1   **Title:** Motor resonance is modulated by an object's weight distribution

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Abstract

Transcranial magnetic stimulation (TMS) studies showed that corticospinal excitability (CSE) is modulated during observation of object lifting, an effect termed 'motor resonance'. Specifically, motor resonance is driven by movement features indicating object weight, such as object size or observed movement kinematics. We investigated in 16 humans (8 females) whether motor resonance is also modulated by an object's weight distribution. Participants were asked to lift an inverted T-shaped manipulandum with interchangeable center of mass after first observing an actor lift the same manipulandum. Participants and actor were instructed to minimize object roll and rely on constrained digit positioning during lifting. Constrained positioning was either collinear (i.e., fingertips on the same height) or noncollinear (i.e., fingertip on the heavy side higher than the one on the light side). The center of mass changed unpredictably before the actor's lifts and participants were explained that their weight distribution always matched the actor's one. Last, TMS was applied during both lift observation and planning of lift actions. Our results showed that CSE was similarly modulated during lift observation and planning: when participants observed or planned lifts in which the weight distribution was asymmetrically right-sided, CSE recorded from the thumb muscles was significantly increased compared to when the weight distribution was left-sided. During both lift observation and planning, this increase seemed to be primarily driven by the weight distribution and not specifically by the (observed) digit positioning or muscle contraction. In conclusion, our results indicate that complex intrinsic object properties such as weight distributions can modulate activation of the motor system during both observation and planning of lifting actions.

Highlights

- Motor resonance is observation-induced activity in the observer's motor system
- We used a dyadic lifting task of objects with asymmetrical weight distribution
- We investigated which movement features modulate motor resonance
- Motor resonance is modulated by the object's weight distribution
- Motor resonance is driven by observed and planned digit positioning

Keywords: Action observation; object lifting; dyadic interaction; motor planning; motor resonance

Introduction

Skilled object manipulation not only relies on tactile feedback but also on anticipatory mechanisms (Johansson & Westling, 1988). However, predictive lifting errors are made when object weight is wrongly estimated (e.g., lifting an opaque box with unexpected amount of filling). In this situation, individuals update their 'sensorimotor memory' which contains short-term associations between previous hand-object experiences and the visual object properties (Baugh et al., 2012). As such, the sensorimotor memory can be flexibly updated and subsequently used for predicting object weight and planning skilled hand-object interactions.

Importantly, skilled hand-object interactions not only require accurate planning for object weight but also for weight distribution. For instance, when having to avoid content spill, object roll has to be minimized by generating appropriate compensatory torque to offset external torque induced by an unbalanced weight distribution. Lukos et al. (2007) showed that individuals can update their sensorimotor memory for an object's weight distribution, in turn enabling them to predictively generate appropriate compensatory torque. In addition, Fu et al. (2010) showed that individuals scale their fingertip forces in function of their digit positioning: when digit positioning is constrained, individuals scale their fingertip forces in function of these fixed contact points. Conversely, when digit positioning is unconstrained, individuals scale their fingertip forces accurately in function of their self-chosen contact points. However, it has been argued that, in line with these findings, object lifting with unconstrained digit positioning relies on both predictive and feedback driven mechanisms: although fingertips are positioned based on the initial motor command, trial-to-trial variability in actual positioning is induced by contextual and executional noise. As a result, the initially planned fingertip forces need to be updated in function of tactile feedback about the actual positioning (Mojtahedi et al., 2015). To end, it is interesting to note that Lukos et al. (2013) showed that when individuals plan to lift an object with unpredictable weight distribution, they rather rely on a motor command based on their previous lift (i.e., sensorimotor memory) than on a generic or 'neutral' motor command. Combined, these studies show that individuals can plan dexterous manipulation of objects with complex properties but that they do rely on sensorimotor integration based on their previous trial (Lukos et al., 2013) and on real-time haptic feedback during lift execution (Mojtahedi et al., 2015).

Performing hand-object interactions modulates activity within the motor system during both motor execution and planning (for a review see Hannah, 2020). That is, corticospinal excitability (CSE) probed with transcranial magnetic stimulation (TMS) over the primary motor cortex (M1) is modulated during motor tasks. Loh et al. (2010) showed that when individuals plan to lift an object, CSE initially

reflects the weight of the previously lifted object. However, when a visual cue indicates that the object

weight has changed, CSE modulation is altered and becomes representative of this new weight (Loh et

al., 2010). In Davare et al. (2019), participants were instructed to grasp and lift an asymmetrical weight

distribution either with constrained or unconstrained digit positioning. They showed that CSE was

increased when motor predictability was lower (i.e., unconstrained positioning). Noteworthy, these

effects were only present after object contact but not during reaching. As such, their findings suggest

that CSE modulation during early contact does not only reflect sensorimotor integration of the previous

trial but also on-line feedback about digit placement.

Parikh et al. (2014) showed that when individuals plan to grasp (but not lift) an object, CSE is

decreased when planning to exert high compared to low force. In addition, the authors argued that,

considering these effects were not altered by paired-pulse TMS, that CSE modulation appears to be

driven by inputs from regions outside M1. To investigate the role of the somatosensory cortex (S1) in

predictive lift planning, Parikh et al. (2020) asked participants to lift an asymmetrical weight distribution

after virtually disrupting either M1 or S1 with repetitive TMS. They found that when digit positioning is

constrained, force planning relies on memory retrieval in M1. In addition, when digit positioning is

unconstrained, grasp planning relies on memory retrieval of digit positioning in M1 and on integrating

haptic feedback regarding digit positioning to generate appropriate load forces in S1. Taken together,

these studies show that M1 and S1 are involved in sensorimotor integration during the planning and

execution of hand-object interactions (Loh et al., 2010; Parikh et al., 2014) and, in particular, also on

objects with an asymmetrical weight distribution (Davare et al., 2019; Parikh et al., 2020).

Although execution of hand-object interactions is pivotal in rapidly updating the sensorimotor

memory, other studies have shown that humans are able to generate similar representations during

observation of object lifting as well. For instance, Meulenbroek et al. (2007) demonstrated that when

two individuals incorrectly predict an object's weight, the second individual will make a smaller lifting

error after observing the first individual making one. These findings have been supported by other

studies (Buckingham et al., 2014; Reichelt et al., 2013). In addition, it has been shown that observation of

skilled lifting can improve predictive object lifting as well, albeit in a smaller manner than observing

lifting errors (Rens et al., 2020, 2021; Rens & Davare, 2019)

Akin to motor execution, observing hand-object interactions not only alters sensorimotor

representations but also activates the observer's motor system (for a review see Naish et al. 2014).

Fadiga et al. (1995) were the first to demonstrate that CSE is similarly modulated during the execution

and observation of hand actions. They argued that the motor system is potentially involved in action

understanding through a bottom-up mapping ('mirroring') of observed actions onto the same cortical

areas involved in their execution (for a review see: Rizzolatti et al., 2014). Consequently, action

observation-driven modulation of CSE has been termed 'motor resonance'. With specific interest to

observation of object lifting, Alaerts et al. (2009, 2010a, 2010b) demonstrated that motor resonance is

modulated by observed movement features indicating object weight, such as intrinsic object properties

(e.g., size), muscle contractions and movement kinematics. Specifically, CSE is increased when observing

lifts of heavy compared to light objects. Critically, other studies have shown that these motor resonance

effects are not robust. For instance, Buckingham et al. (2014) demonstrated, using the size-weight

illusion, that CSE modulation is driven by object size when observing skilled but not erroneous lifts. In

addition, Tidoni et al. (2013) showed that motor resonance is differently modulated when observing an

actor with truthful or deceptive intentions. Last, Rens et al. (2020) demonstrated that motor resonance

is easily biased by differences within the contextual setting even though the observed lifting actions are

the same. In addition, they showed that this bias was generated by top-down inputs from the posterior

temporal sulcus.

Although Alaerts et al. (2010) and Buckingham et al. (2014) highlight that motor resonance can

reflect simple object properties, it is unknown whether it can reflect more complex ones such as an

object's weight distribution. Due to the importance of accurately estimating an object's weight (i.e.,

avoid damage) and weight distribution (i.e., avoid content spill), it is plausible that they rely on similar

mechanisms for estimating these properties. However, as minimizing object roll requires a valid digit

position-force coordination pattern, the observer's motor system should integrate both observed digit

positioning and forces (rather than solely encoding force) for accurately encoding the weight

distribution. As such, in this study we wanted to investigate whether the observer's motor system

encodes (a) observed force scaling, (b) observed digit positioning or (c) the weight distribution, as

indicated by a combination of these features. We asked participants to grasp and lift an inverted T-

shaped manipulandum with interchangeable (left, middle or right) center of mass after first observing an

actor lift the same manipulandum. Participants and actor were required to minimize object roll during

lifting by generating appropriate compensatory torque. The center of mass changed unpredictably

before the actor trials, but participants were informed that they would always lift the same weight

distribution as the actor. As such, participants could potentially estimate the object's center of mass

during lift observation and use this information to predictively plan their own lifts. For lifting the

asymmetrical weight distributions, we constrained digit positioning to two distinct possibilities: (1)

Placing the fingertips on the same height ('collinear positioning') or placing the finger on the heavy side

higher than the one on the light side ('noncollinear positioning'). Importantly, when compensatory

torque is generated for an asymmetrical weight distribution, the fingertip on the heavy side generates

more force when using a collinear positioning compared to a noncollinear one (Fu et al., 2010). By

relying on these constrained digit positionings and the associated force requirements for skilled lifting,

we hypothesized that we could disentangle whether the motor system encodes force or digit positioning

during lift observation. For doing so, we used TMS to probe CSE during lift observation and planning.

