CHAPTER 5

The effect of overcommitment and reward on muscle activity, posture, and forces in the arm-wrist-hand region – a field study among office workers


ABSTRACT

Background. Office workers with high levels of overcommitment and low levels of reward are thought to be more prone to arm-wrist-hand symptoms, possibly through a higher internal physical exposure.

Objective. The aim of this study was to examine the effects of high overcommitment and low reward on forearm muscle activity, wrist posture and kinematics, and forces applied to computer input devices, during computer work in an actual work setting.

Methods. For 120 office workers with four different levels of overcommitment and reward (low-high, high-high, low-low, high-low), we continuously measured wrist extensor muscle activity, wrist posture and kinematics, and forces applied to the keyboard and mouse for two hours during their daily work.

Results. Wrist velocities and accelerations in radial-ulnar direction were higher for workers with high overcommitment compared to workers with low overcommitment, while their wrist range of motion was similar, possibly indicating a higher work pace. Wrist extensor muscle activity and forces applied to the keyboard and mouse were not increased by high overcommitment and/or low reward.

Conclusion. Overall, our findings provide little support for the proposed pathway, of high overcommitment and low reward in the development of arm-wrist-hand symptoms through a higher internal physical exposure.
Effects of overcommitment and reward on physical exposures in the arms, wrists, and hands during computer work

1. INTRODUCTION

Computer work has become a key element in daily work for many people and the number of people working with a computer continues to grow. Symptoms in the arm-wrist-hand region are a common problem among computer workers in industrialized as well as developing countries (e.g. 1;2). Along with serious consequences for the individual involved, arm-wrist-hand symptoms have also been associated with productivity loss and sick leave (3;4), causing high costs for societies and employers.

Arm-wrist-hand symptoms have a multifactorial origin. Besides physical risk factors, such as posture during computer work (5), work-related psychosocial factors and individual factors have been identified as independent risk factors for developing arm-wrist-hand symptoms (6-8). Potential work-related risk factors of psychosocial nature include high time pressure, low decision authority, low task variation, job satisfaction, high efforts, and low reward (1;9-12), and examples of individual risk factors are previous disabling symptoms and overcommitment (10;12;13).

In a recent longitudinal study among computer workers on risk factors for developing arm-wrist-hand symptoms, "low reward" and "high overcommitment" turned out to be important, changeable factors that were both significantly related to arm-wrist-hand symptoms (10). However, it is likely that these factors do not work independently in their effect on arm-wrist-hand symptoms. The personality trait overcommitment, which is defined as excessive work-related commitment and a high need for approval (12), might strengthen the effect of low reward and, when both factors act together, the strongest effects on musculoskeletal health can be expected (12).

It has been proposed that work-related psychosocial and individual risk factors lead to upper extremity symptoms through a higher internal physical exposure during computer use, such as increased muscle activity, higher forces on the keyboard and mouse, more adverse postures, and high repetition of movements (5;14;15), resulting in musculoskeletal damage that could accumulate over time. A recent systematic review (16) has indeed shown that work-related psychosocial stressors led to a higher muscle activity in the forearm region. In addition, other studies found indications for increased forces applied to the keyboard or computer mouse (15;17-19) and more harmful wrist kinematics (18;19) as a result of adverse work-related psychosocial factors. Only one study examined the effects of an individual risk factor on internal physical exposure in the arm-wrist-hand region (20). This study indicated that high motivation, which is a characteristic of the personality trait overcommitment, was associated with increased muscle activity in the wrist extensors during an attention demanding computer task.

It should be noted that all studies included in the aforementioned systematic review (16) and the studies cited above were performed in laboratory settings. Furthermore, the review indicated that induced stress during realistic computer tasks revealed a smaller increase in muscle activity than during more constrained simulated computer tasks. Possibly realistic computer tasks –even when performed in a laboratory setting- allow for adaptive motor strategies to cope with the stress. This raises the question whether
work-related psychosocial and individual factors indeed result in a higher internal physical exposure in the field. Therefore, to better understand the injury mechanisms of work-related psychosocial stressors and individual factors in a realistic work setting, field studies on this topic are needed, but are currently scarce (16).

