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Combined action observation and motor imagery facilitates visuomotor adaptation in children with Developmental Coordination Disorder

Short title: AO+MI FOR CHILDREN WITH DCD

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Highlights

Combined action observation and motor imagery facilitates visuomotor adaptation

Updating of the internal forward model can be advanced by an AO+MI intervention

Internal modelling deficits in children with DCD are reflected in their eye-movements

AO+MI intervention improved eye-hand coordination and movement kinematics

AO+MI has potential as an intervention technique for use with children with DCD
Abstract

The internal modelling deficit (IMD) hypothesis suggests that motor control issues associated with Developmental Coordination Disorder (DCD) are the result of impaired predictive motor control. In this study, we examined the benefits of a combined action observation and motor imagery (AO+MI) intervention designed to alleviate deficits in internal modelling and improve eye-hand coordination during a visuomotor rotation task. Twenty children with DCD were randomly assigned to either an AO+MI group (who watched a video of a performer completing the task whilst simultaneously imagining the kinaesthetic sensations associated with action execution) or a control group (who watched unrelated videos involving no motor content). Each group then attempted to learn a 90° visuomotor rotation while measurements of completion time, eye-movement behaviour and movement kinematics were recorded. As predicted, after training, the AO+MI group exhibited quicker completion times, more target-focused eye-movement behaviour and smoother movement kinematics compared to the control group. No significant after-effects were present. These results offer further support for the IMD hypothesis and suggest that AO+MI interventions may help to alleviate such deficits and improve motor performance in children with DCD.

Keywords: Internal model deficits, motor learning, mental simulation, eye-movements, eye-hand coordination, visuomotor rotation
1. Introduction

Developmental coordination disorder (DCD) is a neurodevelopmental disorder that is estimated to affect between 1.7% and 6% of children worldwide (American Psychiatric Association [APA], 2013). The condition is categorised as a marked impairment in the development of motor coordination that interferes with activities of daily living. These impairments are below the level expected for the child's chronological age and must not be attributable to other neurological conditions, sensory problems, or low intelligence (APA, 2013). While the aetiology of DCD is not fully understood, one suggestion is that these motor control issues are the result of impaired predictive motor control, stemming from disrupted cognitive representations of movement. This has been labelled as the internal modelling deficit (IMD) hypothesis (Wilson & Butson, 2007; Wilson, Ruddock, Smits-Engelsman, Polatajko & Blank, 2013).

According to Wolpert (1997), internal models are neural representations of the external world that are used to calculate and adjust movements by predicting their expected sensory consequences. These predictions are made by comparing the body's current state to an efference copy of the motor command, which contains predicted movement trajectories and associated bodily sensations (Kawato, 1999). As typical sensorimotor learning develops, the incongruence between predicted and actual movement sensations are diminished or are used to guide skilful online adjustments, increasing movement coordination. Conversely, difficulty in the generation or implementation of predictive models of action leads to slow, effortful, inaccurate, and uncoordinated movements that are overly dependent on visual feedback (Deconinck et al., 2006; Wilson et al., 2013). These difficulties are characteristic of children with DCD (for a review, see Adams, Lust, Wilson, & Steenbergen, 2014) and are commonly observed in visuomotor adaptation tasks and through deficits in motor imagery ability.

Visuomotor adaptation is a form of sensorimotor learning that consists of participants learning to adapt, or correct for, an external (often visual) perturbation. One example of this is through visuomotor rotation tasks where the motion of a cursor is rotated by a given angle with respect to the motion of the mouse controlling it. The rate of adaptation to this rotation is a measurement of the direct-effects of the development of an internal model between motor movements and the spatial goal of the task (Wang & Lei,
The examination of after-effects (where the rotation is taken away) is a measure of how established the internal model actually is (Krakauer, 2009), with greater after-effects suggesting a more well-established internal model. After-effects are the unintentional remains of compensatory strategies used to adapt to a novel visuomotor workspace that are present when the performer is reintroduced to an environment in which the use of such strategies is not necessary (Ong & Hodges, 2010).

Using a line drawing task on a digitised tablet, Kagerer, Bo, Contreras-Vidal and Clark (2004) asked children with and without DCD to perform a 45° visuomotor rotation task and examined both direct-effects and after-effects. Results revealed that children with DCD were less affected by the visuomotor rotation and showed no after-effects. This suggested that they had a less well-defined internal model compared to the typically developing children. In a follow-up study, using a more complex 60° visuomotor rotation, Kagerer, Contreras-Vidal, Bo and Clark (2006) showed that children with DCD updated their internal model more effectively during exposure to an abrupt 60° visuomotor rotation compared to a more gradual rotation (i.e., increasing rotations of 10° every 21 trials until a rotation of 60° was achieved). These results suggest that the adaptation process in children with DCD is mediated by the complexity of the visuomotor perturbation, due to an impaired capacity to use small error signals to modify an internal model. Similar findings have also been reported in prism adaptation experiments, in which visual feedback is displaced using prism glasses that deflect vision laterally during throwing tasks (Brookes, Nicolson, & Fawcett, 2007; Cantin, Polatajko, Thach, & Jaglal, 2007).

