# Vorschlag 1:

Eyes on the Classroom – How Teacher Expertise Shapes Gaze Behavior and Classroom Management

# Vorschlag 2:

Who Masters Classroom Disruptions - Expertise Differences in Classroom Management and Gaze Behavior in a Laboratory Setting

# Abstract

*Keywords:* classroom management, professional vision, eye-tracking, expertise differences, managing classroom disruptions

# Introduction

Classroom disruptions, such as students chatting with peers or nervously clicking their pens, can significantly interfere with learning by consuming valuable instructional time, reducing student engagement, and impairing academic achievement (Keller, 2014; Chow et al., 2024; Kraft & Monti-Nussbaum, 2021; Marder et al., 2023). Effectively managing such disruptions is therefore a key component of successful classroom management (Helmke, 2022).

A central element of this ability is teachers’ *professional vision* – the process of noticing, interpreting, and responding to relevant events in the classroom based on expert knowledge (Goodwin, 1994; Sherin & van Es, 2009). It links strategic knowledge with concrete action and allows teachers to anticipate and address disruptions proactively (Gold et al., 2016).

Expert teachers differ from novices not only in the amount of classroom experience but also in how they perceive and respond to complex instructional situations. They demonstrate more refined professional vision and possess a broader repertoire of strategies for managing challenges such as disruptions (Berliner, 2001, 2004; Gold & Holodynski, 2015; Emmer & Gerwels, 2006; Barth, 2017).

To gain a better understanding of the cognitive processes behind such expertise, process-based methods like eye-tracking have become increasingly important. Eye-tracking enables researchers to analyze where and how teachers focus their visual attention during instruction (Dessus et al., 2016; van den Bogert, 2016). Results indicate that expert teachers demonstrate more efficient gaze patterns, reflecting more accurate situational assessment (Gegenfurtner et al., 2020; Wolff et al., 2016).

This study examines differences in gaze behavior, disruption noticing, and classroom management between experienced and inexperienced teachers in a controlled micro-teaching unit in which teachers were confronted with a series of scripted classroom disruptions. Using eye-tracking and self-report measures, it investigates how teachers distribute their attention, how quickly they notice different types of disruptions, and whether expertise is reflected in more efficient gaze patterns, higher confidence, lower perceived disruptiveness, and greater strategic knowledge. The study aims to contribute to a deeper understanding of professional vision and its relevance for effective classroom management (Grub, 2023).

**Classroom Disruptions**

As described by Lohmann & Meyer (2003), disruptions are events that interfere with the essential conditions necessary for effective instruction and student participation. These behaviors can generally be categorized into four primary types, based on their nature and impact on classroom dynamics (Lohmann & Meyer, 2003). *Verbal disruptions* include spoken interruptions such as chatting, whispering, or heckling, which can disturb lesson flow and diminish focus. *Physical disruptions* refer to motor restlessness or unnecessary physical activity, like drumming on desks, snipping fingers, or clicking pens, which distract both teachers and students. Indicators of a *lack of eagerness to learn* manifest through disengagement behaviors, such as drawing, resting one’s head on the desk, or using a phone, actions that reduce participation and hinder comprehension. Finally, *aggressive behavior* encompasses hostile actions or emotional outbursts, including yelling, defiance, or physical confrontations, all threatening the classroom’s safety.

A crucial concept in understanding classroom disruptions is salience, which refers to how noticeable a behavior is within the classroom context (Kilbury et al., 2024). Salience directly affects teachers’ ability to notice disruptions – a key prerequisite for effective management. Highly salient behaviors, such as loud outbursts, tend to capture immediate attention, whereas more subtle behaviors may go unnoticed yet can be equally detrimental if left unaddressed over time.

Before any management strategy can be applied, a disruption must first be perceived and interpreted as requiring intervention. This process is complicated by the subjective nature of disruption, shaped not only by the behavior itself but also by how it is evaluated by both teachers and students (Eckstein et al., 2016). This dual perspective underscores the importance of recognizing both overt and subtle signs of disruption in order to respond appropriately and effectively.

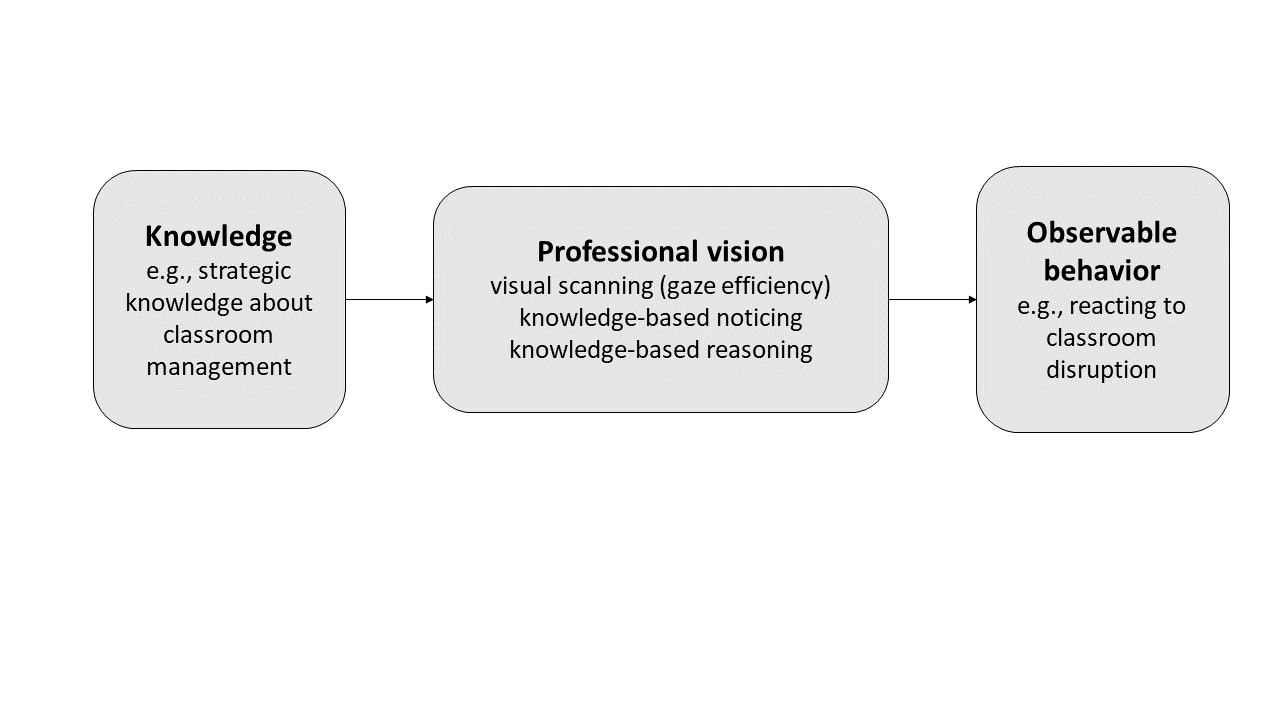
**Competence Development in Classroom Disruption Management**

The management of classroom disruptions is a central component of effective classroom management and requires teachers to develop context-sensitive strategies for responding to disruptive situations (Helmke, 2022). Effective disruption management depends on the ability to deal with the simultaneity and complexity of classroom events (Doyle, 1980), which demands a high level of knowledge and situational awareness.

