

**Ego OR Exo:
Comparing Visual Perspectives on Guidance
Visualisations for Motor Learning**

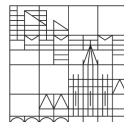
Masterarbeit

vorgelegt von

Stefan Paul Feyer

an der

Universität
Konstanz



Sektion Mathematik und Naturwissenschaft

Fachbereich Informatik und Informationswissenschaft

1.Gutachter: Prof. Dr. Harald Reiterer

2.Gutachter: Dr. Karsten Klein

Konstanz, 2021

Abstract

Motor Learning in Mixed Reality proved to be good. But view reseach in the influence of the perspective on guidance visualisation. This work proposes a study to investigate this. Task is handling physical load. will enable designers of MR ML systems to base their work on on empirical data.

Contents

Abstract	i
Contents	ii
List of Figures	iii
List of Tables	iv
1 Introduction	1
1.1 Outline	2
2 Motor Learning in Virtual Reality	3
2.1 Mixed Reality	3
2.2 Motor Learning	3
2.3 Visual Perspectives	4
2.4 Handling Physical Load - chang to mmh	5
2.5 Ergonomic Risk Measurements	5
2.6 Related Work: Motor Learning in Virtual Reality	6
3 Experiment Design and System Implementation	9
3.1 Experiment Design	9
3.2 E(x g)o- Design and Implementation	18
4 Study Evaluation	32
4.1 Study Evaluation	32
5 Conclusion	33
5.1 System and Study	33
5.2 Outlook	33
6 Attachments	34
6.1 Study Documents	34
References	vii

List of Figures

2.1	Movement classifications by Schmidt et al.	4
2.2	Centricity continuum	5
2.3	Possible perspectives	6
3.1	speed mechanic chart	10
3.2	Study structure	15
3.3	study setting	21
3.4	learner positions	22
3.5	Positions of exo-centric GVs	24
3.6	sub task rating	28
3.7	sub task rating	29

List of Tables

2.1	Overview seminar evaluation	7
2.2	Detailed analysis of related work in seminar thesis.	8
3.1	Mechanics for Motor Learning in Virtual Reality	11
3.2	Description of sub-tasks	14
3.3	Description of tasks	16
3.4	logging detail	25

1 Introduction

The acquisition of movements is a crucial part of human development. Learning movements empowers to be more efficient, faster and more exact. The capability of enhanced movements enables the learner to survive from the very beginning. The process of learning movements is called Motor Learning. Nowadays, Motor Learning is still crucial. Especially for tasks like sports, arts or the ergonomic handling of physical load.

Most movements we learn by voyeurism and mimicking: watching and trying it out by yourself. Mastering a movement is performed best with an experienced teacher. A teacher is hardly replaceable because of immediate visual, audible and haptic feedback on a performed movement. However, if a teacher is not available, for example, based on the location or economic reasons, other sources can be used to learn movements. For example, YouTube¹, TikTok², and other video platforms have become a great source for learning videos with a wide range of purposes. The downside of videos is the two dimensional (2D) experience of a three dimensional (3D) movement. Mixed Reality (MR) can provide this experience in 3D. Furthermore, MR can provide feedback on the performed movement and has the ability for interactions with the virtual guidance visualisation. MR already proved to be a suitable environment for Motor Learning for tasks like dancing [1–5], sports [6, 7], Rehabilitation [8–12], arts [13–20] and others [21, 22].

In the real world, where the student and teacher are real persons, the student sees the teacher, for example, in front of himself/herself. This perspective is called the exo-centric visual perspective. Nevertheless, if we move from the real world to the virtual world of MR, we are no longer restricted to the exo-centric visual perspective. The teacher can be rendered inside the student's body, allowing the student to see the teacher from an ego-centric perspective. The change from the exo-centric to the ego-centric visual perspective potentially influences Motor Learning, shown by previous research; for example, AR-Arm [13] lets the learner experience the movements from an ego-centric perspective. YouMove [1] teaches dance from an exo-centric perspective. OneBody [19], Light Guide [22], MR Dance Trainer [5], Free Throw Simulator [6], Training Physical skills [7], Sleeve AR [11] and Thai Chi Trainer [20] use both visual perspectives. However, only OneBody, LightGuide and TaiChi Trainer found a difference between the visual perspectives. Furthermore, none of these works investigated how the visual perspective influences the performance of the learner. Another topic where MR could be a valuable helper is the ergonomic conduction of movements while handling physical load [23, 24]. Handling physical load in the correct ergonomic conduct in working routines can prevent injuries in everyday life. However, a kinaesthetics teacher is not always accessible, for example, for economic reasons. The influence of the visual perspective on a virtual guidance visualisation teaching the handling of physical load in mixed reality is sparsely investigated **todo: is there a source?**. Especially, locomotion movements like walking in the ego-centric perspective is left out. The lack of research in the influence of the visual perspective on a virtual guidance visualisation, especially for handling physical loads, shows the necessity of investigations on:

RQ1: How does the visual perspective on a virtual guidance visualisation influence Motor Learning in Virtual Reality environments?

To answer this main research question RQ1, several aspects have to be taken into account: accuracy of movements, transfer of information of how to move, the visual focus of the learner and last but not least, the personal preference of the learner. Therefore, to answer the main research question RQ1, it is necessary to answer the following sub-research questions:

¹<https://www.youtube.com/>, accessed 17.2.2021

²<https://www.tiktok.com/>, accessed 17.2.2021

RQ1.1 How does the visual perspective on a virtual guidance visualisation influence movements' accuracy?

RQ1.1.1 How does the visual perspective on a virtual guidance visualisation influence movements' accuracy of the own body?

RQ1.1.2 How does the visual perspective on a virtual guidance visualisation influence the accuracy of handling physical load?

RQ1.1.3 How does the visual perspective on a virtual guidance visualisation influence sub-tasks' accuracy?

RQ1.2 Does the visual perspective on a virtual guidance visualisation influence the transfer of ergonomic principles?

RQ1.3 How does the visual perspective on a virtual guidance visualisation influence the learner's visual focus?

RQ1.4 What is the subjective personal preference of the learner for the visual perspectives?

A detailed discussion of the research questions can be found in 3.1.

The answers to these research questions will enable designers of VR Motor Learning training systems to choose a suitable visual perspective on an empirical basis.

1.1 Outline

This work proposes a study design to answer the research question. To design this study on a solid basis, the theoretical foundations are laid in chapter 2 with a closer look on Motor Learning (section 2.2), visual perspectives (section 2.3) and Mixed Reality (section 2.1). These sections result in the scope and parameters of the study design. Section 2.6 investigates previous works and illustrates the conceptual delimitation of this work from what has already been investigated. Chapter 2 concludes with a research contribution statement, clarifying the Empirical Contribution and Artifact Contribution of this work.

For the proposed study, a system had to be designed to produce data to answer the research questions. This system is called E(x|g)o. The design and implementation is described in section 3.2 followed by the design of the study itself in section 3.1.

E(x|g)o and the study design have been evaluated in a pilot study. The results of the evaluation are depicted in chapter 4. Furthermore, this chapter suggests improvements in the study design in section 5.2. This work concludes in chapter 5 with an outlook on how E(x|g)o can be enhanced and expanded as well as used for further investigations.

long long label asdad asfiojh aojgf oijgoias gjoasig

2 Motor Learning in Virtual Reality

This chapter provides the theoretical background of Mixed Reality, Motor Learning, Visual Perspectives and handling physical load. These are the most important aspects that serve as the foundation for the proposed study design. Finally, an analysis of related work is provided. This chapter gives insights into how this work is informed and differentiated by other researchers.

2.1 Mixed Reality

Milgram and Kishinho [25] describe Mixed Reality for visual displays on a continuum (see seminar thesis chapter 2.3). Virtual Reality is purely digital, and thereby the environment is blocked entirely. In Augmented Reality, the environment is visible and augmented with digital elements. During Motor Learning, the visual perception of the own body is desirable. Though, the approach of augmenting the real-world body with a virtual guidance visualisation is promising. However, today's AR-technology provides a small field of view. A solution to this could be the video see-through technology, but it is limited by latency and distortion.

The body's perception can also be achieved by tracking the learner's body and render it over the learner's physical body. Though, the visual perception of the learner's body can be established in VR. Consequentially, this work will focus on Motor Learning in Virtual Reality.

2.2 Motor Learning

Motor Learning is achieved through instruction, trying, imitation or a combination of them [26]. The process of Motor Learning can be divided into three parts: cognitive stage, associative stage and autonomous stage. In the cognitive stage, training methods are most efficient, and the performance gain is the highest among the stages [26]. Tasks that belong to this stage are thereby best suited for a study. A detailed description of the stages can be found in the preceding seminar thesis chapter 2.2.

Movements can be classified by two means: by the *particular movements* and based on the *perceptual attributes*. Based on the *particular movements*, the classification is described by a continuum, compare figure 2.1 left. On the extremes of the continuum are discrete movements and continuous movements. Between these extremes, serial movements are located. Discrete movements are too short for an evaluation. Continuous movements do not have a recognisable beginning, and thereby they are not suitable for the study in question either. Serial movements are chained discrete movements with a recognisable beginning and end. This allows a task decomposition and an evaluation of particular sub-tasks. Discrete movements are widely used for research in Motor Learning, for example [16, 17, 22]; therefore, the study task design is based on discrete movements.

The classification based on the *perceptual attributes* is also represented by a continuum and includes the environment in which the movement is performed, compare figure 2.1 right. At the extremes of the continuum, open skills and closed skills are located. For closed skills, the environment is predictable, while in open skills, the environment is not predictable. The study aims to analyse the learners' performance of following a movement and

not how they can adapt to environmental changes. Thereby, this study's task must be located on the left-hand side of the continuum: closed skills.

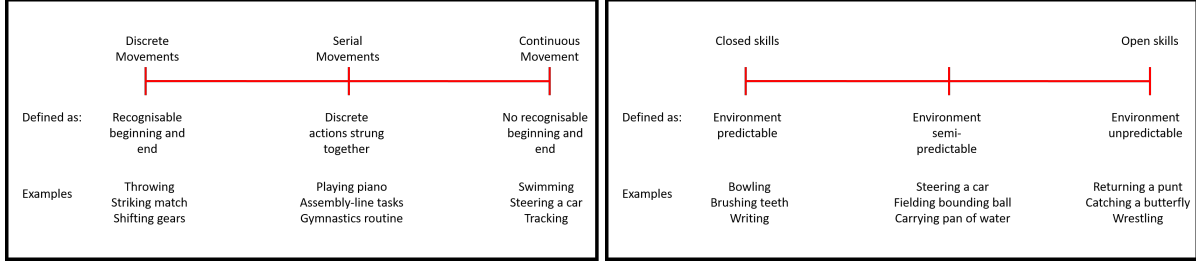


Figure 2.1: Movement classification by *particular movements* (left) and *perceptual attributes* by Smift at al. [26]

2.2.1 Measurements for Motor Learning

The movements of a teacher and the movement of a learner differ. To assess the difference between the two movements, two main classes of measures can be applied [26]: *measures of error for a single object* and *measures of time and speed*. *Measure of error for a single object* represent the degree to which the target movement is amiss. Schmidt et al. [26] provide five *error measures* to calculate this error, compare seminar thesis chapter 2.2. *Constant Error* is the most common measure in related work to determine the difference between the movement of the learner and the movement of the teacher, for example [1, 2, 9, 19, 20, 22, 27]. Constant Error is defined as the average error between the movement of the learner and the movement of the teacher and is described as

$$CE = \frac{\sum_i (x_i - T)}{n} \quad (2.1)$$

with x_i : actual value, T : target value, n : number of values [26].

The basic idea of *measures of time and speed* is that a performer who can accomplish more in a given amount of time or who can accomplish a given amount of behaviours in less time is more skilful. In related work, this measure is mostly assessed by the task completion time, for example [19, 22, 27].

2.3 Visual Perspectives

Wang and Milgram [28] describe visual perspectives by the *centricity continuum* 2.2. On the left extreme on the continuum, the ego-centric visual perspective is located; on the right extreme, the exo-centric visual perspective can be found, while the middle part represents tethered visual perspectives. By moving from the left to the right, the so-called *tethering distance* increases. The *tethering distance* describe the distance of the anchor point of the eyes to the object to control. In this work, the object to control is the human-shaped guidance visualisation (avatar). Furthermore, Wang and Milgram distinguish tethered visual perspective in dynamic and rigid. A detailed description is given in the seminar thesis chapter 2.1. Given a scenario where one learner mimics the movement of one teacher, five different visual perspectives are possible:

- **Ego-centric:** the avatar of the teacher is located inside the body of the avatar of the learner; the learner sees the guidance visualisation inside the own body, compare figure 2.3 top left.

