Applications

Christina Mittag*, Regina Leiss, Katharina Lorenz and Thomas Seel

Development of a home-based wrist range-of-motion training system for children with cerebral palsy

Handgelenk-Trainingssystem zum spielerischen Heimtraining für Kinder mit unilateraler Zerebralparese

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Abstract: This paper presents the proof-of-concept of a home-based gamified wrist rehabilitation training system for children with cerebral palsy (CCP). We describe the user-centered design process of this system, which is composed of a wrist-worn inertial measurement unit (IMU) and a tangible device with an embedded IMU. The system employs a quaternion-based algorithm for automatic real-time estimation of the range of motion (RoM) covered by adduction/abduction and flexion/extension motions of the wrist. Experimental validation shows that the RoM can be determined with sufficient accuracy to control a game and that the algorithm is applicable in CCP. A serious game, which uses the presented algorithm and enables feedback as well as motivating stimuli, is implemented and evaluated by physiotherapists.

Keywords: rehabilitation engineering, home-based motor training, intelligent biomedical systems, inertial sensor fusion, motion state estimation, user-centered design, human-technology interaction, serious game design

Zusammenfassung: Der Beitrag stellt ein Handgelenk-Trainingssystem zum spielerischen Heimtraining für Kinder mit unilateraler Zerebralparese (CCP) vor. Wir beschreiben den nutzerzentrierten Entwicklungsprozess dieses Systems, welches aus einem Inertialsensor (IMU) am Handgelenk sowie einem Handstück mit einem eingebetteten IMU besteht. Kernelement des Systems ist ein Quaternion-basierter Algorithmus zur automatisierten Echtzeit-Schätzung des Bewegungsumfangs (RoM) von Adduktion/Abduktions- bzw. Flexions-/Extensionsbewegungen des Handgelenks. Die experimentelle Validierung zeigt, dass der RoM mit für die Spielsteuerung hinreichender Genauigkeit ermittelt werden kann und der Algorithmus für CCP anwendbar ist. Auf Grundlage dieses Algorithmus wurde ein Spiel implementiert und von Physiotherapeuten evaluiert, das den Kindern sowohl Feedback als auch motivierende Stimuli bietet.

Schlagwörter: Rehabilitationstechnik, häusliches Bewegungstraining, intelligente biomedizinische Systeme, Inertialsensorfusion, Bewegungszustandsschätzung, Nutzerzentriertes Design, Mensch-Technik-Interaktion, Entwurf von Trainingsspielen

1 Introduction

Cerebral palsy – a nonprogressive motor disorder in children [12] – is the most common significant childhood motor impairment in Europe with a prevalence of about 2 of 1000 in newborns [17]. Children with unilateral cerebral palsy (CCP), the predominant form with 54% [6], suffer from the reduced ability to use one body side. These patients need lifelong training to improve, or at least maintain, the function of the affected side. Learned non-use leads to less arm activity, and thereby participation in everyday life may be reduced.

The BMBF-funded project SHArKi (Multimodal, sensor-supported hand and arm training for children) strives to enhance these children’s upper-limb motor function by developing a system with daily-life feedback of arm usage in combination with motivating games for home exercises. The present article focuses on the latter and
describes the development of a game suitable for home-based therapy in CCP.

Self-motivation is the key to increase the amount of arm movements at home. This can be achieved by using serious games or virtual reality. Several studies investigated motivational aspects, compliance, feasibility, and therapy outcome of serious games and virtual reality for upper-limb training in clinical settings and at home. The used systems range from commercial rehabilitation products (e.g., Armeo® Spring, Pablo®) [2, 16] to entertainment products (e.g., Nintendo Wii) [1, 3, 20] or custom-made rehabilitation systems [4, 5, 10]. Commercial rehabilitation products have the advantage of providing certain therapeutic movements and detailed feedback, but due to high prices and complicated setting they are not suitable for home training. Studies on CCP using entertainment products [1, 7, 18] report positive feedback from the included patients and good compliance, whereas the therapeutic outcome in home setting studies remains unclear [1, 7]. Berry et al. provide further insight into the movement pattern by analyzing the play session with an optical tracking system and conclude that “[…] the child’s motivation to succeed in the game may at times conflict with the therapeutic goals of the activity” [20]. Entertainment products have the advantages of general acceptance in CCP and peers, high motivation to move as well as affordability. However, they do neither train specific movements nor do they try to avoid unwanted movements which may increase the misalignment due to spasticity.

