**# Introduction**

In educational contexts, there’s a huge interest in exploring whether heart rate (HR) measures serve as reliable indicators for teachers’ stress during teaching settings [@karner2021teachers; @wettstein2020ambulatory]. Prior studies showed that teacher-centered activities and typical stressors lead to increased teacher HRs in teaching settings. However, these studies often relied on expensive and invasive electrocardiographs (ECG) to measure teachers’ HR [@sperka1995; @scheuch1997psychophysische; @donker2018; @junker2021; @huang2022class]. Using affordable, highly accepted, and non-invasive instruments like wrist-worn fitness trackers [@ferguson2015] could be a valuable tool for analyzing teachers’ HR and the factors contributing to teachers’ stress in everyday teaching situations.

This is particularly crucial given the high stress levels in the teaching profession [@johnson2005experience; @montgomery2005meta]. Classroom disruptions, for example, are one of the major stressors in teachers’ daily work [@boyle1995structural; @aloe2014multivariate]. According to @lazarus1990theory transactional model of stress and coping, the extent of the strain depends on the subjective appraisal of a stressor, which involves considerations about available resources to deal with it. It is, therefore, particularly important for teachers to have sufficient professional and personal resources at their disposal [@cramer2018belastung]. For instance, research has shown that professional knowledge about effective classroom management, including dealing with classroom disruption, reduces the risk of teacher stress [@klusmann2012berufliche]. Professional experience is one factor in how professional knowledge is acquired [@ericsson2006influence]. Consequently, experienced teachers have more effective classroom management skills to deal with classroom disruptions [@wolff2015keeping].

However, there is still a call for research collecting physiological measures of stress to contribute to a better understanding of how stressors like classroom disruptions affect teachers’ stress responses [@wettstein2021]. Thus, teachers’ use of wrist-worn fitness trackers in educational research offers transformative potential by providing detailed data, emphasizing the critical need to monitor teachers’ health and support efforts aimed at preventing stress among teachers. Therefore, this study explored the use of wrist-based fitness trackers as a tool to assess teachers’ HR as an indicator of stress before, during, and after a teaching session in which typical classroom disruptions occurred. Teachers’ HR data were triangulated with teachers’ appraisals of classroom disruptions and their teaching experience.

**## Fitness trackers as a method to assess stress**

Wearables, defined as electronic devices worn directly on the body or integrated into clothing or accessories, serve as versatile solutions [@godfrey2018z], gathering data like location, movements, and vital signs [@cheng2017underlying]. Fitness trackers, a popular wearable technology [@park2020user], provide data on physical activity and cardiovascular parameters such as HR, supporting personalized fitness goals [@nuss2021effects] and stress management [@hao2018chrv]. Their affordability and ease of use have contributed to their widespread use in healthcare, entertainment, and fitness [sinha2019taxonomy]. But also in education, fitness trackers offer added benefits for formal and informal learning environments for both students and teachers [@de2017towards]. Yet, few studies focused on their significance for teachers. However, unlike clinical devices, fitness trackers could offer continuous and less intrusive data collection such as heart rate (HR) over time [@godfrey2018z], aligning with the increasing popularity and acceptance of wearables among the general population [@peng2022acceptance]. Furthermore, in contrast to self-report questionnaires on stress [@chaplain2008; @liu2020] that are prone to biases like social desirability [@razavi2001self] or recall errors [@van2016accuracy], ambulatory assessment methods [@trull2013ambulatory; @wettstein2020ambulatory] like HR measurements can overcome these limitations by offering objective insights into teachers’ stress levels, e.g., by monitoring teachers’ HR without disrupting teaching [@donker2018; @runge2020].

One important health parameter assessed by nearly all wrist-worn fitness trackers is HR [@scalise2018wearables]. HR indicates the number of heartbeats within one minute and is typically expressed as beats per minute (BPM) [@berntson2007cardiovascular; @hottenrott2007]. At rest, the average HR of adults typically ranges between 60 and 80 BPM [@sammito2015guideline]. HR can be detected and measured in different ways using sensors, for example, based on electrocardiogram (ECG) or photoplethysmography (PPG) [@mukhopadhyay2017wearable]. While ECG sensors offer precise measurements by detecting the heart’s electrical activity, their intrusive nature and requirement for direct skin contact may limit their suitability [@kranjec2014non], particularly in educational settings. On the other hand, PPG is another uncomplicated and inexpensive technique to measure HR, commonly found in most fitness trackers [@castaneda2018review]. This optical method assesses HR by flashing green or red lights to measure changes in blood volume [@allen2007photoplethysmography].

Physiologically, HR is regulated and influenced on short-time intervals by the sympathetic and the parasympathetic nervous system [@pham2021]. An increase in the activity of the sympathetic system results in HR being speeded up (“fight or flight”) [@taelman2009influence]. In contrast, increased activity of the parasympathetic has the effect of slowing down the HR (“rest and digest”) [@battipaglia2015]. Stress or mental and physical strain are factors that directly influence HR and lead to an increase in it [@custodis2014heart]. They represent an important physical and emotional stress indicator, as an increased workload is associated with increased HR [@sachs2014]. Therefore, an increase in HR can be regarded as an indicator of increasing stress, and a decrease as an indicator of decreasing stress [@kyriacou1978]. Thus, fitness trackers offer a cheap and unobtrusive way of monitoring a wearer’s stress level.

**## Teacher stress**

The teaching profession is one of the most stressful professions compared to other occupational groups, facing a variety of stressors during everyday work [@smith2000; @herman2020; @schult2014belastet]. According to @kyriacou1978, teacher stress can be defined as a negative affective response, often accompanied by physiological changes such as increased HR, triggered by job-related demands, and perceived as threatening to one’s self-esteem or well-being. Coping mechanisms help to reduce the perceived threat.

This definition of teacher stress is based on the transactional stress model by Lazarus and colleagues [@lazarus1981stressbezogene; @lazarus1984stress], which was modified and tailored to the teaching-learning environment by @kyriacou1978. In general, the transactional stress model [@lazarus1990theory] highlights the interaction between an individual and the environment, whereby stress refers to a person’s subjective reaction to an event (a stressor) that exceeds their adaptive resources.

**Figure 1**

*A model of teacher stress (adapted from van Dick 2006, p.37, modified by the author)*

Fig. 1 shows, in a simplified way, how classroom events affect teachers’ stress levels, according to the model of teacher stress: When potential stressors (e.g., classroom disruptions) occur during teaching (1), teachers intuitively judge how disruptive the event is in a primary appraisal (2). If potential stressors are judged as threatening, i.e., as actual stressors (3), teachers consider whether they have sufficient resources for coping with the stressor (4). Teachers utilize these resources in trying to cope with the stressor, e.g., by employing classroom management strategies (5). In cases where coping fails, stress ensues, often accompanied by physiological reactions like increased HR (6). As part of the coping process, and dependent on its outcomes, teachers re-evaluate the stressor (7).