In line with Alaerts et al. (2010), we hypothesized that, if the observer's motor system encodes

force exertion by the fingertips, motor resonance should be significantly increased when observing

skilled lifts on the asymmetrical weight distributions with collinear compared to noncollinear positioning.

For instance, when the index finger is on the heavy side, it generates more force in the collinear

condition than in the noncollinear one. As such, CSE recorded from the index finger should be larger in

the collinear compared to noncollinear condition. In contrast, if motor resonance is primarily modulated

by digit positioning, we would expect that motor resonance, when observing lifts on the asymmetrical

weight distribution, would be increased when observing lifts with noncollinear compared to collinear

positioning.

Methods

Participants

16 individuals participated in the present study (8 females; mean age = 24 ± 3 years). The Edinburgh

Handedness Questionnaire (Oldfield, 1971) revealed that all participants were strongly right-handed (>

90). Prior to participation, participants were required to fill in a TMS safety screen questionnaire based

on Rossi et al. (2011). Moreover, all participants had normal or corrected-to-normal vision, were free of

neurological disorders and had no motor impairments of the right upper limb. Participants gave written

informed consent and were financially compensated for their time. The protocol was in accordance with

the Declaration of Helsinki and was approved by the local ethical committee of KU Leuven, Belgium.

177   ----------

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179   ----------

Data acquisition

For the present study, we used the same custom-built carbon fiber 'inverted T-shape' grip-lift

manipulandum (Arsalis, Belgium; for all object dimensions see: Figure 1) as in one of our previous studies

(Rens et al., 2021). The manipulandum consisted of a horizontal basis and a vertical block to which two

3D force/torque (F/T) sensors were attached. A cover plate (height × width: 140 × 53 mm) with a central

protruding surface (height × width: 140 × 20 mm) was attached to each F/T sensor to block view on the

sensors. On each protruding surface, we attached two pieces of fine-grained sandpaper (p600) (height x

width: 20 x 20 mm). Both actor and participants were only allowed to place their fingertips on these

constrained locations. The distance between the bottom side of the upper piece and the upper side of

the lower piece of sandpaper on the same cover plate was 20 mm. As a result, the vertical distance

between the center points of the same-sided pieces was 40 mm. This vertical distance was based on a

preliminary investigation which showed that when lifting the asymmetrical weight distribution and

keeping the basis horizontal, the load force difference between the two fingertips was zero if the

fingertips were placed non-collinearly and distanced 40 mm from each other. The manipulandum's

horizontal basis was divided into three compartments enabling the placement of 3D-printed cuboids that

were visually identical (height x width x depth: 55 x 35 x 40 mm). One cuboid was filled with lead

particles and weighted 4.24 N, the other two were hollow and weighted 0.24 N each. Combined with the

manipulandum, the total weight amounted to 8.67 N. The external torque (i.e., torque induced by the

object's weight distribution) could be changed by inserting the heavy cuboid in the left, center or right

compartment and amounted to -245, 0 or + 245 Nmm, respectively (for the calculation of these values

see Rens et al., 2021).

For collecting lifting-related parameters, we used two ATI mini-40 SI-40-2 F/T sensors (force

range: 40, 40 and 120 N for x-, y- and z-axes respectively; force resolution: 0.01 N; torque range: 2 Nmm;

torque resolution: 0.0005 Nmm) (ATI Industrial Automation, USA). F/T sensors were calibrated by the

developer in accordance with the applicable QTI procedures. The maximum amount of error for the

force and torque components were 1.50 % and 1.75 % respectively. Both F/T sensors were connected to

a NI-USB 6221 OEM board (National Instruments, USA) which was connected to a personal computer.

Data was acquired using a custom-written MATLAB script (Mathworks, USA) and sampled at 1 kHz.

TMS procedure and EMG recording

General procedure. Electromyography (EMG) recordings were performed using Ag-AgCl electrodes which

were placed in a typical belly-tendon montage over the right first dorsal interosseous muscle (FDI) and

abductor pollicis brevis (APB). A ground electrode was placed over the processus styloideus ulnae.

Electrodes were connected to a NL824 AC pre-amplifier (Digitimer, USA) and a NL820A isolation amplifier

(Digitimer, USA) which in its turn was connected to a micro140-3 CED (Cambridge Electronic Design

Limited, England). EMG recordings were amplified with a gain of 1000, high-pass filtered with a

frequency of 3 Hz, sampled at 3000 Hz using Signal software (Cambridge Electronic Design Limited,

England) and stored for offline analysis. For TMS stimulation, we used a (figure-of-eight; 70 mm)

DuoMAG 70BF coil connected to a DuoMAG XT-100 system (DEYMED Diagnostic, Czech Republic).

For M1 stimulation, the coil was tangentially placed over the head to induce a posterior-anterior

current flow and to elicit motor evoked potentials (MEPs) in both right FDI and APB. The optimal

stimulation site (i.e., 'hotspot') was defined as the position from which MEPs with maximal amplitude

were systematically recorded in both muscles. For finding the hotspot, we initially turned the stimulation

intensity to 40% of the maximum stimulator intensity and increased the intensity step-wise while

searching. The hotspot was marked on top of the scalp. Stimulation intensity (1 mV threshold) for each

participant was defined as the lowest stimulation intensity that produced MEPs greater than 1 mV in

both muscles and in at least four out of eight consecutive trials when stimulating at the predetermined

hotspot (average stimulation intensity = 54 ± 5.6 % of maximum stimulator output). We assessed

baseline (i.e., resting state) CSE before and after the experimental task. For this, participants received 12

TMS pulses at the previously defined stimulation intensity. During baseline assessment, participants

were instructed to stay relaxed and keep their eyes open.

During the experiment, TMS was applied during both the actor (observation) and participant

trials (execution). During actor trials, TMS was applied 300 ms after the actor lifted the object from the

table (for definition of lift-off see: 'Data analysis'). This timing was based on earlier work from our group

(Rens et al., 2020) which showed clear motor resonance effects 300 ms after observed lift-off. In

addition, similar studies have also applied TMS during the observed lifting phase to assess motor

resonance effects during observation of object lifting (Alaerts, et al., 2010; Cretu et al., 2019; Senot et al.,

2011). During participant trials, TMS was applied 400 ± 100 ms (jitter) after object presentation. As

participants were instructed to only start reaching for the object after TMS was applied, the TMS

stimulation was actually applied during lift planning. For participant trials (i.e., action execution) we

decided to stimulate during planning, not execution, as we did not want to interfere with the

participants' lifting performance which would be caused by the unvoluntary muscle contraction due to

M1 stimulation. TMS timing during participant planning was based on Loh et al. (2010). In their study,

participants were instructed to lift a manipulandum with interchangeable. When a visual cue indicated

the veridical object weight, CSE was differently modulated but only when TMS was applied 150 ms after

object presentation. Considering that we used an object with a more complex property (i.e.,

interchangeable weight distribution), we decided to double the latency after which this TMS effect was

present (i.e., 150 to 300 ms) to provide participants with enough time for planning. To ensure that

participants could not anticipate the TMS timing, we decided to include a jitter of 100 ms We decided to

use 300 ms as the lower threshold (thus 400 ms ± 100 ms jitter). To end, although TMS during lift

planning was applied substantially later in our study than in the one of Loh et al. (2010), it should still

elicit CSE modulation; Parikh et al. (2014) used a similar behavioral task and applied TMS at the 'go cue'

which was given 1000 ms after the 'task cue'. Briefly, in their study, CSE was also task-specifically

modulated. As such, their findings indicate that TMS at the 'go cue' probes planning effects on CSE

modulation, irrespective of the timing of the cue itself.

<u>Experimental set-up</u>

Dyadic set-up. As shown in Figure 1C, participant and actor were comfortably seated at a square table

with their lower arm resting on the table. The actor was seated on the left side of the participants so that

the participant and actor were angled 90 degrees towards each other. The manipulandum was

positioned between both individuals so both individuals could comfortably grasp and lift it. When

grasping, both individuals were required to reach with their entire right upper limb causing their elbow

to lift from the table. The manipulandum was distanced approximately 30 cm from each individual. In

addition, participant and actor were asked to place their hand on a predetermined location in front of

them to ensure consistent reaching throughout the experiment. It is important to note that the

manipulandum was positioned as depicted in Figure 1 (i.e., targeted towards the participant). The

manipulandum was rotated slightly (< 45 degrees) based on the participant's preferences and to improve

lifting comfort. Importantly, when the actor lifted the manipulandum it was positioned in this manner as

well. When the actor reached for the manipulandum, their arm would move in front of the participant's

upper body (Figure 1C). Although this orientation likely occluded visual information about the left side of

the manipulandum, we opted for this positioning rather than opposing both individuals for two reasons.