Between these poles of either high-end, cost-intensive rehabilitation possibilities or the entertaining games for home training lies a gap that several authors try to bridge [5, 9, 11, 18]. Most of these approaches are in a functional test stage. All authors include therapists for CCP in the developmental process; Tatla et al. highlight the importance of interdisciplinary development teams, especially including the therapists into the process of developing and evaluating rehabilitation programs as they are the “stakeholders who determine the appropriateness of intervention […] by combining their clinical reasoning with the needs and preferences of their clients.” [19]. The present paper aims at filling the aforementioned gap while following this advice and presents a user-centered design approach for a game that meets the specific needs of CCP: training of a specific therapeutic movement, adaption to the child’s individual abilities and feedback options for the therapists. For the implementation of the individualization, algorithms for automatic adaptation are developed.

Concerning game input options, the systems described above use the Kinect [18], or a specifically developed joystick [11] or grip handles [5], as well as accelerometers or EMG sensors [9]. Inertial measurement units (IMUs) are small and lightweight and therefore a suitable further input option, as for example used in the Pablo system [16]. IMUs measure acceleration, angular velocity and magnetic field in three dimensions. From this raw data, the segment orientation can be estimated by sensor fusion. If the relative orientation between sensors and segments is known, the relative orientation of linked segments can be estimated and used to determine the range of motion (RoM) of the linking joint.

A recent review [22] compares the results in sensor positions, sensor fusion algorithm and RoM calculations for the upper limb generated by different systems. It concludes that the wrist angles can be determined by IMUs with a difference of less than 4 degrees in comparison to robotic or optical systems, but that due to the variability of study settings and algorithms used, further research is needed.

In the present contribution, we propose a setup with an IMU on a wristband, complemented by a cylindrical tangible device that contains an IMU as well, see Figure 1. We introduce a quaternion-based solution for the RoM estimation task that automatically accounts for individual differences in grasps of the tangible. Finally, we present the iterative design process with therapists which led to the implementation of a game concept that enables a playful wrist RoM training for CCP.

Figure 1: Proposed system (wristband and tangible device with IMUs) and local coordinate axes.

The remainder of the article is structured as follows. Section 2 illustrates the developmental process. Section 3 presents results on technical feasibility of the algorithm as well as physiotherapeutic feedback on the game implementation. Key aspects and limitations are discussed in Section 4, and conclusions are presented in Section 5.
2 Material and methods

2.1 Used therapeutic movements

Supplementary to literature research on hand function and therapeutic goals in CCP, a workshop in the “Practice for Neurological Rehabilitation” was conducted as part of the user-centered design approach. 13 physiotherapists discussed hand and arm exercises for fictive patients to identify and structure therapeutic goals and training movements with emphasis on the importance for the child’s daily life. Semi-structured interviews with physiotherapist and pediatric experts completed the process, resulting in a selection of movements to be used for the game, which are not only therapeutically well-founded but also well detectable by IMUs.

Beyond the importance of arm movements, therapists underlined the importance of wrist movements for the motoric development. Several measuring options were found to be inadequate or intolerable for patients, e.g. the use of gloves with spastic hands or the necessity of attaching something to the hand. This led to the development of a technical system using wristbands with IMUs complemented by a cylindrical tangible device that contains an IMU as well. The advantages of robustness and child-friendly use face the disadvantage that, in comparison to a hand mounted IMU, only the orientation of the cylinder’s longitudinal axis can be approximated.

Interviews and observation of the therapeutic work revealed the importance of individual adjustments to the abilities of each child and its temporary performance (e.g., the presence of fatigue). The therapist motivates the child to go beyond his or her limits but without overstraining it. In the development of the serious game for CCP, we try to adopt this and transfer the mode of operation of the therapist into the game’s guidelines and requirements. We aim for the game to be adjustable to the individual and temporary ability of the child and to be challenging without being overdemanding, because we expect a positive gaming experience to be the key contribution to ensure voluntary use at home.