**## HR in teaching-learning contexts**

Prior research, not using fitness trackers but traditional electrocardiography (ECG), has shown that changes in teacher HR can be mapped onto stressors experienced by teachers during teaching. For example, teachers’ HR increased during teacher-centered activities when they had to take an exposed position in student-teacher interaction [@sperka1995; @scheuch1997psychophysische; @donker2018; @junker2021]. @sperka1995 for example recorded the HR of 16 pre-service teachers during their first lesson. The results showed significantly increased psychophysiological activation in terms of an increased HR during teaching. The activation effect was particularly prominent at the beginning of the lesson and decreased over the course of the lesson. The authors interpret this finding as indicating that pre-service teachers’ proactive coping strategies, such as actively managing student interactions, helped in lowering their HR levels. Other ECG studies identified typical stressors, predicting increased HR values, such as class size [@huang2022class], or low student engagement and motivation [@junker2021]. For example, @junker2021 recorded the HR of 40 teachers during a real classroom lesson. They provided evidence that teacher stress, induced by factors like low student engagement (e.g., lack of motivation or interest in tasks) or teacher-centered activities (e.g., teacher-focused classroom activities), resulted in elevated HR.

In addition to ECG studies, there are a few studies that used wrist-worn fitness trackers to investigate HR trends in teaching-learning situations [@Darnell2019; @chalmers2021]. @Darnell2019 for example measured the HR of 15 medical college students listening to lecture classes using wrist-worn devices. The analysis revealed a constant decrease in HR from the beginning to the end of a lecture, whereas the HR peak was reached during active learning sessions (peer-discussion based problem solving). @chalmers2021 examined the usability of the average HR, measured with a fitness tracker, to identify physiological changes during stress-inducing tasks with the Trier Stress Test in a study with 30 medical students and 30 participants from the general population. The average HR increased significantly between the resting and stress phases. Even though the participants in these studies were learners, not teachers, the results are relevant to the study of teacher stress using wearable devices, because the studies showed that a) HR can be effectively recorded using fitness trackers during a whole learning unit, and b) HR changes are in line with the occurrence of activating or stress-inducing tasks.

So far, to the best of our knowledge, only one study has directly assessed teachers’ HR using a wrist-worn fitness tracker during teaching: @runge2020 used a Fitbit fitness tracker to assess HR as an indicator of stress in *N* = 4 in-service teachers. They used the fitness trackers’ recordings to create a profile for each teacher, with the aim of differentiating between teachers reporting higher or lower levels of stress. In particular, it was found that the combination of a high HR, a high number of steps, and short sleep was characteristic of teachers reporting high stress levels. It should, however, be noted that the generalizability of the results is limited due to the small sample size.

In summary, previous studies have revealed that teachers’ (and students’) HR changes with an increase in HR even before stress occurs and peaks in activating phases [@Darnell2019; @chalmers2021], depending on their activities and the stressors they experience, whereby particularly teacher-centered phases led to an increase in HR [@sperka1995; @scheuch1997psychophysische; @donker2018; @junker2021]. Furthermore, it could be shown that HR as an indicator of stress can be assessed using low-cost, non-intrusive fitness trackers [@Darnell2019; @chalmers2021]. However, studies are still lacking, that investigate data from teacher-worn fitness trackers in larger samples to explore the links between teachers’ HR and subjective stressor appraisal or effects of teaching experience.

**## The role of appraisals, coping, and teaching experience in teachers’ stress response**

As shown in Fig. 1., primary and secondary appraisals play a crucial role in teachers’ evaluation of resources. Both appraisal processes are cognitive and depend highly on the person appraising the situation because they concern the implications of information for one’s well-being on a personal level [@lazarus1987]. Appraisal processes in turn are thought to influence the coping strategies teachers choose to deal with stressful events such as classroom disruptions. Direct action techniques as one coping strategy include teachers’ development of professional knowledge, skills, and practices to effectively manage the classroom environment [@kyriacou2001]. Thus, personal resources such as effective classroom management and an active, problem-focused coping style are negatively related to emotional exhaustion and teachers’ self-reported workload stress [@maslach2001job; @clunies2008self]. Accordingly, when resources are lacking and coping fails, negative consequences like burnout and high turnover can arise [@jalongo2006; @unterbrink2007; @aloe2014], highlighting the importance of research on teacher stress.

Furthermore, as shown in Fig. 1, both primary and secondary appraisals are influenced by teachers’ professional experience, shaping their classroom management skills. As professional experience grows, teachers develop cognitive scripts for managing classroom events, resulting in more complex and effective classroom management skills [@wolff2021classroom]. @clunies2008self showed that using reactive management strategies instead of proactive ones is significantly associated with increased teacher stress and reduced student on-task behavior.

Especially beginning teachers face considerable stress and often feel overwhelmed by the demands of teaching [@ophardt2017klassenmanagement; @wolff2015keeping; @klusmann2012berufliche], with many leaving the profession within the first five years [@ingersoll2003]. Research shows less experienced teachers are more susceptible to burnout, underscoring the importance of professional experience in predicting teacher stress [@fisher2011].

However, to date, there is a lack of studies that attempt to take into account the complexity of the transactional stress model [@obbarius2021; @goh2010revised], as often only individual parts of the model are researched (e.g., @zureck2015perfectionism). Particularly in the educational context, there is not yet enough research on the interplay of various factors contributing to the causes of stress with the targeted sample of teachers. @laugaa2008stress, for example, showed in a questionnaire study with *N* = 410 French teachers that perceived stress and coping strategies are important variables in explaining variance in burnout. Despite that, research on teacher stress primarily relied on questionnaires measuring subjective stress experiences, overlooking important data on real-time physiological stress responses [@wettstein2021]. Taken together, effective classroom management and active coping styles reduce stress, but more studies combining self-reported data with objective measurements such as HR are needed to understand the causes of teacher stress, particularly among beginning teachers prone to burnout.

**## Present Study**

The data analyzed in the present study were obtained from in-service and pre-service teachers who participated in a laboratory study as part of a larger project targeting the development of classroom management.

As part of the larger project, participants came to the lab individually and taught a 15-minute, self-prepared micro-teaching unit to a “class” of three actors (trained student assistants). These actors performed nine possibly disruptive, typical classroom events. The actors received standardized instructions on a screen (only visible to them, not to the participant) to perform a classroom event every one and a half minutes, and they performed the same scripted disruptions for all participants. The micro-teaching unit was potentially stressful for the participants, given its unfamiliar setting and the scripted disruptions of participants’ teaching flow. Thus, we were particularly interested in the changes in participants’ HR before, during, and after this micro-teaching unit. We recorded HR data in five phases, with a total duration of approximately two hours: In the *pre-teaching phase*, participants were welcomed, prepared for the following micro-teaching unit, and familiarized with the setting. In the *teaching phase*, participants taught the micro-teaching unit and experienced possibly disruptive classroom events. In the *post-teaching phase*, participants answered several questionnaires. Next, in the *interview phase,* participants engaged in a stimulated recall interview during which they rated for each classroom events how disruptive it had been to them, and how confident they had felt in dealing with it. In the *end phase*, participants answered another questionnaire. These sequences were identical for all participants. During the entire study, participants wore a fitness tracker on their wrist.