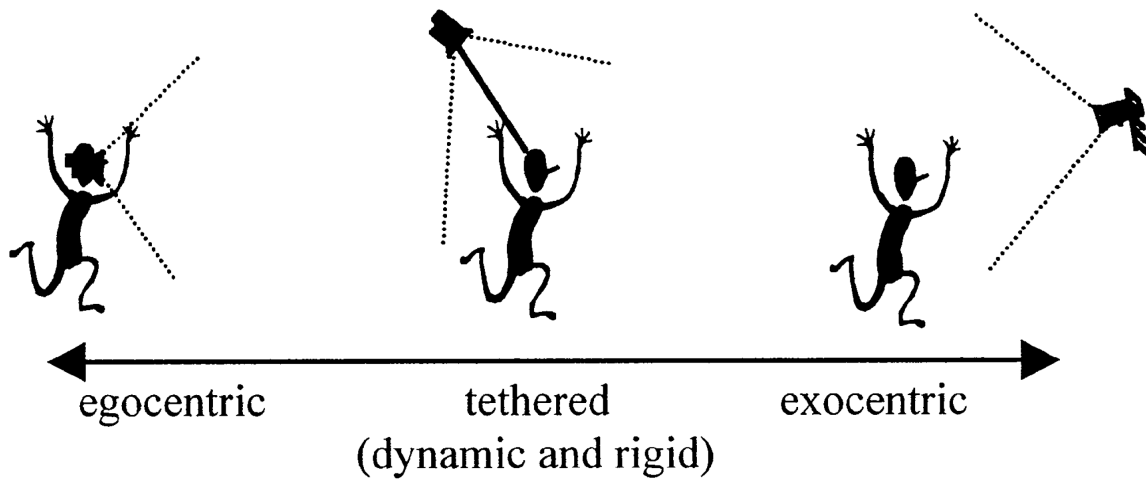


Figure 2.2: Centricity continuum by Wang and Milgram [28]

- **Exo-centric:** the avatar of the guidance visualisation is located outside of the avatar of the learner; the learner sees the guidance visualisation, e.g. in front of him/her, compare figure 2.3 top right.
- **Ego & exo-centric:** the combination of ego-centric and exo-centric. The learner sees the guidance visualisation as well as inside and outside of the own body, compare figure 2.3 middle left.
- **Augmented exo-centric:** the guidance visualisation is located outside of the learner's avatar. Additionally, a virtual copy of the student is located inside the exo-centric guidance visualisation, compare figure 2.3 middle right.
- **Ego & augmented exo-centric:** the combination of the ego-centric visual perspective and the augmented exo-centric visual perspective; the learner sees the guidance visualisation inside the own body, as well as outside. Additionally, a virtual copy of the learner is located inside the exo-centric guidance visualisation, compare figure 2.3 bottom.

2.4 Handling Physical Load - chang to mmh

The handling of physical load is composed of five elemental tasks: lift, lower, push, pull and hold [29]. Additionally, there are non-elemental tasks like turning and sliding, *ibid.*. This work will use a study tasks that include the handling of physical load. Evidently, the task should consist of these elemental tasks. A task that consists of elemental tasks can be generalised to other tasks to a certain extend. To gain a stronger data basis by repeating, multiple elemental tasks can be chained together, to form a so-called Unit-Combined-MMH, *ibid.*. In chapter ?? is described how the elemental tasks become sub-tasks of the study task.

2.5 Ergonomic Risk Measurements

Risk Measurements (RM)

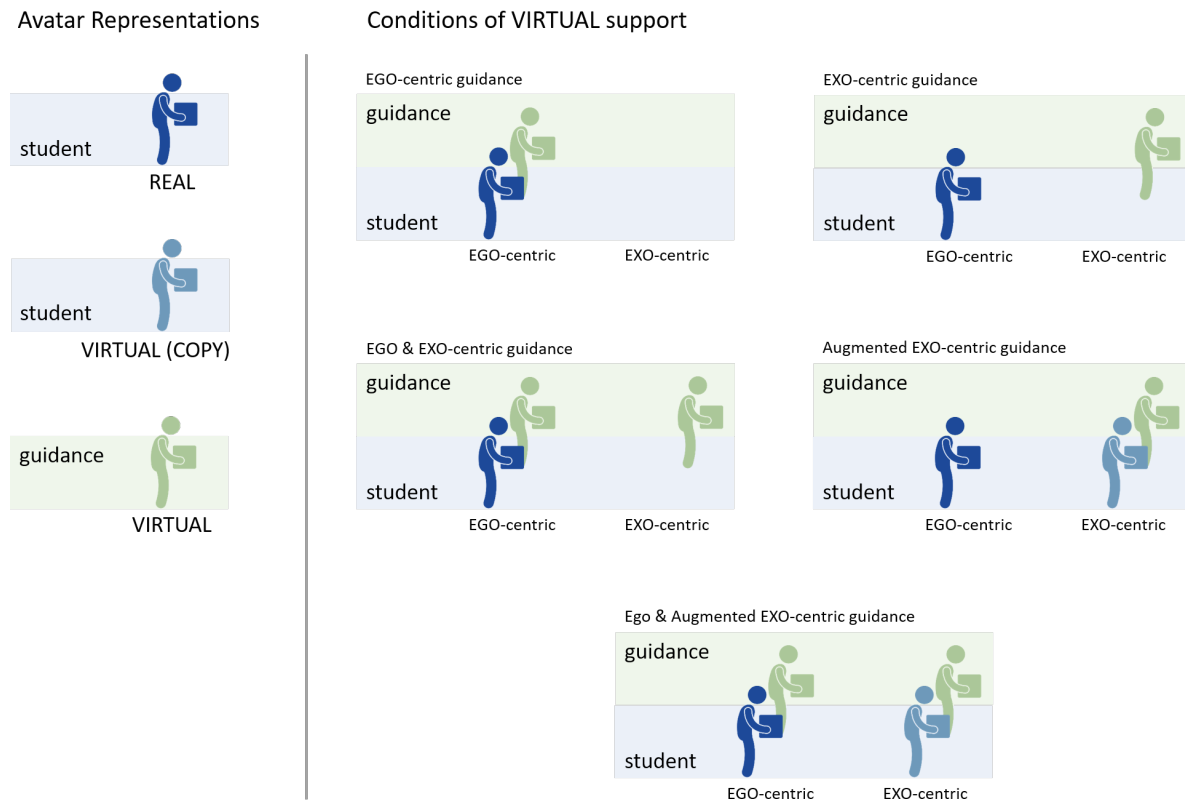


Figure 2.3: Possible perspectives with one real-world student and one real-world teacher.

2.6 Related Work: Motor Learning in Virtual Reality

Training movements in Virtual Reality was investigated previously in several works. The preceding seminar thesis (see chapter 3) provided an overview over 23 (compare table 2.1) of these works and evaluated six of them in detail: Tai Chi Trainer by Chua et al. [20], YouMove by Anderson et al. [1], VR Dance Trainer by Chan et al. [2], OneBody by Hoang et al. [19], LightGuide by Sodhi et al. [22] and Pyhsio@Home by Tang et al. [9]. Special attention was paid to the visual perspective, task, guidance visualisation and their independent and dependent variables they used in their investigations. Finally, the results of these works were concluded. An overview is depicted in table 2.2. These works inform this work in various aspects. Chua et al. used the ego & augmented exo-centric visual perspective, Hoang et al. and Sodhi et al. the ego-centric visual perspective. These visual perspectives proved to be suited for the evaluation of Motor Learning in VR and is adopted for the proposed study design, compare section 2.3. Furthermore, Chan et al. and Chua et al. used high realistic avatars as guidance visualisation, which are used in the proposed study design, compare seminar thesis chapter 3.3. Additionally, recent research indicates that high realism avatars outperform abstract avatars [27, 30]. All authors used a performance measure to evaluate the performed movements of the participants of their studies. Primarily the distance-based measures informed the measures used in the proposed study design.

The above mentioned works do not use the relatively new technology of Vive Trackers in combination with Inverse Kinematics (IK, see project report chapter 2.1 and 2.2). Sra et al. [31] used this technology in 2018 for their system Your Place and Mine to render human-shaped avatars.

The results of related work yielded no clear conclusion about the influence of the perspectives on motor learning. Chua et al. found no difference in the performance between the visual perspectives, Anderson et al. and

2 Motor Learning in Virtual Reality

Ego-centric	Exo-centric	Ego Exo-centric
AR-Arm (Han et al. 2016)	MotionMA (Velloso et al. 2013)	OneBody (Hoang et al. 2016)
Just Follow Me (Yang & Kim 2002)	YouMove (Anderson et al. 2013)	LightGuide (Sodhi et al. 2012)
Gohstman (Chinthammit et al. 2014)	VR Dance Trainer (Jacky Chan et al. 2010)	MR Dance Trainer (Hachimura et al. 2004)
Stylo and Handifact (Katzakis et al. 2017)	Physio@Home (Tang et al. 2015)	Free Throw Simulator (Covaci et al. 2014)
GhostHands (Scavo et al. 2015)	OutSide me (Yan et al. 2015)	Training Physical Skill (Kojima et al. 2014)
	e-Learning Martial Arts (Komura et al. 2006)	SleeveAR (Sousa et al. 2016)
	My Tai-Chi Coaches (Han et al. 2017)	Tai Chi Trainer (Chua et al. 2006)
	Performance Training (Chan et al. 2007)	
	RT Gesture Recognition (Portillo et al. 2008)	
	KinoHaptics (Rajanna et al. 2015)	
	TIKL (Lieberman & Breazeal 2007)	

Dance
Sports
Rehab
Arts
Abstract

Table 2.1: Overview of related work divided by perspective and task

Chan et al. found out that their exo-centric visual perspectives in Virtual Reality outperform traditional video guidance. Hoang et al. and Sodhi et al. conclude that the ego-centric perspective outperforms the exo-centric visual perspective. Nevertheless, an investigation of how the visual perspective influences motor learning was not investigated. Recently, in December 2020, Yu et al. [27] conducted three independent studies to close this gap. In the first study, Yu et al. compared the ego-centric visual perspective and a 2D-mirror for single arm movements. In the second study, they compared the ego-centric and exo-centric visual perspective for Yoga. In the third study, they compared the ego-centric visual perspective with a 3D-mirror for arm movements. Yu et al. conclude their findings in a design guideline for systems training Motor Learning in Virtual Reality: use the ego-centric visual perspective if the type of motion allows, consider alternatives for other types of motions, *ibidem*. In all three studies, the ego-centric visual perspective outperformed the other perspectives if the movement was completely visible from the ego-centric visual perspective. This work, in contrast, focuses on full-body movements that include the handling of physical load. Furthermore, this work provides a third visual perspective, where the ego-centric and exo-centric visual perspective is combined.

2.6.1 Research Contribution Statement

todo: new papers occurred, read them, then write this statement. **notes:** what is done: comparing ego-centric with exo-centric video. comparing ego-centric with exo-centric and the combination, but yielded to no result, because old paper and old pc comparing with mirrors, comparing isolated body parts everyone made his live easy by just looking at stationary movements, mostly containing only some body parts. new: nobody did fullbody movements with locomotion. ego-centric locomotion motion guidance is completely new. related work only investigated on stationary movements. but motor learning is not stationary. body parts is also not stationary. real-world relation poor because of arts dance or abstract. my work is the first one having really a task that is reasonable!

Previous work investigated the differences between the perspectives, but: To my knowledge, there is no investigation on full body movements that include locomotion. Furthermore, there are no investigations that include the handling of physical load. Previous works compared ego-centric Motor Learning with video learning[1, 2], augmented mirrors[19, 27]. The conduction of the proposed study will produce data that serves as a reasonable basis for designers of VR Motor Learning systems choosing a suitable perspectives. This is achieved by an Empirical Research Contribution. The empirical data is gathered by a comparative study between the ego-centric

2 Motor Learning in Virtual Reality

	Tai Chi Trainer	YouMove	VR Dance Trainer	OneBody	LightGuide	Physio@Home
Perspective	Exo-centric, Ego & Augmented Exo-centric	Exo-centric	Exo-centric	Ego-centric, Exo-centric	Ego-centric, Exo-centric	Exo-centric
Task	Tai Chi	Dance (Ballet), abstract	Dance (HipHop)	Martial Arts	Abstract	Shoulder rehab
Guidance Visualisation	hr avatar, wireframe, mimic avatar	Stick figure, mimic avatar	hr figure, mimic avatar	Stick figure, mimic avatar	Indicators, follow/mimic	Indicators
Variables	Perspectives, performance measure	VR/Video, performance	Video/VR, performance	Training method, performance	Visualisations, Perspective, Performance	Visualisation, performance
Results	No difference in performance	VR better than video	VR better than video	Ego better than exo	Ego better than exo	Multi view better than single view

Table 2.2: Detailed seminar thesis evaluation.

visual perspective, the exo-centric visual perspective and the combination. As novelty, the task includes handling of physical load which consists of the elemental tasks of manual material handling. This allows an evaluation of the elemental tasks per visual perspective and can give insights which perspective is suited for specific tasks. Additionally, an artifact contribution is provided by the ego-centric guidance of locomotion movements.

3 Experiment Design and System Implementation

3.1 Experiment Design

3.1.1 Independent Variables: Visual Perspectives

The last chapter pointed out five possible visual perspectives in a scenario with one teacher and one student, compare figure 2.3. All visual perspectives are worth an investigation and a comparable study with all five visual perspectives is desirable. Though, to reduce complexity and the number of participants¹, this work will focus on three visual perspectives.