2.2 Wrist range of motion

An algorithm which calculates the wrist angles for the flexion/extension (F/E) and abduction/adduction (A/A) movement as well as the resulting wrist RoM is needed to provide the real-time game input data as well as offline motion and performance assessment for therapists.

IMUs can be used to determine the orientation of a rigid body with respect to an inertial frame of reference that neither rotates nor accelerates and is aligned with Earth’s gravitational and magnetic fields. This is achieved by algorithms that use the gyroscope for orientation strapdown integration and compensate integration drift by weighted corrections based on accelerometer and magnetometer readings. Orientations are commonly represented by unit quaternions or rotation matrices, and existing algorithms include extended Kalman filter and particle filter implementations, among others. We use the quaternion-based algorithm proposed in [15]. Quaternions are numbers with three imaginary parts i, j, k defined by \(i^2 + j^2 + k^2 = -1\).

They are conveniently represented by a four-dimensional vector

\[ \mathbf{q} = \mathbf{w} + x\mathbf{i} + y\mathbf{j} + z\mathbf{k} \rightarrow [w, x, y, z]^\top. \]  (1)

The conjugate \(\mathbf{q}^*\) of a unit quaternion coincides with its inverse and is obtained by flipping the signs of all imaginary parts. Quaternion multiplication \(\otimes\) follows from the aforementioned definition. If unit quaternions are associated with a rotation axis \(\mathbf{j} = [j_x, j_y, j_z]^\top\) and rotation angle \(\alpha\) via the definition

\[ \mathbf{q} = \cos\left(\frac{\alpha}{2}\right) + (j_x\mathbf{i} + j_y\mathbf{j} + j_z\mathbf{k}) \sin\left(\frac{\alpha}{2}\right), \]  (2)

then for any vector \([A_{x}, A_{y}, A_{z}]^\top\) given in some coordinate system \(A\), the transformation

\[ [0_{Bx}, 0_{By}, 0_{Bz}]^\top = \mathbf{q} \otimes [0_{A_{x}, A_{y}, A_{z}}]^\top \otimes \mathbf{q}^* \]  (3)

yields the vector’s coordinates in a frame \(B\) that is rotated with respect to the frame \(A\) by the angle \(\alpha\) around the axis \(\mathbf{j}\). In the following, \(\mathbf{A}_q\) describes the orientation of the frame \(A\) with respect to \(B\), and \(\mathbf{B}_q^* = \mathbf{A}_q\) describes the orientation of the frame \(B\) with respect to \(A\). Finally, quaternion multiplication corresponds to a concatenation of rotations. Given \(\mathbf{A}_q\) and \(\mathbf{B}_q\), the orientation of \(A\) with respect to \(C\) is

\[ \mathbf{A}_{q} = \mathbf{B}_q \otimes \mathbf{A}_q. \]  (4)

In the present setup, two body frames and two sensor frames are considered. The forearm coordinate system is defined according to the International Society of Biomechanics (ISB) recommendations [23] as illustrated in Figure 1. The forearm IMU is mounted to the wristband such that its sensing axes coincide with the axes of this coordinate system. Due to the cylindrical shape of the tangible device, its longitudinal axis is approximately aligned.
with the mediolateral axis of the hand, and the IMU is embedded into the tangible such that its z-axis coincides with the longitudinal axis. However, also due to the cylindrical shape, the tangible-to-hand orientation around the longitudinal axis is not known.

The proposed algorithm for wrist RoM estimation can be described in three main steps:

Step 1: estimate the relative joint orientation.

The quaternion-based orientation estimation algorithm [15] yields the orientations of the forearm sensor \( F \mathbf{q}(t) \) and the tangible sensor \( T \mathbf{q}(t) \) with respect to a common inertial reference frame \( E \).