Internal modelling deficits have also been evidenced in research examining the motor imagery ability characteristics of children with DCD. Motor imagery is the process of mentally rehearsing actions, typically without overt action or physical output (Jeannerod, 2001). Motor imagery is thought to access the same neural representation of a movement as that used in predictive modelling. This link to internal models is evidenced through research showing that motor imagery activates similar brain regions to those involved in motor skill planning and execution (Hardwick, Caspers, Eickhoff, & Swinnen, 2018), evokes similar eye-movement patterns (Causer, McCormick, & Holmes, 2013) and similar temporal congruence (i.e., mental chronometry) between imagined and executed actions (Guillot, Hoyek, Louis, & Collet, 2012). In accordance with the IMD hypothesis, individuals with DCD...
exhibit impairments in mental chronometry ability (Ferguson, Wilson & Smits-Engelsman, 2015), reduced ability to imagine egocentric transformations of the body (Barhoun et al., 2019), an impairment in the accuracy of motor imagery (Fuchs & Caçola, 2018) and reduced corticospinal excitability during motor imagery (Hyde et al., 2018).

Mental simulation techniques like motor imagery and action observation (i.e., the structured observation of action execution) have been proposed to be effective interventions that target internal model deficits (Adams, Lust & Steenbergen, 2018). These interventions have shown promise in improving movement outcomes in sporting tasks (Cumming & Ramsey, 2009) and for clinical conditions like Parkinson’s disease (Caligiore, Mustile, Spalletta, & Baldassarre, 2017), stroke (Ertelt & Binkofski, 2012; Zimmermann-Schlatter, Schuster, Puhan, Siekierka, & Steurer, 2009) and for children with cerebral palsy (Buccino et al. 2018). It has also been suggested that mental simulation techniques may be beneficial for children with DCD (Adams et al., 2018) and a small number of studies have reported positive outcomes. For example, Wilson, Thomas and Maruff (2002) found that motor imagery training was equally as effective as traditional perceptual motor training for developing motor skills, particularly with children with severe DCD (Wilson, Adams, Caeyenberghs, Thomas, Smits-Engelsman & Steenbergen, 2016). Finally, Adams, Smits-Engelsman, Lust, Wilson and Steenbergen (2017) reported clinically meaningful changes in motor skill proficiency after an intervention that included separate aspects of action observation preceding motor imagery for children with DCD.

Recent research has proposed that combining action observation with concurrent motor imagery of the same action (AO+MI: Eaves, Riach, Holmes, & Wright, 2016; Vogt, Di Rienzo, Collet, Collins, & Guillot, 2013) may lead to improved behavioural outcomes compared to either simulation technique performed in isolation (Bek, Gowen, Vogt, Crawford & Poliakoff, 2019; Romano Smith, Wood, Coyles, Roberts & Wakefield, 2019; Romano-Smith, Wood, Wright & Wakefield, 2018; Scott, Emerson, Dixon, Tayler & Eaves, 2019). The rationale for combining these techniques stems from neurophysiological studies which have identified that AO+MI produces increased activity in cortical areas linked to movement planning and execution, compared to either AO or MI performed separately (e.g., Wright, Williams & Holmes, 2014, for a review see Eaves et al., 2016). Recent evidence has suggested that such activity may be related to specific ways in which action observation
and motor imagery help to develop internal models (Kim, Frank, & Schack, 2017).

Specifically, action observation has been shown to promote the reorganization of
frontoparietal cortex as visual information is mapped onto motor circuits (Apšvalka, Cross, & Ramsey, 2018) and may help to develop the sequencing and timing of basic action concepts (Wright, Wood, Eaves, Bruton, Frank & Franklin, 2018). These basic action concepts are smaller components of mental representations that are related functionally and biomechanically to the successful execution of a motor skill (Frank, Land & Schack, 2013) and are encoded in long-term memory to guide motor skill execution (Schack & Mechsner, 2006). Kinaesthetic imagery has been shown to expedite the development of the internal model by improving the prediction of sensory consequences of the imagined movements (Kilteni, Andersson, Houborg & Ehrsson, 2018). Based on this evidence, and that which suggests children with DCD struggle with visual imagery, it is possible that combining both techniques through AO+MI will provide a more effective intervention that promotes the development of internal models and facilitates motor skill acquisition.

In a recent study that brought these areas together, Marshall, Wright, Holmes and Wood (2019) examined the efficacy of an AO+MI intervention in facilitating adaptation to a visuomotor rotation task in healthy adults. Specifically, participants wore eye-tracking equipment whilst performing an 180° visuomotor rotation task (i.e., leftward movements of the hand resulted in rightward movements of the cursor and vice-versa) at pre-test, during 20 intervention trials, and post-test. Results indicated that, relative to a control group, participants who engaged in AO+MI improved visuomotor adaptation (i.e., reduced task completion time) and alleviated the early reliance on visual feedback to control the cursor movement. This early reliance on visual feedback control is linked to the need to establish effective sensorimotor mapping rules (i.e., an internal model) related to motor commands, sensory outcomes and cursor movement (Sailer, Flanagan & Johansson, 2005). As internal models become established, vision is used in a more feedforward manner (i.e., target-focused) that supports the planning and control of manual action, indicative of task expertise (Land, 2009). Marshall et al.’s (2019) findings indicate that AO+MI interventions can facilitate the development of internal models and that this developmental process can be measured through changes in task-specific eye-movement behaviours.
Despite individuals with DCD exhibiting deficits in the predictive control of eye-movements (e.g., Debrabant, Gheysen, Caeyensberghs, Van Waelveke & Vigerhoets, 2013), no studies have explored eye-movements during the adaptation to visuomotor rotation in children with DCD. This is important as further support for the IMD hypothesis may be gained from an exploration of eye-movement behaviours of children with DCD during adaptation to visuomotor rotation. Furthermore, no studies have explored the efficacy of AO+MI for facilitating this process in this population. As individuals with DCD exhibit poor motor imagery ability, combining action observation with kinaesthetic imagery may be an effective intervention that provides accurate visual and temporal movement cues while enabling cognitive resources to be devoted to the generation of kinaesthetic imagery associated with the observed movement (Eaves et al., 2016). As visuomotor adaptation has been used with children with DCD previously, it is an ideal paradigm to assess the efficacy of AO+MI interventions for improving internal model deficits.