Because overcommitment and reward were found to be important risk factors for developing arm-wrist-hand symptoms and because field studies on this topic are currently scarce, the aim of the present study was to investigate the effect of overcommitment and reward, and their interaction, on physical exposure of the arms-wrists-hands (i.e. wrist extensor muscle activity, wrist posture and kinematics, and forces applied to the mouse and keyboard) during computer work in a realistic work setting. We hypothesized that computer workers with high overcommitment and/or low reward have higher arm-wrist-hand physical exposures during computer work than computer workers with low overcommitment and/or high reward.

2. METHODS

Experimental design and set-up

This study is part of the PROOF study (PRedicting Occupational biomechanics in OFfice workers), which has as overall aim to assess physical exposure during office work in actual work settings. We selected 120 office workers based on a screening questionnaire, indicating highly contrasting levels of overcommitment and reward. During a day representative for “normal workload”, as indicated by the participant, we measured his/her neck-shoulder and forearm muscle activity, postures of the head, neck, torso, and upper extremities, and forces applied to the keyboard and mouse. Only non-obtrusive wireless measurement devices were used. In the present study, the data for the arm-wrist-hand region are presented. Data on the neck-shoulder region have been reported in a separate paper (21), because indications have been found that the underlying mechanisms for developing neck-shoulder or arm-wrist-hand symptoms may be different (17;22).

Before the start of the field measurement, the participant filled out a consent form. All protocols and consent forms were approved by the Harvard School of Public Health Human Subjects Committee, the Medical Ethics Committee of the VU University Medical Center Amsterdam, and the Ethics Committee of the Faculty of Human Movement Sciences of VU University Amsterdam.

Selection of participants

Over 2000 office workers employed at the VU University (8 different departments) and the VU University Medical Center (1 department), in Amsterdam, The Netherlands, were approached as potential participants in the study. In the nine participating departments, each worker received an informative recruitment email and was asked to fill out a short online screening questionnaire containing questions to assess their levels of overcommitment and reward. Overcommitment was assessed using a 6-item subscale from “the need for
effects of overcommitment and reward on physical exposures in the arms, wrists, and hands during computer work

control model” (23) and reward using the reward subscale (11 items) from “the effort-reward imbalance at work model (ERI)” (12).

An average of 35% (247 out of 707) of the approached workers filled out the screening questionnaire. Subjects were recruited at department level. Per department, for each participant the sum scores of overcommitment (range: 6 to 24) and reward (range: 11 to 55) were calculated. Subsequently, we selected participants within four different profiles of the most contrasting levels of overcommitment and reward by using the upper and lower tertiles of our high/low definitions, as illustrated in Figure 1 (24;25), with the intention to create contrast within our total group. Because the recruitment of workers was done in waves across departments, the upper and lower tertile values were calculated for each department separately. In this way, the high and low cut-off values were slightly different across departments. Means and ranges of the cut-off values of all included departments together were as follows: low cut-off value overcommitment 12.2 (11.0-14.0), high cut-off value overcommitment 14.8 (14.0-16.0), low cut-off value reward 46.4 (38.9-49.9), and high cut-off value reward 52.1 (49.9-53.9).

Figure 1. The four defined profiles with most contrasting levels of overcommitment and reward.
Forty-six percent of the workers who completed the screening questionnaire were willing to participate in the field measurements. Based on the partitioning shown in figure 1, about thirty participants were assigned to each group for a total of 120 subjects. These 120 workers were contacted by telephone for participating in the field measurement phase of the project, but also had to meet the following inclusion criteria:

- Office worker, for whom the main work tasks are computer-related.
- Healthy, no pain symptoms in the neck, shoulders, arms, wrists, and/or hands for at least one week prior to the field measurement.
- Contracted to work at least 20 hours per week.
- Comfortable using the mouse with the right hand.

Data collection and data processing

Procedure

For each of the 120 selected office workers we collected approximately two hours of continuous data while the participant performed his/her own work at his/her own work station: right and left wrist extensor muscle activity (extensor carpi radialis), left and right wrist posture (flexion-extension and ulnar-radial deviation), keyboard force, and mouse grip force. See Figures 2a-c for an illustration of the participant set-up. Additionally, a questionnaire identical to the one used in the PROMO study (26), containing potential confounders, was filled out.

Figure 2. Participant set-up. (a) Set-up overview. (b) Participant front view. (c) Participant back view.
Table 1. Participant characteristics of the four groups.