Figure 2.3 shows three main classes of visual perspectives: ego-centric, exo-centric and perspectives which contain both. To answer the research question, it is indispensable to examine at least one of each class. The ego-centric VP is unique and though chosen by default. The exo-centric VP can be realised as purely exo-centric or augmented exo-centric. The combination of ego-centric and exo-centric can be realised as ego & exo-centric or ego & augmented exo-centric. But before the exo-centric VP and the combination can be chosen, a closer look on the mechanics that makes Motor Learning in VR possible is necessary.

Excursion: Mechanics for Motor Learning in Virtual Reality

For teaching movements in Virtual Reality, in the exo-centric visual perspective the following issue arises: the GV can move out of the field of view of the learner by the movement itself. Scenario: the learner and the GV stand side-by-side, the learner sees the GV to the left. The GV now indicates a movement to turn by 90 degrees to the right. As soon as the learner follows this movement, the GV will be located behind the learner after the movement ended. A GV standing behind the learner cannot be seen by the learner.

This issue is solved in existing work with either the restriction of movements [6, 16] or multiple representations of the GV around the learner [17, 20]. The restriction of movements has a strong influence in the task design and is therefore not desirable for the study proposed in this thesis. Consequentially, for exo-centric visual perspectives multiple representations for the GVs on strategic positions around the learner are necessary.

In the ego-centric visual perspective, another issue arises during the teaching of locomotion movements. To understand this issue, two aspects have to be clear before. (1) The nature of an ego-centric GV is to be located inside the learner at any time. (2) A GV indicates movements by moving itself. If the guidance visualisation is about to indicate a movement away from the learner, the GV is moving out of the student's body. But a GV that is outside of the learner's body is no longer ego-centric.

A possible solution is given by the centricity continuum by Wang and Milgram 2.2. Following the nature of the centricity continuum, the tethering distance can be increased by a small amount and the visual perspective can still be classified as ego-centric. But now arises the question, of how far the tethering distance can be increased, with which the perspective still feels ego-centric, but the indication of the movement is considerable. For simplicity, this distance is further called ego-centric tethering distance (ETD). To determine a reasonable ETD,

¹Due to COVID-19 pandemic

3 Experiment Design and System Implementation

a formative study was conducted. During this study, one² non biased³ person was asked to follow movements in the ego-centric visual perspective. The first movement was conducted with an ETD of 5cm. For the following movements the ETD was increased by 5cm each. The subjective assesment of the participant and my observations yielded best for an ETD between 15cm and 30cm. These two values are further called:

$$ETD_{min} = 15cm$$

$$ETD_{max} = 30cm$$

Based on ETD_{min} and ETD_{max} the speed mechanic is developed. The speed mechanic controls the speed of the playback of the GV. At ETD_{min} and below, the animation plays at normal speed. At ETD_{max} the GV stops. Between ETD_{min} and ETD_{max} the animation speed of the GV is linearly interpolated, compare figure 3.1. The speed-mechanic was evaluated by one person⁴. The participant followed the GV in the ego-centric visual

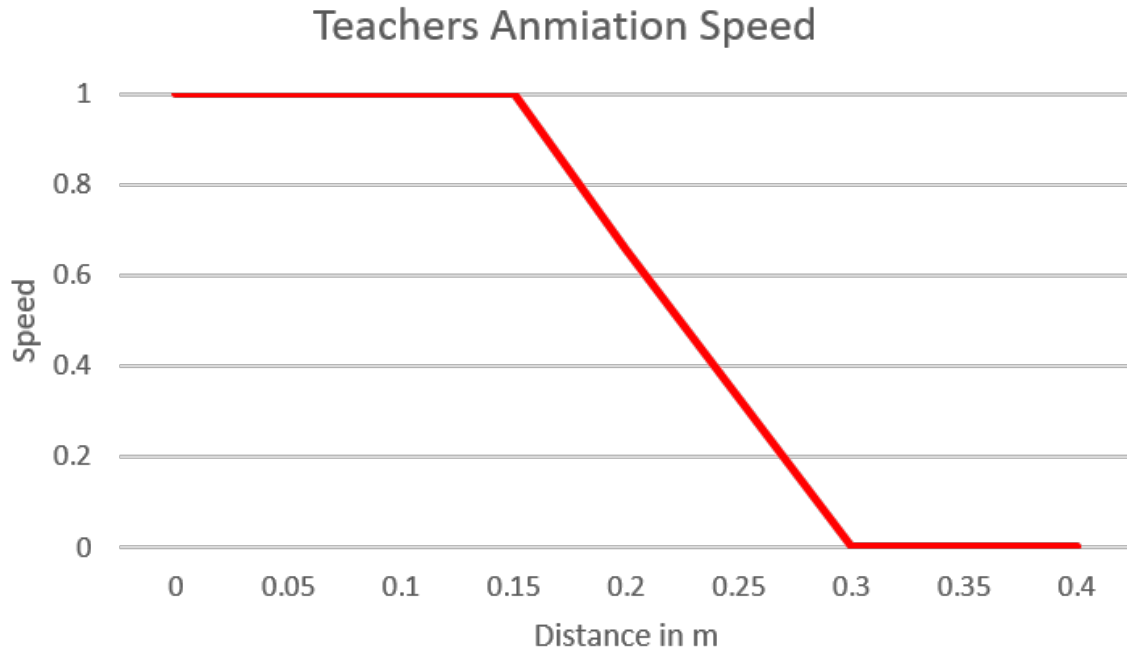


Figure 3.1: speed mechanic chart

perspective. Observations showed that the participant could follow the movement at ease. The opinion of the participant about the speed mechanic was very positive.

With this short excursion, a reasonable decision for the exo-centric VP and the combination can be chosen.

In the ego-centric visual perspective, the learner sees the GV inside the own body. Here, the learner can see the relation of the own body to the GV directly. In the pure exo-centric visual perspective this relation cannot be seen. Thereby, the position of the learner in relation to the GV must be guessed. That, in turn, makes the application

²A larger study was not possible because of the COVID-19 pandemic

³The person had no prior knowledge about the system or motor learning.

⁴Different person than the first one. This person had no prior knowledge about the system nor Motor Learning. A Larger evaluation was not possible because of COVID-19 pandemic.

of the speed mechanic - which is necessary for ego-centric guidance - not possible. A mechanic that is used in all conditions but one could lead to biased data, compare table 3.1. The mechanic of multiple representations

Perspective	Speed Mechanic	Multiple Representations
Ego-centric	Yes	No
Exo-centric	No	Yes
Ego & Exo-centric	Yes	Yes
Augmented Exo-centric	Yes	Yes
Ego & Augmented Exo-centric	Yes	Yes

Table 3.1: Mechanics speed and multiple representations and in which VP they are applied.

does not influence the validity of the study, because the mechanic would solve an issue that does not exist in the ego-centric perspectives. Furthermore, any VP with more than one representation is an exo-centric VP.

In the augmented exo-centric VP, a virtual copy of the learner is located inside the exo-centric GV. The copy lets the learner see the relation of the own body to the GV. Furthermore, augmenting the exo-centric GV with the learner is widely used and evaluated in related work [1, 20]. Consequently, the augmented exo-centric VP will serve as the exo-centric VP.

With the ego-centric and exo-centric VP set, the combination can be chosen. In the ego-centric VP the learner has a direct comparison of the own posture to the posture of the GV in the ego-centric VP. In the augmented exo-centric VP the learner has a direct comparison of the own posture and the posture of the GV in the exo-centric VP. To have the direct comparison from the ego-centric VP AND the exo-centric VP, the ego & augmented exo-centric VP is chosen as the combination. The ego & augmented exo-centric VP is the true combination of ego-centric and augmented exo-centric.

For simplification, the augmented exo-centric VP will be further called exo-centric VP, and the ego & augmented exo-centric will be further called ego & exo-centric VP.

The ego-centric VP, exo-centric VP and the ego & exo-centric VP are the independent variables of the study and form the three study conditions EGO, EXO, EGO & EXO.

3.1.2 Task Design

Hornbæk [32] identified three main types of tasks in HCI studies: representative tasks, simple tasks and tasks that use task specific hypothesis. RQ1 states the main investigation field is Motor Learning. Motor Learning is strongly related to real-world movements, evidently the study task is a representative task.

Real-world tasks that include the handling of physical load, can found in a wide range of activities. For example, a storekeepers job is to clear a palette of cardboard boxes. This task includes to unload the palette, scale the boxes, measure the dimensions of the boxes and finally store them in a rack. Another example is the work at a grinding machine. The worker takes a slug from a shelf and works on it until the slug becomes a workpiece. After that the workpiece is carried to a measurement instrument to be verified. There are plenty of other examples, but these two already clarify that a task which includes the handling of physical load consists out of the elemental tasks for manual material handling: lift, lower, push, pull, hold.

The idea for the study task is to chain these elemental task together to a Unit-Combined-MMH task, that representatively stand for a wide range of tasks that includes the handling of physical load. To achieve this, several aspects have to be taken into consideration: (a) the artifacts with which the learner will interact, (b) a reasonable

task decomposition that allows the investigation of sub-tasks. Furthermore, the study needs a (c) structure. (c) will reveal the necessity of three tasks. These tasks have to be equally (d) complex. This section will subsequently discuss (a-d) and propose the task for the study.

(a) Artifacts

A task that include the handling of physical load obviously need a physical load. In real-world tasks, the physical load can be everything a human can handle. The physical load for this task should fulfil the following criteria. First, the load should have a significant weight, that it is actually perceived as a load, but at the same time, any healthy person with no previous illnesses can handle it without getting injured. Secondly, the physical load should give a enough freedom for interactions. A simple box fulfils the criteria and furthermore has a relation to physical loads of real-world tasks like the handling of parcels. With a physical load, the elemental tasks of lift and lower can be realised by lifting and lowering the box from and to the floor.

Push and pull can be realised by pushing and pulling the box on the floor, but it can feel cumbersome. Moreover, in real-world task pushing and pulling a box is made possible in a more ergonomic height if possible, not least for security reasons. To support push and pull, a table is introduced. This table stands representatively for example the grinding machine or a parcel sorting table.

Finally, the transitions between the elemental tasks have to be supported, to increase the real-world reference. This is achieved by providing a waypoint. The waypoint is a plate on the floor and helps to bring sense in movements. This plate representatively stands for example for a scale or second machine. Walking to a scale or lower a box to the scale on the floor increases the real-world reference more than just an empty place in the room.

(b) Sub-tasks

The goal is to create a Unit-Combined-MMH task with the elemental tasks push, pull, lift, lower and hold. The process of designing the task is complex and happens iteratively.

In the first approach, was a task with four occurrences of every elemental task. For lifting the box from the scale and carry the box to the table, obviously a new task type had to be introduced: carry. Because carry is not an elemental task and for simplicity, elemental tasks and newly introduces task types are from now on referred as sub-tasks. This first approach of designing a task revealed an issue: chaining a given amount of sub-tasks together so that the task is still conductable is hard to achieve. To overcome the inflexibility, a new sub-tasks is introduced: walk. Walk means locomotion without the box in hand. With walk, the box can be pushed from one side of the table, and then be pushed from the other side of the table, which achieves flexibility in task design. Otherwise, on push will always follow pull.

The next approach was a task that consists out of the sub-tasks push, pull, lift, lower, carry, walk and hold. Each sub-task appeared four times. The task was evaluated with one⁵ person. This person had to follow the instructions in the ego-centric VP and exo-centric VP. During the conduction of the task, the person started to look around and correct the own position during the sub-task hold. An interview afterwards showed, that the person thought the GV stopped because his position was too far away from the GV. It became clear, that the speed-mechanic and the sub-task are not compatible. For the participant of the study it is indistinguishable if he/she is too far away from the GV or if it is the sub-task hold. Because of this indistinguishableness, the sub-task hold is excluded from the task as sub-task. However, hold is still part of the whole tasks: between the transitions of the tasks (for example between lift ends and lower starts) is a small pause which is equivalent to hold, but this sequence is too short to log. Furthermore, hold is part of the sub-task carry, where the box is held in front of the body. But hold is not a stand alone sub-task and thus cannot be evaluated isolated.

⁵larger studies covid

3 Experiment Design and System Implementation

A new task was designed with the sub-tasks push, pull, lift, lower, carry and walk. During the design special attention was paid to the magnitude of the movements. For example, every push should be equally far. Lift and lower from and to the scale and lift and lower from and to the table are very different in the magnitude. This resulted in two new sub-tasks: pick and place. Pick means to pick up the box from the table, place means to place the box on the table, while lift and lower the target remained the scale on the floor.

The next stage of the task consists out of the sub-tasks push, pull, lift, lower, carry, walk, pick and place. The task was inspected by a professional physiologist. The physiologist was asked to describe the sub-tasks in detail and perform every sub-task ergonomically. The professional's description of the sub-tasks are listed in table 3.2. From the performance of the professional physiologist and the description of the sub-tasks could be derived several insights. The sub-tasks push and pull are very similar in their conduction. The same applies for the pairs lift and lower as well as pick and place. This meant for the evaluation, that the variations of movements is nearly halved, and the possibility of making mistakes is reduced. Example: for push and pull, one foot has to be placed to the back while the other foot remains under the hip. The hands do the same for every push and pull. If in the participant does perform the sub-task intrinsically correct without the perception of the GV, the study will not measure the influence of the perspective. To increase the number of sources of error, two new sub-tasks are introduced: turn and fold. Turn means to turn the box by 90 degrees on the table, fold means to tilt the box from one side to another. The difference of the hand movement to push and pull is obvious. The difference for the feet results from the fact, that during turn and fold, the weight of the box remains on the table and the force to apply on the box is significantly lower than during push and pull. This results in a different feet placement, which are hip wide under the hip.

