

Supplementary Materials for
Dissecting motor skill acquisition: Spatial coordinates take precedence

Pablo Maceira-Elvira *et al.*

Corresponding author: Friedhelm C. Hummel, friedhelm.hummel@epfl.ch

Sci. Adv. **8**, eabo3505 (2022)
DOI: 10.1126/sciadv.abo3505

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Supplementary Text

Scoring of the behavioral task

We instructed participants to execute the sequence as fast and as accurately as possible, so we tried to score performance in a way that would consider both aspects. Dan and colleagues (77) proposed a measure multiplying the exponential of speed times the exponential of accuracy:

$$PI = \exp\left(-\left(\frac{\mathbf{BlockDuration}}{\mathbf{12}}\right)\right) * \exp\left(-\left(\frac{\mathbf{Errors}}{\mathbf{12}}\right)\right) * \mathbf{100}$$

In their equation, the “12” was the maximum number of correct sequences possible in their task per block, and Errors = 12 - CorrectSequences. We tried using this equation. In our case, it was the time that was fixed (*i.e.* BlockDuration) and the total number of sequences per block varied. We noticed that the equation weighted errors less severely if made at higher speeds, while weighting participants making mistakes at lower speeds more harshly. As data came from people from multiple age groups, some of them moving more slowly (*e.g.*, older adults) we chose not to apply this asymmetric weighting. Additionally, we noticed that the time constant used greatly changed the shape of the learning curve (*e.g.*, using either 90 seconds or 1.5 minutes).

We thought about a hypothetical case in which two persons generated five correct sequences, with the difference that the first person generated a total of 10 sequences (50% accuracy) and the second a total of 5 sequences (100% accuracy). Based on what we instructed in terms of speed and accuracy, we wanted to give higher credit to the person having higher accuracy. Directly multiplying speed and accuracy, namely 10*0.5 for the first and 5*1 for the second person, would yield scores of 5 for both. Therefore, we chose to multiply the number of correct sequences times the accuracy, which would yield 5*0.5 = 2.5 for the first and 5*1 = 5 for the second, placing the desired prime on both parameters. The resulting equation is:

$$\mathbf{score} = \mathbf{CorrectSequences} * \mathbf{PercentCorrect}$$

Correction for individual skill level at the beginning of training

We used a single baseline block as a benchmark (containing an independent sequence, the same for all subjects) for evaluating the initial motor capabilities of each subject. We do not use this block to normalize the performance during training under stimulation, for several reasons. First, it is a different sequence from the sequence people train on, and even if it is similar in structure and equivalent in complexity, previous research shows there is no transfer between different sequences (6), which we also see when comparing the training blocks to the catch blocks in this study (please see *Table S1*). Second, the training kicks in from the very first execution of a sequence (78). Third, the performance averaged over the whole block might

benefit already from the stimulation, so penalizing it with a different sequence executed under different circumstances (i.e., without stimulation) does not seem adequate.

We used the baseline block as reference for initial skill levels. In the first experiment, we found significant differences in the scores of this block between age groups. Nevertheless, we found no significant differences in accuracy (please see *Table S1*). In the second experiment (placebo groups), we found no significant differences in neither the scores nor the accuracy in the baseline block, further supporting the notion of participants starting at the same skill level (please see *Table S2*).

As we confirmed all participants were healthy, we assumed all age groups could improve within the same range of values, allowing us to compare them within the same frame of reference. For this reason, we corrected individual performance by subtracting the score of the first training block. As participants' improvement is in the same order of magnitude, we consider absolute improvement to be an adequate measure.

Catch block scores

In the main manuscript, we mention that we see no generalization of learning to sequences different from the training sequence. The scores below show average scores of all blocks, including the catch blocks (please see *Figure S1*).