The orientation of the tangible sensor with respect to the forearm coordinate system is then obtained by:

\[
\mathbf{T} \mathbf{q}(t) = \mathbf{F} \mathbf{q}(t) \otimes \mathbf{T} \mathbf{q}(t),
\]

and this quaternion is used to transform the tangible coordinate axes \( \mathbf{r} \mathbf{x} = [1 \ 0 \ 0]^\top, \mathbf{r} \mathbf{y} = [0 \ 1 \ 0]^\top, \mathbf{r} \mathbf{z} = [0 \ 0 \ 1]^\top \) into the forearm coordinate system. For \( \mathbf{r} \mathbf{x} \) the calculation reads as follows:

\[
[0 \mathbf{r} \mathbf{x}(t)]^\top = \mathbf{T} \mathbf{q}(t) \otimes [0 \mathbf{r} \mathbf{x}]^\top \otimes \mathbf{T} \mathbf{q}^*(t).
\]

In the further description, all vectors are presented in the forearm frame \( F \).

Step 2: calculate angles that describe the motion.

When the wrist moves, the coordinates of the tangible axes in the forearm coordinate system change. For the A/A estimation, we project the tangible z-axis \( \mathbf{r} \mathbf{z} = [\mathbf{r} \mathbf{z}_x \ \mathbf{r} \mathbf{z}_y \ \mathbf{r} \mathbf{z}_z]^\top \) into the yz-forearm plane by omitting its \( x \)-coordinate and determine the phase angle in that plane using the four-quadrant arctangent function \( \text{atan}_2 \) with unwrap:

\[
a(t) = \text{atan}_2 (\mathbf{r} \mathbf{z}_x(t), \mathbf{r} \mathbf{z}_y(t)).
\]

For F/E, we project the tangible x-axis \( \mathbf{r} \mathbf{x} = [\mathbf{r} \mathbf{x}_x \ \mathbf{r} \mathbf{x}_y \ \mathbf{r} \mathbf{x}_z]^\top \) into the xy-forearm plane and likewise determine its phase angle

\[
\varepsilon(t) = \text{atan}_2 (\mathbf{r} \mathbf{x}_x(t), \mathbf{r} \mathbf{x}_y(t)).
\]

Since the orientation of the tangible with respect to the hand depends on the grasp and is not completely known, the determined angles differ from the true wrist joint angles by an unknown offset. This is a major challenge that results from the robust and easy-to-use design of the system. Nevertheless, if the subject moves back and forth between some minimum and maximum F/E or A/A angle, we can determine the corresponding RoM.

Step 3: define envelope and estimate RoM.

A dynamically adjusting envelope is used to determine the minimum and maximum value that both angles reach during a performed motion. To account for fatigue, the envelope shrinks automatically by a rate \( r = 1/s \) whenever the boundary values are not reached or exceeded. For A/A, the formalism is

\[
a_{\text{up}}(t) = \max \left( a_{\text{up}}(t-1) - rT, \frac{a_{\text{up}}(t-1) + a(t)}{2} \right),
\]

\[
a_{\text{dw}}(t) = \min \left( a_{\text{dw}}(t-1) + rT, \frac{a_{\text{dw}}(t-1) + a(t)}{2} \right),
\]

where \( t-1 \) denotes the previous sampling instant and \( T \) the sampling period. In each time step the algorithm checks whether the reduced value should be taken or if the envelope exceeds the current boundaries. In the second case, the current \( \alpha \) value is set to the mean of the current value and the last boundary value to smooth the envelope function. For F/E the same formalism is used to calculate \( \varepsilon_{\text{up}} \) and \( \varepsilon_{\text{dw}} \). The time-dependent RoM is determined as the difference between \( a_{\text{up}} \) and \( a_{\text{dw}} \) for A/A and \( \varepsilon_{\text{up}} \) and \( \varepsilon_{\text{dw}} \) for F/E:

\[
\text{RoM}_a(t) = a_{\text{up}}(t) - a_{\text{dw}}(t),
\]

\[
\text{RoM}_\varepsilon(t) = \varepsilon_{\text{up}}(t) - \varepsilon_{\text{dw}}(t).
\]

The algorithm should automatically adjust to individual children with different RoM. Therefore, at each time instant, the joint angles are scaled to the current RoM as follows:

\[
\dot{a}(t) = (a(t) - a_{\text{dw}}(t))/(\text{RoM}_a(t)).
\]

\[
\dot{\varepsilon}(t) = (\varepsilon(t) - \varepsilon_{\text{dw}}(t))/(\text{RoM}_\varepsilon(t)).
\]

In summary, the presented algorithm determines the wrist angles for a F/E and A/A movement from two IMUs. Then the angles are scaled to the current RoM (\( \dot{a} \) and \( \dot{\varepsilon} \)) to obtain one parameter for each movement that can be used as game input. The children’s individual RoM can provide feedback for the therapists.