With the introduction of turn and fold all sub-tasks are introduced. A new task was created with all 10 sub-tasks. To assess all sub-tasks multiple times, they appear four times per task. The pair lift and lower and the only in magnitude different pair place and pick relate to each other. To be presented equally in the task among the other sub-tasks they should also only be present four times. Because, lift and lower are measured with the RM (6.2) (see next section) lift and lower was decided to appear three times each, and pick and place one time each. Unfortunately, only one time each pick and place means that all sub-tasks that do not happen at the table had to be conducted in sequence. To regain flexibility in the task, it was decided that pick and place will occur 2 times each. This results in 34 sub-tasks per task. Table 3.2 provides an overview over the sub-tasks and their corresponding description, as well as the occurrences per task.

Sub-task ID	Sub-task description	Professional's description	#of sub-tasks/Task
push	Push box on table	Lunge, feet hip wide, chest out, shoulders back, straight back, lean forward, bend front knee, extend your arms, pressure on front leg, push box by activating back muscles	4
pull	Pull box on table	Lunge, feet hip wide, chest out, shoulders back, straight back, lean forward, bend front knee, extend your arms, pressure on front leg, pull box by activating back muscles	4
turn	Turn box by 90° on table	Feet hip wide, lean slightly forward with straight back, turn box with arm muscles, weight of the box remains on the table	4
fold	Put the box from one side to another on the table	Feet hip wide, straight back, slightly bended arms, depending on the distance to the box: lean over table, no bent knees, weight of the box remains on the table	4
carry	Translation in space with the box in hand	Chest out, straight back, bend elbows to 90°, box near to body, shoulder in neutral-zero	4
walk	Translation in space without the box	"normal walking on their own judgment", straight back	4
lift	Lift up the box from the floor	Approach box as near as possible, weight shifted slightly to the front, bend knees, open legs while going down, stop at the raised heels, lean forward with straight back, lift box with quadriceps (tights), chest out, ellbows aim at ca. 90°	3
lower	Lower box to floor	Head above pelvis, bend knees and open legs, chest out, straight back and head, extend arms	3
place	Put box on table	Paralell hip wide feet, bend knees slightly, lean forward with straight back, lower arms	2
pick	Pick up box from table	Paralell hip wide feet, bend knees slightly, lean forward with straight back, lift with arms, abdominal and back muscles	2
			Total: 34 sub-tasks per task

Table 3.2: Sub-tasks that appear in every task.

(c) Study Structure

In the study three conditions are compared: EGO, EXO and EGO & EXO. The main question of this section is how to assign the participant to the independent variables. The key distinction is between within-subject design and between-subject design [32]. In the within-subject design the participant would experience all VPs. In the between-subject design, the participants would experience only one VP. Within-subject designs typically can detect the differences between the conditions more precisely, *ibid.*. Furthermore, within-subject designs need less participants ⁶ than between subject designs, *ibid.*. For these reasons, the study is conducted in a within-subject design. The within-subject design does also have a drawback: the participants gain experience about

⁶The COVID-19 pandemic makes it hard to find enough participants.

3 Experiment Design and System Implementation

the task and the conditions during the study. This means that one condition is influenced by another condition, which the participant already experienced. Furthermore, there is an asymmetrical carry-over effect between the conditions: EGO & EXO contains condition EGO and EXO and thereby influences EGO and EXO more than EGO and EXO influences EGO & EXO. To reduce the task-related learning effect, three tasks with a nearly equal complexity are created. With this, a participant will face for every condition a different task. The tasks are still similar and the learning effect persists. A reduction of the influence of the learning effect on the outcome can be countered out by counterbalancing the task. Similar to the task, the carry-over effect between the conditions can be also countered by counterbalancing the conditions. Hornbæk proposes in this case to cross the conditions with the task and use a Greco-Latin square. Three conditions and three tasks in a Greco-Latin square results in blocks of nine participants. This block is depicted in figure 3.2. The study should be conducted with at least 4 blocks (4x9=36 participants) **todo: warum????**. The first session of every study is for acclimatisation and is excluded from evaluation.

	Session 1		Session 2		Session 3	
PT	Perspective	Task	Perspective	Task	Perspective	Task
PT1	Ego	T1	Exo	T2	Ego-Exo	T3
PT2	Ego	T3	Exo	T1	Ego-Exo	T2
PT3	Ego	T2	Exo	T3	Ego-Exo	T1
PT4	Ego & Exo	T3	Ego	T1	Exo	T2
PT5	Ego & Exo	T2	Ego	T3	Exo	T1
PT6	Ego & Exo	T1	Ego	T2	Exo	T3
PT7	Exo	T2	Ego-Exo	T3	Ego	T1
PT8	Exo	T1	Ego-Exo	T2	Ego	T3
PT9	Exo	T3	Ego-Exo	T1	Ego	T2

Figure 3.2: Experiment structure: within-subject design in a Greco-Latin square.

(d) Equal Task Complexity

A study participant will face in every condition another task. For the validity of the study it is indispensable that these three tasks have a nearly equal complexity. As described in (b) a task consists of 10 subtasks that occur a specific amount. The main idea to ensure a comparable complexity is to use the sub-tasks for all three tasks in an equal amount. This means, the 34 sub-tasks of task one occur in task two and three, but in a different order. Table 3.3 lists all three tasks. For every task, the sub-task number ST1-ST34 is provided. Every sub-task number stands for a sub-task, which comes with a description and the sub-task ID. Reading the description from top to bottom are the instructions the learner receives from the GV during one condition. The mirror mentioned in the first line is another waypoint, which is necessary for technical reasons and is described in section 3.2.1.

3 Experiment Design and System Implementation

Task 1			Task 2			Task 3		
Sub-task#	Description	ST ID	Sub-task#	Description	ST ID	Sub-task#	Description	ST ID
	start in front of mirror, box on floor			start in front of mirror, box on floor			start in front of mirror, box on floor	
ST1	lift up box	lift	ST1	lift up box	lift	ST1	lift up box	lift
ST2	carry box to table	carry	ST2	carry box to scale	carry	ST2	carry box to table	carry
ST3	place box on table	place	ST3	lower box to scale	lower	ST3	place box on table	place
ST4	push box away	push	ST4	lift up box	lift	ST4	fold box away	fold
ST5	fold box away	fold	ST5	carry box to table	carry	ST5	walk to table center	walk
ST6	walk to left side of the table	walk	ST6	place box on table	place	ST6	turn box left	turn
ST7	fold box to bottom	fold	ST7	push box away	push	ST7	fold box to bottom	fold
ST8	pull box	pull	ST8	walk to right side of table	walk	ST8	push box away	push
ST9	pick up box	pick	ST9	pull box	pull	ST9	walk to right side of table	walk
ST10	carry box to scale	carry	ST10	push box away	push	ST10	pull box	pull
ST11	lower box to scale	lower	ST11	walk to table center	walk	ST11	fold box away	fold
ST12	lift up box from scale	lift	ST12	fold box left	fold	ST12	turn box right	turn
ST13	carry box to table	carry	ST13	turn box right	turn	ST13	push box away	push
ST14	place box on table	place	ST14	fold box to bottom	fold	ST14	walk to table center	walk
ST15	turn box left	turn	ST15	turn box left	turn	ST15	fold box to bottom	fold
ST16	push box away	push	ST16	push box away	push	ST16	turn box left	turn
ST17	pull box	pull	ST17	turn box left	turn	ST17	pick up box	pick
ST18	turn box right	turn	ST18	pull box	pull	ST18	carry box to scale	carry
ST19	fold box away	fold	ST19	fold box away	fold	ST19	lower box to scale	lower
ST20	pull box	pull	ST20	turn box right	turn	ST20	lift up box from scale	lift
ST21	walk to left side of table	walk	ST21	walk left side	walk	ST21	lower box to scale	lower
ST22	pull box	pull	ST22	pull box	pull	ST22	lift up box from scale	lift
ST23	turn box right	turn	ST23	fold box to bottom	fold	ST23	carry box to table	carry
ST24	push box away	push	ST24	push box away	push	ST24	place box on table	place
ST25	fold box to bottom	fold	ST25	walk to table center	walk	ST25	push box away	push
ST26	push box away	push	ST26	pull box	pull	ST26	pull box	pull
ST27	walk to scale	walk	ST27	pick up box	pick	ST27	turn box right	turn
ST28	walk to box on table	walk	ST28	place box on table	place	ST28	walk to right side of table	walk
ST29	turn box left	turn	ST29	pick up box	pick	ST29	pull box	pull
ST30	pick up box	pick	ST30	carry box to scale	carry	ST30	push box away	push
ST31	carry box to (invisible) mirror	carry	ST31	lower box to scale	lower	ST31	pull box	pull
ST32	put box on floor	lower	ST32	lift up box	lift	ST32	pick up box	pick
ST33	lift box up	lift	ST33	carry box to (invisible) mirror	carry	ST33	carry box to (invisible) mirror	carry
ST34	put box to ground	lower	ST34	lower box to ground	lower	ST34	lower box to ground	lower

Table 3.3: tasks

3.1.3 Measures: dependent variables

This work's aim is to answer the main research question RQ1: How does the visual perspective on a virtual GV influence Motor Learning in Virtual Reality. To answer this research question, the proposed study has to generate data that is able to answer the sub-research questions RQ1.1-4. This section will provide the underlying paradigm to every sub-research question and explain which measures are necessary.

RQ1.1 How does the visual perspective on a virtual guidance visualisation influence movements' accuracy?

Paradigma: The more exact the learners' movements match the GV movements, the better the learner could

follow the instruction of the GV. For RQ1.1.1 the limbs of the learner and the limbs of the GV are compared. For RQ1.1.2 the box' accuracy is compared. For RQ1.1.3, both are compared and additionally, the current sub-task is taken into consideration. The accuracy can indicate, how the particular movement is suited for the VP.

RQ1.1.1 How does the visual perspective on a virtual guidance visualisation influence movements' accuracy of the own body?

Measures: (1) Euclidean distance between the learners and GV's hands, feet, head and hip in meters. (2) Angle between learners and GV's the hands, feet, head and hip in degrees.

RQ1.1.2 How does the visual perspective on a virtual guidance visualisation influence the accuracy of handling physical load?

Measures: (3) Euclidean distance between the learners and GV's physical load. (4) Angle between the learners and GV's physical load in degrees.

RQ1.1.3 How does the visual perspective on a virtual guidance visualisation influence sub-tasks' accuracy?

Measures: (1-4), additionally matched to the sub-tasks that is currently performed (5).

(1-4) gives insights to what extent the learner could follow the GV for the whole task. (5) can extract specific sub-tasks for which the learner could follow the GV to a certain extent. For example, in the ego-centric VP the overall accuracy for a task is lower than in the other VPs, but the accuracy for the sub-tasks lift and lower is higher than in other VPs. For this example, measure (5) can extract specific sub-tasks that are performed better or worse than in other VPs.

RQ1.2 Does the visual perspective on a virtual guidance visualisation influence the transfer of ergonomic principles?

Paradigma: the more exact the learners RM matches the GV's RM, the better the ergonomic principles could be transferred.

Measures: (6) Risk Measurements: (6.1) spine bend in degrees, (6.2) squat distance in meters, (6.3) good base in meters **todo: check names**, (6.4) box near body.

Muckell et al. [33] used four RM in their work to assess the performance of the conducted movements. Three of them are assessed in the proposed study, too.

(6.1) spine bend is defined by the difference in degrees between the straight upward vector and the back of the learner. For all sub-task, spine bend should be in a certain window. This window is calculated from the teacher, who recorded the movement. Spine bend indicates, if the learner could perceive the correct posture of his back.

(6.2) squat distance is defined by the distance in meters between the feet. For the sub-tasks lift and lower, the squat distance should be in a certain window. This window is calculated from the teacher, who recorded the movement. For the other sub-tasks, squat distance is not applied, because the knees do not bend in the other sub-tasks. Squat distance indicates, if the learner could perceive that he should bend his knees during lift and lower.

(6.3) good base is defined by the distance in meters between the feet. For the sub-tasks push, pull, turn, fold, lift, lower, pick and place, the squat distance should be in a certain window. This window is calculated from the teacher, who recorded the movement. Good base indicates, if the learner could perceive the correct posture of the feet. Muckell et al. [33] additionally use the RM spine twist in their work. This RM cannot be applied for this study because of the multiple perspective mechanic. The learner has multiple GV around and is free of choice which one to look at. The turn of the head implies spine twist. Though, spine twist has a low validity and reliability.

(6.4) box near body. For the study task design, a professional physiologist was consulted (compare **todo: section**). During the interview, all movements were described in detail, compare **todo: figure**. During the sub-task carry the box should be as near as possible to the body while the elbows should have bend angle of 90 degrees. Unfortunately, the bend angle could not be determined during a study for technical reasons, see chapter **todo: implementation**. Fortunately, the distance between the box and the body can be determined. Box near body, is defined as the distance in meters between the box and the body of the learner. For the sub-task carry, box near body should be in a certain window. The window is calculated by from the teacher who recorded the movement.