Effectively addressing classroom disruptions requires teachers to continuously monitor their environment for behaviors that may interfere with the learning process. Figure 1 presents a simplified model that illustrates how knowledge interacts with perception processes – conceptualized as *professional vision* – and leads to observable teaching behavior. This model is adapted from the competence-to-performance transformation framework proposed by Blömeke et al. (2015) and further developed by Grub (2023).

**Figure 1**

*Model of Professional Competence Development in Classroom Disruption Management (Adapted From Blömeke et al. (2015) and Grub (2023), Modified by The Authors)*



As shown in Figure 1, teachers’ knowledge, such as the strategic knowledge of classroom management, is essential for managing classroom disruptions, as it involves professional judgment in selecting appropriate responses within complex, ambiguous situations (Gold & Holodynski, 2015; Fenstermacher, 1994). As a core element of instructional competence (Barth, 2017; D’Agostino & VanWinkle, 2007; Kunter et al., 2011; Voss et al., 2014), it allows teachers to weigh alternative courses of action and choose strategies that maintain instructional flow while addressing the disruption effectively. This form of knowledge evolves through practical experiences, where declarative knowledge about classroom management is gradually transformed into procedural and situationally applicable expertise (Blömeke et al., 2022; Rauner, 2005). Thus, managing disruptions does not rely solely on knowing rules or routines but on the ability to apply them flexibly and appropriately in context (Gold & Holodynski, 2015).

*Professional vision* serves as a crucial mediating skill between teachers’ dispositions and their classroom behavior. Originally introduced by Goodwin (1994) and adapted to teacher education by Sherin (2007), professional vision encompasses the ability to selectively perceive classroom events and interpret them based on pedagogical knowledge (Van Es & Sherin, 2002). In the initial stage of professional vision, *gaze efficiency* refers to the teacher’s ability to *visually scan* the classroom efficiently and purposefully (Cortina et al., 2015; McIntyre, 2016; Dessus et al., 2016; Huang, 2018; Gegenfurtner, 2020). Thus, gaze efficiency reflects an important facet of professional vision that enables teachers to early perceive relevant classroom events, such as classroom disruptions, supporting proactive classroom management (Grub et al., 2020).

*Knowledge-based noticing* refers to teachers’ ability to selectively attend to and interpret pedagogically meaningful events, such as classroom disruptions, in the classroom (van Es & Sherin, 2002; König et al., 2022). Beyond merely observing, noticing involves making sense of what is seen, guided by prior knowledge and teaching experience (Sherin et al., 2011). Thus, noticing functions as a central link between perception and informed teaching action.

After noticing, teachers engage in *knowledge-based reasoning* that describes how teachers interpret what they notice and use their knowledge to decide how to act, e.g. how to respond to a classroom disruotion (van Es & Sherin, 2002; Gibson & Ross, 2016).

Together, these three processes – gaze efficiency, knowledge-based noticing, and knowledge-based reasoning – constitute a situated, knowledge-driven perceptual cycle that enables teachers to detect, interpret, and respond to classroom disruptions effectively. They bridge the gap between (strategic) knowledge and visible classroom action and are therefore pivotal for the development and demonstration of professional teaching competence.

professional teaching practice The third component of the model, *observable behavior*, refers to the visible application of classroom management strategies, such as the effective reaction to classroom disruptions. Prior research in this area spans several decades and highlights key practices linked to effective classroom functioning. Early studies in the 1950s and 60s focused on teacher leadership styles, while in later decades, research increasingly focused on behaviorist approaches, examining how strategies like positive reinforcement and disciplinary consequences influence student behavior (Brophy, 2006). Observational studies have consistently shown that effective teachers actively prevent misbehavior through clear expectations, structured routines, and consistent responses. Kounin (1970) introduced the concept of “withitness,” emphasizing teachers’ ongoing awareness of classroom dynamics. Similarly, Anderson et al. (1979) found that teachers who explicitly teach rules and reinforce appropriate behavior from the first day of school achieve better academic and behavioral outcomes. Experimental studies have supported both single strategies (e.g., praise, rule instruction) and broader programs like the Good Behavior Game (Barrish et al., 1969), which uses team-based reinforcement to reduce disruptions. Oliver et al. (2011) conducted a targeted meta-analysis on universal, teacher-implemented classroom management practices. Their results revealed a significant reduction in disruptive behaviors, underlining the effectiveness of structured, preventive strategies led by teachers in everyday classroom settings.

In summary, the model illustrates that managing classroom disruptions effectively relies not solely on theoretical knowledge but significantly on the ability to perceive and interpret classroom dynamics in real-time. Integrating these components provides a comprehensive framework for describing and systematically enhancing teacher competence in classroom management.

## Expertise Differences in Classroom Disruption Management

Understanding expertise differences in classroom management requires a systematic assessment of the core professional competencies that underlie teachers’ ability to respond to disruptive behavior. Based on the three-component model of teaching competence (see Figure 1) – strategic knowledge, professional vision, and observable behavior – this section outlines how each of these dimensions can be empirically assessed and linked to professional expertise.

### Teachers’ Strategic Knowledge.

Strategic knowledge refers to context-sensitive, action-oriented knowledge that enables teachers to make informed pedagogical decisions in complex and dynamic situations (Fenstermacher, 1994; Doyle, 2006). Rather than relying on prescriptive rules, teachers with strong strategic knowledge are able to flexibly select appropriate courses of action that align with both instructional goals and students’ behavioral needs (Shulman, 1986; Baumert & Kunter, 2013).

Expertise research has shown that expert teachers differ significantly from novices in how their knowledge is structured and accessed. Experts possess highly organized and interconnected knowledge networks, which allow them to retrieve relevant information quickly and to apply it efficiently to novel situations (Livingston & Borko, 1989; Goldman, 2007). These knowledge structures are shaped through experience and enable teachers to integrate pedagogical concepts with situational demands. In contrast, novices tend to rely on fragmented, less structured knowledge that may lead to rigid or less adaptive decision-making (Shulman, 1987; Ophardt & Thiel, 2013).

Strategic classroom management knowledge develops through the interplay of theoretical learning and practical experience. While novice teachers often view classroom management primarily as discipline and control, expert teachers draw on a repertoire of proactive strategies aimed at establishing norms, preventing disruptions, and maintaining instructional flow (Helmke, 2022; Tynjälä et al., 1997). This development is cumulative and tied to growing professional judgment in navigating complex classroom dilemmas (Berliner, 2001).

Previous efforts to assess classroom management knowledge have either embedded it as a subdimension within broader pedagogical knowledge tests (e.g., COACTIV-R, TEDS-M; Kunter et al., 2013; König et al., 2011) or used licensure-based assessments such as the Praxis Series (Educational Testing Service, 2011). However, these instruments are either not publicly available for research, not specifically focused on classroom management, or lack contextual richness in their scenarios. Moreover, they often emphasize propositional knowledge over strategic reasoning, limiting their utility for expertise research (Gold & Holodynski, 2015).