2.3 Range of motion feasibility evaluation

The presented algorithm automatically determines the individual RoM from the measurements of a forearm IMU and an IMU inside a tangible device. As mentioned above, this setup was chosen to improve usability and avoid unnecessary donning effort. We now investigate whether this approach yields reliable RoM estimates in comparison to the more complex approach of attaching the second IMU...
directly onto the back of the hand. Moreover, the hand-mounted IMU serves as an approximate ground truth, since the accuracy of orientation and joint angle measurements from body-worn IMUs has been demonstrated to lie in the range of a few degrees in numerous studies [13, 14, 21, 22].

Another crucial objective of the algorithm is to be adaptable to changing RoM, which may occur due to fatigue or be dependent on the severity of the patient’s impairment. Therefore, the algorithm is evaluated with respect to the following three key questions, answered by experimental procedures:

1. Which accuracy can be achieved with the tangible in comparison to a hand-mounted IMU?
2. Does the algorithm automatically adapt to changing RoM?
3. Can the algorithm be used to track wrist movements of CCP?

Questions (1) and (2) are addressed in the following experimental trials: Three healthy subjects were equipped with a wristband IMU and a tangible device and wore an additional IMU on the back of the hand for approximate ground truth measurements. The used IMUs are Xsens MTw (Xsens Technologies B.V., Netherlands), and the sampling rate was set to 20 Hz. Subjects performed the following movements while seated at a table:
   a) grasping, max. F/E (5 times), release
   b) grasping, max. A/A (5 times), release.

The entire procedure was repeated five times. One subject was asked to perform a longer trial with changing RoM to analyze the RoM adaptation. Data analysis started when the tangible device was grasped. The RoM was estimated as described above, and ground truth values were determined from the relative orientation between forearm IMU and hand IMU.

Question (3) is addressed by investigating the applicability of the presented algorithm in CCP using two different sets of data from atypical therapeutic wrist exercises called “feed the monster”. The exercise is composed of a complex, task-oriented arm and hand movement: grasp a thin cylindrical card, move the hand under a horizontal bar and release the card through a horizontal resp. vertical slit (Figure 2 shows the exercise presented by a therapist), each version executed six times.

Data was sampled and recorded a) from a therapist during a workshop, presenting “healthy” movements and b) from a child with CP in the course of a longer movement analysis, including the Assisting Hand Assessment (AHA) [8]. The therapist and the child wore the aforementioned IMUs on each hand and both forearms (see Figure 2). The sampling frequency was set to 20 Hz in the therapist workshop and 60 Hz in the assessment with the child to get a higher resolution for further analysis. For this paper the data from the forearm and hand IMU was analyzed during the “feed the monster” exercise with the proposed algorithm.

2.4 Gamification concept and implementation

In addition to technical feasibility, game concepts for CCP must meet therapeutic requirements such as adaptability. Further impairments or reduced concentration due to brain injury need to be considered during the design of the optical game appearance. The motivational aspect must include different challenges and adaptation to fatigue.

In the game ideation phase, the core mechanic of the game was chosen to be an item collecting task in the genre of role playing with an aquatic turtle as avatar. Different approaches to cognitive challenges were developed, and these as well as the motoric challenges were sorted by difficulty. The wrist F/E and A/A movement can be trained separately or in combination. Therefore, two different concepts with different motoric challenges were implemented.

The first game concept (Figure 3, left side) considers only the F/E movement. The swimming turtle moves up and down upon extension and flexion respectively. Up to three cognitive levels were developed for motivation. Whereas in the first level the task is to collect the stars, in the second level the child must additionally avoid water plants. Plants grow at the bottom, and the child must perform a dorsal extension to avoid them. In the third level, more obstacles in the form of swimming fish need to be avoided.
The second concept (Figure 3, right side) uses the scaled F/E angle to move the turtle up and down and the scaled A/A angle to move the turtle to the sides. This game is controlled by 2D movements and has a higher motoric difficulty level. In the first level of the game, the stars will appear stationary on the outer boundaries of the screen. In the second level the stars are moving, and in the third level obstacles in the form of swimming fish need to be avoided.

