

Not fleeting but lasting: Limited influence of aging on implicit adaptative motor learning and its short-term retention.

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Summary

1 In motor adaptation, learning is thought to rely on a combination of several processes. Two of these
2 are implicit learning (incidental updating of the movement due to sensory prediction error) and
3 explicit learning (intentional adjustment to reduce target error). The explicit component is thought to
4 be fast adapting, while the implicit one is slow. The dynamic integration of such a fast and a slow
5 components can lead to spontaneous recovery. That is, after prolonged adaptation of movement to a
6 given perturbation, the learning is extinguished by presenting a perturbation in the opposite
7 direction for a few trials. After such extinction, the learned adaptation can reappear in the absence
8 of any further training, a phenomenon called spontaneous recovery. Trewartha et al. (2014) found
9 that older adults show less spontaneous recovery than their younger controls, indicating
10 impairments in short-term retention of force-field adaptation. This disagrees with evidence
11 suggesting that the implicit component and its retention do not decline with aging.

12 To clarify this discrepancy, we performed a conceptual replication of that result. Twenty-eight
13 healthy young and 20 healthy older adults learned to adapt to a forcefield perturbation in a paradigm
14 known to elicit spontaneous recovery. Both groups adapted equally well to the perturbation. Implicit
15 adaptation of the older subjects was indistinguishable from that of their younger counterparts. In
16 addition, our conceptual replication failed to reproduce the result of Trewartha et al. (2014) and
17 found that the spontaneous recovery was also similar across groups. Our results reconcile previous
18 studies by showing that both spontaneous recovery and implicit adaptation are unaffected by aging.

19 New and Noteworthy:

20 In this study, we tested whether aging influences the ability to learn to counteract a perturbation
21 during reaching movements and to recall previously learned motor memories. In contrast to a
22 previously published paper, we found that the ability of older participants to adapt to a perturbation
23 and to recall motor memories remains unimpaired.

24 Introduction

25 Young healthy adults can adapt to a change in the environment and adapt their reaching
26 performance (1, 2). Such an adaptation process of upper limb movements is studied in the laboratory
27 via perturbation of the visual feedback about the moving direction of the hand (3, 4), by shifting the
28 visual field via prism goggles (5) or by applying a force on the moving arm (6, 7). For any of these
29 perturbations, young participants can readily decrease the effect of the perturbation on their
30 reaching performance through a combination of explicit strategies and implicit adaptation (8–11).
31 Implicit adaptation is the incidental updating of the movement driven by sensory prediction error
32 and occurs gradually (9, 12, 13). Explicit adaptation consists of the application of cognitive strategies
33 to reduce target error and reduces errors rapidly (1). The explicit component contributes more to
34 total adaptation for visuomotor rotation larger than 30° than for force-field adaptation. Indeed, in
35 visuomotor rotation, implicit adaptation amounts to 15-20deg for 45° perturbations (30-50% of
36 compensation, 8). In contrast, learning to counteract a force field is largely an implicit process with
37 only a small explicit component (implicit adaptation around 60-80% of total adaptation 14).

38 Older adults show lower levels of total motor adaptation than young adults (15–29). Recent evidence
39 suggests that this impairment in motor adaptation is specific to the explicit component of adaptation
40 (21–23, 27, 28, 30–32) and that the implicit component of motor adaptation elicited by a visuomotor
41 adaptation and its short-term retention (as measured by the decline in adaptation during one-minute
42 breaks) remains unimpaired in individuals up to 60-70 years old (19, 27, 33–36).

43 Few studies have investigated the effect of age on force-field perturbation (37–41). Little difference
44 in the amount of adaptation to a force-field perturbation has been found between young and older
45 participants (38, 41). Yet, the explicit and implicit components of adaptation have never been
46 measured in these studies. While we know that the contribution of explicit strategies to force-field
47 adaptation is small (14), it is not null. Therefore, it remains unknown whether the implicit component
48 of motor adaptation remains unaffected in older people during a force-field adaptation task.
49 Measuring the explicit and implicit components of motor adaptation is essential in order to gain
50 insight into the source of possible deficits. Interestingly, one study reported a very specific age-
51 related impairment in force-field adaptation. That is, while initial adaptation was unimpaired with
52 age, its short-term retention as measured by spontaneous recovery of adaptation was impaired (41).
53 Furthermore, these authors suggested that this impairment was linked to low explicit memory
54 resources.

55 Spontaneous recovery occurs when motor adaptation to some perturbation A, which is then hidden
56 from view due to adaptation of a second perturbation B, reappear without any additional exposure

to perturbation A (42–47). It suggests that the motor memory of the adaptation to perturbation A is not washed out by adaptation to perturbation B but is retained. Therefore, such spontaneous recovery of motor memories linked to perturbation A represents a proxy for short-term retention of the associated motor memory (47).

The presence of spontaneous recovery indicates the presence of at least two learning processes working on different time scales. One process learns and forgets quickly, while the other is slow (47, 48). In this framework, the spontaneous recovery of motor memory of field A is attributed to the memory trace of the slow learning process, (45, 47). This memory trace is masked by the fast adaptation process during the deadadaptation period. Interestingly, the slow process has been associated with the implicit component of adaptation while the fast process has been linked to the explicit component (45).

Three major concepts reviewed up to this point bear some contradictions: 1) implicit adaptation and its short-term retention are not impaired by aging (27), 2) spontaneous recovery is impaired in older people (41), and 3) the slow process of adaptation, which determines spontaneous recovery, corresponds to the implicit component of adaptation (45). In other words, if implicit adaptation is unimpaired in older people and if it determines spontaneous recovery, then spontaneous recovery cannot be different across age groups. Yet, it is unclear where the contradiction comes from as these different studies have marked differences in protocol, which could affect the results. Implicit adaptation level was obtained using a visuomotor rotation of the cursor feedback (27), whereas its short-term retention was measured in a force-field paradigm (41). As it is known that perturbation type influences implicit adaptation level (8, 14), this difference in perturbation type could be responsible for this discrepancy. There is thus a need to test these three observations within a single experiment. Therefore, we set out to measure both implicit adaptation and its retention via spontaneous recovery in a single force field paradigm in both healthy young and older adults in order to test four different hypotheses, reflecting previous research: 1) implicit adaptation levels at the end of the adaptation period are similar across age groups (27); 2) spontaneous recovery is larger in young than in older participants (41), 3) implicit adaptation at the end of the adaptation period is correlated with the amount of spontaneous recovery (45) and, 4), as suggested by Trewartha and colleagues, spontaneous recovery is related to explicit memory processes such as working memory capacity. We expect that at least one of the three first hypotheses will be violated because, as mentioned above, they are contradicting each other.

Methods

Participants

After signing the informed consent, 28 young adults (19-27, 23 ± 2 , 12 male) and 21 older adults (60-75, 67 ± 4.70 , 10 male) participated in the study. We excluded one older subject from analysis due to an error in task execution (wrong block order). All participants were right-handed as indicated by the Edinburgh inventory (49) and were screened with general health and consumption habits questionnaires. Based on the general health questionnaire, participants with events, diseases or injuries that could affect the control of movement were excluded (e.g. head trauma). Based on the consumption habits, participants using recreational drugs or having hazardous alcohol consumption (more than 21 drinks per week for men or more than 14 for women) were excluded from the study. No participants were excluded for these reasons. The older adults were assessed using the Mini-Mental State Examination and all scored within normal limits (score ≥ 24 , (50)). Approval was obtained by the Ethics Committee Research UZ / KU Leuven. Participants received financial compensation (€10/h).

Sample size was initially planned to reproduce the 20 participants per group as in Trewartha et al. 2014. We first included 20 older participants (recruited 21 but one was excluded due to error in the experiment) and 21 young ones. Upon data analysis, we noticed that, on average, younger people moved faster than older people in this paradigm despite the speed constraints (see below). We recruited an additional 7 young participants that were instructed to move slower to match hand velocity across age groups (as described below in the experimental paradigm).