The aim of this experiment was to extend previous research on visuomotor adaptation and mental simulation in children with DCD by examining the utility of an AO+MI intervention for facilitating visuomotor adaptation and eye-hand coordination. Based on previous evidence (Marshall et al., 2019), it was hypothesised that AO+MI training would help to overcome deficits in internal modelling and produce a significant improvement in visuomotor adaptation task performance, underpinned by the facilitation of more predictive (i.e., target-focused) eye-movement behaviours, shorter cursor path lengths, and smoother movement kinematics. Finally, it was predicted that AO+MI training would produce significant after-effects when participants repeated the task with no rotation applied, indicating the more extensive development of the internal model (Kagerer et al. 2006).

2. Method

2.1 Participants

Twenty children aged 7 to 11 years (13 male, 7 female; age $M = 9.0$, $SD = 1.45$ years) with confirmed or suspected DCD were recruited through local DCD support groups. Potential participants were first screened using the revised version of the Developmental Coordination Disorder Questionnaire (DCDQ: Wilson, Kaplan, Crawford & Roberts, 2007) and those who were identified as potentially having DCD (i.e., scores within the range of 15-
were then invited to a testing session where they also completed the Movement Assessment Battery for Children-2 (MABC-2: Henderson, Sugden & Barnett, 2007). Only children who scored at or below the 5th percentile on the MABC-2 and who, based on parent reports, did not suffer from any other general medical condition known to affect sensorimotor function (e.g., cerebral palsy, hemiplegia, or muscular dystrophy) and had no diagnosis of learning difficulties or ADHD, were asked to take part in the study. Parents and children provided written informed consent and assent, respectively, prior to taking part. The experimental procedures were granted ethical approval by the institutional ethics committee prior to testing.

2.2 Task

Participants performed a virtual radial Fitts task. For this task, a 90° counterclockwise visual feedback rotation was used that resulted in stylus movements along the x-axis producing equivalent cursor movements along the y-axis and vice versa. This rotation resulted in upward movement of the stylus producing rightward cursor movement, rightward stylus movement produced a downward cursor movement, a downward stylus movement produced a leftward cursor movement and a leftward stylus movement produced an upward cursor movement. The goal of the task was to use a stylus to guide a cursor from a central home square to a yellow highlighted target square and then back to the home square (see Figure 1). Six targets were presented sequentially from left to right with the next target becoming highlighted each time the cursor returned to the central square. Based on a similar design used by Heremans et al. (2011), all the target positions were visible throughout the task in an arc radiating out at a distance of 170mm from the central square. One full trial consisted of all six targets being successfully hit and the cursor returning to the central square each time (totalling 12 target hits). Unity3D (Unity Technologies, San Francisco, CA) software was used to present the experimental task, to collect data in relation to cursor movement (80 Hz) and to record task completion time.
Figure 1. Image showing the experimental set-up (a) and the visuomotor adaptation task shown in the AO+MI video (b). The red circle around the cursor square represents the participant’s point of gaze and the yellow squares represent the target squares. The white square in the bottom/centre of the image represents the ‘home’ square.

2.3 Apparatus

Testing was performed on a vertically-oriented Dell ST2220T touchscreen monitor (Dell, Round Rock, TX) with a 480 mm x 270 mm visual display, situated 210 mm from the edge of the table where the participant was seated (Figure 1). Eye-movements were monitored using ETG 2w eye tracking glasses and iView ETG 2.7 software (SMI, Teltow, Germany). The system comprises a pair of lightweight glasses that track participants’ binocular eye-movements at a sampling rate of 60 Hz with a gaze position accuracy of 0.5°. The eye tracking glasses were calibrated for each participant prior to each trial by instructing them to fixate on points on a calibration grid that represented the spatial arrangement of the target sequences. If, during the session, the quality of the calibration was deemed by the experimenter to have deteriorated then the calibration procedure was repeated before testing continued.

2.4 Procedure
2.4.1 Pre-test: No rotation

Participants were first calibrated to the eye-tracker before performing two practice trials (totalling 24 target hits) of the task with no visuomotor rotation applied in order to familiarise themselves with the stylus, goal of the task, and experimental set-up. Participants then performed three pre-test trials (totalling 36 target hits) of the task with no visuomotor rotation applied which would be compared to any after-effects post-intervention. Throughout each phase of the experiment, the number of trials and target hits was based on those used by Kagerer et al. (2006) as this study used a similar visuomotor task to investigate visuomotor adaptation in children with DCD. Participants were instructed to perform the task as quickly and accurately as possible on each trial.

2.4.2 Pre-test: Rotation

Once participants had completed their practice trials, they then performed one trial (totalling 12 target hits) of the task with the $90^\circ$ visuomotor rotation applied. Prior to starting this trial, participants were informed that, although the task looked the same and still had the same goal, the cursor would move differently. Each participant was given a maximum of three minutes to hit all of the presented targets. If all the targets had not been hit during this time, 180 seconds was recorded as the trial completion time, along with the
number of targets successfully hit. Of the 20 participants, 14 reached the 180 second limit on the pre-test ($M = 164.85$, $SD = 26.61$). The three-minute maximum allowed for some control over the amount of exposure participants had to the novel visuomotor environment prior to training. Immediately after completing the pre-test rotation trial, participants started the training intervention to which they had been randomly assigned.