The aims of the present study were twofold:

(1) The first research goal was to investigate whether HR measures assessed by wrist-based fitness trackers were a suitable and effective method for mapping teachers’ HR over the course of the five-phase lab study, including the time before, during, and after the potentially stressful micro-teaching unit.

First, we expected the participants to show an initial increase in their HR, followed by a peak during the micro-teaching unit and a decrease for the remaining phases. In addition, we examined whether z-standardization of the participants’ mean HR could serve as a useful method to account for individual differences in baseline HR. We expected to observe the same trends in both standardized and non-standardized mean HR values.

Second, five representative 10-minute intervals were selected from the five phases (see also Figure 2: pre-teaching interval (I1), teaching interval (I2), post-teaching interval (I3), interview interval (I4), end interval (I5)). We examined HR level and change during these intervals in order to test the hypotheses that a) teachers would show the highest mean HR during the micro-teaching unit (I2), compared to all other phases (\*\*Hypothesis 1a\*\*), and b) that teacher HR would increase while they were preparing for teaching during the pre-teaching interval (I1), but decrease in all of the following intervals, because of habituating to (I2) and recovering from (I3-I5) the potentially stressful micro-teaching unit (\*\*Hypothesis 1b\*\*).

(2) We further explored whether teaching experience would make a difference in how teachers reacted to the classroom disruptions. In line with the research on teacher expertise and teacher stress reviewed above, we expected more experienced teachers might have better classroom management strategies, and thus better resources for coping. In addition, we were interested in finding a relationship between teachers’ appraisals of the classroom events (disruptiveness; confidence in dealing with them) and teachers’ HR, beyond the explanatory power of teaching experience. Specifically, we expected lower HR levels for teachers with more teaching experience, particularly during the micro-teaching unit (\*\*Hypothesis 2a). We expected higher HR levels for teachers who felt more disrupted, regardless of their teaching experience (\*\*Hypotheses 2b\*\*). At the same time, we expected lower HR levels for teachers who felt more confident in dealing with the events, regardless of teaching experience (\*\*Hypothesis 2c\*\*). Lastly, we hypothesized that each of the three predictors (teaching experience, disruption appraisal, confidence appraisal) uniquely contributes to explaining variance in teachers’ HR levels (\*\*Hypothesis 2d\*\*). In addition, we exploratively examined analogous hypotheses for the *changes* in HR.

**# Method**

**## Participants**

The sample consisted of *N* = 84 pre- and in-service teachers from Germany, who were recruited via personal contact, email lists, and flyers. The data of three participants was lost due to failed data transmission, yielding an analysis sample of *n*total= 81 (*n*total = 52 women, *n*total = 29 men), including 40 pre-service (*n*pre-service = 28 women, *n*pre-service = 12 men) and 41 in-service teachers (*n*in-service = 24 women, *n*in-service = 17 men). Participants had a mean age of 30.95 years (*SD* = 10.90; range: 19-60) and an average teaching experience of 5.64 years (*SD* = 9.46; range: 0-37).

**## Setting and Procedure**

The study was carried out following the ethical standards and the approval of the University’s Institutional Review Board. All participants were informed in detail about the aims of the study before testing. Participation was voluntary and only took place after written consent had been given. Participants were not rewarded in any way.

**Figure 2**

*Procedure of the two-hour-long study, consisting of five phases with five representative 10-minute intervals*

**

Each participant came to the lab for a period of approximately two hours in total, and each participant underwent the same phases (see Fig. 2): In the *pre-teaching phase*, the experimenter welcomed the participants and helped them put on the fitness tracker. This was followed by a warm-up session to familiarize the participants with the laboratory setting and the class. This phase took about 10-15 minutes and participants spent this time mostly standing or slowly walking around. During the *teaching phase*, the participants held their self-prepared, micro-teaching unit to a class of three trained actors who performed nine, potentially disruptive, classroom events (e.g., chatting with a neighbor, heckling, looking at the phone; see Table ## in the supplementary material for an overview and categorization of all events; and Fig## in the supplementary material for a depiction of the laboratory setting of the micro-teaching unit). In preparation for the micro-teaching unit, the topic and class level could be freely chosen by the teachers. The type of course was to be an introductory lesson and the prescribed social form required supervised individual work or frontal teaching. The micro-teaching unit lasted about 15-20 minutes. Participants spent this time mostly standing or slowly walking around. While teaching, participants wore eye-tracking glasses, and additionally, their lessons were recorded by cameras. After having completed the micro-teaching unit, in the *post-teaching phase*, participants were seated at a desk and filled in questionnaires for approximately 10-15 minutes: a brief computer-based questionnaire assessing sociodemographic data (e.g., teaching experience, gender, studied school type, studied school subjects, extracurricular teaching activities), and a short knowledge test irrelevant to the present study. In the *interview phase*, participants engaged in a Stimulated Recall Interview (SRI). During the SRI, participants watched the video of their own lesson from the ego perspective, recorded through the eye-tracking glasses, and indicating their gaze point. The experimenter stopped the video each time one of the nine classroom events happened and asked five open and three closed questions per event. Two of the closed questions are relevant to this study: the disruption and the confidence rating (see Measures). The interview lasted about 45-60 minutes, and participants’ position was seated. The *end phase* lasted about 10-15 minutes and participants answered another questionnaire irrelevant to this study, again in a seated position.

**## Measures**

**### Heart Rate Data and Heart Rate Intervals**

To measure teachers’ HR, we used the wrist-based fitness tracker Fitbit Charge 4. In line with the manufacturer's instructions [@fitbitnd], the device was attached a finger’s width above the participants’ nondominant hand’s wrist bone. The tracker works by flashing green LEDs hundreds of times per second, using light-sensitive photodiodes to catch the reflected light, to calculate the volume changes in the capillaries. From this, the tracker calculates how many times the heart beats per minute. HR measurements are generated at least every 15 seconds[[1]](#footnote-1). The raw data that can be extracted from the tracker contains the time stamps of all measurements and the estimated HR in BPM for each time stamp. To account for individual differences in the baseline HR, we first z-standardized the BPM values from the unstandardized mean HRs.

Since we aimed to explore teachers’ HR between different strain phases and to keep measurement intervals comparable within phases, we aggregated HR over a representative 10-minute interval within each phase. Previous research has indicated that 10-minute intervals are a useful duration for analyzing PPG data [@lu2008can]. The intervals were selected based on the following rules: The pre-teaching interval (I1) comprised the first 10 minutes after the fitness tracker had been put on. The teaching interval (I2) started two minutes after the teacher had started the teaching unit. This interval was of the highest relevance to our study. We explicitly chose an early 10-minute interval within the *teaching phase*, as previous studies revealed that the beginning of a lesson is essential and demanding regarding teacher-student interaction [@donker2018quantitative; @claessens2017positive]. The post-teaching interval (I3) started immediately after the end of the teaching unit. The interview interval (I4) was defined as the mid-10 minutes between the end of the teaching unit and the time point when the fitness tracker was taken off so that all participants were being interviewed during this interval. The end interval (I5) comprised the last 10 minutes before the fitness tracker was taken off.