Experimental paradigm for the adaptation task

Participants made center-out, reaching movements in the horizontal plane while holding a robotic handle (KINARM End-Point Lab, BKIN Technologies). Hand position was represented by a white cursor on a display and vision of the hand was occluded. Movement trajectories were sampled at 1000 Hz. At the beginning of each trial, participants had to move their cursor to a starting point in the middle of the screen, after which a target appeared on one of eight possible locations (*Figure 1*) spaced ten cm away from the starting point. Participants were instructed to slice through the target by making a rapid, smooth reaching movement, avoiding any corrections. Once the movement amplitude exceeded ten cm, cursor position froze, providing feedback about movement accuracy and movement time. Movement times within 200 to 350ms resulted in a green cursor (for 5 young participants we increased the time window to 400 – 550ms and for another 2 to 300 – 400ms to keep average movement times the same for both age groups). Too slow movements caused the cursor to turn blue and too fast movements resulted in an orange cursor. After feedback, the starting point appeared, a new trial started, and the participants had to move their hand back to the starting position. On each trial, two points could be earned: one for hitting the target (hand crosses through

the target) and one for applying the right speed. Participants were encouraged to collect as many points as possible.

The task started with 72 baseline trials with reaches towards eight possible targets presented in pseudo-random order (9 cycles of 8 different targets) and a white cursor (Figure 1). Participants then continued with a perturbation phase (trials 73 – 281), during which a viscous force field (12 Ns/m) was applied perpendicular to hand velocity and the cursor had a red color (cued trials). Subjects received additional instructions, which were: “Initially, your cursor was a white dot, but from now on, your cursor can turn red. At that moment, something special will happen, but you still have to try to do the same thing, slice through the target with your cursor. A warning sign will be shown each time your cursor changes color”. Interspersed with these perturbation trials were 20 trials with a white cursor (uncued trials), located at one of four possible locations. While the white cursor could be considered as a cue, we decided to adopt the terminology of Morehead and colleagues (2015). In uncued trials, participants could move straight ahead without interference of the force field and were occasionally reminded of this. Participants were thus aware that uncued trials were similar to baseline trials. For these trials, they were still instructed to bring the cursor to the target. The goal of these trials was to measure the implicit after-effect over a single trial by requiring participants to disengage any strategies that they might be using.

From trial 282 to 305, the force field was reversed, washing out the adaptation to the first perturbation (Deadaptation in Fig.1). Despite this change in the direction of the perturbation, the instruction to the participants remained unchanged: they were instructed to bring the cursor to the target.

Lastly, retention was tested during an error-clamp phase (trials 306 – 369). In these trials, hand trajectory was constrained to a straight line from the starting point to the target, by guiding the handle between two stiff virtual walls (51). Visual feedback was provided during these trials. Throughout the baseline and perturbation phase, we pseudo-randomly introduced error-clamp trials to measure forces participants applied on the robot (8 during baseline and 20 during the perturbation phase). While all 8 targets were used during the retention phase, only 4 of them were used for the error-clamp trials during the baseline and perturbation phase. The direction of the force field, clockwise or counterclockwise, was counterbalanced across subjects and three one-minute breaks were provided (dashed grey vertical lines on Fig.1, after trials 54, 153 and 253).

Experimental paradigm for the visual working memory task

Given the importance of working memory for the explicit component of motor adaptation (28, 52) and its potential link with the spontaneous recovery of motor memory (41), we decided to measure

working memory capacity in all participants. It was quantified with a computer-based task (27, 28, 52, 53). Sixteen white lined squares were presented in a circular array with, in the middle, a white fixation cross. Three to six red circles were presented for two seconds randomly each in one of the squares. The array disappeared leaving only the fixation cross for three seconds, where after the squares returned with a question mark placed in one of them. Subjects had to indicate whether the probed location had contained a red circle. After three seconds of response time, a new trial began. Participants completed 48 trials (12 trials/condition) after eight practice trials. Two participants did not perform this task.

Data processing and analysis

The x and y positions of the handle and x and y forces exerted on the handle were recorded at 1000 Hz. To combine the data from subjects who started with a clockwise force field with those who started with a counterclockwise force field, all the signs of position and force data in the x-direction for the clockwise condition were flipped.

For each baseline or perturbation trial, lateral deviation from the optimal trajectory from starting point to target was calculated at peak velocity. Total adaptation level was quantified as the lateral deviation of the last 80 cued trials during the perturbation phase. Implicit learning was quantified as the average of the lateral deviation over the last 12 uncued trials during that phase.

In the error-clamp trials, the force subjects exerted on the channel walls (perpendicular to the heading direction) at peak velocity was used as a measure of adaptation. A second method we used to quantify learning in error-clamp trials was to compute the slope of the relationship between ideal force during reaching and the exerted lateral force (41, 47). All trials from the perturbation phase onward were corrected for baseline error per target location. Retention of implicit learning, spontaneous recovery, was calculated as the average of the last 48 trials during the error-clamp phase. These outcome measures were controlled for peak hand velocity, because movement speed influences adaptation level (7). In addition, the slope of the relationship between ideal force and actual generated force was used as a control measure for the level of spontaneous recovery (adaptation index, 41).

Working memory capacity is calculated using the following formula: $K = S * (H - F)$. K is the memory capacity, S is the size of the array, H is the observed hit rate and F is the false alarm rate (54). To estimate K, we used the decision tree used by Vandevorde et al. (28) and published as supplementary material by Saenen et al. (53) and available at <https://doi.org/10.6084/m9.figshare.23535396.v1>.

For statistical testing, we used t-tests with unequal variance in all tests. All statistical tests were also reproduced with non-parametric tests but the results between the parametric and non-parametric tests never differed in their conclusion. Effect sizes (Robust Cohen's d) and their confidence interval (computed with bootstrap with 5000 iterations) were obtained from the meanEffectSize function in Matlab. ANCOVA's were performed with the aocool function in matlab (with the model 'parallel lines'), fitting a separate line to each group, but constraining these lines to be parallel as we did not expect a different relationship between the covariates and the dependent variables as a function of age. For all the analyses, the α -level was set at 0.05.

To test for the absence of age-related difference in adaptation (hypothesis 1), we compared the average lateral deviation at the end of the adaptation period over the last 80 cued field trials (analysis 1), the average exerted force (measured at peak velocity) during the last 12 cued clamped trials of the adaptation phase (analysis 2) and the implicit adaptation level measured as the average lateral deviation over the last 12 uncued trials (analysis 3) between young and older participants with an independent t-test with unequal variance. Additionally, an ANCOVA was used to check for any influence of hand velocity on these outcomes. The outcome was set as dependent factor and hand velocity for these specific trials were used as covariate. For each of the analyses, we also performed a Bayesian independent T-test with a Cauchy distribution as prior (width of 0.707).

In analysis 4, we tested age-related differences in spontaneous recovery level measured either as the average force exerted at peak velocity or as the average adaptation index computed over the last 48 clamped trials of the error-clamp period. These two outcomes were submitted to the same statistical tests as in analysis 1. The force data was also submitted to a Bayesian independent Samples T-test in JASP to test how compatible this data was with the hypothesis that spontaneous recovery was larger in young than in older participants (one-sided t-test). The selected prior for this analysis was the default Cauchy prior (Scale=0.707). The Bayesian analysis was performed in JASP (55).

To test possible correlation between adaptation levels at the end of the perturbation phase and during the error-clamp period (hypothesis 3), implicit adaptation levels at the end of the perturbation period (from analysis 3) and spontaneous recovery levels (from analysis 4) were correlated via multilevel correlation (Analysis 5) from the correlation package in R (56). The different levels corresponded to the different age groups.

In analysis 6, an additional ANCOVA was used with spontaneous recovery level (from analysis 4) as dependent factor and implicit adaptation level at the end of learning (from analysis 3) as covariates.

In analysis 7, we conducted a Bayesian independent t-test in JASP (55). The prior was centered on the effect size reported in the original study by Trewartha et al (d=0.8) and followed a Cauchy

221 distribution. We used three different scales for the prior in order to test the sensitivity of our results
222 to the width of the prior distribution. In this analysis, we test the hypothesis that the difference in
223 spontaneous recovery level between age groups is equal to $d=0.8$.