2.4.3 Intervention Groups

AO+MI. Participants in the AO+MI group (six male, four female; age $M = 9.0$, $SD = 1.56$ years) performed motor imagery of executing the task whilst they simultaneously observed a series of videos of a novice, typically developing, adult performer completing the same visuomotor rotation task. The video series consisted of three videos recorded at different stages of the learning experience as they performed 50 trials of the task. These stages were determined based on the number of trials completed by a child of similar age and were identified as: Early (trials 1 to 10), Mid (trials 11 to 30), and Late (trials 31 to 50). Each video was selected to represent the natural progression of adaptive behaviour as the child became more accomplished at the task (see Table 1 for a visualisation of the cursor path associated with these stages). The use of a series of videos for the AO+MI intervention was included as established models of motor imagery recommend that the motor imagery experience should adapt as learning progresses to reflect a learner’s level of physical experience (Holmes & Collins, 2001). In addition, visuomotor adaptation studies using observational learning have also used videos that show progressive changes in the model’s performance (Lei, Bao & Wang, 2016). Each video was filmed from the same first-person perspective, recorded from the scene camera of the eye-tracker, and showed only the touchscreen monitor and the novice performer’s hand moving the stylus over the screen in order to guide the cursor to each target (see Figure 1b). At the start of each video, a motor imagery script was presented in written form on the screen along with an audio-recorded narration. This script was slightly different for each video in order to reflect the adaptations made by the novice performer as their training progressed (see Table 1). Only kinaesthetic imagery instructions were provided because visual information was provided in the video, typical of AO+MI interventions (Eaves et al., 2016).

After each AO+MI trial, participants immediately performed a physical practice trial as previous research has suggested that observational learning alone is not enough to
update an internal model of the visuomotor environment and at least some amount of physical practice is required (Ong & Hodges, 2010; Ong, Larssen & Hodges 2012; Lei et al., 2016). This resulted in this intervention consisting of 21 AO+MI trials (totalling 252 target hits) and 21 physical practice trials (totalling 252 target hits), separated into three blocks of practice (see Figure 2). Rest periods (~ 2 mins) were given after every block and the eye-tracking equipment was checked for calibration before the start of each trial.

**Control.** Participants in the control group (seven male, three female; age $M = 9.0, SD = 1.41$ years) watched 42 second clips of a nature documentary that contained no human motor content (Scott et al., 2019) followed by an immediate physical practice trial. The duration of video clips was chosen in order to represent a total viewing time that was equivalent to the total duration of the AO+MI videos. These trials were also divided into three blocks of seven video and immediate physical practice trials and in total, participants in this group physically performed 21 trials of the task (totalling 252 target hits). Rest periods (~ 2 mins) were given after every block and the eye-tracking equipment was checked for calibration before the start of each trial.

**2.4.4 Post-test: Rotation**

Each participant completed a final rotation trial (totalling 12 target hits) as a post-test that was identical to the pre-test conditions. Each participant was again given a maximum of three minutes to hit all of the presented targets.

**2.4.5 Post-test: No Rotation**

Participants performed three trials of the task (totalling 36 target hits) with no visuomotor rotation, identical to pre-test conditions, to assess the presence of any after-effects. After this was completed, participants and their parents were debriefed and thanked for their participation.
Table 1. AO+MI instructions for each training stages of the intervention and the plotted
cursor paths of the model to give an illustration of the kinematic information represented in
each action observation video.

<table>
<thead>
<tr>
<th>Stage</th>
<th>Instructions</th>
<th>Plotted cursor path</th>
</tr>
</thead>
<tbody>
<tr>
<td>Early</td>
<td>“I am watching the video on the screen. The hand in the video is mine and I am making the movements that I see. I can feel myself holding the pen and I can feel my arm and hand moving the cursor to the yellow squares”</td>
<td>![Early Cursor Path]</td>
</tr>
<tr>
<td>Mid</td>
<td>“I am watching the video on the screen. The hand in the video is mine and I am making the movements that I see. I can feel myself holding the pen and I can feel my arm and hand moving the cursor in circles towards the yellow squares”</td>
<td>![Mid Cursor Path]</td>
</tr>
<tr>
<td>Late</td>
<td>“I am watching the video on the screen. The hand in the video is mine and I am making the movements that I see. My movements are steady and accurate. I can feel myself holding the pen and I can feel my arm and hand moving the cursor in oval patterns towards the yellow squares”</td>
<td>![Late Cursor Path]</td>
</tr>
</tbody>
</table>

3. Measures

3.1 Completion time
The time taken (in seconds) to finish the entire trial (12 target hits), from leaving the home square at the start to returning to the home square after hitting the sixth target, was used as a measure of completion time.

3.2 Target-locking score
Each pre-test and post-test trial for each participant was analysed using the BeGaze 3.7 software (SMI, Teltow, Germany). In addition, the 1st, 3rd, 5th, and 7th trials from each training block were also analysed. Targets were defined as the six outboard target squares and the central home square. Fixations were defined as gaze dispersed over less than 3° of visual angle for a minimum of 80ms. A target-locking score was then calculated by
subtracting the percentage of cursor fixation time from the percentage of target fixation
time to create a ratio measure of the allocation of visual attention. This method has
previously been used to determine the gaze control of participants performing visuomotor
adaptation tasks (Marshall et al., 2019), surgical tasks (Wilson, McGrath, Vine, Brewer,
Defriend & Masters, 2010), and tasks involving the control of a prosthetic hand (Parr, Vine,
Harrison & Wood, 2018; Parr, Vine, Wilson, Harrison & Wood, 2019). Using this method, a
more positive score reflects more time fixating on targets whereas a negative score reflects
more time spent fixating the cursor. A score of ‘0’ reflects equal time spent fixating the
cursor and targets and represents a ‘switching strategy’.