To address this gap, Gold and Holodynski (2015) developed a model-based SJT specifically designed to assess teachers’ strategic knowledge of classroom management. Their instrument includes realistic classroom scenarios that reflect the three major facets of classroom management (monitoring, managing momentum, and establishing rules and routines). Validation studies have demonstrated the SJT’s sensitivity to expertise: in-service teachers significantly outperform preservice teachers, especially in scenarios requiring proactive and nuanced management strategies (Gold & Holodynski, 2015). In contrast to dichotomous formats, their SJT asks participants to rate the effectiveness of each response, allowing for a more nuanced assessment. These findings support the claim that strategic knowledge is not only a core component of classroom management competence but also a reliable indicator of professional expertise.

### Teachers’ Professional Vision.

To assess teachers’ professional vision, understood as their ability to perceive and interpret relevant events in the classroom, eye-tracking has become an established process-based measurement method. It captures attentional behavior in real time and enables inferences about underlying cognitive mechanisms such as noticing and attention control, which typically occur unconsciously and cannot be verbalized reliably (Grub et al., 2020; Seidel & Stürmer, 2014; van den Bogert et al., 2014).

Eye-tracking data are commonly analyzed using fixation-based metrics, as fixations are considered behavioral indicators of cognitive processing (Just & Carpenter, 1976). A fixation is defined as the moment when the eyes remain still over a visual object or person for a brief period, indicating focused attention (Holmqvist et al., 2011). Fixations are characterized by their frequency (number of fixations) and duration (average time in milliseconds), both of which can be used to infer how visual information is processed (Gegenfurtner et al., 2018; Negi & Mitra, 2020).

The relationship between these two parameters is often summarized in the Gaze Relational Index (GRI), a derived metric that divides average fixation duration by total fixation count (Gegenfurtner et al., 2020; Lowe & Boucheix, 2016). Lower GRI values typically indicate short fixations combined with frequent gaze shifts and are associated with more efficient, top-down driven scanning behavior. Higher values suggest longer fixations with fewer shifts, indicating a slower, potentially more cognitively loaded processing pattern (Gegenfurtner et al., 2020).

In the context of classroom management, these indicators have been used to differentiate expertise levels. Studies consistently show that expert teachers exhibit more dynamic and efficient gaze behavior than novices (Cortina et al., 2015; McIntyre, 2016; Dessus et al., 2016; Huang, 2018; Gegenfurtner et al., 2020). For instance, Cortina et al. (2015), for example, investigated attention distribution among 12 mentor-mentee teacher pairs, finding that mentors distributed their gaze more evenly across students, while mentees focused on fewer individuals. This suggests that expert teachers maintain broader classroom awareness, though the study’s small sample size limits its validity. In another study, McIntyre (2016) compared 40 teachers from England and Hong Kong, finding that experts focused more on students, while novices were more distracted by teaching materials and non-relevant stimuli. Huang (2018) analyzed mobile eye-tracking data from 25 expert-novice teacher pairs, finding that experts distributed their gaze more broadly across the classroom, while novices focused on fewer objects. Experts also exhibited shorter, task-relevant fixations, reducing distractions and improving situational awareness. Dessus et al. (2016) extended these findings by analyzing the gaze behavior of two novice and two expert teachers during full-length math lessons in primary school using mobile eye-tracking. Although the study had a small sample (*N* = 4), it provided detailed insights into gaze distribution and attentional strategies. Experienced teachers showed lower attentional lability, i.e., fewer rapid gaze shifts, and distributed their attention more evenly across students. Novices, in contrast, tended to focus disproportionately on a few individuals, regardless of their behavioral or academic characteristics. These differences were associated with higher CLASS scores among expert teachers, suggesting a link between gaze regulation, classroom climate, and instructional quality.

In a reanalysis of two studies, Gegenfurtner et al. (2020) found that experts generally showed lower GRI values than novices, especially in familiar task contexts. While differences were not always statistically significant, the GRI proved useful for comparing visual processing patterns across expertise levels and task demands. Its application in classroom settings is still limited and calls for further research.

Beyond fixation-based measures, time to first fixation (TTFF) has also emerged as a relevant metric in classroom management research (e.g., Yamamoto & Imai-Matsumura, 2013; Keskin et al., 2023). TTFF captures the latency between the onset of a salient classroom event (e.g., a disruption) and the first fixation on the relevant target (e.g., the disruptive student). Although TTFF is often interpreted as a marker of situational awareness, findings are mixed. In a study by Yamamoto & Imai-Matsumura (2013), teachers who noticed student misbehavior did not fixate earlier on the target students than those who missed it. Similarly, Keskin et al. (2023) found no significant differences in TTFF related to student background. These results suggest that TTFF alone may not reliably reflect expertise but should be interpreted alongside complementary measures such as fixation frequency and distribution.

Despite growing interest, research on the GRI in teaching contexts remains limited. Several studies have highlighted its potential as a standardized efficiency indicator but note that its interpretation depends on task complexity and instructional context (Gabel et al., 2023; Kosel et al., 2023). Furthermore, many eye-tracking studies in teacher research suffer from small sample sizes (e.g., Pouta et al., 2021) or methodological limitations, such as low ecological validity due to stationary setups (Stahnke & Blömeke, 2021) or low experimental control in authentic classrooms (McIntyre et al., 2020).

Nevertheless, the combination of fixation-based measures (number and duration of fixations, GRI) and event-based indicators such as TTFF offers a robust toolkit for assessing teachers’ professional vision in classroom management. These metrics provide access to real-time, process-oriented data and allow for fine-grained comparisons of expertise differences in how teachers attend to and process classroom events.

### Teachers’ Observable Behavior in Managing Disruptions.

Research has shown that novice teachers differ significantly from experienced teachers in their classroom management, particularly in handling disruptions (Leinhardt & Greeno, 1986; Berliner, 1987; Bromme, 1992). While experienced teachers can rely on established routines and cognitive schemas that help them manage disruptions efficiently (Thiel et al., 2012; Berliner, 2004), novice teachers often struggle with complex student behavior and tend to lose sight of overall classroom dynamics (Barth, 2017; Thiel et al., 2012). Experts use their monitoring skills to detect disruptions early and intervene effectively (Emmer & Gerwels, 2006), whereas novices frequently overlook such signals or react too late. Novices also tend to rely on less effective reprimands, interrupting the lesson flow (Westerman, 1991; Thiel et al., 2012), and focus on quick fixes (Swanson, O’Connor & Cooney, 1990). Experienced teachers, by contrast, avoid escalating interactions and may employ techniques such as ostentatious ignoring, using paraverbal and nonverbal cues to signal that inappropriate behavior has been noticed but will not be addressed directly (Thiel et al., 2012).

## Present Study

This study addresses previous methodological limitations by employing mobile eye-tracking in micro-teaching units within a highly standardized learning environment, using a comparatively large sample. This approach balances authenticity and experimental control by capturing real teaching-learning interactions while enhancing standardization through scripted behavioral instructions.

The study examines differences in gaze behavior, self-reported classroom management, and strategic knowledge between experienced and inexperienced teachers. Data were collected in the laboratory-based study *Professional Vision of Novice and Expert Teachers* (ProVisioNET), where both in-service (experienced) and pre-service (inexperienced) teachers participated in a controlled micro-teaching unit involving classroom disruptions. Eye-tracking technology, self-report measures, and strategic knowledge assessments were used to explore how expertise influences visual attention, disruption noticing, and management strategies.