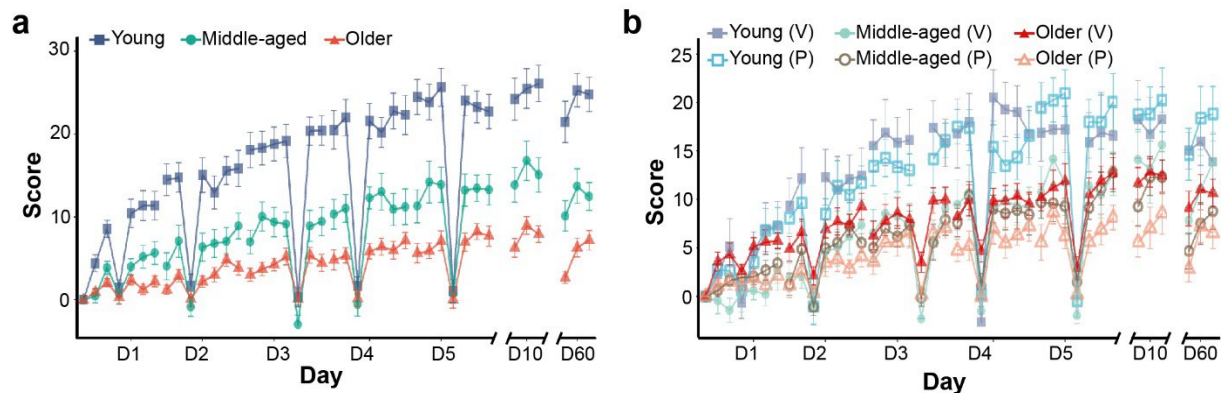


Figure S1. Primary outcome of all blocks (i.e. including catch blocks). **a)** Average score for all blocks of the first experiment (i.e. training without stimulation). The vertical lines represent the standard error of the mean. **b)** Average scores for all blocks in the second experiment (i.e. training with either verum (V) or placebo (P) stimulation), with vertical bars describing the standard error of the mean. The error bars depict the standard error of the mean.

It is clear to see that the improvement seen in the training sequence does not transfer to other sequences. Scores were not significantly different between catch blocks (*Table S1 and Table S2*).

Speed and accuracy

In the main article, we only show the normalized speed and accuracy for each group. The reason is that we were interested in discussing the dynamics of both parameters, and intended to show both processes in the same plot. Here, we show the average speed and accuracy for each group. We describe the speed as the total number of sequences generated within a block (*Figure S2*).