<table>
<thead>
<tr>
<th>Age (years)</th>
<th>Gender (M/F)</th>
<th>Body height (cm)</th>
<th>Body weight (kg)</th>
<th>Work history (years computer work)</th>
<th>Time of day field measurement</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mean SD Range Total number</td>
<td>Mean SD Range Total number</td>
<td>Mean SD Range Total number</td>
<td>Mean SD Range Total number</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Low OC, high RW group (N=31)</td>
<td>37 13 24-62 M:10 / F:21 177.7 9.4 163.0-195.0 72.2 10.5 53.0-91.0</td>
<td>0-5 years: 14 &gt;5 years: 17</td>
<td>Morning: 18 Afternoon: 13</td>
<td></td>
<td></td>
</tr>
<tr>
<td>High OC, high RW group (N=29)</td>
<td>39 12 23-62 M:9 / F:20 173.3 10.1 157.0-195.0 75.5 14.1 56.0-112.0</td>
<td>0-5 years: 12 &gt;5 years: 17</td>
<td>Morning: 17 Afternoon: 12</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Low OC, low RW group (N=30)</td>
<td>40 11 26-60 M:8 / F:22 175.0 11.1 153.0-204.0 72.9 14.9 49.0-123.0</td>
<td>0-5 years: 13 &gt;5 years: 17</td>
<td>Morning: 18 Afternoon: 12</td>
<td></td>
<td></td>
</tr>
<tr>
<td>High OC, low RW group (N=30)</td>
<td>43 10 28-63 M:8 / F:22 175.2 8.9 161.0-199.0 75.6 13.0 54.0-103.0</td>
<td>0-5 years: 7 &gt;5 years: 23</td>
<td>Morning: 16 Afternoon: 14</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

F: female; M: male; OC: overcommitment; RW: reward; SD: standard deviation
Muscle activity

Muscle activity in the left and right extensor carpi radialis (ECR) was measured using surface electromyography with a wireless logger system (Mega WBA, Mega Electronics LTD, Kupio, Finland). The electrodes (12 millimeter diameter Ambu Bluesensor N-00-S surface electrodes) were mounted over the muscle bellies of interest with 20 millimeter inter-electrode spacing, following the guidelines for extensor carpi radialis surface EMG of Basmajian (27). Data were recorded at 1000 samples per second after amplification (bandwidth of 10-500 Hz), then smoothened using a 3Hz second-order, zero phase, low-pass Butterworth filter, and down-sampled to 40 samples per second by using a mean filtering procedure.

For each participant 3 MVC’s were collected from the right and left ECR muscle, with 1 minute of rest in between contractions. MVC’s were collected by asking the participants to form a fist with their hand and to radially deviate and extend their hand against resistance applied by the experimenter. Each muscle’s MVC was the highest 1-second average of the EMG amplitudes collected from the three measurements.

EMG data were expressed as a percentage of a participant’s maximum voluntary contraction (MVC). Median (P50) wrist extensor muscle activity and variability (P90-P10) were calculated for each participant.

Wrist posture and kinematics

We measured left and right wrist flexion-extension and radial-ulnar deviation, using twin axis electrogoniometers (Model SG65, Biometrics Ltd, Gwent, UK), which were mounted over the back of the hand and the forearm with the wrist in a neutral position (i.e. hands pronated, middle metacarpal aligned with the midline of the forearm and backplane of the hand aligned with backplane of the forearm). Data were captured at 1000 samples per second with the same wireless data logging system used to collect the muscle activity data.

The participant’s neutral wrist postures were recorded at the beginning of the experiment and all wrist postures were expressed as deviations from the neutral position. Data were filtered through a 5Hz second-order, low-pass Butterworth filter and then down-sampled to 40 samples per second.

Median (P50) wrist postures and range of motion (P90-P10) were calculated for each participant and expressed in wrist angles (degrees). Wrist kinematics (i.e. wrist joint velocities and accelerations) were calculated by digitally differentiating the posture data, and were summarized in the root mean squared (RMS) value.