**### Teaching Experience**

The participants’ teaching experience was assessed as a part of sociodemographic data. Participants stated their work experience in years (excluding the traineeship year).

**### Subjective appraisal of the classroom events and coping processes**

The subjective disruption and confidence appraisals were assessed during the SRI on an 11-point rating scale, ranging from 0 to 10. For the current analysis, ratings were averaged across the nine classroom events for each participant, as we were not interested in the effects of individual classroom events, but in the general stressfulness of the *teaching phase* for each participant, specifically, in the aggregated effect of all potentially stressful events (disruption rating) and the mean level of subjective coping (confidence rating).

**## Data analysis**

We conducted all analyses with R [@RStudio2020]. Graphics were created using ggplot2 (v3.3.3; Wickham, 2016).

\*\*Research goal 1\*\*. The first research goal included mapping teachers’ HR before, during, and following a micro-teaching unit over the course of a five-phase lab study.

Regarding the teachers’ HR trend, we displayed the HR trend over the course of the entire study. To account for individual differences in the baseline HR, we visually compared unstandardized and standardized mean HR trends over the course of the entire two-hour study.

For testing Hypothesis 1a, which examined mean differences of the HR levels across the selected five intervals, we initially conducted a one-way ANOVA with repeated measures as an omnibus test. The dependent variable comprised the standardized HR mean for each interval. To identify the highest HR level, we subsequently conducted *t*-tests with planned contrasts as post-hoc tests, accompanied by the effect size *d* [@cohen1988new]. Specifically, we tested the differences between the teaching interval (I2) and the other four intervals. Note that standardized mean HR was calculated at the subject level of *n* = 81 participants (see Table 1).

For testing Hypothesis 1b, which examined the HR changes within each interval, we conducted a linear estimation of the increase or decrease in standardized HR values over time. To this end, we used fixed intercept fixed slope regression models [@gelman2006data] for each interval to estimate intercepts and linear slopes for all individuals which were then averaged across individuals.[[2]](#footnote-2) Mean slope and mean intercept estimates are based on all values at all measurement points per interval for all participants (see Table 2). Mean slope and mean intercept values represent the unstandardized coefficients and need to be interpreted as *b*.

\*\*Research goal 2\*\*. Addressing our second research goal, we examined the effects of teaching experience and subjective appraisal of disruptive classroom events on teachers’ HR levels during the five phases.

To test hypothesis 2a, we examined the effect of teaching experience on participants’ HR levels (i.e., standardized mean HR) for each of the five intervals using linear regression models with teaching experience as the sole predictor. To test hypotheses 2b and 2c, we separately augmented the model by either teachers’ disruption appraisal (Hypothesis 2b) or confidence appraisal (Hypothesis 2c) as predictors, while controlling for the shared variance with teaching experience. To test hypothesis 2d, we examined the effects of all three predictors in one regression model. Furthermore, we repeated these steps to explore the effects of teaching experience and subjective appraisals on *changes* in teachers’ HR (i.e., mean slopes) at each interval. Please note: HR levels and changes in the pre-teaching interval (I1), were not predicted by the disruption and confidence appraisal because the classroom events had not taken place yet.

**# Results**

**## Research goal 1: Mapping HR over study phases**

The first part of our first research goal was to map the participants’ overall HR trend and explore whether z-standardization of participants’ HR is a useful method to account for individual differences in the baseline HR. Means, standard deviations, and range of teachers’ unstandardized and standardized HR are shown in Table 1. Fig. 3 a. and b. display the unstandardized and standardized HR, respectively.  HR initially increased, peaked, and then decreased, with the unstandardized and standardized HR graphs showing high similarity. Thus, for all further analyses, we used participants’ standardized HR values. Note that the study exceeded the planned duration of two hours for some participants. To avoid distortion when mapping the HR over the course of the study (see Fig. 3), the end was set at two hours.

**Table 1**

*Mean HR (*M*), standard deviations HR (*SD*), and Range of Teachers’ HR Over the Course of the Entire Study and the Five Intervals (Unstandardized in BPM/z-standardized)*

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| Interval | *M HR* | *SD HR* | Min | Max | |
| Overall Course of 2h | 90.09/0.041 | 15.76/0.991 | 512/-4.03 | 164/4.56 | |
| Pre-Teaching Interval (I1) | 96.28/0.48 | 14.11/0.88 | 56/-3.56 | 139/3.24 | |
| Teaching Interval (I2) | 100.80/0.85 | 16.23/0.77 | 63/-2.18 | 164/4.37 | |
| Post-Teaching Interval (I3) | 93.61/0.27 | 14.01/0.76 | 60/-2.17 | 150/3.06 | |
| Interview Interval (I4) | 82.32/-0.72 | 11.85/0.74 | 51/-2.51 | 132/4.39 | |
| End Interval (I5) | 77.95/-1.07 | 11.14/0.57 | 502/-2.68 | 120/2.96 | |
| 1 Please note that *M* and *SD* of the overall course were subject to rounding differences in the statistic software RStudio [@RStudio2020].  2 Please note that the overall course duration was limited to two hours, as this was the planned duration of the study and only some participants exceeded this planned duration, to which the deviations of the minimum values in the overall course and the end interval can be attributed. | | | | |

**Figure 3**

*Overall course of the HR with the unstandardized HR in BPM shown in Fig. a. and the z-standardized HR shown in Fig. b. for the planned 2-hour study*