RM (6.1-4) are different from accuracy measurements (1-5), because they are independent from the position of the learner and the GV. For example, in the exo-centric VP, a learner cannot percept the correct position where he/she should stand. The learner thereby stands 15cm away from the position he/she should stand. The overall accuracy is thereby lower. But the learner could percept the positioning of his/her feet correctly. In this case the RM (6.3) are fulfilled while the accuracy is biased.

RQ1.3 How does the visual perspective on a virtual guidance visualisation influence the learner's visual focus?

Measures: (7) looking at

Paradigma: the learners visual focus is on the object the learner is looking at.

The learner interacts with a box and has multiple GVs around and inside the learner. Looking at can give insights on which GV the learner is focusing, the frequency of focus changes and the role of the physical load.

RQ1.4 What is the subjective personal preference of the learner for the visual perspectives?

Measures: (8) qualitative data; likert scales, semi-structured interview, digging into occurrences.

The qualitative data serves not only to investigate the learners personal preference, but can be also serves as triangulation method for measures (1-7).

The last measure is the (9) task completion time measured in secondmiliseconds. The speed-mechanic regulates the speed of the animation of the GV. The further the learner is located to the GV, the slower the animation speed of the GV until it stops completely at EDT_{max} . The task completion time can give insights on to what extend the learner could percept the desired position of his/her body in relation to the GV. This measure relates with (1) and can be used for triangulation. Additionally, it is to aspect that the TCT is decreasing from condition to condition, because the participant acimated. By that, TCT could give insights about the learning effect between the conditions. Finally the study is recorded by video. If during the evaluation arises questions about a specifiv topic, the recordings can be consulted.

3.2 E(x|g)o- Design and Implementation

The last section describes an experiment to investigate the influence of the visual perspective on Motor Learning in Virtual Reality. For the conduction of this study, a system is necessary. This system is called E(x|g)o. This section elucidates the development of E(x|g)o. The starting point is the creation of a self-perception of the learner. Section 3.2.1 describes how the learner gets a digital body in the Virtual Reality. After that, in section 3.2.5, the VPs have to be implemented which in the study serve as the conditions. Section ?? deals with the recording of the movements of task 1-3. Section 3.2.9

3.2.1 Self-Perception

There are various options of devices to dive into the Virtual Reality. In the preceeding project report, several devices were evaluated and the decision was made for the Valve Index⁷, because its refreshrate, screen solution and the possibility to wear glasses underneath the head-mounted display (HMD). To determine the position and orientation of the HMD, the so-called Lighthouse⁸ is utilised. A Lighthouse is tracking volume spanned by at least two Base-Stations⁹ at the opposite corners of the room. To improve the tracking and for the avoidance of untracked areas, e.g. under the table, E(x|g) uses four base-stations, one for each corner of the room, to span the Lighthouse.

Till now, the learner can move in an empty virtual world seeing nothing. The next step is to replace the empty virtual world with a meaningful environment. For this, the Game Engine Unity 3D¹⁰ is used. In Unity, a basic room was created. Four light yellow walls, a parquet floor and indirectional lighting. The parquet floor serves a purpose: it has a structure with frequent straight lines which makes it later easier to align the artifacts like table and scale. The room is kept simple to not distract the participant of the study.

The next step is to add the learner's body to this room. To achieve this, the body of the learner needs to be tracked. In the preceding project report multiple full-body tracking systems were compared. The decision was made for Vive Trackers¹¹, because of coordinate system matching, the lower latency and less work-intensive calibration process. The learner wears six Vive Tracker in total. Five of them plus the HMD are necessary for the full-body tracking of the learner. The remainder is necessary for RM (6.1), which is later explained in section 3.2.6. Two trackers are located on the feet, two trackers are located on the hands. The last tracker is located on the lumbar vertebra (back hip). The trackers are attached to the learner by special Vive Tracker Straps¹². The Lighthouse tracks the Vive Trackers and HMD which send their position to the PC (simplified). Here SteamVR¹³ receives the information and forwards it to Unity. In Unity the SteamVR Plugin¹⁴ provides the information in a workable condition. This information is now about to be translated into a rendering of a human-like avatar at the position of the learner's body. This requires several steps. First, an avatar is imported. To create the avatar Reallusion Character Creator 3¹⁵ was used. To match the gender of the participant, a male and a female character were created, wearing the same clothes. Based on the demographic questionnaire, the gender can be set and the participant will see an avatar complying with the participant's gender. Secondly, the information about the trackers' position and orientation in the tracking volume has to be translated into human movements that meet the learner's movement. This is achieved by Inverse Kinematics (IK). IK arises from the field of robotics. A robot arm consists of limbs and joints. Each limb has a specific length and each joint has a specific range of angles to move. The length and the angles are rules. Given an endpoint the robot has to reach with the most outer limb and the rules, the angle of each joint can be calculated to fulfil the goal. This process can be mapped to a human body, too. Unity provides a third-person plugin called FinalIK¹⁶ that is capable for the calculations in question. On the one hand, FinalIK is powerful and unrivalled in functionality compared to other IK tools and though influenced the choice to use Unity for E(x|g) heavily. On the other hand, to match the needs of the study, extensive adjustments were necessary. The preceding project report gives a detailed description of the process, while here a summary is depicted. The main task is to transfer the information from SteamVR to FinalIK in a meaningful way, so that FinalIK animates the learner's body faithfully.

⁷todo: todo

⁸lighthouse

⁹base stations

¹⁰Unity

¹¹Vive trackers

¹²<https://www.google.com/search?q=vive+tracker+straps>, accessed 10.03.2021

¹³steam vr

¹⁴<https://assetstore.unity.com/packages/tools/integration/steamvr-plugin-32647>

¹⁵<https://www.reallusion.com/character-creator/>

¹⁶<https://assetstore.unity.com/packages/tools/animation/final-ik-14290>

SteamVR registers the Vive Tracker in the order they are switched on. To increase the reliability of $E(x|g)_o$, a script was created that assigns the tracker by the hardware ID. The tracker are then assigned to the VRIKCalibrationController. The VRIKCalibrationController matches the tracker with the avatar and resizes the avatar to the learners height. Finally, K is constructed to work with controllers in the users hands. In $E(x|g)_o$, the study participant needs the hands to interact with the box, though the controller are replaced with Vive Tracker on the back on the hands. Shifting the reference points of the hands yields in a faithful representation of the learners hands. The feet needed adjustments in a similar way. Finally, VRIK is able to solve the movements. Solving is the process of translating the tracker information into an animated avatar.**todo: pipeline?** VRIK requires a calibration before use. For calibration, the person attached with the trackers needs to perform a T-Pose in certain direction. To ease the calibration process, a mirror is placed in the room. The participant can be asked to look into the mirror and expand the arms, which leads to the correct orientation of the participant during the calibration process. Immediately with the calibration the mirror disappears. Because the participant is standing in front of the mirror when the task is about to start anyway, the position in front of the mirror is chosen as the starting point and end point of every task.

3.2.2 Artifacts

The learner is able to see the own body in an empty room. The task include the handling of physical load on a table and a scale. To create the table, the scale and the box which will serve as physical load, they need to be constructed physically and digitally.

The first version of $E(x|g)_o$ used a cardboard box (27cm x 26cm x 24cm). During the development it became clear that the size of the cardboard box was too handy to serve as physical load. To determine a suitable size several boxes of different dimensions were tested. With nine different boxes, a set of sub-tasks were performed. The major insight from this test was that the length of the sides of the box should be different, to visually see the direction of the box. Furthermore, the box should be perceived as physical load by having a reasonable size and a certain weight, but not being too heavy to not limit the study participants to strong humans, or being a threat to the bodies of the participants. The sizes were set for 35cmx30cmx25cm. The measures of the box differentiate by 5cm in every dimension. This makes it clear to see the orientation of the box. The final box was constructed from three layered wood with a strength of 27mm. The resulting weight was 5.8kg. To evaluate the weight of the box, one male person¹⁷ was asked to perform every the sub-tasks. Observations revealed no incidences that contradict to use the box as physical load. The person rated the weight as "ok". He had no problems to move the box. The box was painted in three high contrast colours: black, white, red. Each opposite side in the same colour. The painting facilitates the visual perception of the orientation of the box. The digital pendant of the box is a cube in the same colour and size. To translate the position and orientation of the physical box to the virtual one, a Vive Tracker is attached to the box and fixated with a screw. On the plus side of using a screw is the prevention of any relocation of the tracker. The downside is that tremors caused by placing the box on e.g. the table, are transferred directly into the tracker. This causes the tracker to lose tracking. To interrupt the transfer of tremors a shock-absorbing insulation is placed between the tracker and the box.

During the development of $E(x|g)_o$ an office table (120cm x 70cm x 72cm) was used. The size of the table proved to be suitable for the task. The digital pendant to the physical table is a cube in the same size and colour of the tabletop. The position and orientation of the table is assessed by a Vive Tracker. Unfortunately, the tracker is placed inside the "working area" on the top of the table. To shift the tracker out of the working area, a new tabletop was constructed. The new tabletop is out of three layered wood, with increased size of 5cm in each dimension. The tracker is attached on the most outer edge, though out of the working area. To prevent tremors passing from the table into the tracker, a shock absorbing insulation is applied here, too.

The last artifact to create is the scale. The scale is a waypoint in the room where the participants perform lift and lower to the ground. The scale is a rectangular plate of 45cmx45cm, so that the box can be placed on the scale easily. To shift the tracker out of the area where the box will be placed, the plate is extended by 5cm. The

¹⁷A larger evaluation was not possible because of the COVID-19 pandemic. Evaluation with at least one female person is desirable

tracker is attached to the most outer edge of the extension with a screw and shock-absorbing insulation. The digital pendant is a cube with the same dimensions as the physical plate, excluding the extended area.

3.2.3 Study Setting

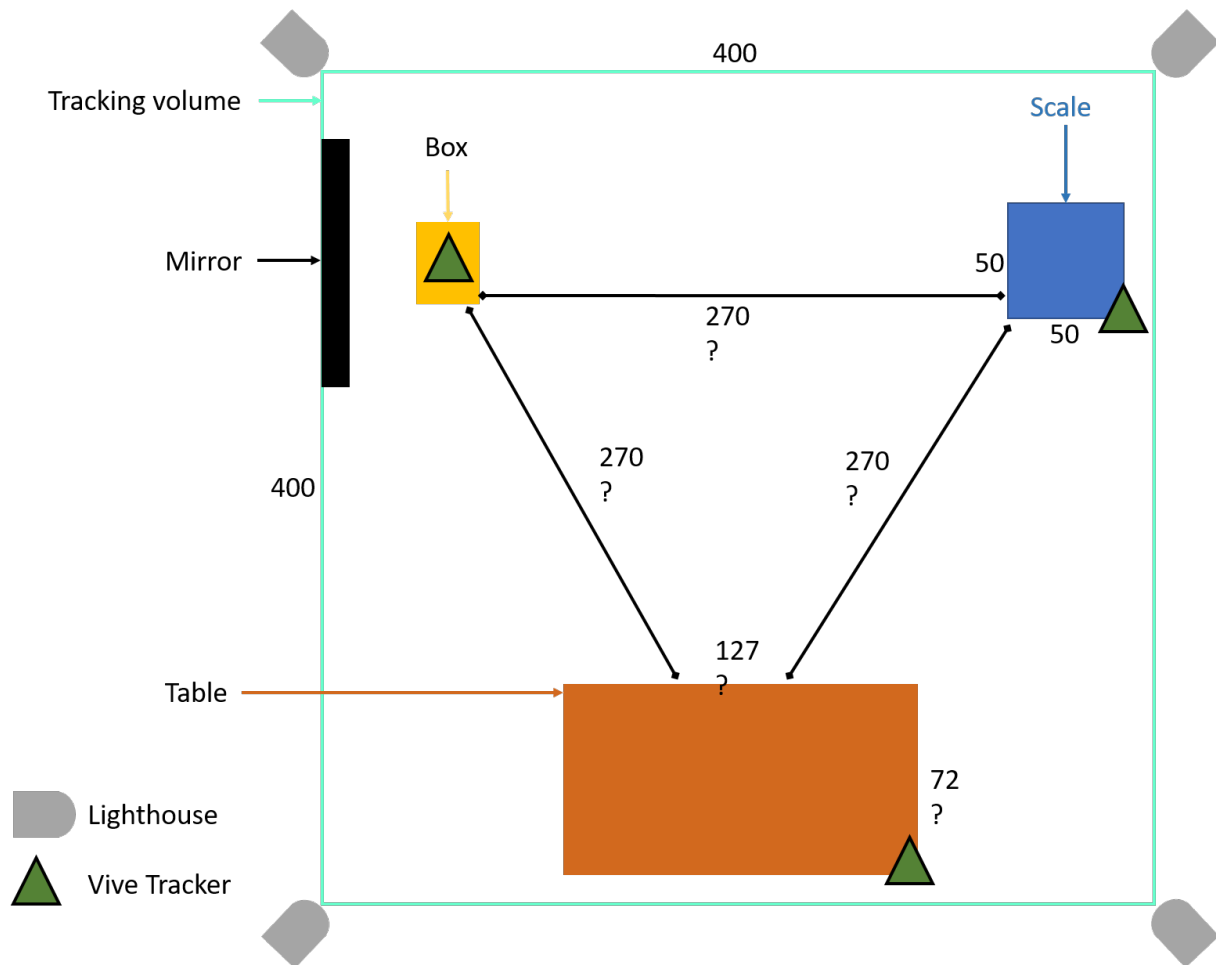


Figure 3.3: study setting

Meanwhile, $E(x|g)$ consists out of a room, avatar representing the learner, t_{bale} , box and scale. This section aims to attain an overview about the alignment of these elements. Figure 3.3 shows the real-world room in which the study is conducted. Figure 3.4 depicts all positions that are described in the task (compare table 3.3). The outer line represents the Lighthouse or tracking volume, which is approximately 400cmx400cm. On the left wall, the mirror is located. In front of the mirror, the starting and end position of the box is seated. Beside the box is position mirror, the start and end point of the learner. The table is placed into the middle of the wall to the left of the mirror. Around the table the positions table center, table right and table left is located. At the opposite wall of the mirror the scale is placed. In front of the scale, is the position scale.