Force

Keyboard force was measured using a keyboard force plate (28), which had three miniature compression load cells (ELFF-B4-10L, Measurement Specialties, Hampton, VA, USA) mounted underneath in a triangular pattern. Each participant’s own keyboard was placed on top of the keyboard force plate.
Mouse grip force was measured using a modified USB mouse with scroll wheel (Model 3902C693, Microsoft, Inc., Redmond, WA (USA), which had three compression load cells (ELW-D1-10L, Measurement Specialties, Hampton, VA, USA) mounted inside, modelled after the design by Johnson and co-workers (29). The instrumented mouse was designed to be operated with the right hand and measured thumb forces applied to the left side of the mouse. All participants were required to use this force-sensing mouse instead of their own mouse.

Keyboard and the mouse force data were collected with USB backplanes (NI cDAQ-9172; National Instruments, Austin Texas) and sampled at 10,000 samples per second, then low-pass filtered at 20Hz (6th order Butterworth filter), down-sampled to 40 samples per second, and normalized to a participant’s maximum voluntary force (MVF). For each participant 3 MVFs were collected per input device, with 1 minute of rest in between. Maximal typing force was measured by having the participant maximally press down the “J” key with their right index finger for 5 seconds, and maximal mouse force by having the participant maximally squeeze the mouse for 5 seconds. Maximum forces were measured whilst adopting similar postures as during actual keyboard or mouse work. The MVF values were the highest 1-second averages of the force signals collected from the three measurements per force-sensing device. For each participant median (P50) and peak (P90) forces on the keyboard and mouse were calculated.

Computer use and potential confounders
Along with the measurements we measured computer input device usage using computer interaction monitoring software. From these data we determined for each participant non-computer activity and computer activity episodes: any period within 30 seconds of pressing a key, clicking a button on the mouse, or moving the mouse (30), whereas a non-computer activity episode was any period without computer input device usage for at least 30 seconds (31).

Subject information, such as age, gender, body height, and work history (i.e. number of years with computer work), was collected through an online questionnaire. Individual anthropometry was measured, including body weight, shoulder width (distance between the left and right acromial processes of each scapula), arm length (acromion-radiale length), hand length (from the distal wrist crease to the tip of digit III of the hand), and hand width (between the heads of the second and fifth metacarpale). Many of these measures were found to be related to physical exposure parameters in the laboratory (32;33).

Statistical analyses
Data were checked for a normal distribution and log-transformed if not normally distributed. We examined the main effects of overcommitment and reward, on median (P50) and range (P90-P10) wrist extensor muscle activity and wrist postures, on root mean square (RMS) wrist velocities and accelerations, and on median (P50) and peak (P90) keyboard and mouse forces. In addition, the interaction effect of overcommitment and reward was
tested, which provides information whether a combination of high overcommitment and low reward results in an enlarged effect. ANOVA for repeated measures was used, with overcommitment (low versus high) and reward (high versus low) as between-subject factors, and body side (left versus right) as a within-subject factor. For keyboard force or mouse force, univariate ANOVAs were used with overcommitment and reward as between-subject factors. Significance level was set at \( p \leq 0.05 \).

To adjust for confounding, we added the afore mentioned potential confounders (age, gender, work history, body height, body weight, BMI, shoulder width, arm length, hand length, and hand width) individually as a covariate to our model for each outcome parameter. If the effect size (i.e. partial eta squared) changed by 10% or more, then a confounder was considered as relevant and was maintained in the adjusted model. All covariates were continuous variables, except for gender (dichotomous) and “work history” (categorical). If two or more covariates were correlated (i.e. \( R^2 > 0.60 \)), only the covariate with the strongest moderating effect was maintained in the adjusted model. An adjusted model for a certain outcome parameter is reported only when one or more relevant confounders for that outcome parameter were identified.

3. RESULTS

Missing data

In the analyses of muscle activity and wrist posture, data of 117 participants were included, in the analyses of the wrist kinematics, 116 participants were included and in the analyses of forces on mouse and keyboard data of 114 participants were included. Data were excluded from the analyses because of technical failures and in one case because only 5 minutes of computer interaction time was measured for that participant.

Wrist extensor muscle activity

The median extensor carpi radialis muscle activity during computer use was slightly above 5% MVC and was not significantly different between participants with low versus high levels of overcommitment or with high versus low levels of reward (Table 2). We also examined the results of overcommitment and reward on the 90th and 10th percentiles of muscle activity, but no differences compared to the median values were found and these data were therefore not presented. Furthermore, we did not find any differences in the variability (i.e. \( P90-P10 \)) of extensor carpi radialis muscle activity between any of the groups. In addition, there were no interaction effects between overcommitment and reward for either median muscle activity and variability in muscle activity. No relevant confounders were identified for wrist extensor muscle activity.
Table 2. Results of median muscle activity and variability of the left and right extensor carpi radialis (n=117).