The described concepts were implemented with the following technical realization. The raw data, consisting of three-dimensional accelerometer, gyroscope and magnetometer data from the forearm IMU and the tangible IMU, are transmitted wirelessly to a Linux laptop, which is executing code that was generated from a custom-made Simulink diagram and sends the game variables and parameters to a web socket. The game itself was implemented in GDevelop, an open-source and easy to use game engine. GDevelop retrieves the data from the web socket.

After implementation, both games were introduced to two experienced therapists for feedback as basis of a successive user-centered validation. A semi-structured interview has been conducted to discuss the following topics: therapeutic goals, target group, game parameter settings, motivation and design aspects.

3 Results

3.1 RoM feasibility

First the differences between hand movement and tangible device as well as the adaptation to different RoM were analyzed. Figure 4 shows the result of the experimental procedure from a healthy subject, performing an F/E movement with different RoMs. The first subplot illustrates the envelope adaption, which shows a fast reaction to an increasing envelope and slower adaption to the decreasing RoM. Because in a game situation the CCP obtains a reward for wider RoM movement, the second subplot marks with crosses, where the current angle was in the lower or upper 20%, corresponding to a scaled angle below 0.2 or above 0.8. A star marks that the current RoM was exceeded. In the third subplot the A/A and F/E RoM are shown as determined by the proposed method and the approximated ground truth determined from the hand mounted IMU. For the F/E movement, the mean RoM difference between hand and tangible device is 7 ± 3 deg for F/E and 12 ± 5 deg for A/A. For the A/A movement, the values amount to 10 ± 12 deg (A/A) and 5 ± 17 deg (F/E).

In a second step the algorithm was used to analyze the wrist movement of a nine-year-old girl with a right side unilateral cerebral palsy. She reached an AHA score of 84 out of 100, so her hemiparesis is mild. Wrist F/E and A/A angles were determined from the forearm and hand IMUs. The current angles for F/E and A/A as well as the RoM of the “feed the monster” exercise with horizontal slot are shown in Figure 5. When grasping a card, ε is minimal and α is maximal, i.e., during grasping the hand is in palmar flexion and abduction. Each time the child releases the card, her individual maximum dorsal extension is reached. The fourth card was dropped on the table, picked up and the way of grasping was corrected by the therapist, which explains the irregularity in the curve.

Table 1 shows the average RoM for the three healthy probands, the therapist and the child with CP. For the F/E movement, the therapist and the child have a similar RoM.
When the healthy subjects were asked to perform a maximum movement, their F/E RoM is about 30–40 degrees higher. For the A/A movement, the patient shows a higher RoM than the therapist or the healthy subjects during the F/E movement. The maximum A/A RoM of the healthy subjects corresponds to the child’s A/A RoM.

The child needs about 55 seconds for 6 repetitions with the horizontal slot and 40 seconds with the vertical slot, the therapist needs about 22 seconds and 23 seconds, respectively.

These different experimental settings demonstrate (1) the feasibility of the IMU-based RoM estimation with a tangible device, (2) the automatic adaptation to time-variant RoM, as well as (3) the applicability of the system for CCP.

### 3.2 Implementation and therapist evaluation

The implemented program calculates the scaled angles (\(\hat{\alpha}\) and \(\hat{\varepsilon}\)) according to the proposed algorithm and sends the calculated angles in real-time to a web socket. The flow experience of the implemented game was good without noticeable time delay. The presented algorithm adapts the game parameters \(\hat{\alpha}\) and \(\hat{\varepsilon}\) to the child’s individual performance by a dynamic envelope. The child’s individual baseline RoM is determined by a calibration movement at the beginning of each play session. The shrinking-envelope feature assures that the subject can continue playing even in the presence of fatigue.