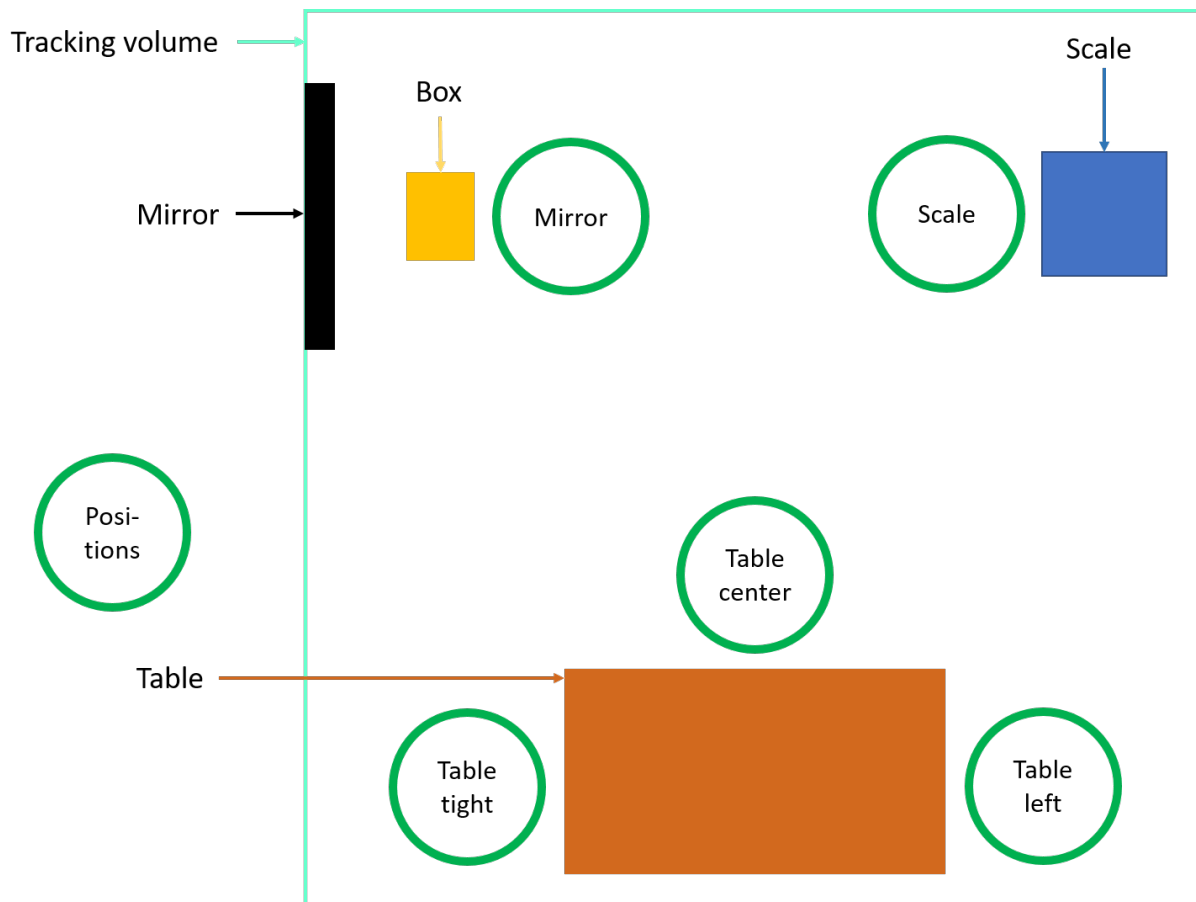


Figure 3.4: learner positions

3.2.4 Guidance Visualisation

The next task is to add the GV to $E(x|g)$ which the learner can mimic. The GV is an avatar like the learners avatar, with the difference, that the motion of the GV is driven by the pre-recorded tasks 1-3. The recording of the tasks was also performed with $E(x|g)$. To use $E(x|g)$ as a recorder, a copy of all trackers and the HMD is created. This recorder-copy is packed as one GameObject with the trackers as childs. The parent GameObject is recorded during the performance of the movements. For the GV, a similar GameObject as the recorder is created and serves as Input for VRIK. A detailed description is given in the project report. In this section the main points of the process are described: the recording of the movements and the resizing of the GV to the size of the learner. The movements in the task have to be performed ergonomically. The measures to evaluate ergonomic movements are the RM. To serve as a strong baseline, a professional for ergonomic movements should record the movements. Because of the COVID-19 pandemic, all approaches to record the movements by a professional failed. Eventually, I myself were trained by a physiologist and recorded the movements by myself. The final recordings were examined by the physiologist. Overall, the movements were rated by the physiologist as "by and large correct". The back is not always straight or in the correct angle. In task 2 during a push and in task 3 during a pull, the feet are misplaced.

With the recording of the tasks at hand, the GV can be animated. For the ego-centric VP it is inevitable, that GV and learner having the exact same size, else the learner cannot percept the GV correctly. Furthermore, the table, box and scale mustn't resize. The solution is to record two sets of object synchronised. The first set contains the

objects that have to be resized, namely the GV, the second set contains objects that mustn't be resized, namely the box, table and scale. The recordings were synchronised by a script, the playing of the animations in E(x|g)o, too. The resizing of the GV takes place in three steps. First, the learner and the GV are calibrated. Then the height difference is measured between learner and GV. In the last step, FinalIK is removed from the GV, the animated GameObject containing all trackers is removed, then FinalIK reattached and calibrated. Thus the learner and the GV have the same size.

3.2.5 Perspectives

To fulfil the studies needs, the VPs have to be implemented.

The ego-centric VP requires besides the learner, one ego-centric GV. The learner needs to stay inside the GV. This is achieved by the speed-mechanic. The distance of the learner is calculated with the help of the tracker at the hip of the learner. The hip is projected to the floor, likewise the GV hip tracker. The projection to the floor is necessary, because the speed-mechanic would apply in the case the GV bends the knees during lift and lower. This restricts the possibility to perform an error by the learner: if the learner does not bend the knees during lift and lower, the GV would stop and remind the learner to bend the knees. To investigate if the learner could percept to bend the knees, the learner must be allowed to do the error. The speed-mechanic relies on the distance between the two projected points on the floor. Additionally, the distance finds application in another functionality. In the ego-centric VP, the learner is located inside the teacher. This means, that the viewport of the learner is inside the head of the GV and let the learner see the inside of the GV head. This leads to distraction by the partly rendered inner head. The solution is to remove the head rendering, if the distance is below EDT_{max} and reinitiate the rendering above EDT_{max} . The rendering is removed by replacing the materials array of the head with an material array that contains only invisible materials.

In the exo-centric VP, four exo-centric GV are located around the learner. The positions of the exo-centric GV were determined after the task was recorded. The difficulty is to determine reasonable positions of the exo-centric GVs. First, in any point in time during the performance of every task, the learner must be able to see a GV by only turning the head. Secondly, the GV and the learner should not move through a table or scale of another GV. The solution to the first part is informed by Chua et al. [20]. Chua et al. chose four representation that are in front, behind, left and right. The latter proved to be impossible, if the exo-centric representations should be near enough to be able to be observed by the learner. This is a limitation of E(x|g)o. The GV need to be shifted to far away from the learner to not cross the artifacts of other GV. In a distance in which no crossing occurs, the movements are barely visible to be mimicked correctly. The exo-centric representations were then placed in a distance, that allows to be observed by the learner, the GV can access all positions without standing in an artifact at that position and that the learner never crosses a table of another GV. Standing at table center: the GV to the left is shifted by two meters to the left, the GV to the right, two meters to the right. The GV in front is shifted by 1.5 meters to the front. The GV in the back is shifted 3 meters to the back. Figure 3.5 shows the positions of the exo-centric VP. Additionally, to the exo-centric GVs a virtual copy of the student needs to be rendered. The virtual copy of the learner is shifted by the same values. In the last VP, the ego & exo-centric VP, the learner has an ego-centric VP, as well as the exo-centric GV with the corresponding virtual copies of the learner. The implementation of copying the learner and the GV and shifting them to the position is rather complex and described in detail in the project report.

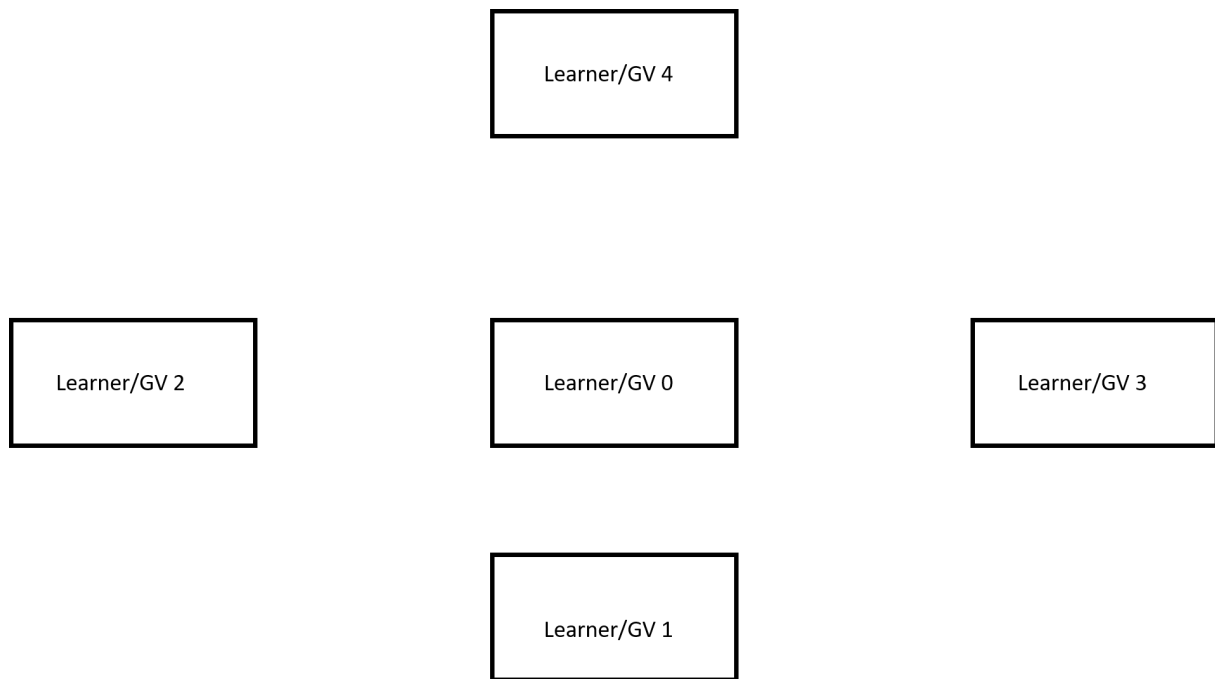


Figure 3.5: shift calc

3.2.6 Quantitative Data Acquisition

Logging ID	Description	Unit	Research question
Elapsed time	Time since the beginning of the task	Milliseconds	RQ1.1.1-3
Current animation Frame	Current frame of the GV animation	frames	RQ1.1.1-3
Subtask ID	The current sub task performed by L	STID	RQ1.1.3
Hip distance	ED between hip of the GV and the hip L	Meters	RQ1.1.1
Left hand distance	ED between left hand GV and left hand L	Meters	RQ1.1.1
Right hand distance	ED between right hand GV and right hand L	Meters	RQ1.1.1
Left foot distance	ED between left foot GV and left foot L	Meters	RQ1.1.1
Right foot distance	ED between right foot GV and right foot L	Meters	RQ1.1.1
Head distance	ED between head GV and head L	Meters	RQ1.1.1*
Box distance	ED between box GV and box L	Meters	RQ1.1.2
Hip angle	ED between hip of the GV and the hip L	Degrees	RQ1.1.1
Left hand angle	ED between left hand GV and left hand L	Degrees	RQ1.1.1
Right hand angle	ED between right hand GV and right hand L	Degrees	RQ1.1.1
Left foot angle	ED between left foot GV and left foot L	Degrees	RQ1.1.1
Right foot angle	ED between right foot GV and right foot L	Degrees	RQ1.1.1
Head angle	ED between head GV and head L	Degrees	RQ1.1.1*, RQ1.3
Box angle	ED between box GV and box L	Degrees	RQ1.1.2
L spine bend	RM spine bend of L	Degrees	RQ1.2
L foot distance	RM base of L	Meters	RQ1.2
L squat distance	RM squat distance of L	Meters	RQ1.2
L hip-box distance	RM elbows L	Meters	RQ1.2
GV spine bend	RM spine bend of GV	Degrees	RQ1.2
GV foot distance	RM base of GV	Meters	RQ1.2
GV squat distance	RM squat distance of GV	Meters	RQ1.2
GV hip-box distance	RM elbows GV	Meters	RQ1.2
L looking at	The object L is looking at	LAID	RQ1.3
Pos x	X position for all 12 trackers	Meters	**
Pos y	Y position for all 12 trackers	Meters	**
Pos z	Z position for all 12 trackers	Meters	**
Rot x	X rotation for all 12 trackers	Meters	**
Rot y	Y rotation for all 12 trackers	Meters	**
Rot z	Z rotation for all 12 trackers	Meters	**
Total 146 columns			

Table 3.4: Detailed overview of logs produced by E(x|g)o per frame. L: learner, GV guidance visualisation, ED: euclidean distance. *head position and rotation is biased in exo-centric conditions because of multiple GV the L can focus on. **All trackers are logged for backup reasons: after the study is conducted a measurement can become interesting that was not of importance before. With these values any measurement can be calculated post-study.