<table>
<thead>
<tr>
<th></th>
<th>Overcommitment</th>
<th>Reward</th>
<th>Interaction</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Low</td>
<td>High</td>
<td>Low</td>
</tr>
<tr>
<td>Mean</td>
<td>p-value</td>
<td>Mean</td>
<td>SE</td>
</tr>
<tr>
<td>Muscle activity</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Median (P50) (% MVC)</td>
<td>5.0 0.3</td>
<td>5.5 0.3</td>
<td>0.24</td>
</tr>
<tr>
<td>Muscle activity</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Variability (P90-P10) (%) MVC</td>
<td>7.7 0.4</td>
<td>8.1 0.5</td>
<td>0.60</td>
</tr>
</tbody>
</table>

SE: Standard error.
Wrist posture and kinematics

Median wrist posture was around 23 degrees extension and ranged between 1.8 and 2.6 degrees ulnar deviation. There were no significant main effects of overcommitment and reward on median wrist posture (i.e. flexion-extension and radial-ulnar deviation) or on wrist range of motion (i.e. P90-P10), and no relevant confounders were identified for wrist posture outcomes.

We found a significantly higher velocity and acceleration in radial-ulnar direction for participants with high overcommitment compared to participants with low overcommitment (Table 3). However, after controlling for confounding (age, body height, body weight, and work history) only wrist velocity (radial-ulnar deviation) remained significantly higher for participants with high overcommitment (Table 3). No interaction effects between overcommitment and reward were found for any of the wrist posture or kinematic outcome variables.

Forces on input device

All force variables were non-normally distributed and therefore log-transformed. There were no main effects of overcommitment or reward on the median (P50) keyboard force and also the interaction between overcommitment and reward was not significant. Peak (P90) keyboard forces were significantly affected by reward, with lower peak keyboard forces for participants with low reward as compared to those with high reward: 1.2% MVF and 1.5% MVF, respectively (see table 4). The main effect of reward remained significant after adjusting for confounders age, gender, hand length, and shoulder width (Table 4).

The median force applied to the side of the mouse (in % MVF) was significantly affected by overcommitment. Participants with low overcommitment squeezed the mouse significantly harder than participants with high overcommitment, 0.3% MVF and 0.1% MVF, respectively. However, this effect was no longer significant after adjusting for confounders (including body weight, body height, and work history) (Table 4). Also a significant interaction was found between overcommitment and reward for median mouse force, which remained after adjusting for confounders (Table 4). Participants with a combination of high overcommitment and low reward showed the lowest mouse squeezing force (Figure 3).
Table 3. Results of median wrist posture and kinematics of the left and right sides (n=117 for posture and n=116 for velocity and acceleration) and results of adjusted models in case relevant confounding was indicated: age, body height, and body weight were included as covariates for velocity, and work history, body height, and body weight were included as covariates for acceleration.

<table>
<thead>
<tr>
<th>Posture: Flexion-extension&lt;sup&gt;A&lt;/sup&gt; Median (P50) (degrees)</th>
<th>Overcommitment</th>
<th>Reward</th>
<th>Interaction Overcommitment / Reward</th>
</tr>
</thead>
<tbody>
<tr>
<td>Posture: Flexion-extension&lt;sup&gt;A&lt;/sup&gt; Range of motion (P90-P10) (degrees)</td>
<td>Low</td>
<td>Mean</td>
<td>22.9</td>
</tr>
<tr>
<td>Posture: Flexion-extension&lt;sup&gt;A&lt;/sup&gt; Range of motion (P90-P10) (degrees)</td>
<td>High</td>
<td>Mean</td>
<td>28.3</td>
</tr>
<tr>
<td>Posture: Radial-ulnar deviation&lt;sup&gt;B&lt;/sup&gt; Median (P50) (degrees)</td>
<td>Low</td>
<td>Mean</td>
<td>2.5</td>
</tr>
<tr>
<td>Posture: Radial-ulnar deviation&lt;sup&gt;B&lt;/sup&gt; Range of motion (P90-P10) (degrees)</td>
<td>High</td>
<td>Mean</td>
<td>4.2</td>
</tr>
</tbody>
</table>
Table 3. Continued