First, Mojtahedi et al. (2017) showed that, when lifting an object together, individuals perform better

when seated next to each other compared to opposed to each other. Second, Alaerts et al. (2009)

showed that modulation of motor resonance is increased when observing actions from a first person

point of view compared to a third person one. For our study, we argued that this side-by-side

configuration would also enhance motor resonance effects in the participants. During the experiment, a

switchable screen (MagicGlass) was placed in front of the participant's face which was transparent

during trials and returned opaque during inter-trial intervals. This screen blocked vision on the

manipulandum when the experimenter would switch the cuboids between compartments, thus making

participants blind to the weight distribution change. Last, one trial consisted of one lifting movement

performed by either the actor or participant, thus being 'participant trials' or 'actor trials'. Trial duration

was 4 seconds and trial onset was indicated by the switchable screen turning transparent. Trial duration

was based on preliminary testing and showed that individuals had sufficient time to reach, grasp, lift and

return the object smoothly at a natural pace. Inter-trial interval was approximately 5 seconds during

which the screen was opaque and the center of mass could be changed. To end, one (female) master's

student performed as the actor for all participants.

<u>Experimental procedure</u>

General procedure. At the start of the session, participants gave written informed consent and were

prepared for TMS (see 'TMS procedure and EMG recording'). Afterwards, the experimenter explained

the experimental task to the participants and gave the following instructions regarding the object lifting

task: (1) lift the inverted T-shape to a height of approximately 5 cm at a smooth pace that is natural to

you. (2) Only use thumb and index finger of the right hand and only place them on the sandpaper pieces.

(3) You are required to use the same digit positioning the actor used in their preceding trial. (4) Keep the

inverted T-shape's base as horizontal as possible during lifting (i.e., 'try to minimize object roll'). (5) The

center of mass in your trials always matches the one in the actor's preceding trial. In sum, the

experimenter explained to the participants that they should try to minimize object roll during object

lifting and that they could potentially rely on the observed lifting performance of the actor to plan their

own lifts. Importantly, participants were instructed to always use the same digit positioning as the actor.

This was done to ensure that we would have enough trials per experimental condition (see below).

After task instructions, participants were allowed to perform 3 practice lifts on the symmetrical

weight distribution and 6 on each asymmetrical weight distribution (left or right). On half of the lifts with

the asymmetrical weight distribution, participants were required to position their fingertips on the same

height, i.e., 'collinear positioning'. In the other half, they were required to place their fingertips

'noncollinearly', i.e., the fingertip on the heavy side was positioned higher than the fingertip on the light

side (left asymmetrical: right thumb higher than right index; right asymmetrical: right thumb lower than

right index). When lifting the symmetrical weight distribution, no compensatory torque should be

generated. Accordingly, for this weight distribution participants were required to always place their

fingertips collinearly as noncollinear positioning would cause participants to automatically generate

compensatory torque. These practice trials were aimed to familiarize participants with the

manipulandum.

Constrained digit positioning. Fu et al. (2010) showed that appropriate compensatory torque can

be generated by many valid digit position-force coordination patterns. As motor resonance is driven by

movement features such as fingertip forces, we wanted to limit the digit positioning options in order to

reduce potential variability in motor resonance effects. We argued that constrained digit positioning and

limited experimental conditions would enable us to find more reproducible results. As mentioned

before, noncollinear positioning was defined based on the vertical difference between fingertips which

caused the load force difference to approximate zero. Collinear positioning was included as we

considered it a 'neutral' condition to contrast the noncollinear one. Because the fingertip on the heavy

side has to generate more force when positioned collinearly compared to noncollinearly, we argued that

these two conditions would allow us to disentangle whether the observer's motor system would encode

observed positioning or observed forces. When using collinear digit positioning, irrespective of the

weight distribution, actor and participant were required to place their fingertips on the upper sandpaper

pieces on each side (Figure 1B). When using noncollinear digit positioning for the asymmetrical weight

distribution, the finger on the heavy side was placed on the upper piece whereas the finger on the light

side was placed on the lower piece. For instance, when the heavy cuboid was inserted in the left

compartment, the thumb and index finger were placed on the upper and lower piece of their respective

side (Figure 1). The actor was instructed to use these digit positioning strategies which ensured that

participants would use the same digit positionings. We opted for these digit positionings (primarily

collinear positioning on the upper pieces instead of the lower ones) to protect the EMG electrodes over

the thumb muscle from rubbing over the manipulandum's base.

Experimental task. After task instructions, participants performed the object lifting task with the

actor. The actor would change the inverted T-shape's center of mass and verbally declare that she would

execute the next trial. Verbal declaration took place before switching of the cuboids. After trial

completion by the actor, participant performed 3 back-to-back trials with the same weight distribution

and using the same digit positioning as the actor. We decided to have participants perform 3 repetitions

based on Fu et al. (2010). To ensure that participants could not rely on sound cues potentially indicating

the new weight distribution, the experimenter always removed and replaced all 3 cubes after randomly

rotating the inverted T-shape prior to each actor trial. These actions were never done before participant

trials as they were explained that the center of mass in their trials would always match the one of the

actor's preceding trial.

During the object lifting task, the experimenter and participants performed 20 transitions from

the middle compartment (i.e., symmetrical weight distribution) to each side (i.e., asymmetrical weight

distributions). When the experimenter lifted the new asymmetrical weight distribution (e.g., left-sided),

she would use collinear positioning in 10 lifts and noncolinear positioning in the other 10. Trial amount

per experimental condition was based on Senot et al. (2011) who used 10 trials per TMS condition when

investigating motor resonance effects during lift observation. After 4 trials were performed on the

asymmetrical weight distribution (i.e., 1 actor and 3 participant trials), the experimenter would change

the object's weight distribution again. Most of these transitions were 'normal' and consisted of the actor

changing the heavy cuboid back to the middle compartment for washing out the internal representation

for the asymmetrical weight distribution. However, 10 transitions were 'catch transitions' in which the

asymmetrical weight distribution was first changed to the other side (e.g., left to right asymmetrical) and

only then to the middle compartment to wash out the internal representation for asymmetrical. We

included these catch transitions to ensure that participants would not anticipate the typical change from

asymmetrical to symmetrical even though we still wanted to wash out the internal representation for

asymmetrical. Catch transitions were inserted equally after each lifting sequence on each asymmetrical

weight distribution. Considering that we had a 2 (side: left or right asymmetrical) by 2 (positioning:

collinear or noncollinear) design when lifting the asymmetrical weight distribution, either 2 or 3 catch

transitions were performed after each lifting sequence (e.g., 2 catch transitions after the lifting sequence

with collinear positioning on the right asymmetrical weight distribution). The actor randomly decided

which digit positioning to use on the new asymmetrical weight distribution in the catch transitions. Last,

in both the normal and catch transition to symmetrical, the standard amount of trials (1 actor and 3

participant's trials) were performed after each weight distribution change.

The object lifting task was split over 4 experimental blocks with a short break between blocks.

For each participant, transition order was pseudo-randomized within each experimental block. As such,

each experimental block contained 5 transitions to left and right asymmetrical. In addition, either 2 or 3

of these transitions to each side were performed with one digit positioning type (e.g., collinear) and the

other 3 or 2 transitions with the other digit positioning type (e.g., noncollinear). Transitions to a specific

side (e.g., symmetrical to left-sided asymmetrical) would repeat maximally 2 times back-to-back and

catch transitions were spread equally over all blocks. The full experimental session lasted approximately

2 h.

<u>Data analysis</u>

Behavioral data. Data collected with the F/T sensors were sampled in 3 dimensions at 1000 Hz and

smoothed using a fifth-order Butterworth low-pass filter (cut-off frequency: 15 Hz). For each sensor, grip

force (GF) and load force (LF) were defined as the exerted force perpendicular to the normal force (Y-

direction on Figure 1) and the exerted force parallel to the normal force (X-direction on Figure 1),

respectively. Digit positioning was defined as the vertical coordinate (X-direction on Figure 1) of the

fingertip's center of pressure on each cover plate. The center of pressure was calculated from the force

and torque components measured from the respective F/T sensor relative to its frame of reference,

using formula 1.

$$COP = \frac{(T_y - F_x \delta)}{F_z} \qquad (1)$$

In formula 1, COP = center of pressure, T_y = Torque in the Y-direction, F_x = Force in the X-direction, F_z =

Force in the Z-direction, δ = cover plate thickness (1.55 mm). Compensatory torque was defined as the

net torque generated by an individual to offset the external torque caused by the object's weight

distribution and was calculated with formula 2 (we refer the reader to the supplementary materials of Fu

et al., 2010 for the detailed explanation of the formula).

$$T_{comp} = \frac{d}{2} \times (LF_{thumb} - LF_{index}) + (COP_{thumb} - COP_{index}) \times GF_{average} \qquad (2)$$

In formula 2, T_{comp} = Compensatory torque, d = horizontal distance between the digits (48 mm; Figure 1;

Y-direction), $LF_{thumb/index}$ = Load force generate by the thumb and index finger, respectively, $COP_{thumb/index}$ =

center of pressure of the thumb and index finger, respectively, $GF_{average}$ = averaged amount of GF exerted

by the thumb and index finger.