Participants individually attended a laboratory session, where they conducted a brief micro-teaching unit lasting approximately 15 minutes. The “class” consisted of three trained actors who portrayed students and simulated typical classroom disruptions. During the micro-teaching unit, teachers’ gaze patterns were recorded using eye-tracking technology. Additionally, self-reports on classroom management and strategic knowledge were assessed through a questionnaire, an interview, and a test.

The study addressed five primary aims:

One aim was to examine how teachers distribute their visual attention during classroom interactions in a lab-based micro-teaching unit. We expected students to be the primary focus of teachers’ attention. This exploration was descriptive.

Another aim was to investigate differences in gaze behavior related to noticing abilities between experienced and inexperienced teachers. We hypothesized that, compared to inexperienced teachers, experienced teachers would demonstrate more efficient gaze patterns throughout the micro-teaching unit (**Hypothesis 1a**), focus their attention more frequently on students (**Hypothesis 1b**), and notice disruptions more quickly (**Hypothesis 1c**).

Additionally, we examined whether the type of disruption (i.e., verbal disruptions, physical disruptions, and indicators of lack of eagerness to learn) influenced the speed at which teachers noticed disruptions. We hypothesized that verbal and physical disruptions would be more salient and, therefore, noticed more quickly than indicators of a lack of eagerness to learn (**Hypothesis 2a**). Furthermore, we explored whether teaching expertise had an impact on the disruption noticing speed.

Another aim was to determine whether differences in teaching expertise were reflected in classroom management measures, including self-reports on classroom management (e.g., self-evaluated classroom management, disruptiveness, and confidence ratings) as well as strategic knowledge of classroom management. We expected experienced teachers to outperform inexperienced teachers (**Hypothesis 3a**).

Finally, we examined the relationship between gaze behavior and classroom management measures, expecting these variables to be correlated.

# Method

## Participants

We recruited a total of 84 teachers from Germany (42 pre-service teachers and 42 in-service teachers) through personal contacts, email lists, and flyers. Pre-service teachers were required to be actively enrolled in a teacher education program and to have completed their first internship, while in-service teachers needed to both have fully completed their teacher training (including the preparatory training year, the “Referendariat”) band to be currently working in the teaching profession. Data from two in-service teachers were excluded due to low-quality eye-tracking data, resulting in a final sample of 82 teachers, comprising 42 pre-service teachers and 40 in-service teachers.

The pre-service teachers (*n*= 29 women, *n*= 13 men) had a mean age of 22.80 years (*SD* = 1.90; range: 19–27). On average, they were in their 7th semester (*M* = 6.70, *SD* = 2.60; range: 3–11) and had an average of 9.60 hours (*SD* = 7.20; range: 1–36) of teaching experience through internships completed during their studies.

The in-service teachers (*n*= 24 women, *n*= 16 men) had a mean age of 39.10 years (*SD* = 10.60; range: 26–60) and an average of 11.60 years (*SD* = 11.30; range: 1–38) of teaching experience.

The study adhered to the ethical guidelines put forth by the University of Leipzig and received approval from the University’s Institutional Review Board. Participants were fully informed about the study’s objectives before testing. Their participation was voluntary, not incentivized, and commenced only after written consent.

## Setting and Procedure

Participants individually attended the lab for approximately two hours, following a standardized procedure for which a seminar room was transformed into a classroom. Upon arrival, they were welcomed by the experimenter, introduced to the procedure, and asked to sign the data protection agreement. Participants were then fitted with a binocular Tobii Pro Glasses 2 eye-tracker, adjusted for comfort and vision. After performing an initial calibration of the glasses (for details of the eye-tracker and the calibration, see Appendix A), the experimenter activated and synchronized the recording devices (eye-tracking glasses, four cameras, and an audio recorder) using an auditory signal. This setup phase included a brief introductory game to acclimate participants to the eye-tracking equipment and the three students, which took approximately 10-15 minutes.

After the initial setup, a second calibration was done in a separate room. As soon as the teacher re-entered the classroom, the micro-teaching unit started. Participants were asked to prepare a 15-minute micro-teaching unit on a topic and grade level of their choice. The only requirement was that the unit had to be an introductory micro-teaching unit, and had to consist of supervised individual work and/or frontal teaching. During the unit, three trained actors (playing students) performed scripted classroom disruptions. The students followed prompts that appeared every 1.5 minutes on a screen only visible to them (e.g., chatting with a neighbor, heckling, looking at the phone; see Table A1 in the supplementary material for an overview and categorization of all disruptions; and Figure B1 and B2 in the supplementary material for a depiction of the laboratory setting of the micro-teaching unit). The order of the disruptions and the performing students were fully balanced using Latin Squares. The whole micro-teaching unit was recorded using eye-tracking glasses to capture teachers’ gaze patterns. The micro-teaching unit lasted about 15-20 minutes.

After the teaching session, participants underwent a third calibration. Following this, all recording devices were stopped, and participants filled out a brief computer-based questionnaire (~10-15 minutes) assessing sociodemographic data and a self-evaluation of their classroom management during the micro-teaching unit. A ten-minute break followed, concluding the first part of data collection.

In the study’s second part, participants engaged in a Stimulated Recall Interview (SRI). They watched a video of their own teaching session, recorded through the eye-tracking glasses, while the experimenter paused the video at each classroom disruption. Participants answered five open-ended questions and three rating questions for each disruption, including self-reported *disruptiveness* and *confidence ratings* (see Measures). The SRI lasted approximately 45-60 minutes. Finally, participants completed a Situational Judgment Test (SJT) online, assessing their *strategic knowledge of classroom management*. The questionnaire took approximately 15 minutes to complete, marking the end of the study.

## Measures

***Gaze Behavior Measures***

Gaze behavior was analyzed using predefined Areas of Interest (AOI) to examine how participants distributed their visual attention during the micro-teaching unit. To code and analyze the gaze behavior, we used the software Tobii Pro Lab Analyzer (Version 1.241.54542). A fixation filter was applied, with a threshold set at 30°/sec to identify fixations, as this default fixation filter is recommended for mobile eye-tracking data in the Tobii Lab Analyzer Software (Tobii AB, 2024). Two types of AOIs were defined to structure the data meaningfully: global AOI and event-based AOI. Global AOI were used to assess gaze behavior over the entire video duration and included the following: the *students*, representing gaze points focused on the group of three students; *teacher material*, capturing gaze points directed toward instructional materials such as the board, screen, or other teaching aids; *student desks*, representing gaze points on elements related to students’ desks, including name tags and student materials; and *classroom/others*, which encompassed gaze points directed toward other areas of the classroom that were not associated with the students or teacher. In contrast, the event-based AOI was coded only during specific classroom events, i.e., disruptions caused by students. The AOI labeled *disruptive student* captured gaze behavior directed toward one of the three students who performed a classroom disruption.

#### Average Fixation Duration on AOI.

The average fixation duration was calculated by dividing the total fixation duration (in milliseconds) by the total number of fixations. Fixation data for these calculations were aggregated across all AOI and summarized for both experienced and inexperienced teachers.