<table>
<thead>
<tr>
<th>Overcommitment</th>
<th>Reward</th>
<th>Interaction Overcommitment / Reward</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td></td>
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</tr>
<tr>
<td>Velocity:</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Flexion- extension(^\text{a})</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>RMS (degrees/s)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Velocity:</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Radial-ulnar deviation(^\text{b})</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>RMS (degrees/s)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Velocity:</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Radial-ulnar deviation(^\text{c})</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>RMS (degrees/s)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Adjusted model(^\text{d})</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Acceleration:</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>Flexion-extension(^\text{a})</td>
<td>Low</td>
<td>High</td>
</tr>
<tr>
<td>RMS (degrees/s(^2))</td>
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</table>
### Acceleration:
*Radial-ulnar deviation*  
RMS (degrees/s²)

<table>
<thead>
<tr>
<th></th>
<th>165.4</th>
<th>5.9</th>
<th>181.8</th>
<th>6.0</th>
<th>0.05*</th>
<th>0.034</th>
<th>175.5</th>
<th>6.2</th>
<th>171.7</th>
<th>5.7</th>
<th>0.66</th>
<th>0.002</th>
<th>0.40</th>
<th>0.007</th>
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</table>

### Acceleration:
*Radial-ulnar deviation*  
RMS (degrees/s²)

<table>
<thead>
<tr>
<th></th>
<th>166.4</th>
<th>6.0</th>
<th>180.9</th>
<th>6.1</th>
<th>0.10</th>
<th>0.026</th>
<th>176.4</th>
<th>6.2</th>
<th>170.9</th>
<th>5.7</th>
<th>0.51</th>
<th>0.004</th>
<th>0.46</th>
<th>0.005</th>
</tr>
</thead>
</table>

A: Extension is positive, flexion negative.  
B: Ulnar deviation is positive, radial deviation negative.  
#: Results when adjusted for confounding  
*: p≤0.05.  
RMS: Root mean square; SE: Standard error.
Table 4. Results of median and peak keyboard and mouse forces (n=114) and results of adjusted models in case relevant confounding was indicated. Age, gender, hand length, and shoulder width were included as covariates for peak keyboard force, and body weight, body height, and work history were included as covariates for median mouse force.

<table>
<thead>
<tr>
<th></th>
<th>Overcommitment</th>
<th>Reward</th>
<th>Interaction Overcommitment / Reward</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Low</td>
<td>High</td>
<td>p-value partial eta squared</td>
</tr>
<tr>
<td>Keyboard force Median (P50) (% MVF)</td>
<td>0.0 0.0 to 2.8 0.0</td>
<td>0.0 to 12.1 0.83</td>
<td>0.000</td>
</tr>
<tr>
<td>Keyboard force Peak (P90) (% MVF)</td>
<td>1.4 0.0 to 7.5 1.3</td>
<td>0.1 to 18.2 0.48</td>
<td>0.004</td>
</tr>
<tr>
<td>Keyboard force Peak (P90) (% MVF)</td>
<td>1.4 0.0 to 7.5 1.3</td>
<td>0.1 to 18.2 0.54</td>
<td>0.004</td>
</tr>
<tr>
<td>Adjusted model#</td>
<td>Mouse force Median (P50) (% MVF)</td>
<td>0.3 -0.2 to 2.6 0.1</td>
<td>-0.3 to 4.0 0.02*</td>
</tr>
<tr>
<td>Adjusted model#</td>
<td>Mouse force Median (P50) (% MVF)</td>
<td>0.3 -0.2 to 2.6 0.1</td>
<td>-0.3 to 4.0 0.08</td>
</tr>
<tr>
<td>Adjusted model#</td>
<td>Mouse force Peak (P90) (% MVF)</td>
<td>1.8 0.4 to 9.3 2.2</td>
<td>0.8 to 9.3 0.31</td>
</tr>
</tbody>
</table>

% MVF: percentage maximum voluntary force; *: p≤0.05; #: Results when adjusted for confounding.
4. DISCUSSION

The aim of this study was to examine whether high overcommitment and/or low reward lead to increased internal physical exposure during computer work in an actual work setting. We expected that workers with high overcommitment and/or low reward would have higher wrist extensor muscle activity, more extreme wrist postures, higher wrist velocities and accelerations, and higher forces applied to the keyboard and mouse.