To investigate the effects of lift observation on the performance of the participants we used the

following variables: Digit positioning difference, defined as the difference between the COP of the thumb

and the index finger (positive values indicate a thumb placement higher than that of the index finger),

compensatory torque, total grip force, and load force difference, defined as the difference between load

forces generated by the thumb and index finger (positive values indicate the thumb generating more

load force than the index finger). We included difference in digit positioning to investigate whether actor

and participants placed their fingertips appropriately. However, it is important to note that during data

processing we excluded trials (both force and TMS data) in which fingertips were not placed on the

sandpaper pieces (< 1% of all trials). We considered compensatory torque as our key indicator of

performance as it results from the combination of grip and load forces as well as digit positioning and

because we explicitly asked participants to minimize object roll during lifting ('task goal'). Moreover, Fu

et al. (2010) showed a strong linear correlation between compensatory torque and peak object roll,

arguing its validity as our key indicator of performance. We included total grip force and load force

difference to explore their potential effect on CSE modulation during observation and planning. Last, it is

important to note that, for analysis purposes, we inversed the sign for compensatory torque for the right

asymmetrical weight distribution, which allows for better comparisons regarding performance between

the left and right side. We argued not to do so for load force and digit position differences for

interpretability with respect to motor resonance effects.

In line with Fu et al. (2010), we extracted digit positioning difference at early object contact,

which we defined as total GF > 1 N. Considering that we had no proxy of lift onset (again see Fu et al.

2010), we decided to approximate lift onset by using object lift-off instead. In line with our previous work

(Rens et al., 2020, 2021; Rens & Davare, 2019), we defined lift-off as the time point where total load

force > 0.98 x object weight. As such, we extracted total grip force, load force difference and

compensatory torque at object lift-off.

EMG data. From the EMG recordings, we extracted the peak-to-peak amplitude of the MEP using

a custom-written MATLAB script. All EMG recordings were visually inspected and afterwards analyzed in

the script. Trials were excluded when the MEP was visibly contaminated by noise (i.e., spikes in

background EMG) or when an automatic analysis found that background EMG was larger than 50 µV

(root-mean-square error) in a time window of 200 ms prior to the TMS stimulation. In line with previous

work of our group (Rens et al., 2020) we excluded outliers for each participant separately. Outliers were

defined as values exceeding the mean ± 3 SD's. The total amount of removed MEPs was 1.68 %.

For each participant, all MEPs collected during the experimental task were z-score normalized

for observation and planning separately. Last, we also assessed pre-stimulation ('background') EMG by

calculating the root-mean-square across a 100ms interval ending 50ms prior to TMS stimulation. To end,

we did not normalize baseline CSE measurements to allow for direct comparisons between muscles.

<u>Statistical analyses</u>

For statistical purposes, we did not include catch trials (center of mass change from side to side) and

trials in which the weight distribution changed from asymmetrical to symmetrical. First, we excluded

transitions to the symmetrical weight distribution as we had less than 10 MEPs for these conditions. As

mentioned before, we had 20 transitions to each asymmetrical weight distribution, 10 in which the actor

used collinear positioning and 10 in which she used noncollinear positioning. After performing this lifting

sequence, the weight distribution normally changed to symmetrical. However, as mentioned before, we

also included 10 catch transitions in which one asymmetrical weight distribution changed to the other

(i.e., 5 changes from left to right and 5 from right to left). Due to these 10 catch transitions, we had less

than 10 MEPs for each transition to symmetrical. For instance, when a lifting sequence was performed

on the right asymmetrical weight distribution with noncollinear digit positioning only 7 or 8 normal

transitions were performed in which the weight distribution changed directly back to symmetrical. Given

the limited amount of MEP data for these changes with respect to relevant motor resonance studies

(Alaerts et al. 2010; Buckingham et al. 2014; Senot et al. 2011) and the actor choosing her digit

positioning randomly in these catch transitions, we decided to not include these trials for analysis

purposes. Note that we included lifts for the symmetrical weight distribution lifts in the analysis of the

actor's performance, to compare this against the asymmetrical distributions. Second, we excluded catch

trials for analysis purposes due to the very limited amount we had.

We investigated whether the actor's and participants' performance changed between

experimental blocks. Considering we did not find any learning effects, we collapsed our data across

experimental blocks. All statistical analyses were performed in SPSS statistics version 25 (IBM, USA) and

are described below. For each parameter of interest, we performed a separate analysis. We used linear

mixed models (LMM) considering our coding for the noncollinear condition (left noncollinear: thumb

higher than index; right noncollinear: thumb lower than index). This allowed to consider digit positioning

to be a 'nested' factor within the weight distribution for our analyses (see below) instead of a 'main'

factor as in a typical ANOVA. The reason hereof is that it has been shown that body representations have

different preferential associations between the fingers and their positioning in space (e.g., index finger

on top and thumb on the bottom) (Romano et al., 2017). As such, it is plausible that these preferential

associations modulate CSE, recorded in the FDI (index finger) and APB (thumb) muscles, differently

during lift planning and observation.

For the actor's data (behavioral only), we used the factors SIDE (mid, left and right) and

POSITION (collinear and noncollinear which were coded as described in the preceding paragraph). We

included the actor's collinear lifts on the symmetrical weight distribution (i.e., mid) to investigate

changes in lift performance. Specifically, if the actor correctly planned to lift an asymmetrical weight

distribution, her lift performance should differ significantly from that when she planned to lift the

symmetrical weight distribution (Rens and Davare 2019). We included SIDE as a main effect and nested

POSITION within SIDE (i.e., $POSITION_{SIDE}$).

For the participants' data (behavioral and MEPs), we used the factors SIDE (left and right only),

POSITION (collinear and noncollinear) and REPETITION (first, second and third lift after the center of

mass change). Please note that the factor REPETITION and all its interaction effects described below

were only included during lift planning, not observation, which we investigated separately. SIDE and

REPETITION were included as main effects. Again, POSITION was nested within SIDE. For the participants'

data we also included the interaction effect SIDE X REPETITION as well as REPETITION X POSITION$_{SIDE}$.

Here, we did not include the lifts on the symmetrical weight distribution as lift performance of

participants can be investigated by comparing lifts without tactile feedback (i.e., first lift after lift

observation) to their own lifts on the same object with tactile feedback (second and third lifts). As such,

lift performance is quantified by comparing lifts without and with tactile feedback (Fu et al. 2010;

Reichelt et al. 2013; Rens and Davare 2019). To analyze motor resonance effects during lift observation,

we used the factors SIDE (left and right) and POSITION (collinear and noncollinear). We included SIDE as

a main effect and nested POSITION within SIDE (i.e., POSITION$_{SIDE}$).

481   For both the actor and participant data, we included the intercept in the model. Moreover, for

all participants' analyses we also included the participant's number as random effect. Last, we decided to

include the mixed model covariance structures as first-order autoregressive based on the assumption

that correlation in residuals between factor levels was identical across levels. We used type III sum of

squares and Maximum Likelihood (ML) for mixed model estimation and Bonferroni corrections for

pairwise comparisons. All data in text is presented as the mean ± SEM. P-values < 0.05 are discussed as

statistically different.

488   ----------

489   Figure 2

490   ----------

Results

In the present study, we investigated how CSE is modulated when observing and planning lifts of objects

with an asymmetrical weight distribution. Participants performed an object lifting task in turns with an

actor and were required to lift a manipulandum with interchangeable center of mass as skillfully as

possible, i.e., minimize object roll by generating appropriate compensatory torque. When participants

were required to lift an object with unknown weight distribution, they first observed the actor lift the

new weight distribution. Afterwards, the participants lifted this weight distribution three times.

Importantly, participants were instructed to place their fingertips on the same constrained locations the

actor used. After participants performed their three lifts, the actor changed the weight distribution and

again lifted the new weight distribution. Accordingly, participants could potentially derive critical

information about the object's weight distribution by observing the actor's lifts. During the behavioral

task, TMS was applied during observed lifting (after observed lift-off) and during lift planning in the actor

and participant trials, respectively.

504   Traces of the included lifting parameters, when lifting an asymmetrical weight distribution, can

be found in Figure 2, showing the difference between a single lift of the actor (solid lines) with collinear

(blue) and noncollinear positioning (green). In addition, the dashed traces represent a single first lift of a

participant after observing the actor's lift. As the actor could anticipate the object's weight distribution,

she would predictively generate compensatory torque to minimize object roll (Figure 2A; blue and green

solid traces). When the participants lifted the object after observing the actor, they were able to

generate appropriate compensatory torque after observing noncollinear positioning (green dashed

trace) but not after observing collinear one (blue dashed trace). Accordingly, the difference in lift

performance can be quantified through differences in compensatory torque. As mentioned before,

appropriate compensatory torque is the resultant of a valid digit positioning – force coordination pattern

(Fu et al., 2010). For instance, it can be seen in Figure 2B that in the actor's (blue solid) and participant's

lift (blue dashed), fingertips were positioned similarly. However, as the actor but not the participant

generated appropriate compensatory torque, it can be seen in Figure 2D that the load force difference

increased faster for the actor than the participant. Conversely, when lifting with noncollinear digit

positioning (green traces), the fingertip on the heavy side is positioned higher than the fingertip on the

light side (Figure 2B). Due to this vertical height difference, the fingertip on the heavy side does not have

to generate more load force than the fingertip on the light side (Figure 2D), resulting in a load force

difference close to zero. Last, as can be seen in Figure 2C, the fingertips generate more grip force when

positioned collinearly compared to noncollinear positioning.