3.3 Movement Kinematics

For each trial, cursor movements were filtered using a 2nd order dual lowpass
Butterworth filter with an 8 Hz cut off frequency. The filtered data was then processed with
custom written Matlab 2017b (MathWorks Inc, Natick, MA) routines.

3.4 Total Path length

As children with DCD are thought to persist with ineffective movement strategies
(Biotteau, Chaix & Albaret, 2016), we measured total path length (mm) to gain a
quantifiable representation of the movement strategies that children were using in both
groups. Total path length was calculated between sampled pairs of x and y coordinates
using the following formula where \(x_1, x_2\) and \(y_1, y_2\) represent points along the x and y axes
respectively. The total units of distance (mm) for each sampled point were then summed to
provide a total path length for each trial.

\[
Path \ length = \sqrt{(x_2 - x_1)^2 + (y_2 - y_1)^2}
\]

3.5 Normalised Jerk

For each trial, jerk was calculated as a measure of movement smoothness. As jerk
varies according to both the duration and size of a movement, these data was normalised
using the following formula where \(j\) refers to jerk and \(t\) to time:

\[
Normalised \ jerk = \sqrt{\frac{1}{2} \int dt \ j^2 (t) \times \frac{\text{duration}^5}{\text{length}^2}}
\]
This calculation produces a unit-free measure that can be used to compare movements of
different sizes and durations (Teulings et al., 1997; Kagerer et al., 2006).

### 3.6 After-effects

The presence of after-effects following the adaptation training was assessed by
calculating completion time, target-locking score, total path length, and normalised jerk on
the no rotation trials pre and post intervention. In addition, the length and root mean
square error (RMSE) of the path to the first target was also calculated in order to identify
any initial after-effects before they were washed out over subsequent target hits. RMSE is a
measure of the spatial deviation from a direct vector between home and target (Kagerer et
al., 2004; 2006).

### 3.7 Data analysis

Due to the data for completion time and mean total path length violating the
assumption of normality, these variables were successfully log transformed. Separate 2
(Group: AO+MI, control) x 5 (Time: Pre-test, T1, T2, T3, Post-test) mixed measures ANOVAs
were performed on participant’s completion time, gaze control, mean path length, and
normalised jerk. Significant interactions were followed up with Bonferroni corrected
pairwise comparison that compared each group at each time point (Pre-test, T1, T2, T3,
Post-test). To assess the presence of after-effects, a 2 (Group: AO+MI, control) x 2 (Pre-test
vs. Post-test) mixed measures ANOVA was conducted for pre and post no rotation trials
(Kagerer et al., 2006). For all analyses, where sphericity was violated, Greenhouse-Geisser
corrections were applied. Effect sizes are reported as partial eta squared ($\eta^2_p$), and the
alpha level for statistical significance was set at 0.05.

### 4. Results

#### 4.1 Completion time

The ANOVA revealed significant main effects for time, $F(2.41, 43.35) = 152.45, p <\
.001, \eta^2_p = .89$, and group, $F(1,18) = 11.53, p = .003 \eta^2 = .39$, which were superseded by a
significant interaction effect, $F(2.41, 43.55) = 3.97, p = .020, \eta^2_p = .18$. As expected, post-hoc
comparisons revealed no significant difference between groups at pre-test ($p = .699$) or T1 ($p$
= .172), but the AO+MI group produced significantly faster completion times than the
control group at T2 \((p = .002)\), T3 \((p = .007)\) and post-test \((p = .009)\). These data are presented in Figure 3a.

### 4.2 Target-locking score

The ANOVA revealed significant main effects for time, \(F(2.01, 36.22) = 114.78, p < .001, \eta_p^2 = .86\), and group, \(F(1,18) = 22.89, p < .001, \eta_p^2 = 0.56\), which were superseded by a significant interaction, \(F(2.01, 36.22) = 4.26, p = .022, \eta_p^2 = .19\), for target-locking score. As expected, post-hoc comparisons revealed no significant difference between groups at pre-test \((p = .33)\), but the AO+MI group had a significantly greater TLS at T1 \((p < .001)\), T2 \((p < .001)\), T3 \((p = .002)\) and post-test \((p = .012)\). These data are presented in Figure 3b.

### 4.3 Movement kinematics

All pre-test kinematic data for one participant in the AO+MI group was removed prior to analysis due to technical issues with the touch screen that meant the cursor functioned correctly but the values generated were erroneous.

### 4.4 Total path length

The ANOVA revealed a significant main effect for time, \(F(1.94, 33.01) = 12.53, p < .001, \eta_p^2 = .42\), indicating that both groups produced shorter cursor paths as training progressed. There was no significant main effect for group, \(F(1, 17) = 3.91, p = .064, \eta_p^2 = .18\), and, unexpectedly, no significant interaction was found, \(F(1.94, 33.01) = 2.65, p = .087, \eta_p^2 = .14\). These data are presented in Figure 3c. A visual representation of path length illustrating the strategies that participants typically used is presented in Figure 4.