A simple design without much distraction was realized in GDevelop, which helps the child to focus on the movement task. A primary goal of the game is to motivate a child to perform repeated motions at large amplitudes, which corresponds to \(\hat{\alpha}\) or \(\hat{\varepsilon}\) regularly exceeding 0.8 or falling below 0.2. The F/E game target is to collect 15 stars, which will appear at the top and bottom of the water corresponding to scaled epsilon ranges of 0–0.2 and 0.8–1. When the child gets “lazy” and only collects the lower stars without moving the hand, the bottom stars will disappear and the child has to collect stars at the top. This encourages continuous F/E movement.

The second game uses movements in two directions. In the first level, the stars will appear stationary on the outer boundaries from 0–0.2 and 0.8–1 at both dimensions.

In addition to these automatic adaptions, speed levels and number of obstacles may be adapted to the child’s capability by the therapist, such that the game is perceived as a challenge without being overly demanding.
The three cognitive levels aim to motivate the child to play for a longer time and therefore perform more repetitions.

The presented games were evaluated in the categories therapeutic goals, target group, setting parameters, motivational and design aspects by two experienced physiotherapists who very often train the wrist dorsal extension with their patients. Their affirmative first impression of the game was supported by positively rated details such as: (1) The turtle as a slow but amiable animal offers a “role model” for the children, whose movements are slow as well. (2) The clear design offers only little distraction from the concentration on the game task. (3) In addition to wrist movement, other therapeutic goals such as concentration and reaction, eye-hand-coordination and movement planning are trained. (4) The possibility to play with different cognitive levels offers not only chances for children with different grades of severity, but also potential for fast learners to be increasingly challenged with regard to reaction time and movement planning due to moved elements. (5) The therapists approve of their options to adapt the difficulty level by game parameters such as turtle speed, number of obstacles or size of stars as well as the chance to get feedback on the child’s RoM.

They conclude that the game is feasible for all severity levels and CCP of three years or older, provided that they can control their wrist movement, understand the connection between hand movement and turtle movement and can hold the tangible during game flow.

4 Discussion

This contribution introduces a training system with a balanced trade-off between the characteristics of therapy, requirements for home use and technical feasibility in a user-centered approach. Therapists highlight the importance of wrist movements. However, from a technical point of view, wrist motion cannot be tracked with only one IMU on the distal end of the forearm. Therefore, a tangible device with a second IMU is a child-friendly option for home use.

The proposed tangible device is a small, lightweight and robust solution that can be used even by young children at home, in contrast to many systems described in literature that only meet the therapeutic requirements.

The data from the subject with CP show that the algorithm can be used for this child with a hand and forearm IMU. One limitation is that we have no data from several children with different severity grades and that the IMU was mounted on the hand, not inside a tangible device. However, in Section 3.1 small differences were found between the RoM determined with the tangible and the hand-mounted sensor. The technical realization shows an accuracy of about 5–7 degrees for the F/E movement and 10–12 degrees for the A/A movement. The probands reported that the maximal A/A movement was more difficult and unusual. A further examination shows that with increased ulnar abduction, the index finger is relatively pushed forward, with radial abduction the finger shift is opposite. This might explain a larger RoM in the tangible sensor than in the hand sensor. The implemented algorithm proved robust even with a more complex task performed by a therapist and a child with CP.

The shrinking rate of the envelope margins was set to 1°/s. In the test with healthy probands, it seems that the envelope could be steeper (see Figure 4) to adapt faster to the new RoM. The therapy exercise with the child shows that the envelope is exceeded at each repetition, indicating that the envelope shrinks too fast. The number of repetitions per time seems to be a crucial factor for the envelope shrinking. A healthy subject performs six movement repetitions in less than ten seconds, a therapist needs about 22 seconds for the “feed the monster” task and the child needs approximately twice the time. For a positive game experience, the movement speed of CCP and number of repetitions must be investigated further.

The game concept was developed with regard to motivational aspects, allows multiple motoric and cognitive challenges and is adaptable to the child’s abilities. To individually adapt the game, the therapists not only wish to set the turtle speed and number of obstacles, but the size of obstacles, the number of repetitions and the game length, too.