Section 3.1.3 defined the measures that are necessary to answer the research questions. E(x|g)o must be capable to assess all measures. This section explains how E(x|g)o assesses the measures. An overview about all measures

is listed in table ???. Table ??? lists the logging ID, a description about what the measurement is measuring, the unit in which the measurement is measured and for which research question the measurement is assessed. The quantitative data acquisition can be divided in several classes: (i) accuracy measurements (1-5), (ii) ergonomic measurements (6), (iii) focus measurement (7) and (iv) time measurement (9).

(i) Accuracy Measurements

The accuracy measurements assess the discrepancy between the movements of the learner and the movements of the GV. Accuracy measurements are subdivided in distance-based measures and angle-based measure. Distance-based measures rely on the the Euclidean distance between the body parts of the learner and the body parts of the GV. The reference point for the body part is the tracker which is attached to the body part. The body parts are: hip, left hand, right hand, left foot, right foot, head. The distance between the box of the learner and the box of the teacher is an accuracy measurement, too. Likewise the body parts, the distance between the two boxes is the Euclidean distance between the tracker of the box and the tracker of the GV. Please note, the trackers are not visible to learner during the study. The trackers of the GV are the recorded trackers. Angle-based accuracy measurements assess the discrepancy of orientation between the body parts and box of the learner and the GV, and is measured in angle. The calculation of the angle-based measurements comply with the calculations of the distance-based measurements. Means, the angle between the corresponding trackers are measured. To conclude: distance-based measurements assess the error in the positioning, angle-based measurements assess the error in orientation.

(ii) Ergonomic Measurements

The ergonomic measurements are the four risk measurements: support base, squat, upright stance, and hip-box distance. Support base is the distance between the feet. For push and pull, lift and lower, turn and fold, pick and place each a window in which the distance should be located is defined. The definition of the window is based on the movement of the persons who recorded the movement. The percentage of time the learner is inside the window, is the the outcome of the measurement. For a better understanding imagine the following example: the window for push and pull for the support base is 20cm-30cm. During the performance of push the learners feet distance was inside the window for 90 seconds. The whole performance of push lasted 100 seconds. The RM support base yields in a score of 90%. Because the movements were recorded by myself and not by a professional, the window defined on my own movements would be misleading. Because of this predefined error, the definition of a window is renounced.

The measurement for squat is the distance of the hip and floor. It indicates if the learner bended the knees correctly. It is applied in the sub-tasks lift and lower. Calculations of the RM score of squat complies with support base.

Upright stance is the measurement of the spine bend is correctly. For upright stance, an additional tracker is applied on the back of the student, above the hip tracker. The angle of spine bend is the angle between the upright vector and the vector of the upper hip tracker. Upright stance is applied for push and pull, lift and lower, turn and fold.. The bend angle during pick and place depends heavily on the position of the box on the table and thereby varies. Because of this variation, a window cannot be defined for pick and place and though the RM upright stance is not applied to pick and place. Calculations of the RM score of upright stance complies with the calculations of the preceding RM.

Hip-box distance is calculated by the distance between the hip tracker and the box tracker. it is applied for the sub-task carry. The calculations comply with the preceding RMs. The limitation of hip-box distance is that the measurement is influenced by the circumference of the torso of the learner. An formative evaluation of hip-box distance was not possible due to the COVID-19 pandemic.

(iii) Focus Measurement

The virtual room the learner sees in $E(x|g)$ is filled with tables, boxes, GVs and virtual copies of the learner. To assess on what the learner is focusing during the movements, every object reviewed a name. In every frame, raytracing is performed. The rays' origin is the HDM and expands straight forward. The name of the object first hit by the ray is written into the log file. The name is coded with the position 0-4 (see positions in figure 3.4) and an object identifier (box, table, scale, GV, learner). A test revealed a systematic error by pointing too high. To correct the discrepancy, colliders of the objects were increased. The tables' and scales' collider height is increased by 20cm. The box colliders height is doubled. The learners' and GVs' avatar were wrapped into an capsule collider with a height 200cm and a radius of 30cm. The values were determined by experimentation. To test the values, all sub-tasks were performed and for any point the object which is hit by the ray is displayed. The displayed name complied with the object in focus in nearly any point in time. Using an eye-tracker would increase the accuracy, but was available.

(iv) Time Measurement

The speed-of the animation of the GV is determined by the distance between the learner and the GV. Comparing the time the learner needed to perform the task with the time the task lasts without the speed-mechanic can draw conclusions about the position of the learner to the GV. The tasks differ in the amount of time to performed. The plain-times without the speed-mechanic are: task 1 takes 172128ms, task 2 189040ms and task 3 176668ms. The subtraction of the plain-time from the total time is the additional time the learner needed to perform the movement. The time discrepancy can also be applied to specific sub-tasks. This measurement will mainly used in the evaluation for triangulation.

3.2.7 Qualitative Data Acquisition

The qualitative data assess during a study session relies on two questionnaires. After each condition, the participant is asked to fill the after session questionnaire (appendix A). After all conditions, the participant is interviewed. The guideline of the semi-structured interview is attached in appendix B. The questionnaires aim to assess the participants impressions and opinions about the VPs. In the questionnaires, a different wording is applied to ease understanding. For example, the GV is called virtual teacher.

After Session Questionnaire

The after session questionnaire starts with a question about the subjective overall performance of the learner.

Q1: How accurate did your movements comply with the virtual teacher?

A: Likert scale from one (very good) to 7 (very poor)

- Linked research questions: RQ1.1.1-3, RQ1.4
- Triangulation for (1-4,6)

The answer to this question gives insights how accurate the participant assesses the performed movements. Furthermore, this question can be used to determine, if the qualitative accuracy complies with the subjective

3 Experiment Design and System Implementation

opinion of the participants. The next question aims to assess the users subjective performance for the sub tasks. The participant is asked to fill in a table. Each line represents a sub-task. Each sub-task can be rated on a Likert scale from 1 (very good) to 7 (very poor).

Q2: During the task there were several smaller reoccurring movements, like pulling or lifting the box. Please rate these smaller movements, to what extend you could follow to movements.

A: Likert scale from one (very good) to 7 (very poor) for each sub-task.

- Linked resarch questions: RQ1.1.1-3, RQ1.4
- Triangulation for (1-4,5,6)

Movement	1	2	3	4	5	6	7
Example movement		X					
Pushing the box on the table							
Pulling the box on the table							
Folding the box to a side on the table							
Turning the box on the table							
Lift up the box from the floor							
Lower the box to the floor							
Picking up the box from the table							
Placing the box on the table							
Carrying the box							
Walking without the box							

Figure 3.6: sub task rating

Beside the objective opition about the participants performance, the answers of this question can be used to compare with the qualitative data. Question three aims to assess the subjective accuracy of the participants body parts.

Q3: Please rate to what extend you think you could align your body parts with the teachers body parts.

A: Likert scale from one (very good) to 7 (very poor) for each body part.

- Linked resarch questions: RQ1.1.1-3, RQ1.4
- Triangulation for (1-4,5,6)

Body part	1	2	3	4	5	6	7
Example body part		X					
Legs							
Arms							
Back							

Figure 3.7: sub task rating

This question assesses how good or bad the participant could see the body parts of the GV. The last question is not handed to the participant. It serves as the basis for an semi-structured interview. It gives the possibility to dig into extreme values of the questions answered before and in any incidences occurred during the performance of the task. Furthermore, the participant is encouraged to speak frank about the session.

Q4: (As interview question) Did you have problems to follow the instructions?

- E.g. because you could not see some body parts?
- E.g. bad perception related to the perspective?
- Go into extreme values of this questionnaire!
- Address critical incidences!

A: Take down of participants statements.

Semi-Structured Interview

After all three sessions done, the participant is interviewed. The interview is semi-structured. The guideline contains seven main questions, partly with additional hints to dig deeper or to lower the entry threshold for the participant to start reporting.

Q5: You saw three visual perspectives: ego-centric, exo-centric and the combination. What do you think about these perspectives?

- entry question, encourage participant to talk frank, address interesting mentions

Linked research question: RQ1.4

Q6: Prioritise the perspectives by to what extend you think you could follow the movements. (1 best to 3 worst)

- Why did you decide for the prioritisation?

Linked research question: RQ1.4

Q7: Imagine you want to learn a movement in VR. Which perspective would you use for that?

- Or would you use a totally different one?

Linked research question: RQ1.4

Q8: In which of the three perspectives was the easiest to understand?

- Was there a perspective that confused you?
- Was there a perspective you did not understand right away?
- * What do you think caused the confusion?

Linked research question: RQ1.4

Q9: What do you think are the advantages and disadvantages of the perspectives?

Linked research question: RQ1.4

Q10: Could you see some body parts better or worse in the perspectives?

- What about your legs, arms, back?
- Could you detect that during lift and lower you should squat?
- Could you detect that you should step back during push and pull?

Linked research question: RQ1.4

Q11: Did you miss a feature?

- Dig for improvements for E(x|g)oor experiment design.

Q12: (Space to ask for critical incidences, if any occurred.)

3.2.8 Study Procedure

As soon the participant enters the room, the participant receives a warm welcome to feel comfortable. The process start with a Welcome letter (appendix C), followed by the informed consent and a demographic questionnaire. In the mean time, E(x|g)ois set up by choosing the condition, set the gender of the participant as well as the log is configured with the participant ID and task id. After the demographic questionnaire, a spoken explanation about what is about to happen is given. Then the trackers are attached to the participant. The calibration process is explained: looking in the mirror and extend the arms. An explanation of the perspective is provided. Then, the first session is started. E(x|g)ogets started, the cameras and screen recording are set up and the participant gets the HMD. The participant is asked to stand in front of the mirror to calibrate. For calibration, the key C is pressed at the PC. To identify the cam recordings a sign is held into the cameras. The task is started with the key S. During the session, attention is payed to the cable of the HMD to avoid stumble of the participant. Furthermore, the participant is observed. After the session ended, the HMD is removed. The participant fills in the after session questionnaire. Session two and three are conducted likewise. After all three sessions, the trackers

are removed. The participant is interviewed. The payment is given and the receipt signed. In the very end, the participant is thanked and said goodbye. If appeared, door step talk is appreciated.

3.2.9 Limitations

E(x|g)o and the study is designed for a task that includes the handling of physical load. If the outcome can be applied to movements without a physical load is questionable. Furthermore, the exo-centric GV sometimes walk through artifacts (table, scale) of other GV, which can cause confusion to the study participant. The movements are not recorded by an professional, errors in ergonomic movements are possible. Lastly, only a small number of participants participated in the formative tests to evaluate partial aspects of E(x|g)o. Especially, the hip-box distance is not tested, because multiple persons with different physique would be necessary.

4 Study Evaluation

(5 pages)

4.1 Study Evaluation

4.1.1 Pilot Studies

4.1.2 Questionnaires

4.1.3 Process

4.1.4 Incidences

4.1.5 acclimatisation phase

hüft tracker rutschen hoch. komma durch punkt ersetzen
aufgrund der Pilotstudie beschreiben, welche elemente gut bzw schlecht sind.
Wird gemessen was gemessen werden soll
sind die positionen der lehrer ok
sind tisch und box geeignet
gibt es schwierigkeiten etwas zu verstehen
ist die aklimatisierungsmethode angebracht
wie ist die dauer der durchführung einer session
sind die gestellten fragen am ende zielführend
pausen zwischen den sessions
sind die anweisungen die gegeben wurden zu viel/zu wenig
...
refinements

4.1.6 How to Evaluate

5 Conclusion

(3 pages)

5.1 System and Study

Zusammenfassung der Evaluation des Systems über die Eignung zur Durchführung einer Studie, die Daten generiert, um die Forschungsfrage zu beantworten.