### 4.5 Normalised Jerk

The ANOVA revealed a significant main effect for time, \(F(4, 68) = 11.79, p < .001, \eta_p^2 = .41\), indicating both groups exhibited an increase in movement smoothness throughout the training. A significant main effect for group was also revealed, \(F(1, 17) = 31.98, p < .001, \eta_p^2 = .65\), indicating that the movements of the control group were significantly more jerky \((M = 8.98, SD = 4.02)\) compared to the movements of the AO+MI group \((M = 5.28, SD = 1.77)\). In contrast to our predictions, no significant interaction was found, \(F(4, 68) = .79, p = .536, \eta_p^2 = .05\). These data are presented in Figure 3d.
4.6 After-effects

The ANOVA revealed no significant main effects or interactions between groups for all after-effect variables measured (see Table 2).
Figure 3. Mean completion time (a), mean target-locking score (b), total cursor path length (c) and normalised jerk (d) for both groups across pre-test, training blocks (T1, T2, T3) and post-test.
Table 2. After-effects data showing the completion time, target-locking score and kinematic data (SD) for each intervention group at pre-test and post-test with no rotation present.

<table>
<thead>
<tr>
<th></th>
<th>AO+MI</th>
<th>Control</th>
<th>Inferential Statistics</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Pre</td>
<td>Post</td>
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</tr>
<tr>
<td>Completion time</td>
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<td></td>
<td>Time</td>
</tr>
<tr>
<td>(seconds)</td>
<td>14.66</td>
<td>17.04</td>
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<tr>
<td></td>
<td>(5.90)</td>
<td>(5.31)</td>
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<td></td>
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</tr>
<tr>
<td>Time</td>
<td>14.49</td>
<td>17.27</td>
<td>1,18</td>
</tr>
<tr>
<td>Group</td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Interaction</td>
<td></td>
<td></td>
<td>1,18</td>
</tr>
<tr>
<td>Target-locking score (%)</td>
<td>64.01</td>
<td>61.61</td>
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<tr>
<td></td>
<td>(23.01)</td>
<td>(31.50)</td>
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<tr>
<td></td>
<td></td>
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<td>1,18</td>
</tr>
<tr>
<td>Time</td>
<td>62.34</td>
<td>55.40</td>
<td>1,18</td>
</tr>
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<td>Group</td>
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<td></td>
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</tr>
<tr>
<td>Interaction</td>
<td></td>
<td></td>
<td>1,18</td>
</tr>
<tr>
<td>Total path length (mm)</td>
<td>2749.55</td>
<td>3262.23</td>
<td>1,18</td>
</tr>
<tr>
<td></td>
<td>(171.05)</td>
<td>(1035.77)</td>
<td>1,18</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1,18</td>
</tr>
<tr>
<td></td>
<td>2866.35</td>
<td>2884.36</td>
<td>1,18</td>
</tr>
<tr>
<td></td>
<td>(246.44)</td>
<td>(672.84)</td>
<td>1,18</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td>1,18</td>
</tr>
<tr>
<td>Time</td>
<td>2866.35</td>
<td>2884.36</td>
<td>1,18</td>
</tr>
<tr>
<td>Group</td>
<td></td>
<td></td>
<td>1,18</td>
</tr>
<tr>
<td>Interaction</td>
<td></td>
<td></td>
<td>1,18</td>
</tr>
<tr>
<td>Normalised jerk</td>
<td>3.21</td>
<td>2.02</td>
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<tr>
<td></td>
<td>(0.46)</td>
<td>(0.53)</td>
<td>1,18</td>
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<td></td>
<td>2.99</td>
<td>2.68</td>
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</tr>
<tr>
<td></td>
<td>(0.94)</td>
<td>(1.11)</td>
<td>1,18</td>
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<tr>
<td></td>
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<td></td>
<td>1,18</td>
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<tr>
<td></td>
<td>2.69</td>
<td>2.30</td>
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<tr>
<td></td>
<td>(1.11)</td>
<td>(2.05)</td>
<td>1,18</td>
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<tr>
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<td></td>
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<td>1,18</td>
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<tr>
<td>First path length (mm)</td>
<td>265.76</td>
<td>338.28</td>
<td>1,18</td>
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<tr>
<td></td>
<td>(65.16)</td>
<td>(122.90)</td>
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<td></td>
<td></td>
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<td></td>
<td>254.82</td>
<td>270.23</td>
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<td></td>
<td>(49.99)</td>
<td>(56.05)</td>
<td>1,18</td>
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<td>1,18</td>
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<td></td>
<td>270.23</td>
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<tr>
<td></td>
<td>(56.05)</td>
<td>(56.05)</td>
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<td>First path RMSE (mm)</td>
<td>9.81</td>
<td>12.09</td>
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<td>(0.55)</td>
<td>(4.05)</td>
<td>1,18</td>
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<td></td>
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<td></td>
<td>1,18</td>
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<td>10.45</td>
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<td>(2.99)</td>
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<td>10.94</td>
<td>10.94</td>
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<td></td>
<td>(2.99)</td>
<td>(2.99)</td>
<td>1,18</td>
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<tr>
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<td>1,18</td>
</tr>
</tbody>
</table>
Figure 4. A visual representation of cursor paths and respective completion times (seconds) produced by two participants during the AO+MI (top) and control (bottom) training interventions. These participants were chosen as their post-test completion times were similar to the overall group mean completion times for each AO+MI ($M = 53.76$, $SD = 22.34$) and Control group ($M = 26.15$, $SD = 14.82$) at post-test.

AO+MI

<table>
<thead>
<tr>
<th></th>
<th>Pre-test</th>
<th>T1</th>
<th>T2</th>
<th>T3</th>
<th>Post-test</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>180 secs</td>
<td>76.00 secs</td>
<td>47.17 secs</td>
<td>40.62 secs</td>
<td>28.72 secs</td>
</tr>
</tbody>
</table>

Control

<table>
<thead>
<tr>
<th></th>
<th>Pre-test</th>
<th>T1</th>
<th>T2</th>
<th>T3</th>
<th>Post-test</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>180 secs</td>
<td>120.61 secs</td>
<td>72.56 secs</td>
<td>56.26 secs</td>
<td>54.93 secs</td>
</tr>
</tbody>
</table>
5. Discussion

The aim of this experiment was to extend previous research on visuomotor adaptation and mental simulation in children with DCD by examining the benefits of an AO+MI intervention for facilitating visuomotor adaptation and eye-hand coordination.