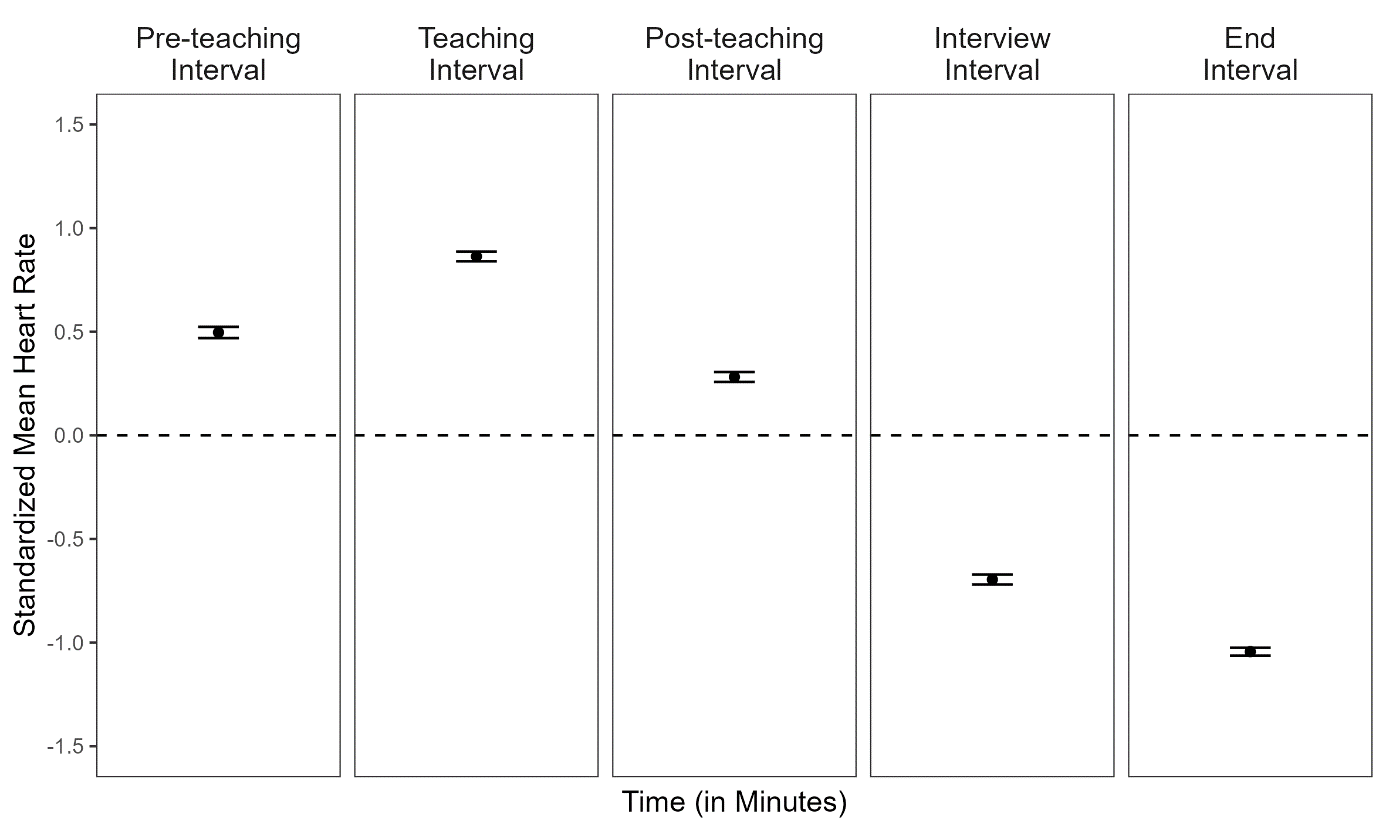


*Note:* The shadow around the line represents the 95% confidence interval. The confidence interval shown refers to the HR measurement points during the entire study period.

We first tested the hypothesis that teachers showed the highest standardized mean HR peaked during the micro-teaching unit, compared to all other phases (Hypothesis 1a). Repeated measures ANOVA revealed that the standardized mean HR differed statistically significantly between intervals, *F*(4, 400) = 257.50, *p* < .05, *f* = 1.60 (large effect). Post-hoc contrasts indicated that, as hypothesized, the standardized mean HR was significantly higher in the teaching interval (I2) than in all other phases (see also Fig. 4). Specifically, it was higher than in the pre-teaching interval (I1; *t*(1) = 32.71, *p* < .05, *d* = 0.82; large effect), the post-teaching interval (I3; *t*(1) = 32.00, *p* < .05, *d* = 1.34; large effect), the interview interval (I4; *t*(1) = 453.47, *p* < .05, *d* = 3.37; large effect), and the end interval (I5); *t*(1) = 511.89, *p* < .05, *d* = 4.68; large effect).

**Figure 4**

*Standardized mean HR for the five intervals*



*Note:* The dotted line represents the grand mean. Error bars represent the 95% confidence interval around the mean.

Next, we examined HR changes within each interval to test for the hypothesis that HR would increase in the pre-teaching (I1) and decrease in the post-teaching (I3), interview (I4), and end intervals (I5; Hypothesis 1b). The mean intercepts and mean slopes, complemented by their standard deviations for each interval, are shown in Table 2; the graphical representation of the slopes is displayed in Figure 5. The slope mean of the pre-teaching interval (I1) was significantly positive, indicating an increase in HR, as hypothesized. Further, the slope means of the teaching interval (I2) and post-teaching interval (I3) were significantly negative, indicating a decrease in HR. For the last two intervals, the interview interval (I4) and end interval (I5), the slope mean was negative but did not differ significantly from zero.

**Table 2**

*Descriptive statistics* *(*n, M, SD*)* *for the mean intercepts and the mean slopes for the different intervals for all individuals*

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| Interval | n1 | *M (SD)* | | *p* | |
|  |  | Intercept | Slope | Intercept | Slope |
| (1) Pre-teaching interval | 6896 | 0.052 (0.820) | 0.085\* (0.133) | .57 | < .05 |
| (2) Teaching interval | 7150 | 1.025\* (0.690) | -0.039\* (0.108) | < .05 | < .05 |
| (3) Post-teaching interval | 6664 | 0.549\* (0.547) | -0.060\* (0.101) | < .05 | < .05 |
| (4) Interview interval | 6287 | -0.617\* (0.614) | -0.022 (0.070) | < .05 | .01 |
| (5) End interval | 5990 | -1.004\* (0.500) | -0.012 (0.074) | < .05 | .14 |
| *Note.* \* *p* < .05  1All measurement points per interval for all participants. Note that the variation in *n* stem from the variation in the number of collected data points by the fitness tracker. | | | | | |

**Figure 5**

*Graphical display of the mean slopes of the standardized mean HR for each interval*



**## Research goal 2: Prediction of standardized mean HR and mean slopes with teaching experience and self-report data**

Correlations among standardized mean HR/mean slopes, teaching experience, disruption appraisal, and confidence appraisal are presented in Table 3. Correlations between standardized mean HR/mean slopes and the other variables were mostly very small and statistically non-significant, except for the pre-teaching interval (I1), in which mean slope and teaching experience correlated negatively (*r*=-.27), and the interview interval (I4) in which mean HR and teaching experience correlated positively (*r*=.24). Correlations between teaching experience and appraisals were substantial: more experienced teachers had lower disruption appraisals (*r*=-.36), and higher confidence appraisals (*r*=.44). Moreover, the two appraisal variables were negatively correlated (*r*=-.37).

Concerning the effect of teaching experience on participants’ HR levels (i.e., standardized mean HR) for each of the five intervals (Hypotheses 2a-d), teaching experience significantly predicted standardized mean HR only in the interview interval (Table 4, Interview Interval, Model 1), indicating a higher standardized mean HR for teachers with more teaching experience. This relationship is, in fact, in the opposite direction predicted by Hypothesis 2a.

Neither adding disruption appraisal while controlling for the shared variance with teaching experience (\*\*Hypothesis 2b\*\*) nor adding confidence appraisal while controlling for the shared variance with teaching experience (\*\*Hypothesis 2c\*\*) revealed any significant effects on teachers’ standardized mean HR.

When considering the effects of the three predictors in concert (\*\*Hypothesis 2d\*\*), standardized mean HR was significantly predicted only by disruption appraisal, and only in the post-teaching interval (Table 4, Post-Teaching Interval, Model 4), indicating a higher standardized mean HR for teachers who felt more disrupted by the classroom events, when controlling for all other factors.

Concerning the explorative investigation of the effects of teaching experience and subjective appraisals on *changes* (i.e., mean slopes) in teachers’ HR, teaching experience significantly predicted the mean slope in the pre-teaching interval(Table 4, Pre-Teaching Interval, Model 1), indicating a less steep HR increase in teachers with more teaching experience. For all other intervals, the prediction was not significant.