523   ----------

524   Figure 3

525   ----------

<u>Actor's lift performance</u>

The actor's lift performance should have been consistent due to (a) our constrained positioning

conditions and (b) her changing the weight distribution. However, we decided to analyze her

performance for verification and to quantify which observed movement features might have driven

motor resonance effects. We expected that the actor would comply with task instructions and would

always lift the manipulandum skillfully (i.e., generate appropriate compensatory torque).

532   *Compensatory torque at lift-off.* As the actor was not blinded to the weight distribution change,

she generated significantly more compensatory torque when lifting the asymmetrical weight

distributions (left = 232.56 ± 4.22 Nmm; right = 189.03 ± 3.39 Nmm) compared to the symmetrical one

(mean = 6.90 ± 1.38 Nmm; *both p < 0.001*) (SIDE: $F_{(2,80)}$ = *1135.70; p < 0.001*). Although the actor was

instructed to lift the asymmetrical weight distributions as skillfully as possible, she failed to do so when

using collinear positioning: as shown in Figure 3A, the actor generated more compensatory torque when

lifting asymmetrical weight distributions with noncollinear positioning (between adjacent bars: *each p <*

0.001) (POSITION$_{SIDE}$: $F_{(2,80)}$ = *518.184; p < 0.001*). Noteworthy, the performance when lifting

noncollinearly was similar between asymmetrical weight distributions (Figure 3A; between green bars: *p*

= 1.00). Last, the actor's lift performance was the least skillful when collinearly lifting the right

asymmetrical weight compared to all other asymmetrical conditions (*all p < 0.001*).

Digit positioning at early object contact. Although digit positioning was constrained to small

sandpaper pieces and we removed trials in which fingertips were not positioned correctly, we provide

these results for clarity. Please note that positive values for difference in digit positioning indicate that

the thumb was positioned higher than the index finger and negative values indicate that the index finger

was positioned higher than the thumb. Briefly, when the actor was instructed to use noncollinear

positioning, she appropriately placed her fingertips further apart compared to the collinear conditions

(Figure 3B; green compared to blue bars: *all p < 0.001*) and differently than the other noncollinear

condition (between green bars: *p < 0.001*) *(*POSITION$_{SIDE}$: $F_{(2,80)}$ = *2342.74, p < 0.001).* Noteworthy, digit

positioning also differed significantly between all collinear conditions (*all p < 0.001)* which indicates that

the weight distribution caused subtle differences in the collinear digit positioning. However, considering

how small the differences between collinear conditions are (Figure 3B; Mid col = 4.55 ± 0.24 mm; Left col

= 9.86 ± 0.25 mm; Right col = 0.32 ± 0.36 mm), it is unknown to which extent they were visible to the

participants.

Total grip force at lift-off. When the actor lifted the asymmetrical weight distributions

noncollinearly, she used less grip force than when lifting all weight distributions collinearly (Figure 3C;

green compared to blue bars: *all p < 0.001*) (POSITION$_{SIDE}$: $F_{(2,80)}$ = *405.68; p < 0.001*). Importantly,

differences between all conditions were significant (*all p < 0.001*) suggesting that the actor used a unique

amount of grip force in each condition. It also important to note that the actor used, on average, more

grip force when lifting the right asymmetrical weight distribution (mean = 24.12 ± 0.63 N) compared to

the left one (mean = 19.77 ± 0.52 N; *p < 0.001*) *(*SIDE: $F_{(2,80)}$ = *46.13; p < 0.001*).

Load force difference at lift-off. Please note that these values are expressed as the difference in

load force between the thumb and index finger therefore positive values indicate that the thumb

exerted more load force than the index finger and negative values indicate that the index finger exerted

more load force than the thumb. First, when the actor lifted the asymmetrical weight distributions

collinearly, she generated significantly more load force with the fingertip on the heavy side (left = 5.43 ±

0.18 N; right = -8.91 ± 0.20 N) compared to lifting the same weight distribution noncollinearly (Figure 3D;

between adjacent bars: p < 0.001) and compared to all other conditions *(all p < 0.001)* ($POSITION_{SIDE}$:

$F_{(2,80)}$ = 877.52; p < 0.001). Second, we found no evidence that the load force difference was different

when the actor lifted the asymmetrical weight distributions noncollinearly *(*Figure 3D; between green

bars: *p = 0.09).*

573   ----------

574   Figure 4

575   ----------

<u>Participants' lift performance</u>

We included the participants' lifting parameters to investigate their lift performance after lift

observation (first lift) and haptic feedback (second and third lifts) and how it might have affected CSE

modulation during lift planning. Based on our previous work (Rens et al., 2021), we expected that lift

performance would be suboptimal after lift observation. That is, participants would perform better in

their second and third lifts compared to their first ones.

Compensatory torque at lift-off. Our results revealed that there were large differences in lift

performance between digit positioning conditions. When participants used noncollinear positioning,

they overcompensated for the asymmetrical weight distribution and generated more compensatory

torque (left = 319.50 ± 5.68 Nmm; right = 321.57 ± Nmm; Figure 4A green lines) than required (245

Nmm). In contrast, when lifting collinearly, participants could not generate appropriate compensatory

torque (left = 133.35 ± 6.73 Nmm; right = 48.14 ± 4.14 Nmm; Figure 4A blue lines). Although lift

performance did not differ between the noncollinear conditions (*p = 1.00*), it did differ between

noncollinear and collinear ones *(all p < 0.001*). In addition, participants performed worse when

collinearly lifting the right asymmetrical weight distribution compared to all other conditions *(all < 0.001)*

(*$POSITION_{SIDE}$: $F_{(2,179)}$ = 1571.60; p < 0.001*). When participants lifted the asymmetrical weight

distributions noncollinearly, they did not generate more compensatory torque over repetitions (*between*

green connected scatters: all p = 1.00). This indicates that their lift performance did not differ

significantly after lift observation (first repetition: left = 323.56 ± 6.44 Nmm, right = 320.41 ± 10.26 Nmm)

and after having tactile feedback (second repetition: left = 316.34 ± 5.86 Nmm; right = 323.23 ± 9.22

Nmm; third repetition: left = 315.89 ± 7.07 Nmm; right = 321.07 ± 7.99 Nmm) (REPETITION X

$POSITION_{SIDE}$: $F_{(4,179)}$ = 4.99; p < 0.001). In contrast, when using collinear positioning, participants

performance improved over repeated lifts. This increase was significant from their first (left = 113.78 ±

8.57 Nmm, right = 20.26 ± 3.71 Nmm) to second lift (left = 139.44 ± 7.84 Nmm, right = 56.60 ± 4.82

Nmm) (*both p < 0.009*), but not significant from their second to third lift (left = 146.83 ± 5.02 Nmm, right

= 67.57 ± 6.69 Nmm) (*both p > 0.54*) (Figure 4A blue connected scatters). To end, when participants used

collinear positioning, the difference in compensatory torque was also statistically different between the

first and third lifts for each side (left: *p = 0.005;* right: *p < 0.001*).

604   *Digit positioning at early object contact.* As participants were instructed to imitate the actor's

constrained digit positioning and we excluded trials with incorrect digit positioning, we primarily report

digit positioning for transparency reasons. As shown in Figure 4B, participants did not change their digit

positioning over the multiple repetitions within each condition (REPETITON X POSITION$_{SIDE}$: *$F_{(4179)}$ = 0.75;*

p = 0.56). This indicates that they adhered to the constrained digit positionings. Logically, when lifting the

left asymmetrical weight distribution noncollinearly, participants placed their thumb higher than their

index finger (mean = 48.22 ± 0.1.06 mm) and vice versa for the right asymmetrical weight distribution

(mean = -42.45 ± 0.84 mm; *p < 0.001)*. These noncollinear positionings also differed significantly from

those when lifting collinearly *(all p < 0.001)* (POSITION$_{SIDE}$: *$F_{(2,179)}$ = 3605.93; p< 0.001*). Noteworthy,

collinear digit positioning for the left (mean = 5.80 ± 0.67 mm) and right (mean = 1.62 ± 0.54 mm)

asymmetrical weight distributions also differed significantly (*p = 0.003;* effect of POSITION$_{SIDE}$).

Although this difference is significantly different, it is unlikely that participants planned their digit

positioning differently between sides in the collinear condition as this difference is only ± 4 mm and the

constrained locations for digit positioning were only 20 mm wide. Arguably, it is more likely that this

small difference between sides is driven by skin deformation due to external torque caused by the

weight distributions (Kalra et al., 2016) or by slipping of the fingertips (Johansson & Westling, 1984).