224 In analysis 8, working memory capacity was compared between young and older participants with an
225 independent t-test with unequal variance.

226 In analysis 9, we investigate the possible association (hypothesis 4) between working memory
227 capacity (from analysis 8) and spontaneous recovery levels (from analysis 4) via multilevel correlation
228 from the correlation package in R (56). The different levels corresponded to the different age groups.

229 All data can be found on the RDR repository of the KU Leuven: <https://doi.org/10.48804/KMGKLH> All
230 analysis scripts can be found at: [10.5281/zenodo.8284036](https://doi.org/10.5281/zenodo.8284036)

Results

Force-field adaptation does not decline with aging

The aim of this experiment was to measure the impact of aging on implicit adaptation and its short-term retention through spontaneous recovery. Participants made center-out reaching movements towards targets, while adapting to a force field that pushed their hand away perpendicular to the heading direction (cued trials with red cursor). With practice, subjects gradually decreased their error over the course of learning (*Figure 2A*). Total adaptation level at the end of the perturbation phase was similar across age groups (*Figure 2C*, mean \pm SD, young: 2.23 ± 1.43 mm, older: 2.29 ± 1.87 mm, Analysis 1: $t(34.07) = -0.13$, $p = 0.89$, Cohen's $d = -0.011$ CI=[-0.64,0.62]) and at the end of the deadadaptation period (trials 298 to 305, $t(45.25) = 0.75$, $p = 0.46$, Cohen's $d = 0.378$ CI=[-0.19, 1.02]).

Given the importance of hand speed in force-field adaptation, we checked that the hand speed was comparable across the two groups. At the end of the adaptation period, hand velocity was comparable (Fig.2C, young: 0.55 ± 0.15 m/s, older: 0.56 ± 0.12 m/s, $t(45.35) = -0.156$, $p = 0.87$, $d = 0.053$, CI=[-0.46,0.76]). Controlling movement speed did not change the outcome of the analysis of the lateral deviation at the end of the adaptation period (ANCOVA, $F(1,45) = 0.0161$, $p = 0.899$). The corresponding Bayesian analysis suggested that there was moderate evidence of an absence of difference (BF = 0.293).

The force that participants exerted against the perturbation built up as participants learned to counteract the perturbation (*Figure 2B*). In error-clamp trials, the exerted force reached similar levels at the end of the perturbation phase for both groups (*Figure 2E*, mean \pm SD, young: 3.6 ± 1.2 N, older: 3.4 ± 1 N, Analysis 2: $t(44.15) = 0.46$, $p = 0.65$, Cohen's $d = 0.16$, CI=[-0.42,0.79]). For those trials, we also did not find any evidence that the velocity varied across age groups ($t(44.34) = -0.89$, $p = 0.38$, Cohen's $d = -0.104$, CI=[-0.60,0.52]). Therefore, controlling for hand speed did not change the results (Analysis 2, ANCOVA: $F(1,45) = 1.84$, $p = 0.18$). The corresponding Bayesian analysis suggested that there was moderate evidence of an absence of difference (BF = 0.315).

In some trials (uncued trials), we warned the participants that the force field would be turned off in order to force participants to stop using any strategy to compensate for the perturbation and to measure implicit adaptation (8). In these trials, perpendicular error increased with continued learning (*Figure 2A*). Participants made reaching movements to four different target locations (ordinal directions, see figure 1). These trials were randomly presented throughout the perturbation phase, but in a fixed sequence. For some reason, participants from both groups exhibited different amounts of lateral deviation in function of target direction (*Figure 2A*). However, this effect of target direction was identical between the two age groups. We averaged the responses of the last 12 uncued trials

and compared these between our age groups (*Figure 2D, Analysis 3*). We could not find evidence for a difference in implicit adaptation between young and older adults (mean±sd, young: -11.96±3.70mm, older: -10.91±3.09mm, Analysis 3: $t(44.76) = -1.07$, $p=0.29$, Cohen's $d = -0.24$, $CI = [-0.83, 0.33]$). In those trials, we did not find any evidence that hand speed differed across groups (young: 0.573±0.162 m/s, older: 0.609±0.149 m/s, $t(43.00) = -0.785$, $p=0.4368$, Cohen's $d = -0.076$, $CI = [-0.56, 0.61]$). This result remained the same after the implicit adaptation level was controlled for movement speed (Analysis 3: ANCOVA: $F(1,45)=2.18$, $p=0.146$). The corresponding Bayesian analysis suggested that there was anecdotal evidence of an absence of difference ($BF = 0.45$).

No evidence that spontaneous recovery declines with aging

At the end of the experiment, lateral deviation of each movement was clamped to zero, ensuring participants would always hit the target. This enabled us to measure the retention of implicit adaptation without interference of trial-by-trial learning. Exerted force increased over time in the same direction as during the perturbation phase, characteristic of spontaneous recovery (*Figure 3A*). The average response of the last 48 clamp trials was compared between age groups (*Figure 3B*) and we did not find any evidence for a difference between young and older adults (median, young: 0.88±0.73N, older: 1.01±1.5N, Analysis 4: $t(25.46) = -0.35$, $p=0.73$, Cohen's $d = 0.036$, $CI = [-0.54, 0.67]$). Note that this result is independent of which trials are analyzed. Performing a trial-by-trial analysis as in Trewartha et al, no between-group differences remained significant after correction for multiple comparisons ($p<0.05/64$). Similarly, analyzing all 64 trials from the spontaneous recovery period provided the same statistical results ($t(25.44) = -0.367$, $p=0.72$, Cohen's $d = 0.003$, $CI = [-0.58, 0.71]$). We did not find any evidence that hand speed differed across groups (young: 0.53±0.14 m/s, older: 0.55±0.09 m/s, $t(45.99) = -0.4$, $p=0.69$, Cohen's $d = 0.034$, $CI = [-0.46, 0.78]$), indicating we succeeded in this aim.

Controlling movement speed did not change the result (Analysis 4: $F(1,45)=0.138$, $p=0.71$). In addition, the adaptation index (*Figure 3C*), which was used in Trewartha et al. (41), did not differ between age groups either (young: 0.15 ± 0.1, older: 0.17 ± 0.23, Analysis 4: $t(24.20) = -0.335$, $p=0.7402$, Cohen's $d = 0.014$, $CI = [-0.56, 0.65]$).

To confirm these results, we performed a Bayesian analysis on the force data in order to test how compatible our data was with the idea that the spontaneous recovery level was larger in young than in older participants. There was moderate support ($BF=4.44$) for the idea that the spontaneous recovery level was not larger in young than in older participants.

If the level of spontaneous recovery is linked to the level of adaptation at the end of the learning period, the absence of age-related difference in spontaneous recovery is consistent with the absence

of age-related difference in implicit adaptation in this study (Fig.2). Indeed, we expect that people with more implicit adaptation at the end of learning would exhibit more spontaneous recovery, resulting in a positive correlation between the two. Therefore, we pooled the data for all participants and correlated both measures while taking the two different groups into account (*Figure 4, Analysis 5*). A significant positive correlation was found between the level of implicit adaptation and the level of spontaneous recovery ($N = 48$, $r = 0.55$, $t(46)=4.42$, $p < 0.001$).

Given this correlation, it might be that small differences in implicit adaptation level at the end of the learning period can mask age-related effects in spontaneous recovery. That is, if older participants had slightly larger implicit adaptation levels, it could compensate for a decrease in spontaneous recovery. Therefore, we compared spontaneous recovery across age groups while controlling for implicit adaptation levels (*Analysis 6*). Yet, this additional analysis further confirmed our previous result and did not provide any evidence that spontaneous recovery level was smaller in older participants ($F(1,45)=1.317$, $p=0.2571$). If anything, marginal means obtained in the ANCOVA tended to indicate that, when controlling for implicit adaptation levels, older adults tended to exhibit more spontaneous recovery than younger adults (young: $0.88N\pm0.73$; older: 1.009 ± 1.5 , mean \pm SD).

Combining the data of the original study and of the present conceptual replication favor the null hypothesis.