Based on the assumption that the impairments associated with DCD are the result of deficits in internal modelling, it was predicted that a dual-simulation technique incorporating the simultaneous performance of both AO and MI would facilitate the development of internal models, improve visuomotor adaptation, and optimise both eye-movement behaviour and movement kinematics. The results of this experiment provide some support for these hypotheses. First, as predicted, the AO+MI training group produced a significant improvement in task performance (i.e., quicker completion times) compared to the control group. In fact, the AO+MI group performed significantly quicker than the control group by the second training block (T2) and maintained this advantage in the subsequent training block (T3) and post-test phase (Figure 3a). These results are the first to demonstrate that AO+MI interventions can aid visuomotor adaptation and support previous research that has shown beneficial effects of AO+MI on performance outcomes generally (Bek et al., 2019; Romano-Smith et al., 2018; 2019) and within the DCD population specifically (Scott et al., 2019).

Further evidence that AO+MI helped to develop internal models is reflected in the eye-movement data. As eye-movement patterns are shaped by internal models (Hayhoe & Ballard, 2005), it was expected that any changes in the internal model would be reflected in changes in eye-movement behaviour. As predicted, the eye-movements of the AO+MI group progressed from being predominately used as a feedback resource (i.e., watching the cursor movement) to becoming a feedforward resource (i.e., target-focused) as children became more skilled at the task. Whereas both groups exhibited a predominantly ‘cursor-focused’ visual strategy at pre-test (target-locking score of approximately -60%), the AO+MI group became almost totally ‘target-focused’ at post-test (target-locking score of approximately 40%). In contrast, the control group were unable to progress much beyond a switching strategy between the cursor and target by post-test (target-locking score just above 0%; Figure 3b). Interestingly, the AO+MI group surpassed the development of the control group after the first training block.
These changes in eye-movement behaviours are consistent with previous research in visuomotor learning (e.g., Sailer et al., 2005) and with recent research showing similar benefits of AO+MI training on visuomotor rotation in typically developing adults (Marshall et al., 2019). This early reliance on slower (visual) feedback control is linked to the need to establish effective sensorimotor mapping rules (i.e., an internal model) relating to motor commands, sensory outcomes and cursor movement (Sailer et al., 2005). As skill progresses and sensorimotor mapping rules are developed, cursor movement is controlled by proprioceptive modes of control and vision is freed-up to focus on targets ahead of time (Marshall et al., 2019). Task-specific (goal-directed) eye-movements of this nature support the planning and control of manual action and are indicative of top-down attentional control and task expertise (Land, 2009). Interestingly, children with DCD have shown an inability to develop optimal, task-specific, eye-movement strategies unless explicitly trained to do so (Miles, Wood, Vine, Vickers & Wilson, 2016; Wood et al., 2017; Slowinski et al., 2019), as evident in our control group. This reliance on vision to monitor movements aligns with evidence from neurological studies that suggests that children with DCD display increased cortical activity in areas related to visuospatial processing and conscious movement control compared to typically developing peers (Zwicker, Missiuna, Harris, & Boyd, 2010). This shows that deficits in internal modelling are reflected in eye-movement behaviours of children with DCD and that the exploration of eye-movements during motor skill learning may provide an insight into internal model development in this population.

The findings from the kinematic data were less clear. Significant interaction effects in the kinematic variables, corresponding to those seen in the performance and eye movement data, were predicted. No significant interactions were present. In fact, no differences were found in the total path length between groups, indicating that participants used similar path lengths to hit the targets. However, on inspection of the examples of movement strategies used between groups (Figure 4), a number of qualitative differences are evident. First, both groups initially used a strategy almost exclusively based on vertical and horizontal cursor movements. These movements are typical of an early ‘exploratory’ stage of learning in visuomotor adaptation tasks (Sailer et al., 2005) and are thought to represent individuals freezing degrees of freedom in order to simplify the movement problem. The AO+MI intervention facilitated participants to change this strategy to a more...
optimal one (which more than halved their task completion time at T1), whereas the children in the control group seemed to persist with this inefficient strategy almost until the post-test phase. This persistence with an ineffective strategy is typical of children with DCD (Biotteau et al., 2016).

In terms of movement smoothness, the AO+MI group were predicted to exhibit significant reductions in jerk after the intervention. This would indicate a better developed internal model, more effective movement planning and, consequently, more smoothly controlled actions. Although the differences elicited by the AO+MI intervention failed to produce a significant interaction, it is clear that the intervention had different, albeit not significant, effects on each intervention group (Figure 3d). This was somewhat reflected in the significant main effect for group that suggested that the AO+MI group participants had significantly less jerk compared to control group participants. Although no group differences were present at pre-test, it is clear that the AO+MI group experienced an increase in the smoothness of their movement (i.e., decreased jerk) throughout the training and post-test compared to the control group. Based on this, and our findings from the performance and eye movement data, it is possible that the AO+MI intervention facilitated more effective movement planning and smoother cursor movement owing to a more substantially developed internal model.