620   *Total grip force at lift-off.* As shown by the green line plots in 4C, participants scaled their grip

forces similarly when using noncollinear positioning for both weight distributions and all repetitions *(all p*

= 1.00) (REPETITION X POSITION$_{SIDE}$: *$F_{(4,179)}$ = 5.53; p < 0.001*). In contrast, this lifting consistency was

lower when using collinear digit positioning. When participants lifted the right asymmetrical weight

distribution collinearly, participants significantly increased their grip forces from their first lift (mean =

23.95 ± 0.95 N) to their second one (mean = 27.92 ± 0.77 N; *p < 0.001*) but not from their second to third

one (mean = 28.74 ± 0.74 N; *p = 1.00*). The difference between the first and third lift was also statistically

different (*p < 0.001*). When lifting the left asymmetrical weight distribution collinearly, participants did

not significantly increase their grip force from their first (mean = 24.10 ±0.98 N) to second lift (mean =

26.40 ± 0.98 N; *p = 0.51*) and also not from their second to third lift (mean = 27.45 ± 0.86 N; *p = 0.86*).

However, the amount of grip force participants generated in this condition was statistically different

between the first and third lift ($p = 0.05$). By and large, these findings indicate that when participants

lifted the asymmetrical weight distributions collinearly, they increased their grip force over the three

repetitions. To end, as shown in Figure 4C, the amount of grip forces applied when lifting collinearly or

noncollinearly differed significantly for both weight distributions and all repetitions *(all p < 0.001).*

635    *Load force difference at lift-off.* Based on our experimental set-up and constrained positioning,

we expected that noncollinear positioning would lead to a load force difference close to zero.

Conversely, collinear positioning would lead to the fingertip on the heavy side scale its load force larger

than the finger on the light side. As shown in Figure 4D, when participants used noncollinear positioning,

the load force difference was significantly different for the left (mean = -1.65 ± 0.24 N) and right (mean =

2.82 ± 0.24 N; $p < 0.001$) asymmetrical weight distributions (POSITION$_{side}$: *$F_{(2,179)}$ = 667.79; p < 0.001*). This

difference was not only present during their first lift (left = -1.98 ± 0.30 N; right = 3.13 ± 0.21 N) but

persisted during their second (left = -1.53 ± 0.25 N; right = 2.71 ± 0.26 N) and third lifts (left = -1.44 ± 0.25

N; right = 2.61 ± 0.32 N) *(between sides: all p < 0.001)* (REPETITION X POSITION$_{SIDE}$: *$F_{(4,179)}$ = 3.69; p =*

0.006). Furthermore, our results provide no evidence that participants, in the noncollinear condition,

scaled their load forces differently over multiple repetitions for each side (*between repetitions for each*

side: all p = 1.00). As such, our results do not indicate that participants altered their load force difference

based on tactile feedback.

648    In line with our expectations, our results show that when participants used collinear positioning,

they exerted more load force with the finger on the heavy side (left = 3.01 ± 0.21 N; right = -5.10 ± 0.38

N; between sides: *$p < 0.001$*). Moreover, when using collinear positioning for each weight distribution,

the load force difference became larger from the first (left = 1.67 ± 0.26 N; right = -3.95 ± 0.37 N) to the

second lift (left = 3.34 ± 0.23 N; right = -5.41 ± 0.42 N). Noteworthy, this difference was significant for the

left asymmetrical weight distribution *(p < 0.001)*, but not for the right asymmetrical one *(p = 0.06).* These

findings indicate that participants altered their load force difference based on tactile feedback, in

particular, when lifting the left asymmetrical weight distribution. This effect further increased from the

second to third lift for the left (mean = 4.02 ± 0.21 N, *$p < 0.001$*) but not right (mean = -5.93 ± 0.47 N, *$p =$*

0.26) asymmetrical weight distribution. To end, the difference between the first and third lift for each

side was also statistically different (left: *$p < 0.001$*; right: *$p = 0.01$*).

659                  ----------

660                 Figure 5

661                  ----------

<u>Corticospinal excitability during baseline.</u>

Before and after participants performed the experimental task, we assessed baseline (resting state) CSE.

These results are shown in Figure 5. We used a two-by-two repeated measures ANOVA with factors

MUSCLE (FDI and APB) and TIME (pre- and post-experiment). Briefly, the analysis did not indicate that

CSE was differently modulated between the FDI (mean = 1.50 ± 0.24 mV) and APB (mean = 1.15 ± 0.22

mV) (main effect of MUSCLE; $F_{(1,15)} = 1.00; p = 0.33$). No main effect of TIME was found either ($F_{(1,15)} =$

$2.17; p = 0.16$) as CSE did not differ significantly between pre- (mean = 1.15 ± 0.15 mV) and post-

experiment (mean = 1.49 ± 0.22 mV). Noteworthy, the analysis revealed significance for the interaction

effect MUSCLE X TIME ($F_{(1,15)} = 7.55; p = 0.015$) although post-hoc exploration did not reveal significant

differences between measurements (Figure 5). In sum, our results provide no evidence that baseline CSE

was differently modulated before or after the experiment and between the FDI and APB muscles.

673   ----------

674   Figure 6

675   ----------

<u>Corticospinal excitability during lift observation</u>

As mentioned before, we hypothesized that modulation of motor resonance would be driven by specific

observed movement features (i.e., digit positioning or muscle contraction).

First dorsal interosseus muscle (FDI). As shown in Figure 6A, both the effect of SIDE and

POSITION$_{SIDE}$ were not significant *(both F < 0.94, both p > 0.91)*. As such, our results provide no evidence

that corticospinal excitability was differently modulated during the different observation conditions

when being recorded from the FDI muscle.

Abductor pollicis brevis (APB). In contrast to our findings for the FDI muscle, both the effects of

SIDE and POSITION$_{SIDE}$ were significant *(both F > 5.19, both p < 0.008)*. As can be seen in Figure 6B, when

participants observed the actor lifting the right-sided asymmetrical weight distribution, CSE was

significantly larger (mean = 0.15 ± 0.06) compared to when participants observed lifts on the left-sided

weight distribution (mean = -0.07 ± 0.04) *(p = 0.013; main effect of SIDE)*. Although the nested effect

POSITION$_{SIDE}$ was significant, post-hoc comparisons did not reveal significant differences between

conditions *(all p > 0.06)*. In sum, these findings indicate that, CSE was significantly facilitated when

observing lifts with an asymmetrical weight distribution that was right-sided.

<u>Corticospinal excitability during lift planning</u>

Although this study was aimed to investigate motor resonance effects, we expected that CSE modulation

would be similar (i.e., driven by the same movement features) as during lift observation.

First dorsal interosseus muscle (FDI). All effects were not significant *(all F < 1.42, all p > 0.24)*. As

such, our results provide no evidence that CSE was differently modulated during the different lift

planning conditions when being recorded from the FDI muscle (Figure 6C).

Abductor pollicis brevis (APB). As shown in Figure 6D, in line with our findings for CSE modulation

during lift observation, the effect of SIDE was significant $(F_{(1,192)} = 16.23, p < 0.001)$ although all other

effects were not *(all F < 2.06, all p > 0.13)*. When participants planned to lift the right asymmetrical

weight distribution (mean = 0.11 ± 0.04), CSE was significantly larger compared to when they planned to

lift the left-sided one (mean = -0.08 ± 0.03) *(p < 0.001; main effect of SIDE)*. As such, these findings

indicate that, irrespective of digit positioning, CSE was significantly facilitated when planning to lift a

right asymmetrical weight distribution.

<u>Background EMG during the experiment</u>

To ensure that between-group and between-condition differences were not driven by differences in

hand relaxation during lift observation and planning, we investigated potential differences in background

EMG. For this we used the same statistics as the ones we used for lift observation and lift planning.

Background EMG during lift observation. For both the FDI and APB muscle, all effects were not

significant *(all F < 0.68, all p > 0.42)*. As such, these findings provide no evidence that background EMG

different significantly between conditions when participants observed lifts on the asymmetrical weight

distributions.

Background EMG during lift planning. For both the FDI and APB muscle, all effects were not

significant *(all F < 0.69, all p > 0.60)*. As such, these findings provide no evidence that background EMG

different significantly between conditions when participants planned to lift the asymmetrical weight

distributions.

Discussion

In the present study, we investigated how CSE is modulated during observation of object lifting (i.e.,

'motor resonance'). Although we were initially interested in motor resonance effects only, we also

recorded CSE during lift planning. Participants were asked to lift a manipulandum after observing an

actor lift it. The object's center of mass could be changed by placing a heavy cuboid in one of three

compartments (Figure 1A). The object's weight distribution could be 'symmetrical', by inserting the

heavy cuboid in the middle compartment, or left- and right-sided 'asymmetrical', by inserting the heavy

cuboid in the left or right compartment, respectively. Participants and actor were instructed to lift the

object skillfully, i.e., generate appropriate compensatory torque to minimize object roll. Moreover, when

lifting the asymmetrical weight distributions, digit positioning was constrained to two specific digit

positioning strategies and, as mentioned before, TMS was applied during both observation and planning.