#### Fixation Number per Minute on AOI *Students*.

The fixation number per minute was calculated by summing all fixations within the AOI *students* and dividing by the session duration in minutes. This yielded an individual fixation rate per participant, which was then summarized for experienced and inexperienced teachers.

#### Gaze Relational Index.

The Gaze Relational Index (GRI) was calculated as a composite metric to evaluate the efficiency of participants’ scanning behavior during the micro-teaching unit. The GRI was derived by dividing the average fixation duration (in milliseconds) by the fixation number per minute.

#### Time to First Fixation on AOI *Disruptive Student*.

The time to first fixation (TTFF) measured the time (in seconds) for participants to fixate on the AOI *disruptive student* after a disruption onset. Values were extracted by identifying the first fixation timestamp relative to the disruption onset, excluding invalid cases (TTFF = 0[[1]](#footnote-2) or >30s[[2]](#footnote-3)). Raw values were converted from milliseconds to seconds and log-transformed for normalization. To assess overall responsiveness to classroom disruptions, log-transformed TTFF values were averaged across all disruptions per participant and then summarized for experienced and inexperienced teachers.

### Self-Evaluated Classroom Management

After the micro-teaching unit, teachers answered a questionnaire using five items from a validated questionnaire (Helmke et al., 2013) and eleven self-developed items derived from the research literature (Kiel et al., 2013; Kounin, 2006; Marzano, 2007) to assess their self-evaluated classroom management. The questionnaire was a 4-point Likert scale (1 = strongly disagree; 4 = strongly agree).

### Disruptiveness and Confidence Ratings

The disruptiveness and confidence ratings were assessed during the SRI on an 11-point rating scale, ranging from 0 (not at all disrupting/confident) to 10 (extremely disrupting/confident). Ratings were averaged across the nine classroom disruptions for each participant to capture a general sense of how disruptive the classroom disruptions were during the micro-teaching unit and how confident participants felt in handling them.

### Strategic Knowledge of Classroom Management

Teachers’ strategic knowledge of classroom management was assessed using a Situational Judgment Test (SJT; Gold & Holodynski, 2015) via an online questionnaire on SoSci Survey. Participants graded five to six action alternatives for twelve teaching scenarios in which classroom disruptions were discussed on a six-point Likert scale (grade 1 = “very good” to grade 6 = “unsatisfactory”). As the SJT was originally designed for primary schools, adjustments were made to enable the use of the SJT for all types of schools in the *ProVisioNET* study. Due to their general applicability, all twelve scenarios and answer options were adopted and only the names of the class levels were removed from the questions - except for scenario 6, where this information was essential.

## Data analysis

The data were analyzed using R (RStudio Team, 2020, Version 2024.12.0) and IBM SPSS Statistics (Version 29). Graphics were created using ggplot2 (Wickham, 2016).

To examine teachers’ visual attention distribution (Aim 1), fixation durations[[3]](#footnote-4) for different AOI (e.g., students, disruptive student, teacher material) were analyzed descriptively. Mean proportions were compared between experienced and inexperienced teachers using independent-sample *t*-tests, with effect sizes reported as Cohen’s *d* (Cohen, 1988).

To assess differences in gaze behavior related to noticing abilities (Aim 2), we calculated the GRI during the micro-teaching unit, fixation number per minuteon AOI *students*, and TTFF on AOI *disruptive student*. Group differences were tested with t-tests, and effect sizes were reported.

To investigate how disruption type affects noticing speed (Aim 3), a 2 × 3 repeated-measures ANOVA was conducted, with expertise (experienced vs. inexperienced teachers) as a between-subject factor and disruption type (verbal, physical, lack of eagerness) as a within-subject factor. Bonferroni-adjusted post-hoc comparisons were performed for significant effects.

To examine classroom management differences (Aim 4), independent-sample *t*-tests compared self-evaluated classroom management, disruptiveness and confidence ratings, and strategic knowledge.

Finally, to explore the relationship between gaze behavior and classroom management (Aim 5), Pearson correlations were computed between gaze efficiency metrics (GRI, fixation number per minuteon AOI *students*, TTFF on AOI *disruptive student*) and classroom management measures. Statistical significance was assessed using *p*-values, with effect sizes reported.

# Results

## Gaze Distribution for Areas of Interest

The descriptive results for experienced and inexperienced teachers to investigate teachers’ gaze distribution are displayed in Figure 1.

**Figure 1**

*Average Fixation Duration Percentages by Area of Interest (AOI) and Teacher Experience Group for the Entire Micro-Teaching Unit*

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*Note.* The bar chart displays the percentage of fixation durations for five AOI (students, teacher material, student desk, classroom/others, and disruptive student) during the micro-teaching unit. Data are presented separately for experienced and inexperienced teachers, with error bars indicating 95% confidence intervals.

As suggested, both groups exhibited the highest percentages of fixation durations in the AOI *students*, while no statistically significant differences were observed between experienced and inexperienced teachers for any AOI (see Appendix C, Table C2 for *t*-test results and effect sizes).

## Gaze Behavior

Means, standard deviations, and range of experienced and inexperienced teachers’ gaze behavior measures are shown in Table 1.

**Table 1**

*Means, Standard Deviations, And Range of Experienced and Inexperienced Teachers’ Gaze Behavior Measures*

|  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Gaze Behavior Measures | Experienced | | | | Inexperienced | | | |
| *M* | *SD* | Min | Max | *M* | *SD* | Min | Max |
| GRI for micro-teaching unit | 5.32 | 2.21 | 2.22 | 11.72 | 6.08 | 2.15 | 2.14 | 10.80 |
| Fixation Numbers in AOI *Students* | 98.53 | 19.21 | 49.07 | 135.81 | 91.21 | 14.52 | 68.99 | 123.37 |
| TTFF (log) in AOI *Disruptive Student* | 0.28 | 0.69 | –1.81 | 1.51 | 0.30 | 0.65 | –1.06 | 2.21 |
| *Note*. GRI = Gaze Relational Index, AOI = Area of Interest, TTFF = Time to First Fixation (log-transformed). Values represent the mean (*M*), standard deviation (*SD*), and range (minimum and maximum values) for each measure. Fixation numbers in AOI *Students* refer to the total number of fixations directed at students per minute. TTFF values represent the log-transformed time (in seconds) until teachers first fixated on a disruptive student. | | | | | | | | |

First, experienced teachers exhibited more frequent but shorter fixations, leading descriptively to a lower GRI than inexperienced teachers across the entire micro-teaching session (**Hypothesis 1a**). However, this difference was not statistically significant, *t*(80) = –1.57, *p* = .12.

Second, experienced teachers directed their gaze toward AOI *students* more often than inexperienced teachers (**Hypothesis 1b**), and this difference was statistically significant, *t*(80) = 1.96, *p* = .05, *d* = 0.43 (small effect).

Third, experienced teachers detected disruptions descriptively slightly faster, as reflected in their shorter noticing speed for the AOI *disruptive student*[[4]](#footnote-5) (**Hypothesis 1c**), but with no statistically significant difference, *t*(80) = –0.14, *p* = .89.

