
<td>0.171 (1.40)</td>
<td>1.413*** (4.02)</td>
<td>0.181 (1.42)</td>
<td>1.726*** (3.93)</td>
<td>0.002 (0.01)</td>
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<td>Work autonomy</td>
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<td>0.185*** (3.90)</td>
<td>-0.056 (-0.65)</td>
<td>0.211*** (4.65)</td>
<td>-0.247** (-2.05)</td>
<td>0.236*** (3.95)</td>
</tr>
<tr>
<td>Lack of time</td>
<td>-0.114 (-1.03)</td>
<td>0.114*** (3.24)</td>
<td>-0.123 (-0.97)</td>
<td>0.144** (2.62)</td>
<td>-0.240 (-1.50)</td>
<td>0.181*** (3.04)</td>
</tr>
<tr>
<td>Production targets</td>
<td>-0.106 (-0.59)</td>
<td>0.173*** (2.66)</td>
<td>0.087 (0.41)</td>
<td>0.205** (2.71)</td>
<td>-0.290 (-1.19)</td>
<td>0.209** (2.17)</td>
</tr>
<tr>
<td>Performance assessment</td>
<td>0.062 (0.27)</td>
<td>0.290*** (2.66)</td>
<td>0.062 (0.24)</td>
<td>0.236** (2.17)</td>
<td>-0.299 (-0.87)</td>
<td>0.175 (1.41)</td>
</tr>
<tr>
<td>Social support</td>
<td>-0.020 (-0.35)</td>
<td>-0.086*** (-3.91)</td>
<td>-0.055 (-0.87)</td>
<td>-0.107*** (-4.94)</td>
<td>0.012 (0.13)</td>
<td>-0.128*** (-3.65)</td>
</tr>
<tr>
<td>Blue collar</td>
<td>0.253 (1.05)</td>
<td>-0.200** (-2.26)</td>
<td>0.273 (1.00)</td>
<td>-0.177* (-1.77)</td>
<td>0.673* (1.76)</td>
<td>-0.193 (-1.22)</td>
</tr>
<tr>
<td>Piece rate</td>
<td>0.090 (0.38)</td>
<td>-0.090 (-0.92)</td>
<td>0.070 (0.25)</td>
<td>-0.089 (-0.88)</td>
<td>0.252 (0.57)</td>
<td>-0.067 (-0.38)</td>
</tr>
</tbody>
</table>

Control variables | Yes | Yes | Yes | Yes | Yes | Yes |

N | 18804 | 18804 | 16585 | 16585 | 11656 | 11656 |
R² | 0.12 | 0.14 | 0.11 | 0.12 |

Source: 2010-EWCS, own calculations. Notes: Average marginal effects from Poisson regression in Columns (1) and (2), coefficient estimates from OLS regressions in Columns (3)-(6). The dependent variables are the number of sickness absence days in columns (1), (3) and (5) and the number of sickness presenteeism days in column (2), (4) and (6), both including those with zero days. All variables shown in the table except adverse working conditions, lack of time, work autonomy, social support, are dummies (see Table 1 for descriptive statistics). Sociodemographic variables as well as country and industry dummies are included as control variables but are not shown. The sociodemographic controls comprise sex, having children, partnership status, age categories, financial problems and educational status. The health control variables include self rated health status (3 dummies), 14 different symptoms and diseases from which the respondent suffered during the last 12 months as well as an index measuring mental health. The first two estimations rely on the full sample of 18,804 observations. Columns (3) and (4) rely on the observations from the EU-27 countries, i.e. 16,585 observations, and columns (5) and (6) rely only on observations that have at least one sickness day, be it absence or presenteeism (11,656 observations). T-statistics based on standard errors clustered on the country level are in parentheses. * p < 0.1, ** p < 0.05, *** p < 0.01.
References


Hansen, Claus D. and Johan H. Andersen, “Going ill to work – what personal circumstances, attitudes and work-related factors are associated with sickness presenteeism?,” *Social Science and Medicine*, 2008, 67 (6), 956–964.