The effect size for the difference in force used between young and older subjects during spontaneous recovery in the study of Trewartha was $d=0.8$ (personal communication from Trewartha). We use this effect size as a prior with Cauchy distribution. In this case, a Bayes Factor (BF_{10}) larger than 1 would favor the effect size found in the original study (favoring the hypothesis that the difference in spontaneous recovery between young and old participants is as big as claimed by Trewartha et al. ($d=0.8$)). A Bayes Factor smaller than 1 would indicate that the effect is smaller than in the original study (favoring H_0). As a sensitivity analysis, we tested different widths for the prior distribution (narrow 0.5, medium: 0.707, wide: 1). In all cases, the posterior was closer to 0 than the prior, with Bayes Factor (BF_{10}) yielding substantial evidence ($BF_{10}>3$, 57) that the difference in spontaneous recovery levels should be smaller than $d=0.8$ ($BF_{01} = 6.89$ for a default prior width = 0.707, $BF_{01} = 7.33$, narrow prior with SD = 0.5; $BF_{01} = 7.3$, wide prior with SD = 1.414). Overall, the hypothesis that the effect of age on spontaneous recovery level is smaller than 0.8 was 6 to 7 times more likely than an effect size of 0.8. In other words, the Bayesian analysis favored the hypothesis that the actual effect of aging on spontaneous recovery was smaller than that of the original study with a median effect size of 0 and a confidence interval of [-0.58, 0.56].

No correlation between explicit adaptation level and working memory capacity score

In the study of Trewartha et al. (2014), spontaneous recovery level were linked to cognitive processes such as explicit memory (their Fig.7). Therefore, we checked whether we could link any aspects of spontaneous recovery to explicit memory processes such as working memory capacity. In our sample, we tested working memory capacity in all our participants except two young adults. Older adults exhibited lower working memory capacity than younger adults (Analysis 8, $t(42.26) = 3.5$, $p = 0.0011$, Cohen's $d = 1.16$, $CI = [0.48, 2]$). Yet, this does not seem to affect the amount of spontaneous recovery as this was similar across age groups (Fig.3). In addition, we did not find any evidence that the amount of spontaneous recovery was correlated with the working memory capacity score (Analysis 9, Fig. 6, $N = 46$, $r = -0.05$, $t(44) = -0.36$, $p = 0.72$). This questions the link between memory processes and spontaneous recovery of motor adaptation.

Discussion

In this study, we tested whether aging influenced the ability to adapt reaching movements accordingly when movements were perturbed. Participants reached to targets while a force field perturbed their movements in an adaptation period. In some catch trials, participants were cued that the force field would be turned off in the subsequent trial (8). Any error in reaching direction in these trials was attributed to implicit adaptation. After a short de-adaptation period with a reversed force field, spontaneous recovery of motor memories of the adaptation period was tested by guiding the hand directly towards the targets in error clamp trials (47). Across age groups, we observed little difference in performance in this task. Both total adaptation and implicit adaptation were not impaired in older adults compared to their younger controls. In addition, we failed to replicate the observation of Trewartha and colleagues (41) and found that spontaneous recovery remained also unaffected by aging. Yet, implicit adaptation and spontaneous recovery levels were correlated independently of age groups, suggesting that spontaneous recovery is linked to the memory of implicit adaptation (45). In contrast to Trewartha et al. (41), we could not find any evidence that spontaneous recovery of motor memories were linked to memory processes.

Our results potentially resolve the contradiction that spontaneous recovery, but not implicit adaptation, was impaired with aging (27, 41). Indeed, the slow process of adaptation is believed to reflect the implicit process (12, 13) and the spontaneous recovery is linked to this slow process (45). It was therefore surprising that some studies found that the implicit component of adaptation was not affected by aging (22, 27, 33) but that the spontaneous recovery was (41), given that they come from the same process (45, 47).

The absence of age-related impairment in spontaneous recovery implies that as we age, we do not get more forgetful of movements in the short-term. Indeed, spontaneous recovery is a measure of short-term retention of the slow implicit process. Following the two-state model (47), the spontaneous recovery results from the rapid decay of the fast state to zero in the error-clamp phase, while the slow process still contains a memory trace of the motor memory acquired during the first adaptation phase. This is consistent with the correlation between the amount of implicit adaptation during learning and the amount of spontaneous recovery (Fig. 5). This is also consistent with the findings of McDougle et al (45). Therefore, a decrease in spontaneous recovery could be due either to worse implicit adaptation during learning (which we did not find) or smaller retention rate (58). The absence of age-related difference in spontaneous recovery suggests that there is no evidence for age-related deficits in either implicit adaptation or retention rate. This finding is in contrast to the results reported by Trewartha et al (41) who found lower spontaneous recovery in older people.

Previous studies that quantified short-term retention in old and young adults gave mixed results. No deficit in short-term retention measured after a one-minute break was reported in a visuomotor rotation task (27). These authors investigated retention of visuomotor adaptation in two different adaptation paradigms. First, one-minute breaks were inserted during regular visuomotor rotation paradigm. In this case, there was no evidence of a difference in retention level of total adaptation between young and old participants. Second, they used one-minute breaks during task-irrelevant clamped feedback paradigm that is known to elicit pure implicit adaptation (1, 13, 59, 60). In this case again, there was no evidence for a deficit in short-term retention of implicit adaptation. However, one other study that measured the explicit component of visuomotor rotation by asking participants to report their aiming direction found that older participants exhibited worse retention of implicit adaptation (58). Yet, it is unclear why people would change the explicit report of their aiming direction in a one-target task after a one-minute break if nothing happened during the break. Similarly, an age-related deficit in the retention did occur in a gait adaptation paradigm (24). These authors suggested that the implicit, and not the explicit component of adaptation was impaired, because larger forgetting was observed in older adults independently of whether a cognitive distraction was presented during the gait adaptation period or not. Such a cognitive distractor would have the ability to reduce the contribution of the explicit component. For this reason, the observed effect was indirectly attributed to the implicit component of adaptation.

Possible sources of discrepancy with the study of Trewartha.

The fact that our results differ from the study of Trewartha et al. (41) might stem from one of the small differences in protocol between our studies even though we tried to use a very similar protocol

395 to theirs. Yet, they differed in several aspects.

396 The experimental design of the forcefield task here used 8 radial targets from a central start position.
397 In contrast, the Trewartha study used alternating movements between two targets, with forces only
398 applied to movements in one direction. The impact of target number on age-related differences in
399 implicit motor adaptation (or absence thereof) remains unknown. Our protocol had a more
400 extensive adaptation period (209 trials vs. 118 in Trewartha et al.), deadadaptation phase (24 vs. 15)
401 and retention phase (63 vs. 22), while the baseline and de-adaptation phases were similar (baseline
402 73 vs. 52 trials and de-adaptation 24 vs 20). The longer adaptation period might have resulted in
403 more opportunity for the participants to learn the force field implicitly, which might have concealed
404 a learning deficit in the older adults that is then later reflected in the spontaneous recovery period.
405 However, Trewartha et al. did not observe any difference in adaptation level during learning.

406 The type of movement and allowed movement speed also differed across the studies. While our
407 participants had to slice through the target, those from Trewartha et al. had to stop on the target.
408 We allowed for a greater variability in hand velocity, allowing faster movements (0.3 – 0.5 m/s vs 0.3-
409 0.4 m/s in the Trewartha et al.). Yet, the hand velocity was matched between our group of young and
410 old participants. Because older adults tended to move slower, we asked a few young participants to
411 perform the experiment while adapting the accepted speed range. On average, our age groups
412 moved with the exact same velocity.

413 One additional difference lies in our sample. Trewartha et al. showed that, within their sample,
414 participants who scored high on an explicit memory task had better spontaneous recovery. So
415 maybe, our sample of older people all had very good cognitive memories. Yet, our sample of older
416 participants had worse working memory capacity than younger participants.

417 Finally, Trewartha measured explicit and implicit components in explicit and implicit memory tasks
418 and compared the results between age groups. Older adults scored less in both the explicit and
419 implicit memory tasks. We measure implicit adaptation within our adaptation paradigm during
420 learning and working memory in a separate task. This test of implicit adaptation could also have
421 influenced the outcomes of the study. We found no difference in implicit level even though older
422 adults had worse working memory capacity.