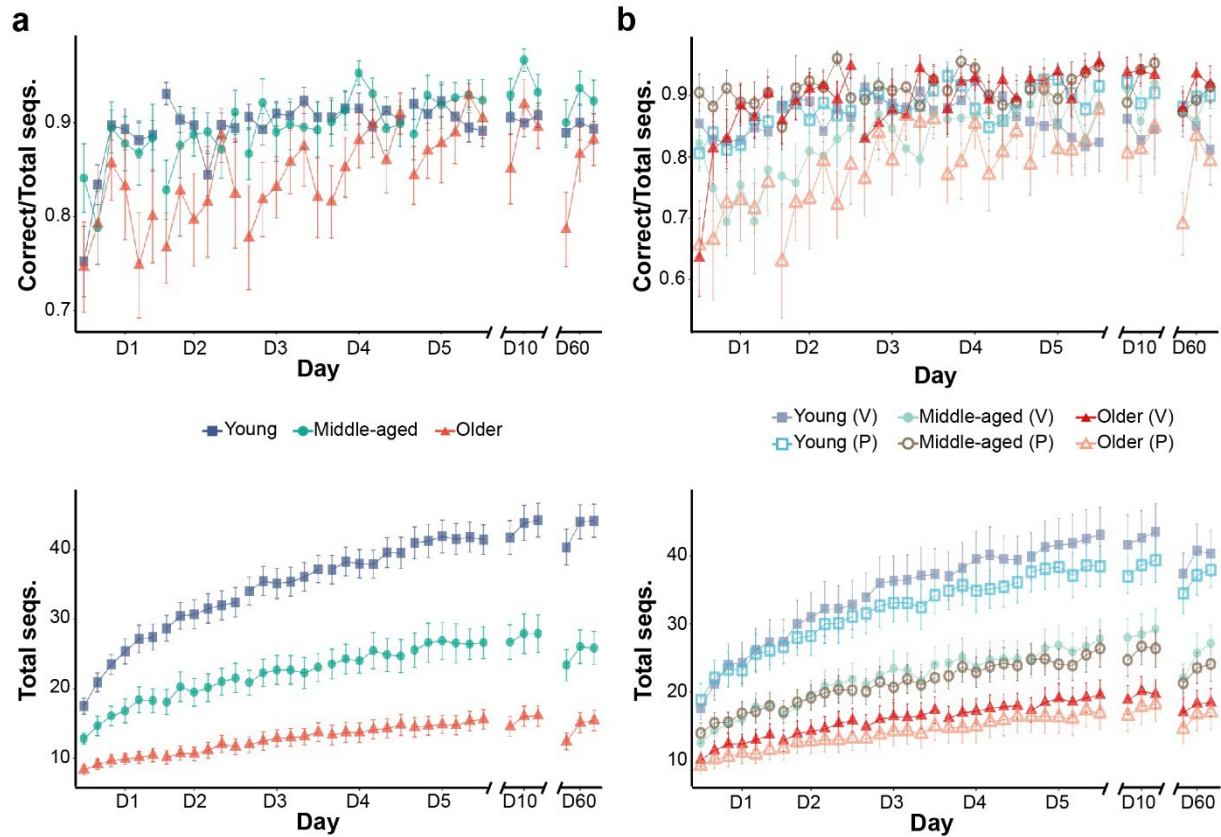


Figure S2. Group average accuracy and speed. Accuracy (top) and speed (bottom) in the first (a) and second (b) experiments. We defined speeds to be the total number of correct sequences generated within a block, while accuracy corresponds to the ratio of correct sequences to total sequences in each block. The error bars depict the standard error of the mean. V: Verum; P: Placebo.

Uncorrected scores

In the main manuscript, we present centered average scores for all groups of both experiments, which was necessary to compare the scores statistically. *Figure S3* shows the uncentered scores. Interestingly, the unstimulated groups and the groups receiving placebo stimulation look almost exactly the same, and the absence of an effect of atDCS in young and middle-aged adults is more evident.

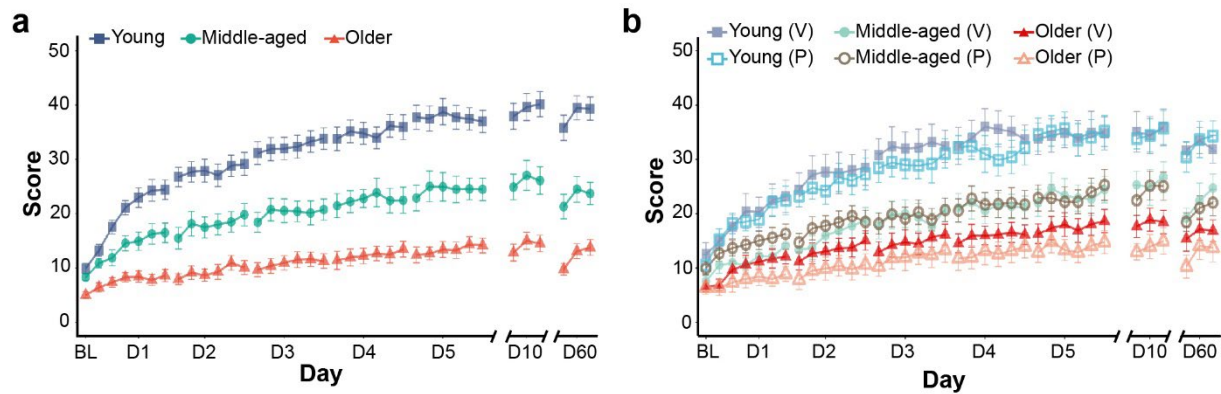


Figure S3. Uncorrected scores for both experiments. The scores obtained by the three groups of the first experiment (a) match those obtained by the groups receiving placebo stimulation in the second experiment (b). Please notice that without the correction, the verum and placebo groups of both young and middle-aged adults are even more similar, highlighting the absence of an effect reported in the main manuscript. In contrast, older adults receiving verum stimulation outscore the unstimulated and the placebo groups of older adults. V: Verum; P: Placebo.