The absence of the expected after-effects may undermine our conclusion that AO+MI facilitated the development of an internal model. In fact, both groups exhibited less jerk when the rotation was taken away – probably reflecting an overall learning effect or acclimatisation to the equipment. The lack of the expected after-effects is, however, consistent with the results of other studies that have also found no after-effects despite successful visuomotor adaptation (e.g., Ong & Hodges, 2010; Lei et al., 2016). In studies of children with DCD, both Kagerer et al. (2004) and King et al. (2011) also reported no significant after-effects when using a similar visuomotor rotation task. In fact, to date only one study has shown some evidence of significant after-effects in children with DCD during visuomotor rotation adaptation (Kagerer et al., 2006). While the presence of after-effects is considered evidence for the formation of an internal model, it is uncertain whether the absence of after-effects necessarily means that no internal model was actually developed. For example, previous visuomotor adaptation studies have suggested that the internal
model can be updated even in the absence of after-effects (Wang & Lei, 2015). Based on our after-effects data, the extent to which AO+MI facilitated the development of the internal model is unclear. However, when considering the direct-effects data (i.e., performance, eye-movements, and kinematics) it is reasonable to suggest that the direct-effects observed in the current experiment provide preliminary support for the formation and ongoing updating of an internal model in children with DCD.

Some limitations of this experiment need to be considered prior to endorsing AO+MI as an effective intervention. First, the sample used in the experiment was relatively small. Previous motor imagery studies conducted by Wilson et al. (2002; 2016) employed group sizes of 18 and 12 participants respectively. However, it is important to note that Wilson et al. (2002) included participants who scored at or below the 50th percentile on the MABC test with only 11 children below the 15th percentile, whilst their replication study used the criteria of the 10th percentile (Wilson et al., 2016). In the present study, only children who scored at or below the 5th percentile on the MABC-2 test were included in data analysis. The more stringent inclusion criterion in this study was selected in order to provide a more representative sample of the DCD population as it is these individuals who benefit most from mental simulation interventions (Wilson et al., 2016). However, due to heterogeneous nature of DCD and the high movement variability associated with the condition, it is possible that this small sample size had a negative influence on the quality of the kinematic data. It is therefore clear that further studies are needed with larger samples sizes before the efficacy of AO+MI interventions for the DCD population can be established. Second, the task used was a 2D computer-based task and it is evident that the beneficial performance effects seen here may not transfer to more complex tasks like those required for activities of daily living. Finally, this study did not have a delayed retention test and, therefore, a more thorough examination of the long-term effects of this intervention is required in order to examine AO+MI-induced motor skill consolidation over a longer period.

Despite these limitations, this research offers several theoretical and practical implications that could facilitate future research. Theoretically, these findings offer some support for the IMD hypothesis and extend existing literature by showing, for the first time, that AO+MI can be used to alleviate deficits in the development of internal models in children with DCD. These results show that the AO+MI group successfully integrated visual-
spatial information from the AO+MI training into their own physical practice and this process facilitated the rate of their adaptation. The action observation component may have allowed participants to map visual information onto motor circuits in order to enhance motor performance (Apšvalka et al., 2018) and helped to develop basic action concepts related to the timing and sequencing of cursor movement (Wright et al., 2018). The kinaesthetic imagery component has been shown to update the proprioceptive components of the internal model that subsequently improve movement planning and control (Kilteni et al., 2018). The development of more elaborate proprioceptive control is indicative of more expert-like motor control that allows vision to be allocated as a feed-forward resource to guide action ahead of time (Sailer et al., 2005), thereby improving performance. Taken together, it is plausible that combining two mental simulation techniques during AO+MI provided a beneficial effect for the formulation and development of internal models of movement control. Without such training, the control group adapted to the visuomotor rotation significantly more slowly, had a less target focused eye-movement strategy, and less effective movement kinematics.

Additionally, DCD is often characterised as a motor learning disorder despite much evidence to the contrary (see Biotteau et al., 2016 for a review). Whilst motor learning for children with DCD is slower than for typically developing children, the present study again demonstrates that while children with DCD may struggle with formulating effective movement strategies themselves, they are well equipped to incorporate or mimic (e.g., Scott et al., 2019; Slowinski et al., 2019) strategies once they are exposed to them. Although our data suggest that AO+MI may be a suitable intervention for this purpose, further examination of the potential neural mechanisms underpinning these effects is needed in future research (Zwicker et al., 2010), and an examination of the additive effects of each action observation and motor imagery component would be important for the design.

From a practical perspective, AO+MI interventions appear to offer a suitable adjunct to the physical practice of motor skills for children with DCD. Consequently, AO+MI may be a suitable technique for parent-led interventions that can be performed at home using digital technologies. Previous research with clinical populations has evidenced the benefits of such an approach for learning activities of daily living (Bek et al., 2018), and parental
involvement has been highlighted as a key factor in ensuring the success of the interventions for DCD (Morgan & Long, 2012). Future research should therefore explore the feasibility of this approach for children with DCD. Finally, while it is difficult to isolate the contribution of the individual action observation or motor imagery components, combining these techniques into a single intervention may be of particular practical benefit to children with this condition. As individuals with DCD exhibit poor motor visual imagery ability, combining action observation with kinaesthetic imagery may be an effective intervention that provides accurate visual and temporal movement cues while enabling the limited cognitive resources synonymous with the condition (Alloway, 2011) to be devoted to the generation of kinaesthetic imagery associated with the observed movement (Eaves et al., 2016).

In conclusion, these results support the IMD hypothesis as a possible explanation for the coordination impairments associated with DCD and suggest that AO+MI interventions may help children with DCD to overcome such difficulties. Future research with individuals with DCD should examine the efficacy of AO+MI interventions for more complex movements (e.g., sports skills), and for improving functional movements required for activities of daily living.

Declaration of Competing Interest

The authors declare no conflict of interest

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https://doi.org/10.1016/j.neulet.2005.10.009