## Effect of Disruption Type and Expertise on Disruption Noticing Speed

For **Hypothesis 2a**, a significant main effect of disruption type was found, *F*(1.94, 141.49) = 68.05, *p* < .05, *η²* = .34, indicating that the type of disruption influenced how quickly teachers noticed it. Consistent with our hypothesis, post-hoc comparisons revealed that verbal disruptions were detected significantly faster than both physical disruptions, *t*(73) = 6.33, *p* < .05, *d* = 0.62 (medium effect), and lack of eagerness disruptions, *t*(73) = 11.09, *p* < .05, *d* = 1.23 (large effect). Furthermore, physical disruptions were noticed faster than lack of eagerness disruptions, *t*(73) = -5.72, *p* < .05, *d* = 0.62 (medium effect). The main effect of expertise, *F*(1, 73) = 0.03, *p* = .86, and the interaction between expertise and disruption type, *F*(1.94, 141.49) = 1.28, *p* = .28, were not significant, suggesting that expertise did not influence noticing speed, nor did it interact with disruption type.

## Expertise Differences in Classroom Management Measures

Table 2 presents descriptive statistics and internal consistency reliabilities for all classroom management measures.

**Table 2**

*Means, Standard Deviations, Range and Internal Consistency Reliability (McDonalds’ Omega, ω) of Experienced and Inexperienced Teachers’ Classroom Management Measures*

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Classroom Management Measures | Experienced | | | | Inexperienced | | | | ω |
| *M* | *SD* | Min | Max | *M* | *SD* | Min | Max |
| Self-evaluated Classroom Management | 3.25 | 0.31 | 2.38 | 3.88 | 3.06 | 0.30 | 2.56 | 3.62 | .82 |
| Disruptiveness Rating | 4.81 | 2.90 | 0 | 10 | 5.55 | 2.81 | 0 | 10 | .84 |
| Confidence Rating | 8.44 | 1.68 | 2 | 10 | 7.18 | 2.04 | 0 | 10 | .88 |
| Strategic Knowledge | 0.75 | 0.08 | 0.54 | 0.88 | 0.72 | 0.13 | 0.10 | 0.91 | .89 |
| *Note*. Values represent the mean (*M*), standard deviation (*SD*), and range (minimum and maximum values), along with McDonald’s Omega (ω) coefficient, which indicates the internal consistency reliability for each measure. *Self-evaluated classroom management* refers to participants’ assessment of their classroom management competencies. The *disruptiveness rating* indicates the extent to which participants judged the disruptions as disruptive, while the c*onfidence rating* measures their confidence in managing them. S*trategic knowledge* measures teachers’ knowledge of classroom management strategies. | | | | | | | | | |

Independent-sample *t*-tests examined whether teaching expertise influenced classroom management measures (**Hypothesis 3a**). As expected, experienced teachers evaluated their classroom management competencies significantly higher than inexperienced teachers, *t*(80) = 2.78, *p* < .05, *d* = 0.62 (medium effect). They also reported disruptions as less disruptive, *t*(80) = –2.57, *p* < .05, *d* = –0.57 (medium effect), and greater confidence in managing them, *t*(80) = 5.63, *p* < .05, *d* = 1.24 (large effect). However, strategic knowledge of classroom management did not differ significantly between groups, *t*(80) = 1.00, *p* = .32, suggesting expertise was not associated with higher scores on the situational judgment test (SJT).

## Correlation Between Gaze Behavior Measures and Classroom Management Measures

Table 3 presents Pearson correlation coefficients between gaze behavior measures and classroom management measures, separately for experienced (below the diagonal) and inexperienced teachers (above the diagonal).[[5]](#footnote-6)

**Table 3**

*Correlations Between the Gaze Efficiency Measure and Classroom Management Measures for Experienced (Below Diagonal) and Inexperienced (Above Diagonal) Teachers*

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| Variable | (1) | (2) | (3) | (4) | (5) | (6) | (7) |
| (1) GRI | – | –.44\* | .11 | –.01 | .16 | .15 | –.14 |
| (2) Fixation Number Per Minute on AOI *Students* | –.53\* | – | –.39\* | .39\* | –.49\* | –.22 | .14 |
| (3) TTFF (log) on AOI *Disruptive Student* | .05 | –.29 | – | –.20 | .32\* | –.01 | .01 |
| (4) Disruptiveness Rating | .03 | –.20 | .02 | – | –.02 | .01 | –.15 |
| (5) Confidence Rating | 0 | .06 | –.06 | –.51\* | – | .48\* | –.19 |
| (6) Self-evaluated Classroom Management | .11 | 0 | –.18 | –.11 | .33\* | – | –.21 |
| (7) Strategic Knowledge | –.36\* | .16 | –.34\* | –.19 | –.03 | .12 | – |
| *Note.* The table presents Pearson correlation coefficients between gaze behavior measures (GRI = Gaze Relational Index, fixation number per minute on AOI s*tudents,* and TTFF (= log-transformed time to first fixation on AOI d*isruptive student*), and classroom management measures (self-evaluated classroom management, disruptiveness and confidence ratings, and strategic knowledge) for inexperienced teachers (above diagonal) and experienced teachers (below diagonal).  Statistically significant correlations (*p* < .05) are marked with an asterisk (\*).  GRI = Gaze-Relational Index; TTFF = Time to First Fixation on AOI d*isruptive student* (log-transformed). | | | | | | | |

For experienced teachers, lower GRI values (indicating more efficient gaze behavior) were significantly associated with higher fixation frequency on students and greater strategic knowledge. Additionally, more frequent fixations on students correlated with higher confidence in handling disruptions. Longer TTFF on disruptive students was linked to lower strategic knowledge.

For inexperienced teachers, greater fixation numbers on students were negatively correlated with self-evaluated classroom management and confidence ratings. Additionally, higher confidence in handling disruptions was positively associated with self-evaluated classroom management.

# Discussion

## Key Findings

Our study investigated …

Overall, our findings indicate that …

Our findings are consistent with prior research that illustrates the …

## Limitations and future directions

While the laboratory setting of the study allowed for a controlled implementation of stressors and high internal validity, it was not an authentic classroom environment, raising questions about its external validity. Most importantly, the teacher and their students did not have a shared history, and only a very thin basis for establishing a positive teacher-student relationship, which is a core characteristic of effective classroom management (Beaty-O’Ferrall et al., 2010; Rüedi, 2014).

In addition, the micro-teaching unit was only about 15 minutes long, and thus much shorter than a regular school lesson, providing less opportunities for experienced teachers to build up an engaging lesson.

Finally, the onset of disruptive student behavior was scripted, following an experimental time schedule, which was not affected by the behavior of the teacher. Thus, the setting may have masked effects of teaching experience by providing too little opportunities of experienced teachers to demonstrate their true classroom management skills.

In subsequent studies, it would therefore be insightful to …

## Conclusion

This study investigated …

In summary, our study contributes to the understanding …

# References

Barth, V. L. (2017). *Professionelle Wahrnehmung von Störungen im Unterricht*. Springer VS. https://doi.org/10.1007/978-3-658-16371-6

Beaty-O’Ferrall, M. E., Green, A., & Hanna, F. (2010). Classroom Management Strategies for Difficult Students: Promoting Change through Relationships. *Middle School Journal (J1)*, *41*(4), 4–11.