423 For all these reasons, our study represents a conceptual replication of the study by Trewartha et al.
424 2014 and not a direct/exact replication. If any of the factors identified as differences between our
425 study and that of Trewartha is responsible for the difference in outcomes, it means that the age-
426 related effect on spontaneous recovery, if it exists, is highly sensitive to the experimental conditions.
427 By employing multiple methodologies, conceptual replications provide a robustness test of the

findings. Our study suggests that the generalizability and robustness of the original results should be considered with caution. The age-related difference in spontaneous recovery found by Trewartha and colleagues might be true, but is likely dependent on the experimental conditions.

Do explicit/cognition or implicit adaptation relate to spontaneous recovery

Beyond the technical differences, there are also differences in the theoretical approaches between the two studies. While Trewartha focused on the role of cognition on the spontaneous recovery, we believe that implicit motor adaptation modulates spontaneous recovery. Indeed, Trewartha and colleagues found that people who had “good” explicit memory had higher levels of spontaneous recovery. Our attempt at a conceptual replication of this correlation failed as we did not find any evidence that working memory capacity was linked to spontaneous recovery. The result of Trewartha and colleagues is at odds with the study of Keisler and colleagues (61) who showed that a secondary cognitive task disrupted the fast process but not the slow process responsible for spontaneous recovery (45). They are also incompatible with the finding that patients with early-stage Alzheimer's Disease exhibit a deficit in the fast process of motor adaptation but not in spontaneous recovery or in the slow process (62). Our results rather agree with the results of Keisler than with those of Trewartha. Indeed, we found that the amount of implicit adaptation measured during learning correlated with the level of spontaneous recovery across participants. Future studies could sample older participants based on their cognitive profiles (e.g., selecting older adults with more prominent age-related impairments in working memory or set shifting in order to isolate how specific cognitive attributes affect specific aspects of motor adaptation.

Statistical view on this absence of replication

Our study and the study of Trewartha provide conflicting results. Our Bayesian analyses aimed at reconciling those conflicting results. The Bayesian analysis suggests that, given our data, the influence of age on the spontaneous recovery of motor memories is very likely much smaller than what was reported by Trewartha and colleagues (41). Yet, the Bayesian analysis does not prove that there is no effect. It estimates that the effect size lies somewhere in an interval between -0.6 (medium effect size of larger spontaneous recovery for older people) and 0.55 (medium effect size for a larger spontaneous recovery for younger people).

Yet, beyond such statistical arguments, our results are well aligned with the observation that the spontaneous recovery of motor memories depends on the slow implicit component of motor adaptation (45, 47, 61) and that this component is not affected by aging (20–22, 27, 29, 31, 34, 38, 39). The results of Trewartha and colleagues are at odds with this theory, which motivated our conceptual replication attempt.

Limitations of the study

In this study, we measured the amount of implicit adaptation by asking people to disengage from learning as instructed by a color cue and a warning sign. While other studies have used a similar approach (8, 27), this method assumes that participants can effectively disengage explicit strategies, which we cannot check. In addition, the implicit component of motor adaptation is quantified as the distance participants deviated from the straight trajectory. However, short-term retention was measured by the force that was exerted perpendicular to the heading direction. This difference in units makes direct comparison between the two measures difficult. Another way of separating implicit from explicit learning is described by Schween et al. (14). Implicit adaptation level was determined with 'No Push'- trials, where participants were instructed to ignore the force field and to not push against it. Total adaptation, including the explicit component, was measured in 'Push'-trials, which had an extra reminder to push against the force field. The difference in exerted force is attributed to the explicit component of motor adaptation. The results indicate that in a 'Push'-trial, participants apply more force and in a 'NoPush'-trial less force, as compared to a regular trial. Therefore, our study could have benefited from such an assessment.

Our Bayesian analysis suggests that the maximum effect size should be much smaller than anticipated based on the study of Trewartha. This means that we only have 60% power to detect an effect if there is one of $d=0.6$. This should motivate future studies to include more participants as we now have a better estimate of the possible effect size range.

Finally, the group of older participants exhibited much more inter-subject variability than the group of younger participants. This is typical in aging studies but would need to be tackled to get a better estimate of the spontaneous recovery of these older participants.

Conclusion

We attempted a conceptual replication of the effect of age on spontaneous recovery as demonstrated by Trewartha and colleagues but could not replicate their results as we failed to find evidence for a difference in spontaneous recovery between young and old participants. The current results are more in line with the idea that spontaneous recovery depends on the retention of implicit adaptation and that implicit adaptation is not affected by aging.

References

1. **Morehead JR, Orban de Xivry J-J.** A Synthesis of the Many Errors and Learning Processes of Visuomotor Adaptation. .

- 493 2. **Shadmehr R, Smith MA, Krakauer JW.** Error correction, sensory prediction, and adaptation in
494 motor control. *Annu Rev Neurosci* 33: 89–108, 2010. doi: 10.1146/annurev-neuro-060909-
495 153135.
- 496 3. **Krakauer JW, Ghez C, Ghilardi MF.** Adaptation to visuomotor transformations: consolidation,
497 interference, and forgetting. *J Neurosci* 25: 473–8, 2005. doi: 10.1523/JNEUROSCI.4218-
498 04.2005.
- 499 4. **Orban de Xivry J-J, Lefèvre P.** Formation of model-free motor memories during motor
500 adaptation depends on perturbation schedule. *J Neurophysiol* 113: 2733–41, 2015. doi:
501 10.1152/jn.00673.2014.
- 502 5. **Welch RB.** Adaptation to prism-displaced vision: The importance of target-pointing. *Percept*
503 *Psychophys* 5: 305–309, 1969. doi: 10.3758/BF03209569.
- 504 6. **Lackner J, DiZio P.** Rapid adaptation to Coriolis force perturbations of arm trajectory. *J*
505 *Neurophysiol* 72: 299, 1994.
- 506 7. **Shadmehr R, Mussa-Ivaldi F a.** Adaptive representation of dynamics during learning of a motor
507 task. *J Neurosci* 14: 3208–24, 1994.
- 508 8. **Morehead JR, Qasim SEE, Crossley MJJ, Ivry RB.** Savings upon Re-Aiming in Visuomotor
509 Adaptation. *J Neurosci* 35: 14386–14396, 2015. doi: 10.1523/JNEUROSCI.1046-15.2015.
- 510 9. **Taylor JA, Krakauer JW, Ivry RB.** Explicit and implicit contributions to learning in a sensorimotor
511 adaptation task. *J Neurosci* 34: 3023–3032, 2014. doi: 10.1523/JNEUROSCI.3619-13.2014.
- 512 10. **Taylor JA, Ivry RB.** Flexible Cognitive Strategies during Motor Learning. *PLoS Comput Biol* 7:
513 e1001096, 2011. doi: 10.1371/journal.pcbi.1001096.
- 514 11. **Welch RB, Choe CS, Heinrich DR.** Evidence for a three-component model of prism adaptation. *J*
515 *Exp Psychol* 103: 700–5, 1974.
- 516 12. **Mazzoni P, Krakauer JW.** An implicit plan overrides an explicit strategy during visuomotor
517 adaptation. *J Neurosci* 26: 3642–5, 2006. doi: 10.1523/JNEUROSCI.5317-05.2006.
- 518 13. **Morehead JR, Taylor JA, Parvin D, Ivry RB.** Characteristics of Implicit Sensorimotor Adaptation
519 Revealed by Task-irrelevant Clamped Feedback. *J Cogn Neurosci* 26: 1–14, 2017. doi:
520 10.1162/jocn_a_01108.
- 521 14. **Schween R, McDougle SD, Hegele M, Taylor JA.** Assessing explicit strategies in force field
522 adaptation. *J Neurophysiol* 123: 1552–1565, 2020. doi: 10.1152/jn.00427.2019.
- 523 15. **Aucie Y, Harket HM, Sombric CJ, Torres-Oviedo G.** Older adults generalize their movements
524 across walking contexts more than young during gradual and abrupt split-belt walking. .
- 525 16. **Bakkum A, Gunn SM, Marigold DS.** How aging affects visuomotor adaptation and retention in a
526 precision walking paradigm. *Sci Rep* 11: 789, 2021. doi: 10.1038/s41598-020-80916-8.
- 527 17. **Bock O.** Components of sensorimotor adaptation in young and elderly subjects. *Exp Brain Res*
528 160: 259–63, 2005. doi: 10.1007/s00221-004-2133-5.