Motor chunk estimation from behavioral data

A common approach at extracting chunking patterns in the execution of discrete sequence production tasks is to average the inter-key intervals (IKIs) and detect statistically significant increases, assumed to separate adjacent chunks (7, 9, 40). Acuna and colleagues (79) implemented a probabilistic approach meant to estimate the likelihood of an individual performing a certain chunking pattern based on the IKIs and the errors generated during the execution of the task, as well as the correlation of these two components. The model they propose uses expectation maximization to estimate the most likely chunking pattern generated on each trial (*i.e.* sequence execution). For each trial (t), the model uses prior knowledge on the chunking pattern executed in the preceding trial ($t-1$). This approach stands on an assumption that makes its use difficult in our dataset. The model assumes that improvement is steady and consistent during training, with similar chunking patterns between adjacent trials. This could be the case in young adults, with average scores monotonically increasing between days. In contrast, middle-aged and older adults often show diminished performance at the beginning of a training session with respect to the previous day. On the other hand, the use of a catch block presented in the middle of training sessions often disrupted performance in the subsequent training block (as reported by several participants and as observed in the “dips” in the learning curves of each day, *Figure S1*), which probably had an impact on the executed chunking pattern. Additionally, errors in older adults, more frequent than in young, would likely obstruct finding the chunking pattern implemented by a participant, as errors cause participants to slow down (80). Furthermore, the probabilistic approach enforces the notion of chunks eventually being fully concatenated by the end of training, while chunk formation is likely constrained by the computational cost of retrieving a certain amount of sequence elements (41).

For these reasons, we decided to use the approach proposed by Song & Cohen (70). In their method, chunking strategies are detected by applying a k-means clustering algorithm to the IKIs, forcing two clusters to label IKIs as either “fast” or “slow”.

Each sequence had nine IKIs, with the first one reflecting the interval between the last key press of the previous sequence and the first key press of the current sequence. After removing incorrect sequences from each block, we normalized the IKIs of each sequence to the total duration the sequence (*i.e.* divided each IKI by its sequence duration), to account for the gradual increase in speed during training. After normalization, we applied the K-means clustering algorithm to sequences of each block (Sklearn, <https://scikit-learn.org/>) enforcing the notion of two clusters being present (*i.e.*, “fast” and “slow”), labeling the IKIs of each sequence

based on their proximity to them (*Figure S4a*). The outcome of this step was a chunking pattern for each individual sequence (*Figure S4b*).

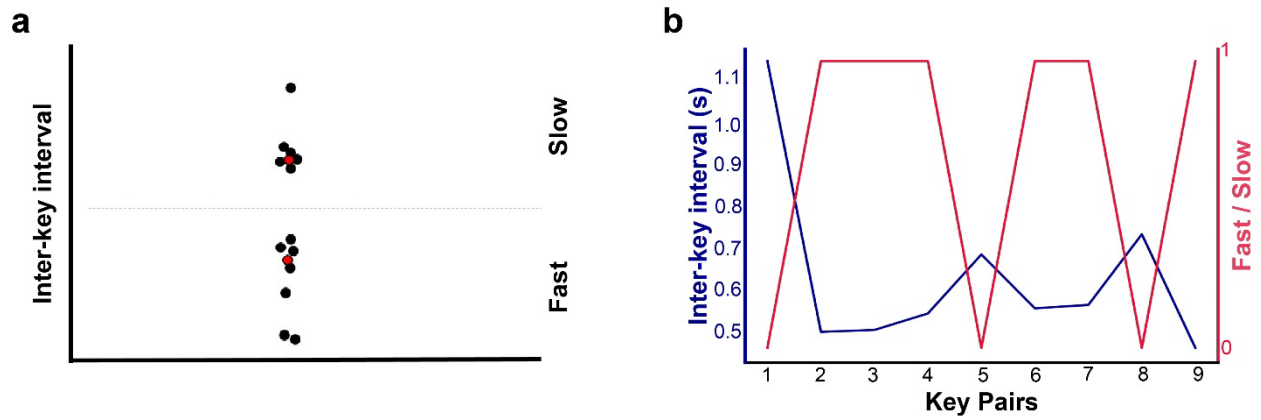


Figure S4. Extraction of chunking patterns from inter-key intervals (IKIs). *a* We applied *k*-means clustering on IKIs from each block, defining two cluster centroids (i.e. “fast” and “slow”). *b* After estimating the centroids, we labelled each IKI of the block as either “1” (i.e. fast) or “0” (i.e. slow), interpreting adjacent “1”s as intervals belonging to the same motor chunk.