Blömeke, S., Gustafsson, J.-E., & Shavelson, R. J. (2015). Beyond dichotomies: Competence viewed as a continuum. *Zeitschrift Für Psychologie*, *223*(1), 3–13. https://doi.org/10.1027/2151-2604/a000194

Chow, J. C., Sayers, R., Fu, Y., Granger, K. L., McCullough, S., Kingsbery, C., & Morse, A. (2024). A Systematic Meta-Review of Measures of Classroom Management in School Settings. *Assessment for Effective Intervention*, *49*(2), 60–74. https://doi.org/10.1177/15345084231208671

Cohen, J. (1988). Statistical power for the behavioural sciences. Hilsdale. *NY: Lawrence Erlbaum*, *58*(1), 7–19.

Eckstein, B., Grob, U., & Reusser, K. (2016). Unterrichtliche Devianz und subjektives Störungsempfinden. Entwicklung eines Instrumentariums zur Erfassung von Unterrichtsstörungen. *Empirische Pädagogik (EP)*, *30*(1), Article 1.

Gold, B., & Holodynski, M. (2015). Development and Construct Validation of a Situational Judgment Test of Strategic Knowledge of Classroom Management in Elementary Schools. *Educational Assessment*, *20*(3), 226–248. https://doi.org/10.1080/10627197.2015.1062087

Grub, A.-S. (2023). *PRONOEA - Professional vision of novice and expert teachers* [doctoralThesis, Saarländische Universitäts- und Landesbibliothek]. https://doi.org/10.22028/D291-39788

Grub, A.-S., Biermann, A., & Brünken, R. (2020). *Process-based measurement of professional vision of (prospective) teachers in the field of classroom management. A systematic review*. https://doi.org/10.25656/01:21187

Helmke, A. (2022). *Unterrichtsqualität und Professionalisierung: Diagnostik von Lehr-Lern-Prozessen und evidenzbasierte Unterrichtsentwicklung* (1. Auflage). Klett Kallmeyer.

Helmke, A., Schrader, F.-W., Helmke, T., Lenske, G., Pham, G., Praetorius, A.-K., & Ade-Thurow, M. (2013). *Basisfragebogen EMU - Evidenzbasierte Methoden der Unterrichtsentwicklung*. Unterrichtsdiagnostik. http://www.unterrichtsdiagnostik.info/downloads/fragebogen/

Keskin, Ö., Gabel, S., Kollar, I., & Gegenfurtner, A. (2023). Relations between pre-service teacher gaze, teacher attitude, and student ethnicity. *Frontiers in Education*, *8*. https://doi.org/10.3389/feduc.2023.1272671

Keskin, Ö., Seidel, T., Stürmer, K., & Gegenfurtner, A. (2024). Eye-tracking research on teacher professional vision: A meta-analytic review. *Educational Research Review*, *42*, 100586. https://doi.org/10.1016/j.edurev.2023.100586

Kiel, E., Frey, A., & Weiß, S. (2013). *Trainingsbuch Klassenführung*. Verlag Julius Klinkhardt.

Kilbury, M., Böhnke, A., Haase, S., & Thiel, F. (2024). The development and validation of a video tool for capturing teachers’ noticing in salient and non-salient classroom disruptions. *Computers in Human Behavior Reports*, *16*, 100481. https://doi.org/10.1016/j.chbr.2024.100481

Kounin, J. S. (2006). *Techniken der Klassenführung*. Waxmann Verlag.

Kraft, M. A., & Monti-Nussbaum, M. (2021). The Big Problem With Little Interruptions to Classroom Learning. *AERA Open*, *7*, 23328584211028856. https://doi.org/10.1177/23328584211028856

Kunter, M., Baumert, J., & Blum, W. (2011). *Professionelle Kompetenz von Lehrkräften: Ergebnisse des Forschungsprogramms COACTIV*. Waxmann Verlag.

Lohmann, G., & Meyer, H. (2003). *Mit Schülern klarkommen: Professioneller Umgang mit Unterrichtsstörungen und Disziplinkonflikten*. Cornelsen-Scriptor.

Marder, J., Thiel, F., & Göllner, R. (2023). Classroom management and students’ mathematics achievement: The role of students’ disruptive behavior and teacher classroom management. *Learning and Instruction*, *86*, 101746. https://doi.org/10.1016/j.learninstruc.2023.101746

Marzano, R. J. (2007). *The Art and Science of Teaching: A Comprehensive Framework for Effective Instruction*. ASCD.

Onkhar, V., Dodou, D., & de Winter, J. C. F. (2024). Evaluating the Tobii Pro Glasses 2 and 3 in static and dynamic conditions. *Behavior Research Methods*, *56*(5), 4221–4238. https://doi.org/10.3758/s13428-023-02173-7

RStudio Team. (2020). *RStudio: Integrated Development Environment for R*. RStudio, PBC.

Rüedi, J. (2014). Zur Bedeutung positive Beziehungen für die Klassenführung und den Umgang mit Unterrichtsstörungen. *Beziehungen in Schule Und Unterricht. Teil*, *3*, 105–126.

Tobii AB. (2024). *Tobii Pro Lab User Manual v 24.21*. https://go.tobii.com/tobii\_pro\_lab\_user\_manual

Voss, T., Kunter, M., Seiz, J., Hoehne, V., & Baumert, J. (2014). Die Bedeutung des pädagogisch-psychologischen Wissens von angehenden Lehrkräften für die Unterrichtsqualität. *Zeitschrift für Pädagogik*, *60*(2), 184–201.

Wickham, H. (2016). *ggplot2: Elegant Graphics for Data Analysis*. Springer-Verlag New York. https://doi.org/10.1007/978-0-387-98141-3

Yamamoto, T., & Imai-Matsumura, K. (2013). Teachers’ Gaze and Awareness of Students’ Behavior: Using An Eye Tracker. *Comprehensive Psychology*, *2*, 01.IT.2.6. https://doi.org/10.2466/01.IT.2.6

Zierer, K. (2015). *Jahrbuch für Allgemeine Didaktik 2015: Thementeil: Klassenmanagement / Klassenführung - Perspektiven, Befunde, Kontroversen*. wbv Media GmbH & Company KG.

# Appendix

# Appendix A

**Eye-tracking apparatus and calibration**

Teachers wore a binocular Tobii Pro Glasses 2 eye-tracker during the micro-teaching unit to record eye-tracking data. The system consisted of a wearable head unit and a recording unit. The head unit was a measuring device with different sensors. A high-definition scene camera captured a full HD video of the teacher’s field of vision. An integrated microphone recorded surrounding sounds. Infrared light illuminators supported the eye-tracking sensors which recorded the eye orientation to capture the teacher’s gaze point. The videos were recorded with a sampling rate of 50 Hz in a video resolution of 1920 x 1080 at 25 frames per second. The scene camera had a field of view of 90 degrees in 16:9 format (82 degrees horizontal and 52 degrees vertical) and a frame dimension of 179 x 159 x 57 mm (width x depth x height). The recording unit is a compact computer that manages the head unit. It captures and saves eye-tracking data, audio, and scene camera footage on a removable SD memory card.