- 529 18. **Buch ER, Young S, Contreras-Vidal JL.** Visuomotor adaptation in normal aging. *Learn Mem* 10:
530 55–63, 2003. doi: 10.1101/lm.50303.
- 531 19. **Cisneros E, Karny S, Ivry RB, Tsay JS.** Differential Aging Effects on Implicit and Explicit
532 Sensorimotor Learning. *bioRxiv*: 2024.07.02.601091, 2024.
- 533 20. **Cressman EK, Salomonczyk D, Henriques DYP.** Visuomotor adaptation and proprioceptive
534 recalibration in older adults. *Exp Brain Res* 205: 533–544, 2010. doi: 10.1007/s00221-010-2392-
535 2.
- 536 21. **Hegele M, Heuer H.** Adaptation to a direction-dependent visuomotor gain in the young and
537 elderly. *Psychol Res* 74: 21–34, 2010. doi: 10.1007/s00426-008-0221-z.
- 538 22. **Heuer H, Hegele M.** Adaptation to visuomotor rotations in younger and older adults. *Psychol*
539 *Aging* 23: 190–202, 2008. doi: 10.1037/0882-7974.23.1.190.
- 540 23. **Li N, Chen G, Xie Y, Chen Z.** Aging Effect on Visuomotor Adaptation: Mediated by Cognitive
541 Decline. *Front Aging Neurosci* 13: 742928, 2021. doi: 10.3389/fnagi.2021.742928.
- 542 24. **Malone LA, Bastian AJ.** Age-related forgetting in locomotor adaptation. *Neurobiol Learn Mem* ,
543 2015. doi: 10.1016/j.nlm.2015.11.003.
- 544 25. **Seidler RD.** Differential effects of age on sequence learning and sensorimotor adaptation. *Brain*
545 *Res Bull* 70: 337–46, 2006. doi: 10.1016/j.brainresbull.2006.06.008.
- 546 26. **Sombric CJ, Torres-Oviedo G.** Cognitive and Motor Perseveration Are Associated in Older
547 Adults. *Front Aging Neurosci* 13: 183, 2021. doi: 10.3389/fnagi.2021.610359.
- 548 27. **Vandevoorde K, Orban de Xivry JJ.** Internal model recalibration does not deteriorate with age
549 while motor adaptation does. *Neurobiol Aging* 80: 138–153, 2019. doi:
550 10.1016/j.neurobiolaging.2019.03.020.
- 551 28. **Vandevoorde K, Orban de Xivry J-J.** Why is the explicit component of motor adaptation limited
552 in elderly adults? *J Neurophysiol* 124: 152–167, 2020. doi: 10.1152/jn.00659.2019.
- 553 29. **Vandevoorde K, Orban de Xivry J-J.** Does proprioceptive acuity influence the extent of implicit
554 sensorimotor adaptation in young and older adults? .
- 555 30. **Bock O, Girgenrath M.** Relationship between sensorimotor adaptation and cognitive functions
556 in younger and older subjects. *Exp Brain Res* 169: 400–406, 2006. doi: 10.1007/s00221-005-
557 0153-4.
- 558 31. **Hegele M, Heuer H.** Age-related variations of visuomotor adaptation result from both the
559 acquisition and the application of explicit knowledge. *Psychol Aging* 28: 333–9, 2013. doi:
560 10.1037/a0031914.
- 561 32. **Wolpe N, Ingram JN, Tsvetanov KA, Henson RN, Wolpert DM, Cam-CAN, Rowe JB.** Age-related
562 reduction in motor adaptation: brain structural correlates and the role of explicit memory.
563 *Neurobiol Aging* 90: 13–23, 2020. doi: 10.1016/j.neurobiolaging.2020.02.016.
- 564 33. **Huang J, Gegenfurtner KR, Schütz AC, Billino J.** Age effects on saccadic adaptation: Evidence
565 from different paradigms reveals specific vulnerabilities. *J Vis* 17: 9, 2017. doi: 10.1167/17.6.9.

- 566 34. **Reuter E-M, Leow L-A, Carroll TJ.** Task feedback processing differs between young and older
567 adults in visuomotor rotation learning despite similar initial adaptation and savings. .
- 568 35. **Tsay JS, Asmerian H, Germine LT, Wilmer J, Ivry RB, Nakayama K.** Predictors of sensorimotor
569 adaption: insights from over 100,000 reaches. *bioRxiv*: 2023.01.18.524634, 2023.
- 570 36. **Vachon CM, Modchalingam S, Hart BM 't, Henriques DYP.** The effect of age on visuomotor
571 learning processes. *PLOS ONE* 15: e0239032, 2020. doi: 10.1371/journal.pone.0239032.
- 572 37. **Cesqui B, Macrì G, Dario P, Micera S.** Characterization of age-related modifications of upper
573 limb motor control strategies in a new dynamic environment. *J NeuroEngineering Rehabil* 5: 31,
574 2008. doi: 10.1186/1743-0003-5-31.
- 575 38. **Huang HJ, Ahmed AA.** Older adults learn less, but still reduce metabolic cost, during motor
576 adaptation. *J Neurophysiol* 111: 135–144, 2014. doi: 10.1152/jn.00401.2013.
- 577 39. **Kitchen NM, Miall RC.** Adaptation of reach action to a novel force-field is not predicted by
578 acuity of dynamic proprioception in either older or younger adults. *Exp Brain Res* 239: 557–574,
579 2021. doi: 10.1007/s00221-020-05997-3.
- 580 40. **Reuter E-M, Pearcey GEP, Carroll TJ.** Greater neural responses to trajectory errors are
581 associated with superior force field adaptation in older adults. *Exp Gerontol* : #pagerange#,
582 2018. doi: 10.1016/j.exger.2018.05.020.
- 583 41. **Trewartha KM, Garcia A, Wolpert DM, Flanagan JR.** Fast But Fleeting: Adaptive Motor Learning
584 Processes Associated with Aging and Cognitive Decline. *J Neurosci* 34: 13411–13421, 2014. doi:
585 10.1523/JNEUROSCI.1489-14.2014.
- 586 42. **Coltman SK, Cashaback JGA, Gribble PL.** Both fast and slow learning processes contribute to
587 savings following sensorimotor adaptation. *J Neurophysiol* 121: 1575–1583, 2019. doi:
588 10.1152/jn.00794.2018.
- 589 43. **Ethier V, Zee DS, Shadmehr R.** Spontaneous recovery of motor memory during saccade
590 adaptation. *J Neurophysiol* 99: 2577–83, 2008. doi: 10.1152/jn.00015.2008.
- 591 44. **Kojima Y, Iwamoto Y, Yoshida K.** Memory of learning facilitates saccadic adaptation in the
592 monkey. *J Neurosci* 24: 7531–9, 2004. doi: 10.1523/JNEUROSCI.1741-04.2004.
- 593 45. **McDougle SD, Bond KM, Taylor JA.** Explicit and Implicit Processes Constitute the Fast and Slow
594 Processes of Sensorimotor Learning. *J Neurosci* 35: 9568–9579, 2015. doi:
595 10.1523/JNEUROSCI.5061-14.2015.
- 596 46. **Sarwary AME, Wischniewski M, Schutter DJLG, Selen LPJ, Medendorp WP.** Corticospinal
597 correlates of fast and slow adaptive processes in motor learning. *J Neurophysiol* 120: 2011–
598 2019, 2018. doi: 10.1152/jn.00488.2018.
- 599 47. **Smith MA, Ghazizadeh A, Shadmehr R.** Interacting Adaptive Processes with Different
600 Timescales Underlie Short-Term Motor Learning. *PLoS Biol* 4: 1035–1043, 2006. doi:
601 10.1371/journal.pbio.0040179.
- 602 48. **Kording KP, Tenenbaum JB, Shadmehr R.** The dynamics of memory as a consequence of
603 optimal adaptation to a changing body. *Nat Neurosci* 10: 779–786, 2007. doi: 10.1038/nn1901.