Labelling each sequence in this way results in many different patterns. *Figure S5a* shows an example of such variability, in which each histogram bin corresponds to a different sequence pattern. As sequences are binary (i.e. consist of ones and zeros), we converted them to decimal to represent them as a single number.

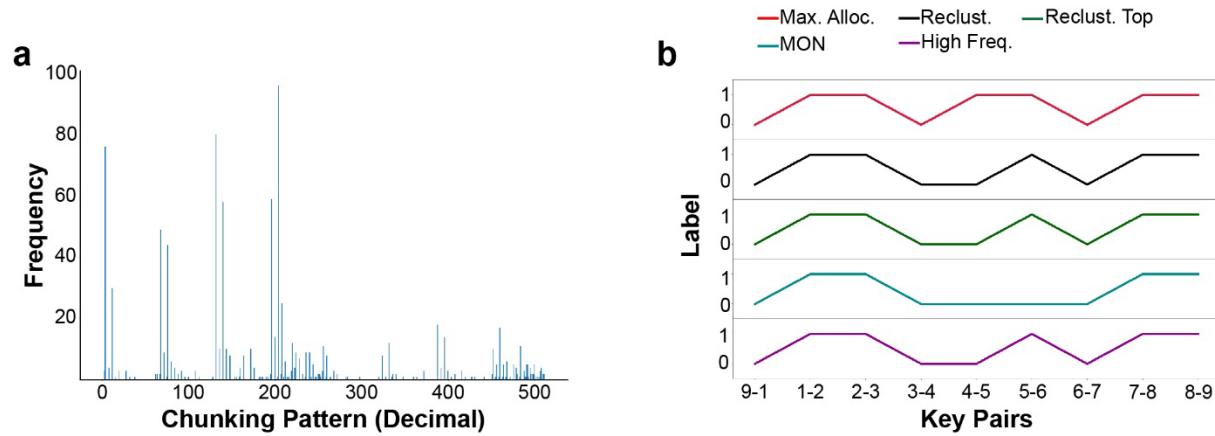


Figure S5. Chunking patterns generated by a single subject. a) Binary chunking patterns (e.g. [011011011]) are represented in decimal form (e.g. to $(0)2^8+(1)2^7+(1)2^6+(0)2^5+(1)2^4+(1)2^3+(0)2^2+(1)2^1+(1)2^0=219$). **b)** Output from each criterion for pattern selection. The criteria are called, from top to bottom: “Maximum allocation”, “Reclustering”, “Reclustering top”, “More-often-than-not” and “Highest frequency”. Please refer to the Methods section in the main article for further details.

Because each participant generates multiple sequences, we defined the criteria described in the main text to determine a single pattern for each participant, on each day of training. *Figure S5b* shows the output from the five criteria for a single participant on the first day of training. As some patterns are slightly different from one criterion to the next, we perform a majority vote using the five criteria. In this example, the resulting pattern would be [0 1 1 0 0 1 0 1 1].

SICI measurements

In the main text, we mention that we used TMS to measure intra-cortical inhibition at rest in all participants of the second experiment. We applied the SICI paradigm before and after the first training day, to quantify the interneuronal GABAergic inhibition within the primary motor cortex, directly involved in the learning and execution of the motor sequence. We repeated these measurements on the fifth and sixtieth days. We show the group results in *Figure S6*.