Malposition of the hand due to spasticity results in a palmar flexion in combination with abduction or adduction, hence the A/A movement should only be trained from a neutral wrist position.

The difficulty level also depends on the arm posture – such as holding the arm high, supination of the arm or opening of the fingers. These difficulty steps are another example for the trade-off between therapeutic wishes and technical feasibility. Assuming the forearm sensor has an approximately fixed position on the dorsal aspect of the forearm, a supination movement can be easily detected whereas it remains unclear if the arm is placed on a table or stretched out.

In terms of feedback, the therapists would like to know points of fatigue, number of repetitions and the error rate (ratio collected stars to total amount of stars) in each direction, in addition to the overall RoM.
Although the therapists rate the game as motivating for CCP, they suggest adding further motivational elements, e.g., direct feedback to the CCP on its own improvement or a reward system. Ideas are visual elements e.g., the turtle is dancing after each game, or adaption options for the turtle avatar e.g., change the turtle’s color. They propose a further cognitive challenge where only stars from one color should be collected and also recommend four, instead of two, sequential levels with increasing difficulty: F/E movement (as described), A/A movement, F/E movement and A/A movement (as described) and F/E movement and A/A movement with the forearm in supination.

We plan to implement the proposed features so that the game may be played by CCP with even more motivating experience.

5 Conclusions

This contribution presents a novel training system and a quaternion-based algorithm for real-time estimation of the wrist range of motion from the measurements of inertial sensors at the forearm and inside a tangible device. The user-friendly solution to use a tangible device results in additional uncertainty due to different possibilities to grasp a rotationally symmetric form. The estimation algorithm was designed to compensate this uncertainty, and it was demonstrated experimentally that the method yields reliable RoM estimates in healthy subjects and CCP. Wrist movement in CCP can be analyzed and used as game input by automatic adjustment to the patient’s performance via an auto-adaptive envelope.

The technical realization of a serious game for CCP wrist training is described. Several user-interactions during the development process gave an inside view of the therapeutic setting and helped to understand the specific therapeutic needs and requirements of a home system for CCP. A positive therapist feedback and their remarks and recommendations on movements and motivation will help to further improve the system. This underlines the importance of continuous involvement of therapists, pointed out by Tatla et al. [19].

Subsequent to the present proof-of-concept for a home use wrist movement game for CCP, the next development steps will include a user-test with CCP for adaption of relevant parameters such as envelope steepness and game velocity as well as analysis of motivational aspects on the game design.

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Ethical approval: The research related to human use complies with all the relevant national regulations, institutional policies and was performed in accordance with the tenets of the Helsinki Declaration. Ethical approval for the test was given by the Technical University Berlin Ethic Committee.

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Bionotes

Christina Mittag
Chair of Medical Engineering, Technische Universität Berlin, Berlin, Germany
christina.mittag@tu-berlin.de

Christina Mittag received her M. Sc. in biomedical engineering from Technische Universität Berlin and is currently working as research assistant in the medical engineering lab at Technische Universität Berlin.
Regina Leiss
Chair of Medical Engineering, Technische Universität Berlin, Berlin, Germany
regina.leiss@tu-berlin.de

Regina Leiss obtained her engineering diploma in environmental technology at Technische Universität Berlin. Through her many years of work in research management she broadened her experience and is now guiding interdisciplinary teams in the medical engineering lab at Technische Universität Berlin.

Katharina Lorenz
Chair of Medical Engineering, Technische Universität Berlin, Berlin, Germany
katharina.lorenz@tu-berlin.de

Katharina Lorenz is product designer with focus on interaction design with a participatory design approach. She received her diploma at Berlin Weißensee Art College and is currently working as research assistant in the medical engineering lab at Technische Universität Berlin and at German Research Center for Artificial Intelligence.

Thomas Seel
Control Systems Group, Technische Universität Berlin, Berlin, Germany
seel@control.TU-berlin.de

Thomas Seel studied engineering cybernetics at OvGU Magdeburg and UC Santa Barbara and received the Ph. D. from Technische Universität Berlin in 2016. Being a member of the control systems group at Technische Universität Berlin, he teaches courses in sensor fusion and control engineering and manages the group’s research focuses learning control systems and inertial sensor networks.