**Table 3**

*Correlations between standardized mean HR/mean slopes and the predictor variables of teaching experience (TE), disruption appraisal (DA), and confidence appraisal (CA) for the five intervals*

|  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- |
| Variable | Pre-Teaching Interval | Teaching  Interval | Post-Teaching  Interval | Interview  Interval | End  Interval |
| TE | − .17/− .27\* | .11/−.02 | − .04/−.03 | .24\*/−.20 | .04/.11 |
| DA | − .01/.16 | − .20/.08 | .20/−.14 | − .13/.01 | .04/.12 |
| CA | − .10/− .18 | .06/.09 | .04/−.03 | .09/−.19 | − .07/.13 |
| *Note.* TE = teaching experience, DA = disruption appraisal, CA = confidence appraisal,  \* *p* < .05. | | | | | |

**Table 4**

*Multiple linear regression of standardized mean heart rate and mean slopes predicted by teaching experience, disruption appraisal, and confidence appraisal for the five intervals*

|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
|  |  | *Dependent variable: standardized mean HR and mean slopes* | | | | | | | | | | | | | | | | | |
|  | Model 1 | | | | | Model 2 | | | | | Model 3 | | | | | Model 4 | | | |
|  | Std. mean HR | | | Mean slope | | Std. mean HR | | Mean slope | | | Std. mean HR | | Mean slope | | | Std. mean HR | | Mean slope | |
|  | β (SE) | *p* | β (SE) | | *p* | β (SE) | *p* | | β (SE) | *p* | β (SE) | *p* | | β (SE) | *p* | β (SE) | *p* | β (SE) | *p* |
| **Pre-teaching Interval (I1)1** |  |  |  | |  |  |  | |  |  |  |  | |  |  |  |  |  |  |
| Teaching  Experience | -.17  (.005) | .12 | -.27\*  (.002) | | <.05 |  |  | |  |  |  |  | |  |  |  |  |  |  |
| R2 | .030 |  | .071 | |  |  |  | |  |  |  |  | |  |  |  |  |  |  |
| **Teaching Interval (I2)** |  |  |  | |  |  |  | |  |  |  |  | |  |  |  |  |  |  |
| Teaching  Experience | .11  (.002) | .34 | -.02  (.001) | | .83 | .04  (.005) | .73 | | .01  (.001) | .96 | .10  (.006) | .42 | | -.08  (.001) | .54 | .05  (.006) | .67 | -.05  (.001) | .72 |
| Disruption  Appraisal |  |  |  | |  | -.18  (.041) | .13 | | .08  (.010) | .50 |  |  | |  |  | -.19  (.042) | .13 | .12  (.010) | .34 |
| Confidence  Appraisal |  |  |  | |  |  |  | |  |  | .01  (.046) | .92 | | .12  (.011) | .34 | -.04  (.047) | .76 | .15  (.012) | .24 |
| R² | .012 |  | .000 | |  | .040 |  | | .015 |  | .012 |  | | .010 |  | .042 |  | .031 |  |
| ∆ R² |  |  |  | |  | .028 |  | | .015 |  | .000 |  | | .010 |  | .030 |  | .031 |  |
| **Post-teaching Interval (I3)** |  |  |  | |  |  |  | |  |  |  |  | |  |  |  |  |  |  |
| Teaching  Experience | -.04  (.005) | .70 | -.03  (.001) | | .80 | .04  (.005) | .76 | | -.09  (.001) | .44 | -.08  (.006) | .55 | | -.02  (.001) | .89 | -.01  (.006) | .91 | -.07  (.001) | .61 |
| Disruption  Appraisal |  |  |  | |  | .22  (.040) | .07 | | -.18  (.009) | .14 |  |  | |  |  | .25\*  (.041) | <.05 | -.20  (.010) | .12 |
| Confidence  Appraisal |  |  |  | |  |  |  | |  |  | .08  (.045) | .55 | | -.03  (.011) | .83 | .14  (.046) | .27 | -.08  (.011) | .54 |
| R2 | .002 |  | .001 | |  | .043 |  | | .020 |  | .006 |  | | .002 |  | .058 |  | .023 |  |
| ∆ R2 |  |  |  | |  | .041 |  | | .019 |  | .004 |  | | .001 |  | .056 |  | .022 |  |
| **Interview Interval (I4)** |  |  |  | |  |  |  | |  |  |  |  | |  |  |  |  |  |  |
| Teaching  Experience | .24\*  (.006) | <.05 | -.20  (.001) | | .07 | .22  (.006) | .06 | | -.23  (.001) | .06 | .25\*  (.006) | <.05 | | -.14  (.001) | .25 | .23  (.007) | .07 | -.17  (.001) | .18 |
| Disruption  Appraisal |  |  |  | |  | -.05  (.045) | .66 | | -.08  (.006) | .52 |  |  | |  |  | -.06  (.047) | .61 | -.12  (.007) | .34 |
| Confidence  Appraisal |  |  |  | |  |  |  | |  |  | -.02  (.050) | .85 | | -.13  (.007) | .29 | -.04  (.052) | .76 | -.16  (.007) | .20 |
| R2 | .058 |  | .040 | |  | .060 |  | | .050 |  | .058 |  | | .054 |  | .061 |  | .069 |  |
| ∆ R2 |  |  |  | |  | .002 |  | | .010 |  | .000 |  | | .014 |  | .003 |  | .029 |  |
| **End Interval (I5)** |  |  |  | |  |  |  | |  |  |  |  | |  |  |  |  |  |  |
| Teaching  Experience | .04  (.004) | .70 | .11  (.001) | | .32 | .07  (.005) | .58 | | .18  (.001) | .13 | .09  (.005) | .46 | | .07  (.001) | .58 | .10  (.005) | .43 | .12  (.001) | .33 |
| Disruption  Appraisal |  |  |  | |  | .06  (.035) | .60 | | .19  (.007) | .12 |  |  | |  |  | .04  (.037) | .76 | .23  (.007) | .07 |
| Confidence  Appraisal |  |  |  | |  |  |  | |  |  | -.11  (.039) | .38 | | .10  (.008) | .43 | -.10  (.041) | .44 | .16  (.008) | .22 |
| R2 | .002 |  | .013 | |  | .005 |  | | .053 |  | .012 |  | | .025 |  | .013 |  | .078 |  |
| ∆ R2 |  |  |  | |  | .003 |  | | .040 |  | .010 |  | | .012 |  | .011 |  | .065 |  |
|  | *Note*. Coefficients are standardized. Effects of teaching experience and appraisals on teachers’ standardized mean HR are displayed for the five intervals.  In Model 1, standardized mean HR and mean slopes were predicted only by teaching experience. In Model 2, solely disruption appraisal was added as a predictor. In Model 3, solely confidence appraisal was added as a predictor. In Model 4, all three predictors were considered in concert.  1 We calculated only Model 1 for the pre-teaching interval because the classroom events had not yet occurred in this interval.  \* *p* < .05. | | | | | | | | | | | | | | | | | | |

**# Discussion**

**## Summary of key findings**

Our study aimed to investigate how data collected from a wrist-worn fitness tracker could shed light on the effects of stressors, such as classroom disruptions, on teachers’ stress responses before, during, and after teaching sessions. Therefore, we measured teachers’ HR over the course of a five-phase lab study, including a micro-teaching unit with potentially disruptive classroom events. Moreover, we examined whether variance in HR measures can be explained by teachers’ teaching experience and self-reported appraisal (disruption and confidence appraisal) of the classroom events.