Zusammenfassung, was gut und schlecht ist bei der Studienaufführung.
Reflexion und Contribution, inkl. zu erwartende empirische Contribution

5.2 Outlook

Was kann noch evaluiert werden mit diesem System?

anderer Task ohne physical load, sitzend zur Bedienung von Maschinen, Realismusgrad der Avatare, Anzahl Avatare, Position von Avataren, Geschwindigkeit der Animationsanleitung...

Wer hat welchen Nutzen von der Beantwortung der Forschungsfrage: Designer von Motorlearning VR-Systemen.

Bezug zu Erweiterungen der Implementierung

Possible improvement: DTW <https://towardsdatascience.com/dynamic-time-warping-3933f25fcdd>

6 Attachments

6.1 Study Documents

References

- [1] Fraser Anderson et al. “YouMove: Enhancing Movement Training with an Augmented Reality Mirror”. In: *Proceedings of the 26th Annual ACM Symposium on User Interface Software and Technology*. UIST ’13. St. Andrews, Scotland, United Kingdom: Association for Computing Machinery, 2013, 311–320. ISBN: 9781450322683. DOI: 10.1145/2501988.2502045. URL: <https://doi.org/10.1145/2501988.2502045>.
- [2] J. C. P. Chan et al. “A Virtual Reality Dance Training System Using Motion Capture Technology”. In: *IEEE Transactions on Learning Technologies* 4.2 (2011), pp. 187–195. DOI: 10.1109/TLT.2010.27.
- [3] Shuo Yan et al. “OutsideMe: Augmenting Dancer’s External Self-Image by Using A Mixed Reality System”. In: *Proceedings of the 33rd Annual ACM Conference Extended Abstracts on Human Factors in Computing Systems*. CHI EA ’15. Seoul, Republic of Korea: Association for Computing Machinery, 2015, 965–970. ISBN: 9781450331463. DOI: 10.1145/2702613.2732759. URL: <https://doi.org/10.1145/2702613.2732759>.
- [4] Jacky Chan et al. “Immersive Performance Training Tools Using Motion Capture Technology”. In: *Proceedings of the First International Conference on Immersive Telecommunications*. ImmersCom ’07. Bussolengo, Verona, Italy: ICST (Institute for Computer Sciences, Social-Informatics and Telecommunications Engineering), 2007. ISBN: 9789639799066.
- [5] K. Hachimura, H. Kato, and H. Tamura. “A prototype dance training support system with motion capture and mixed reality technologies”. In: *RO-MAN 2004. 13th IEEE International Workshop on Robot and Human Interactive Communication (IEEE Catalog No.04TH8759)*. 2004, pp. 217–222. DOI: 10.1109/ROMAN.2004.1374759.
- [6] Alexandra Covaci, Anne-Hélène Olivier, and Franck Multon. “Third Person View and Guidance for More Natural Motor Behaviour in Immersive Basketball Playing”. In: *Proceedings of the 20th ACM Symposium on Virtual Reality Software and Technology*. VRST ’14. Edinburgh, Scotland: Association for Computing Machinery, 2014, 55–64. ISBN: 9781450332538. DOI: 10.1145/2671015.2671023. URL: <https://doi.org/10.1145/2671015.2671023>.
- [7] Taihei Kojima et al. “Training Archived Physical Skill through Immersive Virtual Environment”. In: *Human Interface and the Management of Information. Information and Knowledge in Applications and Services*. Ed. by Sakae Yamamoto. Cham: Springer International Publishing, 2014, pp. 51–58. ISBN: 978-3-319-07863-2.
- [8] Eduardo Velloso, Andreas Bulling, and Hans Gellersen. “MotionMA: Motion Modelling and Analysis by Demonstration”. In: *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems*. New York, NY, USA: Association for Computing Machinery, 2013, 1309–1318. ISBN: 9781450318990. URL: <https://doi.org/10.1145/2470654.2466171>.
- [9] Richard Tang et al. “Physio@Home: Exploring Visual Guidance and Feedback Techniques for Physiotherapy Exercises”. In: *Proceedings of the 33rd Annual ACM Conference on Human Factors in Computing Systems*. New York, NY, USA: Association for Computing Machinery, 2015, 4123–4132. ISBN: 9781450331456. URL: <https://doi.org/10.1145/2702123.2702401>.
- [10] Vijay Rajanna et al. “KinoHaptics: An Automated, Wearable, Haptic Assisted, Physio-therapeutic System for Post-surgery Rehabilitation and Self-care”. In: *Journal of Medical Systems* 40 (Dec. 2015). DOI: 10.1007/s10916-015-0391-3.
- [11] Maurício Sousa et al. “SleeveAR: Augmented Reality for Rehabilitation Using Realtime Feedback”. In: *Proceedings of the 21st International Conference on Intelligent User Interfaces*. IUI ’16. Sonoma, California, USA: Association for Computing Machinery, 2016, 175–185. ISBN: 9781450341370. DOI: 10.1145/2856767.2856773. URL: <https://doi.org/10.1145/2856767.2856773>.

- [12] Maureen Holden et al. "Virtual Environment Training Improves Motor Performance in Two Patients with Stroke: Case Report". In: *Journal of Neurologic Physical Therapy* 23 (1999). URL: https://journals.lww.com/jnpt/Fulltext/1999/23020/Virtual_Environment_Training_Improves_Motor.13.aspx.
- [13] Ping-Hsuan Han et al. "AR-Arm: Augmented Visualization for Guiding Arm Movement in the First-Person Perspective". In: *Proceedings of the 7th Augmented Human International Conference 2016*. AH '16. Geneva, Switzerland: Association for Computing Machinery, 2016. ISBN: 9781450336802. DOI: 10.1145/2875194.2875237. URL: <https://doi.org/10.1145/2875194.2875237>.
- [14] U. Yang and G. J. Kim. "Implementation and Evaluation of "Just Follow Me": An Immersive, VR-Based, Motion-Training System". In: *Presence* 11.3 (2002), pp. 304–323. DOI: 10.1162/105474602317473240.
- [15] Nicholas Katzakis et al. "Stylo and Handifact: Modulating Haptic Perception through Visualizations for Posture Training in Augmented Reality". In: *Proceedings of the 5th Symposium on Spatial User Interaction*. SUI '17. Brighton, United Kingdom: Association for Computing Machinery, 2017, 58–67. ISBN: 9781450354868. DOI: 10.1145/3131277.3132181. URL: <https://doi.org/10.1145/3131277.3132181>.
- [16] Taku Komura et al. "E-Learning Martial Arts". In: *Proceedings of the 5th International Conference on Advances in Web Based Learning*. ICWL'06. Penang, Malaysia: Springer-Verlag, 2006, 239–248. ISBN: 3540490272. DOI: 10.1007/11925293_22. URL: https://doi.org/10.1007/11925293_22.
- [17] Ping-Hsuan Han et al. "My Tai-Chi Coaches: An Augmented-Learning Tool for Practicing Tai-Chi Chuan". In: *Proceedings of the 8th Augmented Human International Conference*. AH '17. Silicon Valley, California, USA: Association for Computing Machinery, 2017. ISBN: 9781450348355. DOI: 10.1145/3041164.3041194. URL: <https://doi.org/10.1145/3041164.3041194>.
- [18] Otniel Portillo-Rodriguez et al. "Real-Time Gesture Recognition, Evaluation and Feed-Forward Correction of a Multimodal Tai-Chi Platform". In: *Haptic and Audio Interaction Design*. Ed. by Antti Pirhonen and Stephen Brewster. Berlin, Heidelberg: Springer Berlin Heidelberg, 2008, pp. 30–39. ISBN: 978-3-540-87883-4.
- [19] Thuong N. Hoang et al. "Onebody: Remote Posture Guidance System Using First Person View in Virtual Environment". In: *Proceedings of the 9th Nordic Conference on Human-Computer Interaction*. NordiCHI '16. Gothenburg, Sweden: Association for Computing Machinery, 2016. ISBN: 9781450347631. DOI: 10.1145/2971485.2971521. URL: <https://doi.org/10.1145/2971485.2971521>.
- [20] Philo Tan Chua et al. "Training for physical tasks in virtual environments: Tai Chi". In: *IEEE Virtual Reality, 2003. Proceedings.* 2003, pp. 87–94. DOI: 10.1109/VR.2003.1191125.
- [21] J. Lieberman and C. Breazeal. "TIKL: Development of a Wearable Vibrotactile Feedback Suit for Improved Human Motor Learning". In: *IEEE Transactions on Robotics* 23.5 (2007), pp. 919–926. DOI: 10.1109/TR0.2007.907481.
- [22] Rajinder Sodhi, Hrvoje Benko, and Andrew Wilson. "LightGuide: Projected Visualizations for Hand Movement Guidance". In: *Proceedings of the SIGCHI Conference on Human Factors in Computing Systems*. CHI '12. Austin, Texas, USA: Association for Computing Machinery, 2012, 179–188. ISBN: 9781450310154. DOI: 10.1145/2207676.2207702. URL: <https://doi.org/10.1145/2207676.2207702>.
- [23] Maximilian Dürr et al. "NurseCare: Design and 'In-The-Wild' Evaluation of a Mobile System to Promote the Ergonomic Transfer of Patients". In: *Proceedings of the 2020 CHI Conference on Human Factors in Computing Systems*. CHI '20. Honolulu, HI, USA: Association for Computing Machinery, 2020, 1–13. ISBN: 9781450367080. DOI: 10.1145/3313831.3376851. URL: <https://doi.org/10.1145/3313831.3376851>.
- [24] Maximilian Dürr et al. "KiTT - The Kinaesthetics Transfer Teacher : Design and Evaluation of a Tablet-based System to Promote the Learning of Ergonomic Patient Transfers". In: *Proceedings of the ACM Conference on Human Factors in Computing Systems (CHI 2021)*. New York: ACM, 2021. ISBN: 978-1-4503-8096-6. DOI: 10.1145/3411764.3445496.
- [25] Paul Milgram and Fumio Kishino. *A TAXONOMY OF MIXED REALITY VISUAL DISPLAYS*. 1994.

- [26] R.A. Schmidt and T.D. Lee. *Motor Control and Learning: A Behavioral Emphasis*. Human Kinetics, 2005. ISBN: 9780736042581.
- [27] X. Yu et al. “Perspective Matters: Design Implications for Motion Guidance in Mixed Reality”. In: *2020 IEEE International Symposium on Mixed and Augmented Reality (ISMAR)*. 2020, pp. 577–587. DOI: 10.1109/ISMAR50242.2020.00085.
- [28] Wenbi Wang. “Dynamic Viewpoint Tethering: Controlling a Virtual Camera for Effective Navigation in Virtual Environments”. In: *CHI '01 Extended Abstracts on Human Factors in Computing Systems*. CHI EA '01. Seattle, Washington: Association for Computing Machinery, 2001, 93–94. ISBN: 1581133405. DOI: 10.1145/634067.634124. URL: <https://doi.org/10.1145/634067.634124>.
- [29] R. Rajesh. “Manual Material Handling: A Classification Scheme”. In: *Procedia Technology* 24 (2016). International Conference on Emerging Trends in Engineering, Science and Technology (ICETEST - 2015), pp. 568–575. ISSN: 2212-0173. DOI: <https://doi.org/10.1016/j.protcy.2016.05.114>. URL: <https://www.sciencedirect.com/science/article/pii/S2212017316302031>.
- [30] Maximilian Dürr et al. “EGuide: Investigating Different Visual Appearances and Guidance Techniques for Egocentric Guidance Visualizations”. In: *Proceedings of the Fourteenth International Conference on Tangible, Embedded, and Embodied Interaction*. TEI '20. Sydney NSW, Australia: Association for Computing Machinery, 2020, 311–322. ISBN: 9781450361071. DOI: 10.1145/3374920.3374945. URL: <https://doi.org/10.1145/3374920.3374945>.
- [31] Misha Sra, Aske Mottelson, and Pattie Maes. “Your Place and Mine: Designing a Shared VR Experience for Remotely Located Users”. In: *Proceedings of the 2018 Designing Interactive Systems Conference*. DIS '18. Hong Kong, China: Association for Computing Machinery, 2018, 85–97. ISBN: 9781450351980. DOI: 10.1145/3196709.3196788. URL: <https://doi.org/10.1145/3196709.3196788>.
- [32] Kasper Hornbæk. “Some Whys and Hows of Experiments in Human–Computer Interaction”. In: *Found. Trends Hum.-Comput. Interact.* 5.4 (June 2013), 299–373. ISSN: 1551-3955. DOI: 10.1561/11000000043. URL: <https://doi.org/10.1561/11000000043>.
- [33] Jonathan Muckell, Yuchi Young, and Mitch Leventhal. “A Wearable Motion Tracking System to Reduce Direct Care Worker Injuries: An Exploratory Study”. In: *Proceedings of the 2017 International Conference on Digital Health*. DH '17. London, United Kingdom: Association for Computing Machinery, 2017, 202–206. ISBN: 9781450352499. DOI: 10.1145/3079452.3079493. URL: <https://doi.org/10.1145/3079452.3079493>.