- 604 49. **Oldfield RC.** The assessment and analysis of handedness: The Edinburgh inventory.
605 *Neuropsychologia* 9: 97–113, 1971. doi: 10.1016/0028-3932(71)90067-4.
- 606 50. **Folstein MF, Folstein SE, Mchugh PR.** "MINI-MENTAL STATE" A PRACTICAL
607 METHOD FOR GRADING THE COGNITIVE STATE OF PATIENTS FOR THE CLINICIAN*. Pergamon
608 Press, 1975.
- 609 51. **Scheidt RA, Reinkensmeyer DJ, Conditt MA, Rymer WZ, Mussa-Ivaldi F a.** Persistence of motor
610 adaptation during constrained, multi-joint, arm movements. *J Neurophysiol* 84: 853–62, 2000.
- 611 52. **Christou AI, Miall RC, McNab F, Galea JM.** Individual differences in explicit and implicit
612 visuomotor learning and working memory capacity. *Sci Rep* 6, 2016. doi: 10.1038/srep36633.
- 613 53. **Saenen L, Verheyden G, Orban de Xivry J-J.** The differential effect of age on upper limb sensory
614 processing, proprioception and motor function. .
- 615 54. **Vogel EK, McCollough AW, Machizawa MG.** Neural measures reveal individual differences in
616 controlling access to working memory. *Nature* 438: 500–503, 2005. doi: 10.1038/nature04171.
- 617 55. **JASP Team.** JASP (version 0.17.2)[Computer software] [Online]. 2023. <https://jasp-stats.org/>.
- 618 56. **Makowski D, Ben-Shachar MS, Patil I, Lüdtke D.** Methods and Algorithms for Correlation
619 Analysis in R. *J Open Source Softw* 5: 2306, 2020. doi: 10.21105/joss.02306.
- 620 57. **Dienes Z.** Using Bayes to get the most out of non-significant results. *Front Psychol* 5: 1–17,
621 2014. doi: 10.3389/fpsyg.2014.00781.
- 622 58. **Bindra G, Brower R, North R, Zhou W, Joiner WM.** Normal Aging Affects the Short-Term
623 Temporal Stability of Implicit, But Not Explicit, Motor Learning following Visuomotor
624 Adaptation. *eNeuro* 8, 2021. doi: 10.1523/ENEURO.0527-20.2021.
- 625 59. **Avraham G, Morehead JR, Kim HE, Ivry RB.** Reexposure to a sensorimotor perturbation
626 produces opposite effects on explicit and implicit learning processes. *PLOS Biol* 19: e3001147,
627 2021. doi: 10.1371/journal.pbio.3001147.
- 628 60. **Kim HE, Parvin DE, Ivry RB.** The influence of task outcome on implicit motor learning. *eLife* 8:
629 e39882, 2019. doi: 10.7554/eLife.39882.
- 630 61. **Keisler A, Shadmehr R.** A Shared Resource between Declarative Memory and Motor Memory. *J*
631 *Neurosci* 30: 14817–14823, 2010. doi: 10.1523/JNEUROSCI.4160-10.2010.
- 632 62. **Sutter K, Wijdenes LO, Beers RJ van, Claassen JAHR, Kessels RPC, Medendorp WP.** Early-Stage
633 Alzheimer’s Disease Affects Fast But Not Slow Adaptive Processes in Motor Learning. *eNeuro*
634 11, 2024. doi: 10.1523/ENEURO.0108-24.2024.

635

636

637 Legends

638 *Figure 1: Paradigm. A change in cursor color indicated presence (cued) or absence (uncued) of a force field. Interspersed*
639 *throughout the baseline and perturbation phase, were error clamp trials. Eight targets (open circles) were displayed, of*
640 *which only four (filled black circles) were used for the uncued trials during the perturbation phase and error clamp trials*
641 *during baseline and perturbation phase.*

642 *Figure 2: Total adaptation level did not decline with aging. A Lateral deviation from the optimal trajectory at peak velocity*
643 *for young (blue) and older adults (orange) over the course of baseline and perturbation phase. Interspersed with these*
644 *perturbation trials were uncued baseline trials (filled circles) where no perturbation was applied. B Exerted force*
645 *perpendicular to heading direction at peak velocity during baseline and perturbation phase for young (blue) and older adults*
646 *(orange). C Lateral deviation during the last 80 field trials from the perturbation phase and the corresponding hand velocity.*
647 *D Lateral deviation during the last 12 uncued trials from the perturbation phase and the corresponding hand velocity. E.*
648 *Exerted force during the last 12 error clamp trials from the perturbation phase and the corresponding hand velocity. For*
649 *panels C, D and E, each dot represents the mean data from one individual. Error bar represents mean and standard error.*
650 *For all panels, data from 28 young and 20 older participants are presented.*

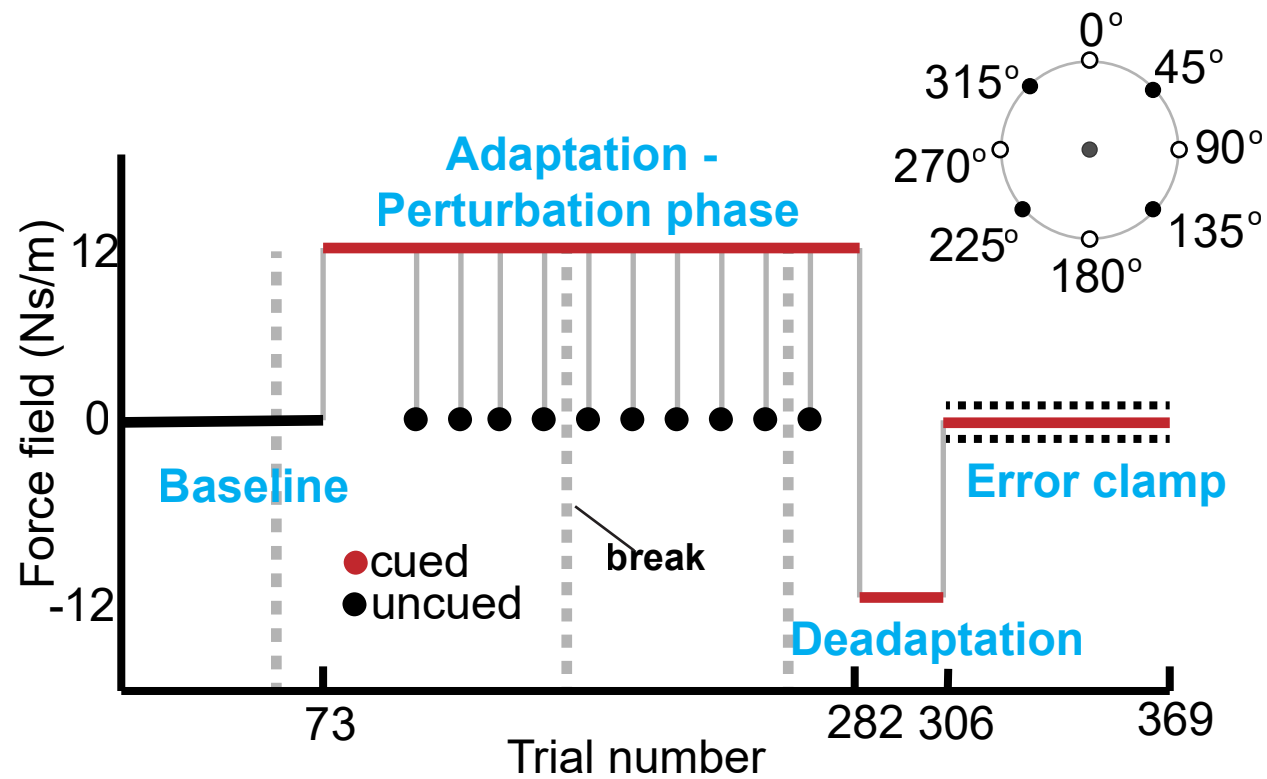
651 *Figure 3: Spontaneous recovery did not decline with aging. A Exerted force at peak velocity during the error-clamp phase for*
652 *young (blue) and older adults (orange). Each dot represents the average force exerted by the individuals from an age group*
653 *for a single trial. B Exerted force during the last 48 trials of the error-clamp phase. C Adaptation index for the last 48 trials of*
654 *the error-clamp phase. For panels B and C, each dot represents the mean data from one individual. Error bar represents*
655 *mean and standard error. For all panels, data from 28 young and 20 older participants are presented.*