The results were twofold: First, as expected, teachers’ HR increased before the micro-teaching unit, peaked during, and progressively decreased after the teaching phase. Second, contrary to our expectations, teachers’ HR could not be predicted by the teaching experience or by the subjective appraisals (disruption appraisal, confidence appraisal) either during the decisive phase - the teaching phase in which the potentially disruptive teaching events took place – or during the other phases.

**## Findings from mapping teachers’ HR over study phases**

Our first research question concerned the effectiveness and suitability of HR measures assessed by wrist-based fitness trackers for mapping teachers’ HR over the course of the five-phase lab study, including the time before, during, and after the potentially stressful micro-teaching unit. Results supported our hypotheses: Firstly, as expected in Hypothesis 1a, standardized mean HR was significantly higher in the micro-teaching unit than in all other phases with large effect sizes (0.82 ≤ *d* ≤ 4.68). This finding is in line with prior studies showing that teachers’ HR varies depending on their activities and encountered stressors, particularly increasing during phases where teachers are in an exposed position [@sperka1995; @scheuch1997psychophysische; @donker2018; @junker2021]. Secondly, teachers’ standardized mean HR increased before the micro-teaching unit but subsequently declined in the following intervals (Hypothesis 1b). Our findings suggest that teachers’ HR increased due to the preparation for the micro-teaching unit and decreased as a process of habituation to the teaching environment and recovery from the potentially stressful micro-teaching unit. This finding corresponds with results from prior studies that investigated HR trends in teaching-learning situations showing that HR changes align with activating events or stress-inducing tasks [@Darnell2019; @chalmers2021]. Moreover, researchers found that wearable sensing devices, like smart wristbands, can effectively capture changes in (students’) HR levels as physiological responses during various activities like lectures, self-tests, presentations, and exams [@francisti2023identification]. Thirdly, results revealed that the standardized and non-standardized mean HR values were comparable (see Fig. 3), i.e., this method ensured that observed differences in HR between individuals were not solely due to inherent differences in baseline HR levels (but see ##Limitations). Taken together, the findings indicate that wrist-worn fitness trackers are a useful tool to map teachers’ HR before, during, and after teaching.

## **Findings from the prediction of standardized mean HR and mean slopes with teaching experience and subjective appraisal ratings**

Regarding our second research question, the linear regression models only partially supported our expectations. Building on the model of teacher stress [@kyriacou1978, see Fig. 2], we hypothesized that more experienced teachers might have better classroom management strategies, and thus better resources for coping. Contrary to expectations, we found no effects of teaching experience or subjective appraisal ratings on the teachers’ HR in the investigated intervals. This finding suggests that experienced teachers still experience stress in teaching situations. The non-significant associations may be the result of the fact that teaching experience did not contribute to resources for coping to prevent stress. Consistent with this finding, recent research suggests that interventions to reduce stress and burnout in teachers need to address multiple levels (individual, individual-organizational, and organizational; @mcintyre2017towards). Merely enhancing teachers’ skills and coping mechanisms might not effectively diminish their stress levels unless changes are made to the organizational context of schools, including factors such as excessive workloads, resource limitations, and unsupportive administrative practices [@eddy2019single]. Moreover, teachers’ professional experience does not appear to be a guarantee for more professional knowledge. Other studies have also shown that teachers with more professional experience do not perform better than their colleagues with less professional experience in terms of pedagogical-psychological knowledge, which also includes effective classroom management [@kirschner2016professionswissen]. These results strongly underline the fact that “experienced” is not synonymous with “expertise”, but rather requires a deliberate practice of teaching “to choose to improve, to learn through […] experience, and to integrate new knowledge into future performances” [@dunn1999deliberate, p. 647].

Recent research also emphasizes the negative impact of work demands on emotional exhaustion but finds no significant effects on physiological measures such as heart rate variability and no moderating effects of job resources [@schmid2020teachers].

Nevertheless, we found some individual effects in some of the investigated intervals. First, when solely considering teaching experience (Hypotheses 2a), HR was predicted only in one of the investigated intervals, showing that teachers with more experience had higher standardized mean HR levels, contrary to expectations (β = .24). One explanation for the higher HR of experienced teachers could be that age correlated strongly with teaching experience (*r* = .94). For instance, researchers showed that older (female) teachers did not experience a decrease in their HR during periods of low stress levels, from which they concluded that recovery from stress was insufficient in the older teachers [@ritvanen2006responses]. Another explanation could be provided by @alhija2015teacher, who found that more experienced teachers reported more stress due to student misbehavior compared to less experienced teachers as a result of less tolerance due to burnout. In other words, experienced teachers already have elevated HRs due to the increased perceived stress level.

Second, adding disruption appraisal (Hypothesis 2b) and confidence appraisal (Hypothesis 2c) while controlling for the shared variance with teaching experience for both appraisals, no effects on teachers’ HR were found. One consideration might be that both single-item scales were not sensitive enough to assess teachers’ appraisal processes. This finding may be explained by the dynamic interaction between individuals and their environment that shapes the continuously changing nature of stress, making it challenging to determine process markers for appraisal and rendering the pursuit of a singular satisfactory measure difficult due to stress’s inherent complexity [@lazarus1990theory].

Third, when all three predictors were considered in concert, teachers’ standardized mean HR was significantly predicted only by disruption appraisal in the post-teaching interval (β = .25). This result is consistent with the idea that teachers who felt more disrupted by classroom events showed higher HRs, underlining @wettstein2021 call for the importance of ambulatory assessment methods, particularly in the context of classroom disruptions for gaining a deeper understanding of teacher stress and its impact on both psychological and biological measures.

Concerning the explorative investigation of the effects of teaching experience and subjective appraisals on *changes* in teachers’ HR, we did not find any effects except for the pre-teaching interval (I1), indicating that teachers with more teaching experience had a less steep HR increase before the teaching took place (β = -.27). This result could be explained by the fact that teachers with more professional experience felt less nervous about the upcoming micro-teaching unit, as it is a habit for them to teach in front of a class.

**## Limitations and future directions**

As with all research, methodological limitations need to be considered.

First, the laboratory setting of the study did not correspond to an authentic classroom environment. The data collected during the micro-teaching unit in a highly standardized situation was based on a fictional setting, resulting in various aspects, such as the lack of a teacher-student relationship. Nevertheless, researchers agree that positive relationships are the basis for effective classroom management and productive handling of classroom disruptions [@ruedi2014; @beaty2010]. For this reason, it cannot be excluded that the artificial laboratory situation led to an increased HR of the subjects due to nervousness and excitement. However, these conditions were identical for all participants, meaning that even if the HR was artificially increased by arousal due to the study design, this was the case for all participants. Nevertheless, in subsequent studies, it would be fruitful to measure teachers’ HR in authentic classrooms to contribute to the external validity of stress in teachers.