Only few of our expectations were supported by the data. As we expected, wrist velocity and acceleration in radial-ulnar direction were higher for highly overcommitted workers as compared to those with low overcommitment. However, we did not find a higher median wrist extensor muscle activity, nor more extreme postures for those with high overcommitment and/or low reward as was expected. Results of mouse and keyboard forces were in contrast with our expectations. We found indications for lower median mouse force for workers with high overcommitment than for workers with low overcommitment, and peak keyboard force was lower for individuals with low reward as compared to those with high reward. Although the hypothesized pathway of arm-wrist-hand symptom development due to workplace and individual stressors via an increased internal loading did receive only limited support in this study, more evidence for this pathway appears to exist for neck-shoulder symptoms. In line with other studies (20;34) we found higher...
trapezius activity for workers with a combination of high overcommitment and low reward and a main effect of overcommitment on median trapezius muscle activity, indicating a higher trapezius activity for those who are overcommitted, as reported by Bruno-Garza et al. (21).

In line with our hypotheses, wrist velocity and acceleration in radial-ulnar direction were higher for workers with high overcommitment compared to workers with low overcommitment. Arm-wrist-hand symptoms can originate from overexertion of forearm muscles, controlling this movement, or from increased friction of tendons and nerves in the carpal tunnel (35;36). High wrist movement velocity and acceleration were indicated as the risk factors that were most consistently associated with arm-wrist-hand musculoskeletal disorders (37). Higher wrist velocity and acceleration in combination with a similar range of motion in the high and low overcommitment groups, may indicate that the highly overcommitted workers had a higher work pace. A higher work pace might increase physical exposure in the arm-wrist-hand region through increased forearm muscle activity, as was also found in a recent paper reviewing experimental laboratory studies on this topic (16). However, in the present study wrist extensor muscle activity was not higher in workers with high overcommitment. This may be because the muscle we measured, the extensor carpi radialis, is primarily a wrist extensor and to a lesser extent also a radial deviator of the wrist.

Unexpectedly, we found that workers with low overcommitment showed indications for a higher median mouse force than workers with high overcommitment, particularly when these workers experienced low reward as well. This finding can be interpreted in different ways. First, it could be that workers with low overcommitment indeed squeezed the mouse harder than the workers with high overcommitment. Second, it could be that the workers in both groups squeezed the mouse equally firm, but that the low overcommitted group worked with the mouse for a larger percentage of the measurement time, since we took the median of the mouse force during total computer interaction time, i.e. keyboard, mouse and idle activity. However, duration of mouse use was comparable in both overcommitment groups. A third explanation could be that the lower mouse forces (% MVF) in workers with high overcommitment were the result of a higher maximum voluntary force and not because of higher absolute forces on the mouse. However, the statistical differences between groups were similar for absolute mouse forces (Newton) and mouse force expressed in % MVF. Hence, it seems that workers with overcommitment indeed used lower forces when using the mouse, possibly because they make faster mouse movements and/or they make quicker changes between the mouse and the keyboard.

Keyboard force results were also unexpected. We found that peak keyboard force was higher for individuals with high reward, indicating that these workers occasionally pressed harder on the keyboard than workers with low reward. However, we do not have a possible explanation for this result.

Force values in the present study were small. To give an indication of the magnitude of the presented forces in % MVF; median mouse forces were around 0.1 N, all peak mouse
forces were between 0.9 N and 1.0 N, all median keyboard forces around 0.0 N, and all peak keyboard forces between 0.5 and 0.6 N. The reason for these relatively low force values was that median and peak forces were calculated across all computer activities, i.e. keyboard, mouse and idle activity to reflect exposure during computer use. Negative values in the ranges of mouse force (Table 4) may appear strange, but were the result of random error of our measurement device. When the mouse was untouched (i.e. zero mouse force) fluctuations in mouse force were equally distributed around zero.