The evaluation of the calibration process followed the guidelines outlined in the Manual of Tobii AB (2024) and Onkhar et al. (2024) for assessing calibration quality. Participants’ gaze was calibrated using a bullseye card that the participant held at arm’s length. A successful calibration was achieved when the participant’s gaze marker sufficiently overlapped with the bullseye for a specified time, based on criteria internally determined by the manufacturer’s software (Tobii AB, 2024). All participants achieved successful calibration, and no participants were excluded due to calibration failure. The robustness of the calibration was further verified through a secondary nine-point calibration. During this step, participants were asked to read numbers from one to nine aloud and direct their gaze at specific fields corresponding to each number. The initial calibration was performed before and the verification calibration before and after each micro-teaching unit.

**Table A1**

*Classification of Nine Typical Classroom Disruptions According to Lohmann & Meyer (2003) Performed in The Micro-Teaching Unit by Actors*

|  |  |  |
| --- | --- | --- |
| Verbal disruptions | Physical disruptions | Lack of eagerness to learn |
| Heckling | Clicking pen | Looking at phone |
| Chatting | Snipping hands | Drawing |
| Whispering | Drumming hands | Head on table |

*Note.* Disruptions were classified based on the typology provided by Lohmann & Meyer (2003)*.* Categories include verbal, physical, and disengagement-related behaviors performed during the micro-teaching unit. The order of the performing actors and the disruptions was fully balanced using Latin squares.

# Appendix B

# Laboratory Setting of The Study

**Figure B1**

*Laboratory Setting of The Micro-Teaching Unit. Ein Bild, das Mobiliar, Stuhl, Kleidung, Schuhwerk enthält.

Automatisch generierte Beschreibung*

*Note*. The setting included three actors as the class (left) and a teacher (participant, right).

**Figure B2**

*Laboratory Setting of The Interview.*

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

Automatisch generierte Beschreibung

*Note*. The experimenter and participant watched the previously taught micro-teaching unit on video.

# Appendix C

**Additional Results for Gaze Behavior Across AOI**

**Figure C1**

*Average Fixation Number Percentages by Area of Interest (AOI) and Teacher Experience Group for the Entire Micro-Teaching Unit with 95% Confidence Intervals*



*Note.* The bar chart illustrates the average fixation number percentage directed at five AOIs (Students, Teacher Material, Classroom/Others, Student Desk, and Disruptive Student) during the micro-teaching unit. Results are presented separately for experienced and inexperienced teachers, with error bars indicating 95% confidence intervals.

**Table C2**

t*-Test Results and Effect Sizes for Fixation Number Percentages (FNP) and Fixation Duration Percentages (FDP) Across AOIs Between Experienced and Inexperienced Teachers*

|  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- |
| Measures | AOI | *t-*value | df | *p*-value | *M* Difference | Cohen's *d* | 95% CI |
| FNP a | Students | 0.01 | 80 | 0.99 | 0.03 | 0.00 | [–4.63, 4.69] |
| FNP | Disruptive Student | –1.32 | 80 | 0.19 | –0.27 | –0.29 | [–0.67, 0.14] |
| FNP | Teacher Material | 0.89 | 80 | 0.37 | 2.32 | 0.20 | [–2.84, 7.48] |
| FNP | Student Desk | –1.17 | 80 | 0.25 | –1.72 | –0.26 | [–4.65, 1.21] |
| FNP | Classroom/Others | –0.43 | 80 | 0.67 | –0.63 | –0.10 | [–3.53, 2.27] |
| FDP b | Students | –0.43 | 80 | 0.67 | –1.26 | –0.10 | [–7.1, 4.58] |
| FDP | Disruptive Student | –1.90 | 80 | 0.06 | –0.48 | –0.42 | [–0.98, 0.02] |
| FDP | Teacher Material | 1.55 | 80 | 0.13 | 4.45 | 0.34 | [–1.28, 10.18] |
| FDP | Student Desk | –1.67 | 80 | 0.1 | –2.89 | –0.37 | [–6.35, 0.56] |
| FDP | Classroom/Others | –0.28 | 80 | 0.78 | –0.30 | –0.06 | [–2.43, 1.84] |
| *Note*. This figure displays the results of *t*-tests and effect sizes (Cohen’s *d*) for fixation number percentages (FNP) and fixation duration percentages (FDP) across Areas of Interest (AOIs) between experienced and inexperienced teachers. AOIs included Students, Disruptive Student, Teacher Material, Student Desk, and Classroom/Others. Positive mean differences (*M* Difference) indicate higher percentages for experienced teachers. Confidence intervals (95%) for the mean differences are presented in brackets.  *p*-value < .05 is considered statistically significant.  a FNP = Fixation Number Percentages  b FDP = Fixation Duration Percentages | | | | | | | |

**Appendix D**

**Correlation Between Gaze Behavior Measures and Classroom Management Measures**

**Table D1**

*Correlations Between the Gaze Behavior Measures and Classroom Management Measures Across All Participants*

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| Variable | (2) | (3) | (4) | (5) | (6) | (7) |
| (1) GRI | –.50\* | .08 | .06 | –.02 | .07 | –.23\* |
| (2) Fixation Number Per Minute on AOI *Students* | – | –.34\* | 0 | –.13 | –.06 | .16 |
| (3) TTFF (log) on AOI *Disruptive Student* | –.34\* | – | –.07 | .12 | –.01 | –.12 |
| (4) Disruptiveness Rating | 0 | –.07 | – | –.36\* | –.13 | –.03 |
| (5) Confidence Rating | –.13 | .12 | –.36\* | – | .49\* | –.06 |
| (6) Self-Evaluated Classroom Management | –.06 | –.10 | –.13 | .49\* | – | –.05 |
| (7) Strategic Knowledge | .16 | –.12 | –.03 | –.06 | –.05 | – |
| *Note.* The table presents Pearson correlation coefficients between gaze behavior measures (GRI = Gaze Relational Index, fixation number per minute on AOI *students*, and TTFF = log-transformed time to first fixation on AOI *disruptive student*) and classroom management measures (self-evaluated classroom management, disruptiveness and confidence ratings, and strategic knowledge) across all participants. Lower GRI values indicate more efficient gaze behavior. Statistically significant correlations (*p* < .05) are marked with an asterisk (\*). | | | | | | |

1. As our focus was on gaze behavior directed toward the disruptive student, instances where the gaze was already fixated on the disruptive student, when the event started, were excluded from the analysis. [↑](#footnote-ref-2)
2. Fixation times beyond 30 seconds were excluded, as they exceeded the duration of the scripted disruptions. [↑](#footnote-ref-3)
3. The percentage of fixation duration was used as the primary measure, as it offered a more intuitive visualization of attention distribution over time (see Appendix C, Figure C1 for the percentage for fixation numbers). [↑](#footnote-ref-4)
4. Before conducting the analysis, 21.33% of data points were excluded because the participant either did not fixate on the disruption or had fixation times of zero or longer than 30 seconds, exceeding the defined threshold (see Data Analysis). Removing them ensured a focus on meaningful and interpretable gaze detection times. [↑](#footnote-ref-5)
5. See Appendix D (Table D1) for a correlation table including all gaze behavior measures and classroom management measures for all participants. [↑](#footnote-ref-6)