Secondly, we recorded HR as a physiological indicator of stress, whereas heart rate variability (HRV) which is the variation in the interval between consecutive heartbeats, predominantly depending on the external regulation of HR [@rajendra2006heart], would be an even more accurate measure. The reason for this is that HR is under the control of the sympathetic nervous system and can remain largely unaffected when moderate stress is experienced, whereas HRV is predominantly under the influence of the parasympathetic nervous system and therefore also indicates more subtle changes [@wettstein2020ambulatory]. However, the fitness tracker model used for this study was not equipped with a sensor to record HRV at the time. Future studies should consider this measure and use low-cost fitness trackers that measure HRV.

Third, the subjects’ resting HR was not recorded during the study, although this is an important marker to consider inter- and intrapersonal cardiovascular differences in participants [@nanchen2018; @heneghan2019]. For this purpose, it is necessary to record the HR during a resting phase without physical or emotional stress, ideally fifteen minutes before the beginning of the activity, to determine a valid baseline HR. This is a necessary condition when carrying out tests in the laboratory, but in practice, it can be a challenge to achieve these conditions [@sammito2015guideline] due to time constraints and difficulties in the acquisition of participants. To account for individual differences in the baseline HR without measuring the resting HR for 15 minutes, we z-standardized the BPM values from participants’ mean HRs (see ## in the Method Section).

Fourth, when HR is used as a marker to identify an individual’s experienced stress level during an activity, these measurements must be precise and accurate. As a result, there are a large number of scientific publications that deal with the validity and reliability of the parameters measured by fitness trackers [@montoye2017comparative; @muggeridge2021measurement; @gagnon2022; @fuller2020; @jo2016; @hajj2023; @jachymek2021]. Research on the reliability of Fitbit devices for the measurement of HR showed that this brand is accurate in controlled settings, depending on the activity level [@wallen2016accuracy; @hajj2023; @fuller2020; @jo2016]. For example, the Fitbit fitness tracker showed good measurement accuracy during resting phases [@jo2016; @muggeridge2021measurement] and for activities such as walking, jogging, and running [@hajj2023]. Findings in some studies indicate that the Fitbit fitness tracker showed a decrease in accuracy by underestimating the HR, especially at higher exercise intensities such as cycling [@thomson2019heart; @montoye2017comparative; @jo2016; @jachymek2021]. However, @chevance2022accuracy concluded in their systematic review and meta-analysis of Fitbit fitness tracker measures that the underestimation of HR has an acceptable range and that the accuracy depends on the context (e.g., quality of the study, type of activity). @gagnon2022 also summed up in his study that Fitbit devices can be used in research to detect stress-induced HR variation, but they cannot replace an ECG machine when precision is of the greatest importance.

Fifth, we did not assess any personal traits of the participants, such as emotional stability, work commitment, and life satisfaction, which are considered to be potential sources or protective factors of teacher stress [@wettstein2021]. Furthermore, we did not gather any information on the health status of the participants, although factors such as alcohol consumption, fitness level, cardiovascular diseases, etc. have an influence on HR and should be taken into account [@sammito2015guideline]. Future research should incorporate additional information collected to account for the fact that human HR is, in addition to the autonomic nervous system and genetic factors, influenced by numerous external factors such as social, personal, psychological, environmental, and behavioral factors [@wang2022].

**## Practical implications and hands-on tips with fitness trackers**

The results of our study point to several important practical implications for research on teacher stress and health.

Our results suggest that wrist-worn, low-cost, and nonintrusive fitness trackers can be used as a tool in educational settings to record HR. Thus, the increasing availability of HR data through wearable fitness trackers offers teachers the opportunity to self-monitor important mental health indicators such as HR as an indicator of stress, beyond traditional self-reporting or expensive, intrusive ECG devices. Using fitness trackers could enable teachers to strengthen their self-awareness in stressful situations and allow for early self-intervention. Furthermore, the use of fitness trackers in education could help teachers to create a greater awareness of the interplay between teaching practice, and physiological and psychological measures. For example, researchers were able to show that an increased HR during teaching was linked to less effective and sometimes confusing prosody patterns, such as intonation, speaking pace, and pausing [@tobin2016expression]. Research on mental health has shown that to achieve a regular and meaningful use of fitness trackers as mental health monitoring, it would be essential for participants to understand the purpose of using the fitness trackers as well as the meaning of the data [@ng2018]. It would therefore be important to explain the use of fitness trackers and the handling of data to teachers, e.g., with interventions in order to achieve long-lasting benefits with the technology.

Important aspects to consider for future research when using wrist-worn fitness trackers include, noting the time to make the HR measurable at different intervals, e.g., to replicate findings on whether HR differs in teacher-centered and student-centered activities using fitness trackers. Especially teacher-centered activities, historically overlooked in studies on psychological stress, emerged as the most influential factor in inducing stress among educators, with important implications for their psychological and physiological well-being [@junker2021].

Moreover, future research could use low-cost and non-invasive devices to accompany teachers in their everyday school practice to gain insight into teachers’ stress experience during teaching. Even in teacher training, wearable fitness trackers during internships could provide new insights into the stress experience of student teachers. Thus, the evaluation of the data with the participants could provide clues as to which moments in the teaching profession and teacher training are experienced as particularly stressful to discuss and implement possible stress-reducing measures in teacher training. Accordingly, the combination of subjective self-reported data such as interviews or questionnaires and objective measures such as HR would be an important step towards understanding and possibly preventing the development of stress in the teaching profession.

**## Conclusion**

This study investigated whether HR data collected from teacher-worn fitness trackers are suitable to explore links with factors such as subjective stressor appraisal, or effects of teaching experience to achieve a more profound comprehension of stressful transactional processes occurring in the classroom. Our results suggest that the widespread availability of HR data via fitness trackers presents opportunities for teachers to self-monitor stress levels for early intervention. Integrating fitness trackers into teacher training and everyday practice could offer valuable insights into stress experiences, facilitating the development of targeted interventions to support educator well-being. In summary, our study contributes to the understanding of stress in educational settings and underscores the potential of wearable technology in advancing research on teacher well-being.

**APPENDIX**

**Figure XX**

*Setting of the 15-minute micro teaching unit. Note. The setting included three actors as the class (left) and a teacher (right).*

Ein Bild, das Mobiliar, Stuhl, Kleidung, Schuhwerk enthält.

Automatisch generierte Beschreibung

**Figure XX**

*Setting of the interview. Note. The experimenter and participant watched the previously taught unit on video.*

Ein Bild, das Mobiliar, Zeichnung, Entwurf, Tisch enthält.

Automatisch generierte Beschreibung

**Figure XX**











1. The fluctuations in the number of seconds in which the HR was measured are due to the participants' movements, meaning that the device could not measure the HR every second. [↑](#footnote-ref-1)
2. Although this procedure does not account for nonmonotonic progressions in individual HR, a graphical evaluation revealed that the linear estimates corresponded well to the majority of the cases (see XX in the supplementary material). [↑](#footnote-ref-2)