In this study, we found small effects of high overcommitment and no important effects of low reward. Besides the possibility that overcommitment and reward do not really increase internal physical exposure of the arms-wrists-hands, it is also possible that this is caused by the way we measured these concepts. Even though we aimed for high contrasts between the high and low overcommitment and reward groups, the sum scores of both groups were rather close together and indicated a ceiling effect for reward. Possibly the discriminant validity of the effort-reward imbalance (ERI) at work model, was insufficient for our study population, consisting of office workers who were relatively highly educated. The population the ERI model was originally tested in was much more heterogeneous (12). Stronger contrasts between the groups could have resulted in stronger effects of high overcommitment and low reward than we found in our study. We cannot fully rule out the possibility that the limited contrast between groups was the result of selection bias, because we do not know the profiles of reward and overcommitment of the workers who did not respond to the screening questionnaire. However, there were no indications of selection bias for those who filled out the screening questionnaire and were willing to participate in the study; their levels of reward and overcommitment were similar to those of individuals who answered to the screening questionnaire, but were not willing to participate. Another possible reason for finding small effects of high overcommitment and no important effects of low reward might be the participant’s level of perceived stress during the measurement, which was “lower compared to a usual working day” for 53 participants. For the other participants their level of perceived stress was “similar”, except for 2 participants who indicated to perceive “higher” stress compared to a normal working day.”

Several strengths of the present study are noteworthy. This is the first study measuring many different physical exposure parameters during two hours of computer work, while participants actually performed their own work. We used non-obtrusive wireless measurement devices and as a result of this we were able to perform the measurements in the workers’ normal work environment. By this means, the stressors of interest are more natural as compared to the induced stressors in experimental studies. Also, the number of participants enrolled in this study was relatively high. However, some limitations should also be mentioned. A first limitation of this study is that we assumed that the levels of overcommitment and reward at the moment of selecting the participants, were representative for the stress experienced during the physical exposure measurements. However, the questions in the overcommitment and reward scales refer
to a more general state of stress, experienced during a longer period of time, and may vary over days. As a consequence, random noise will have been induced, diminishing the contrasts between groups, and resulting in smaller effects. Although we were partly able to overcome this by using the highest and lowest tertiles, it still might have attenuated our results. Second, we only looked at physical exposure during computer use and were not able to collect sufficient data on non-computer use while at work. Results of physical exposure during non-computer use might be important for understanding the development of arm-wrist-hand symptoms, since variation in biomechanical exposure is thought to be beneficial for musculoskeletal health and might be a remedy against musculoskeletal symptoms (38). Third, it remains to be seen if our conclusions can be generalized to other work settings and populations. The studied stressors and tasks are natural and the study population is representative for computer workers, but the measurements were performed in the public sector only. Especially in the private sector more extreme levels of overcommitment and reward can be expected and possibly stronger effects of high overcommitment and low reward might be found. Lastly, because this study has many outcome parameters, our findings that highly overcommitted workers show higher wrist movement velocities and accelerations, could be based on chance and should therefore be confirmed in future studies.

In our study we focused on effects of two different psychosocial stressors, and their interaction to evaluate physical exposure during computer work, because these psychosocial stressors were found to be the strongest risk factors for developing arm-wrist-hand symptoms. However, in a realistic work setting multiple psychosocial and individual factors are present and because effect sizes of single effects are small, we intent to test the combined effect of multiple factors in a follow-up study. In future studies, it is also of interest to examine work pace (e.g. typing speed, mouse movement velocity, and productivity) additionally to physical exposure, to be able to link both constructs and thereby possibly shed light on the underlying mechanisms in the development of musculoskeletal injury. For the development of intervention programs that prevent for musculoskeletal injury, it is important to know whether work-related psychosocial and individual risk factors increase physical exposure directly, or indirectly through adapted work patterns, such as an increased work pace.

In conclusion, workers with high overcommitment show higher wrist velocity and acceleration in the radial-ulnar direction during computer work, possibly as a result of a higher work pace. Even though this finding provides some support for the hypothesis that arm-wrist-hand symptoms may develop through a higher physical load, no other physical load parameters were increased by high overcommitment and/or low reward. Unexpectedly, we found that workers with low reward squeezed the mouse less firmly compared to workers with high reward. Overall, our findings provide little support for the proposed pathway, of high overcommitment and low reward in the development of arm-wrist-hand symptoms through a higher internal physical exposure.
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