Our results indicate that the participants' lift performance was suboptimal as they were not able

to generate appropriate compensatory torque (Figure 4A). Noteworthy, the actor was also not able to do

so when using collinear digit positioning on the asymmetrical weight distributions. As such, suboptimal

lift performance was likely caused by limitations in our experimental set-up. Noteworthy, our results

indicate that CSE modulation during both lift observation and planning was driven by the object's

asymmetrical weight distribution. That is, when participants observed or planned a lift on the right

asymmetrical weight distribution, CSE recorded from the APB muscle was significantly increased

compared to observing or planning a lift on the left asymmetrical one. ~~During lift observation CSE~~

~~modulation seemed to be primarily driven by an observed change in digit positioning (i.e., the thumb~~

~~moving to the lower constrained location; Figure 4B). During lift planning, CSE modulation seemed to be~~

~~more generally increased for both constrained digit positioning conditions on the right asymmetrical~~

~~weight distribution (Figure 6D).~~ In line with previous work (Alaerts, et al., 2010b; Rens et al., 2020), these

findings indicate that motor resonance can be modulated by an object's weight distribution similarly to

other intrinsic object properties (such as weight).

Previous studies have shown that CSE is modulated during lift observation ('motor resonance').

For instance, Alaerts et al. (2010a, 2010b) showed that motor resonance during lift observation reflects

object weight as indicated by visual object properties (such as degree of filling) or by observed

movement features (such as muscle contraction or movement kinematics). In addition, Buckingham et al.

(2014) showed that motor resonance is driven by object size when observing skilled, but not erroneous

lifts. Last, Rens et al. (2020), showed that motor resonance is modulated by object weight but can be

strongly biased by contextual cues. Our findings corroborate these studies by showing that motor

resonance was modulated by the intrinsic object properties. Specifically, CSE recorded from the APB

muscle was increased when observing lifts in which the object's weight was asymmetrically distributed

to the right side (Figure 4). Considering that the object was visually identical in all weight distribution

conditions, participants could not rely on visual object-related information (e.g., the heavy cube being

visually different). As such, motor resonance should have been solely driven by observed movement

features (such as digit positioning or object muscle contraction) within the observed lifts on the right

asymmetrical weight distribution. However, it is important to note that CSE modulation was only present

in the APB and not in the FDI muscle which might indicate that these effects of complex object

properties (i.e., weight distribution) are relatively weak.

When the actor had lifted the symmetrical weight distribution and would then lift the right

asymmetrical one with noncollinear positioning, she would shift her thumb on the light side to the lower

constrained location whereas the index finger stayed on the same upper one. It is likely that this shift in

digit positioning increased motor resonance when observing noncollinear lifts on the right asymmetrical

weight distribution whereas the other movement features contributed arguably less. That is, when the

actor lifted both asymmetrical weight distributions noncollinearly, she generated similar compensatory

torques (Figure 3A). In addition, the actor used higher grip forces when lifting the asymmetrical weight

distributions with collinear positioning (Figure 3C) and she scaled her thumb load forces higher when

lifting the left-sided asymmetrical weight distribution with noncollinear positioning (Figure 3D). As such,

only when observing noncollinear lifts on the right asymmetrical weight distribution did the thumb

positioning change whereas the other movement features (i.e., force and compensatory torque) in this

condition were similar to those in the noncollinear left asymmetrical condition. Given the change in digit

positioning and the overlap in other movement features, it is likely that motor resonance effects in this

condition were primarily driven by the observed digit positioning.

In contrast to our initial hypotheses motor resonance effects were statistically driven by

differences in the asymmetrical weight distribution rather than constrained positioning conditions

(Figure 6). Accordingly, motor resonance should also have been increased when observing lifts on the

right asymmetrical weight distribution with collinear positioning. There are several factors that could

have driven this increase in motor resonance. First, in this condition, the actor's performance was the

worst as she generated the least amount of compensatory torque (Figure 3A). Second, in this condition

she scaled her grip forces higher than in all other conditions (Figure 3C). Third, the thumb had to

generate almost no load force as this was done by the index finger on the heavy side (Figure 3D). Last,

digit positioning did not deviate strongly from those in the other collinear conditions. As such, motor

resonance effects when observing collinear lifts on the right asymmetrical weight distribution could have

been driven by object roll (due to the inappropriate amount of compensatory torque) or the observed

grip forces. Arguably, the contribution of the observed grip forces on motor resonance is debatable.

Specifically, when the actor lifted the left asymmetrical weight distribution with collinear positioning, she

generated more load forces with her thumb compared to when she used the same positioning for the

right asymmetrical weight distribution (Figure 3D). As such, this 'force interpretation' would suggest that

observed grip forces modulate motor resonance whereas observed load forces do not. Although Alaerts

et al., (2010b) have demonstrated that CSE is modulated during observed squeezing (i.e., visible muscle

contraction: grip forces) and during observed lifting when the hand contraction cannot be seen (i.e.,

movement kinematics: indicative of the load forces), it is still unsure how these force parameters

converge in modulating CSE during observed lifting. To our knowledge, the differential muscle

contractions of grip and load forces on CSE modulation have not been disentangled yet.

To end, we initially hypothesized that motor resonance would be selectively modulated by either

observed digit positioning or muscle contractions. However, our results suggest that modulation of

motor resonance was more generally modulated by movement features indicating the object's right

asymmetrical weight distribution. Indeed, motor resonance was primarily increased when observing a

visible shift in digit positioning in the right noncollinear condition and by the largest object roll, caused by

inappropriate compensatory torque, in the right collinear condition.

Based on previous findings of our group (Rens et al., 2020), it is plausible that digit positioning

and object roll could have modulated motor resonance. In that study, we highlighted that motor

resonance is only modulated by object weight when the actual object weight would always match the

participants' weight expectations. However, when the participants' weight expectations could be

incorrect, motor resonance was rather driven by a mechanism monitoring these expectations. Those

findings have been supported by a reasoning within the review of Amoruso and Finisguerra (2019). They

argued that motor resonance reflects the inner replica of the observed action when observed in isolation

but can be altered by higher-level factors (such as contextual cues) when present. In sum, both these

works indicate that motor resonance can be flexibly driven by different movement features during action

observation. In the current study, participants were asked to minimize object roll during lift observation.

This contextual importance of accurately estimating the weight distribution during lift observation might

have caused motor resonance to be driven by 'salient' movement features indicating the object's weight

distribution. Arguably, it is plausible that participants focused on shifts in digit positioning and visible

object roll to perceive the weight distribution which in turn modulated their motor resonance.

Critically, we did not find similar modulation of CSE recorded from the FDI muscle. This was likely

caused by experimental limitations. When the inverted T-shape manipulandum was placed in front of

the participant, it was rotated (< 45 degrees) according to the participants' preferences. As most

participants found the manipulandum relatively heavy, the rotated position enabled them to reduce

wrist overextension, improving lifting comfort. Due to the object rotation, the index finger was hidden

behind the manipulandum and participants had no vision on the index finger during both lift observation

and execution. These visiblity limations might have completely eradicated any CSE modulation recorded

from the FDI muscle. It is also possible that similar visiblity limitations drove the selective modulation of

motor resonance for the right but not left asymmetrical weight distribution. When the actor lifted the

manipulandum (see dyadic positioning; Figure 1C), she would reach with her arm in front of the

participant blocking vision on the manipulandums left side with her lower arm. Because of these visiblity

limitations, participants might have primarily focused on thumb actions and the right side of the

manipulandum. However, it is important to note that the participant' lift performance for the

noncollinear conditions did not differ between asymmetrical weight distributions. Furthermore, when

using collinear positioning, participants performed worse on the right asymmetrcial weight distribution

compared to the left one. As such, even though motor resonance was selectively modulated for the right

asymmetrical weight distribution, participants perceived both weight distributions similarly during lift

observation as indicated by their own lift performance after lift observation.

In a recent motor resonance study on humans, Cretu et al. (2019) investigated whether motor

resonance effects are present when the hand-object interaction cannot be seen. They found that motor

resonance can be driven by contextual cues but only if they are informative of the hidden action. As

such, Cretu et al. (2019) showed that that relevant information regarding the observed action needs to

be present in order to modulate motor resonance. In addition, we previously showed that motor

resonance is flexibly modulated based on experimental context (Rens et al., 2020). As participants in the

present study had no vision on the index finger during observed object lifting, they may have focused

solely on the thumb thus eradicating motor resonance effects recorded from the FDI muscle. However,

future research is necessary to substantiate the notion that motor resonance can be digit-specifically

modulated based on visibility. Additionally, futures studies could place the actor opposite of the

participant or use transparent objects to ensure both fingers are always visible. We initially decided to

place the actor at the side of the participant as Alaerts et al. (2009) showed that motor resonance effects

are stronger when executed actions are observed from a first person point of view.

With respect to lift execution, our participants were not able to lift the manipulandum skilfully in

the collinear condition, even after haptic feedback (i.e., second and third lifts; Figure 4A). In contrast, in

the noncollinear condition, participants were able to generate appropriate compensatory torque already

in their first lift (Figure 4A) and did not improve from their first to second lifts. Critically, our actor also

performed suboptimally when lifting collinearly. In contrast, in Fu et al. (2010), participants were able to

lift an asymmetrical weight distribution skilfully when using constrained collinear digit positioning. Taken

together, it may not have been feasible in our study to lift the asymmetrical weight distribution skilfully

with collinear digit positioning. It is important to note that even though our manipulandum and the one

of Fu et al. (2010) had similar external torque, ours was slightly heavier. In addition, it is possible that

textural differences (i.e., friction of the graspable surfaces) differed between ours and theirs which led to

differences in performance (Johansson & Westling, 1984). As such, having decreased the object weight,

thus making lifts on the asymmetrical weight distributions less challenging, could have removed the

potential confounding effects of lifting performance differences between conditions.