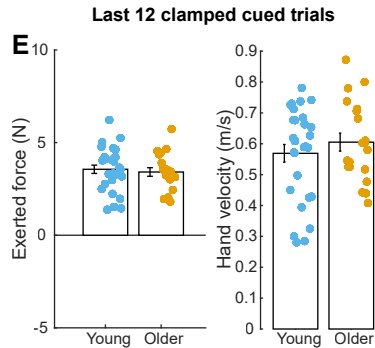
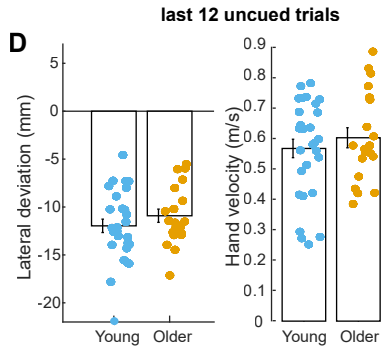
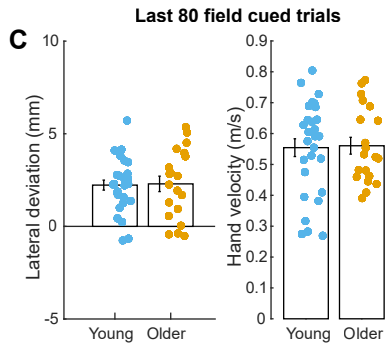
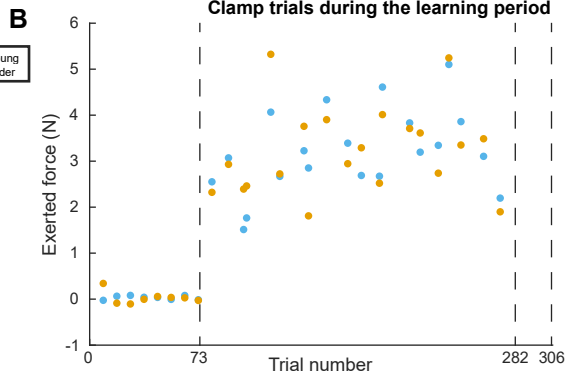
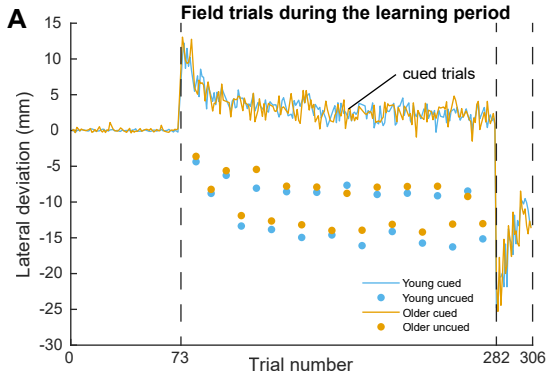
656 *Figure 4: Implicit adaptation (data from figure 2.D) and spontaneous recovery (data from figure 3.B) are correlated (N = 48).*
657 *To facilitate interpretation, implicit level was converted to positive values, such that participants with a higher implicit*
658 *adaptation level have a larger lateral deviation. Each dot represents the data from one individual. Data from both groups*
659 *were combined thanks to multilevel correlation. Regression line was obtained with robustfit method in Matlab.*

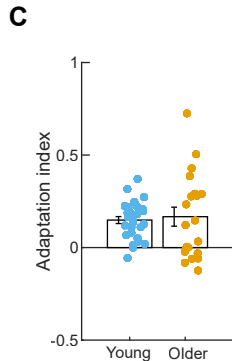
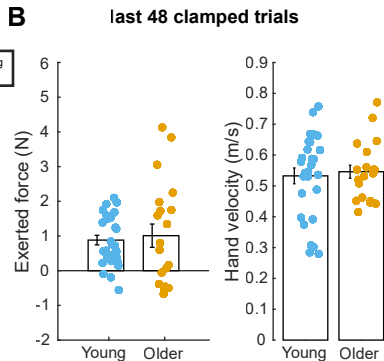
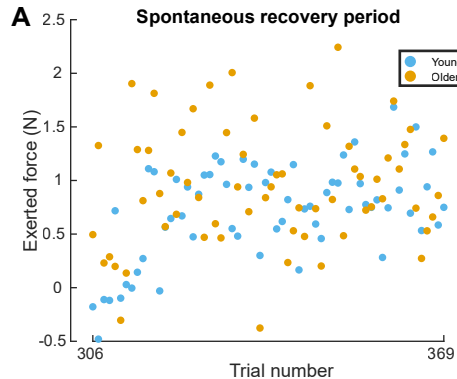
660 *Figure 5: Output of the Bayesian analysis. The Bayesian analysis takes the previous data as the prior (centered on $d=0.8$,*
661 *Trevartha et al. 2014) and computes the posterior based on the data of the present study.*

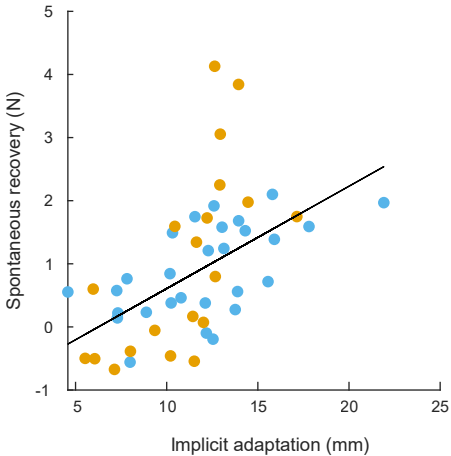
662 *Figure 6: Spontaneous recovery level was not correlated with spatial working memory capacity score (N = 46). Each dot*
663 *represents the data from one individual. Data from both groups were combined thanks to multilevel correlation. Regression*
664 *line was obtained with robustfit method in Matlab.*

665



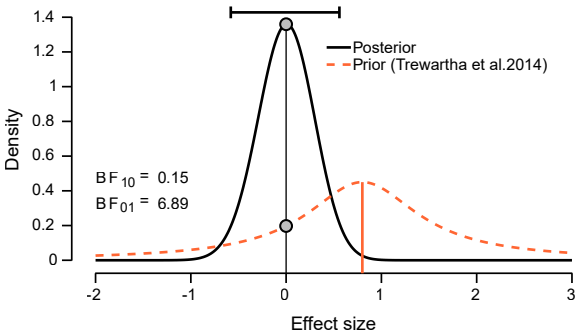


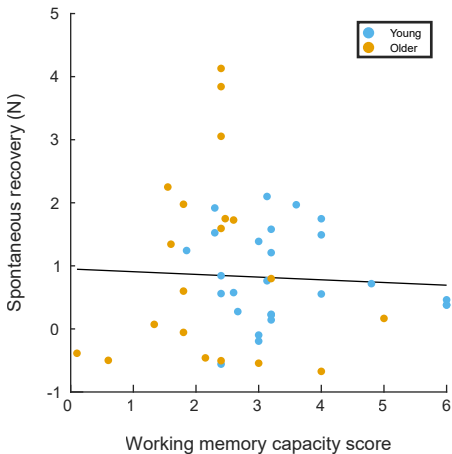




Median: 0

95% CI: [-0.581, 0.559]





Limited influence of aging on implicit adaptation and its short-term retention.