Parikh et al. (2014) showed that when individuals plan to generate high or low forces, CSE during

planning is decreased when planning to generate high forces. Their findings suggest that predictive force

planning is force-dependently modulated akin to work of Loh et al. (2010) on predictive object lifting.

However, in our study this force-dependent modulation seemed to be absent: when participants

planned to lift the right asymmetrical weight distribution, CSE was similarly modulated in both digit

positioning conditions even though they generated more force in the collinear than in the noncollinear

condition. As such, CSE during lift planning was unlikely modulated by planned forces but rather by the

object's weight distribution. However, as lift planning was contaminated by suboptimal performance,

future research is necessary to substantiate whether the motor system can veridically encode an object's

weight distribution during lift planning. Furthermore, Davare et al. (2019) showed that CSE is increased

when lifting an asymmetrical weight distribution with unconstrained compared to constrained digit

positioning. This suggests that CSE modulation is driven by the sensorimotor uncertainty (regarding digit

positioning). However, this effect was only present when TMS was applied during early object contact

and not during mid-reach. With respect to our study, it is unlikely that mechanisms related to

sensorimotor uncertainty affected CSE modulation during lift planning as we constrained digit

positioning in all conditions and we applied TMS during lift planning, not early object contact.

In our study, we decided on our constrained digit positionings with the intention of having the

digit positioning difference and load force difference approximate zero in the collinear and noncollinear

conditions, respectively (in support of this statement see Figure 3 for the actor's performance). However,

as mentioned before, lifting performance was suboptimal for both actor and participants. As such, future

studies could decrease object weight or let participants choose their own constrained digit positionings

to ensure skilled performance. In our previous work (Rens et al., 2021), we found that observing skilled

lifts of asymmetrical weight distributions did not enhance predictive lift planning in the observer. In the

present study, we found that participants generated more compensatory torque after observing skilled

lifts with noncollinear digit positioning, contrasting the findings of our previous study. Arguably, this

difference is driven by the constrained digit positioning in the present study as participants were free to

choose their own digit positioning in Rens et al. (2021). However, more research is required to

investigate the interaction between action observation and constraining motor execution on predictive

lift planning.

888   Even though lifting performance was suboptimal, our results indicate that CSE modulation during

lift planning and observation is not associated with the participants' lift performance. Although CSE was

similarly modulated for collinear and non-collinear lifts on the right center of mass, participants did not

lift the right asymmetrical weight distribution skillfully with collinear digit positioning. Indeed, a

limitation in our study is the erroneous lifting performance on the asymmetrical weight distribution

when using collinear digit positioning: Buckingham et al. (2014) showed that motor resonance can be

differently modulated when observing skilled or erroneous actions. As such, future research is required

to disentangle how motor resonance relates to motor planning and proper execution.

896   Last, it has been shown that S1 is involved in integrating haptic feedback to generate appropriate

load forces when lifting objects (Parikh et al., 2020) and stimulating it during our task could have affected

lift performance. The scalp location for stimulating S1 has been shown to be located approximately 2 cm

lateral and 0.5 cm posterior to the scalp location for stimulating M1 (Holmes et al., 2019; Holmes &

Tamè, 2019). However, confounding stimulation of S1 may be unlikely as we oriented the TMS coil to

induce a posterior-anterior current. In addition, the spread of TMS across neighboring tissue has been

considered relatively small based on modelling (Deng et al., 2013) and single-cell recordings in the

macaque monkey (Romero et al., 2019). In particular, Romero et al., (2019) showed that the spread of

single-pulse TMS was limited to less than 2 mm in diameter. Although we cannot exclude that S1 was

unintentionally stimulated in the present study, it should not have affected conditions differently. That

is, we applied TMS during every trial (i.e., during lift observation and during lift planning on all weight

distributions). As such, potential confounding effects of S1 stimulation should have affected all

conditions equally. To end, to investigate this potential confounding effect of TMS, future studies could

include a 'no stimulation condition' to compare lift performance with or without single-pulse TMS during

lift planning.

911   Motor resonance has been argued to rely on mirror neurons. Mirror neurons are similarly

activated when executing or observing the same action and have been argued to be involved in action

understanding by "mapping" observed actions onto the cortical representations involved in their

execution (Rizzolatti et al., 2014). Mirror neurons are primarily located in M1, the ventral premotor

cortex (PMv) and the anterior intraparietal area (AIP) (Rizzolatti et al., 2014). Importantly, these regions

also constitute the cortical grasping network which is pivotal in planning and executing grasping actions

(for a review see Davare et al. 2011) further substantiating these neurons involvement in action

understanding. Our findings corroborate mirror neuron functioning by showing that CSE was similarly

modulated during lift observation and planning. Interestingly, even though our motor resonance findings

might have been partially driven by the experimental context (Rens et al., 2020), CSE modulation was

similar during planning, which substantiates that the same mechanisms underlied CSE modulation in our

experiment. To end, CSE modulation being similarly increased for the right asymmetrical weight

distribution during both observation and planning indicates that the motor system encoded the same

object-related information during both planning and observation.

In conclusion, the present study investigated how CSE is modulated during the observation and

planning of lifts with an asymmetrical weight distribution. Our findings suggest that CSE modulation

during observation was driven by the object's weight distribution as potentially indicated by digit

positioning and object roll (i.e., inappropriate compensatory torque). During lift planning, CSE was

similarly modulated by the object's weight distribution indicating, in line with previous research (for a

review see Cattaneo & Rizzolatti, 2009), that the same neural mechansisms are involved in CSE

modulation during lift observation and planning. As such, our findings provide further support that the

motor system is involved in the observation and planning of hand-object interactions.

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Figure captions

Figure 1. Experimental set-up. (A) Left: Frontal, side and top-down view of the 'inverted T-shape'
manipulandum with dimensions (in mm). **(B)** Schematic drawing of the manipulandum with the three
compartments indicated with L, C and R standing for left, central and right respectively. In red, the
constrained locations are marked. X, Y and Z indicates the frame of reference for the force/torque
sensors in the vertical component. **(C)** Dyadic positioning at a square table with the manipulandum
placed in between the actor and participant. Figure modified from and reprinted with permission from
Rens et al. (2020b)

Figure 2. Example parameter traces. One lift example of performance on a left-sided asymmetrical
weight distribution for the actor (solid traces) and first lift of a participant (dashed traces). These traces
show the typical evolution of different parameter profiles over time for a lift with fingertips positioned
on the same height ('collinear'; in blue) or with fingertips positioned on different heights ('noncollinear';
in green). **(A)** Compensatory torque (in Nmm). **(B)** Digit positioning difference, i.e., vertical height
difference in center of pressure of the fingertips (position thumb - position index finger; in mm). **(C)** Total
grip force (in N). **(D)** Load force difference (load force thumb – load force index finger; in N). Vertical
dashed line on each figure indicates object contact. We cleared compensatory torque and digit
positioning difference values before object contact as they are highly contaminated by noise.

Figure 3. Lift performance of the actor. The actor's averaged lift performance. The actor used either
collinear (col; blue) or noncollinear (noncol; green) digit positioning to the object. Each scatter dot
represents the averaged performance of the actor as observed by one participant. **(A)** Compensatory
torque (in Nm). **(B)** Digit positioning difference (in mm). **(C)** Total amount of grip force (in N). **(D)** Load
force difference (in N). All data is presented as the mean ± SEM.

Figure 4. Lift performance of the participants. Connected scatterplot showing the averaged lift
performance of the participants for their three lift repetitions. Performance is shown for when the
weight distribution was left (solid scatter) or right (empty scatter**)** asymmetrical. Participants used either
collinear (blue) or noncollinear (green) digit positioning for lifting. **(A)** Compensatory torque (in Nmm).
(B) Digit positioning difference (in mm). **(C)** Total grip force (in N). **(D)** Load force difference (in N). All
data is presented as the mean ± SEM. Vertical lines with a p-value indicate that all scatters above this line
differ significantly from all scatters below. P-values between two connected scatters indicate that those
two differ significantly from each other. For within-condition comparisons, only significant differences
between two subsequent trials are shown (e.g., between trial one and two but not between trial one and
three).

Figure 5. Corticospinal excitability during baseline (resting state). Average motor evoked potential
(MEP) values in mV during baseline before (pre) and after (post) the experiment. MEPs were recorded
from first dorsal interosseous muscle **(FDI) a**nd from the abductor pollicis brevis muscle **(APB)**. Each
colored scatter dot on top of each bar represents the average MEP value for one participant. All data is
presented as the mean ± SEM.

Figure 6. Corticospinal excitability during lift observation and planning. Average motor evoked
potential (MEP) values (z-score normalized) during lift observation **(top row)** and lift planning, pooled for
REPETITION, **(bottom row)** recorded from first dorsal interosseous muscle **(FDI; panels A and C)** and
from the abductor pollicis brevis muscle **(APB; panels B and D)**. The actor and participants used either
collinear (col; blue) or noncollinear (noncol; green) digit positioning to lift the asymmetrical weight

distributions (left or right). Each dot on top of each bar represents the average MEP value for one
participant for that respective condition during lift observation and planning. All data is presented as the
mean ± SEM.

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