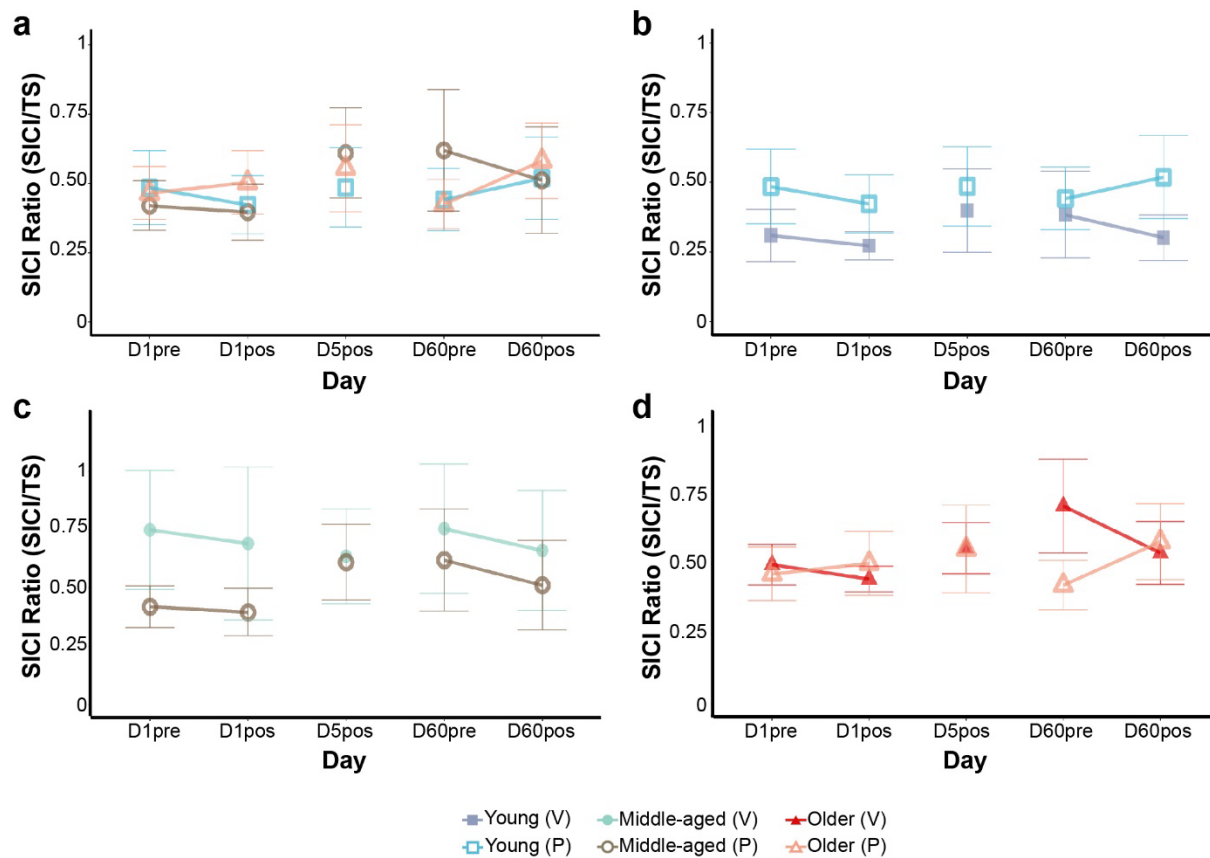


Figure S6. SICI measurements at rest in participants of the second experiment. Inhibitory efficiency, quantified as the ratio of the amplitude resulting from the short-interval intracortical inhibition (SICI) paradigm divided by the amplitude of the test stimulus (TS). Each marker corresponds to the average ratio per group, with its corresponding standard error. **a)** SICI at rest in all groups receiving placebo stimulation. Panels **b** to **d** show SICI measurements at rest in young (**b**), middle-aged (**c**) and older (**d**) adults receiving verum (V) and placebo (P) stimulation. Recordings were performed either before (i.e., “pre”) or after (i.e., “pos”) performing the finger-tapping task.

Self-reports on attention and fatigue

We used a visual analogue scale (VAS) to inquire about the state of attention and fatigue in all participants before and after each training block. This scale is presented as an ungraded line, with attention ranging from “completely attentive” (0) to “completely inattentive” (10), and fatigue ranging from “awake” (0) to “tired” (10).

In *Figure S7* we show the scores reported by the participants over the course of training, averaged per age and stimulation group, with the error bars representing the standard error of the mean. *Figure S8* shows the difference in attention and fatigue between the last block of a training day and the first block of the following training day, plotted against the score difference between the same blocks (i.e., offline learning).

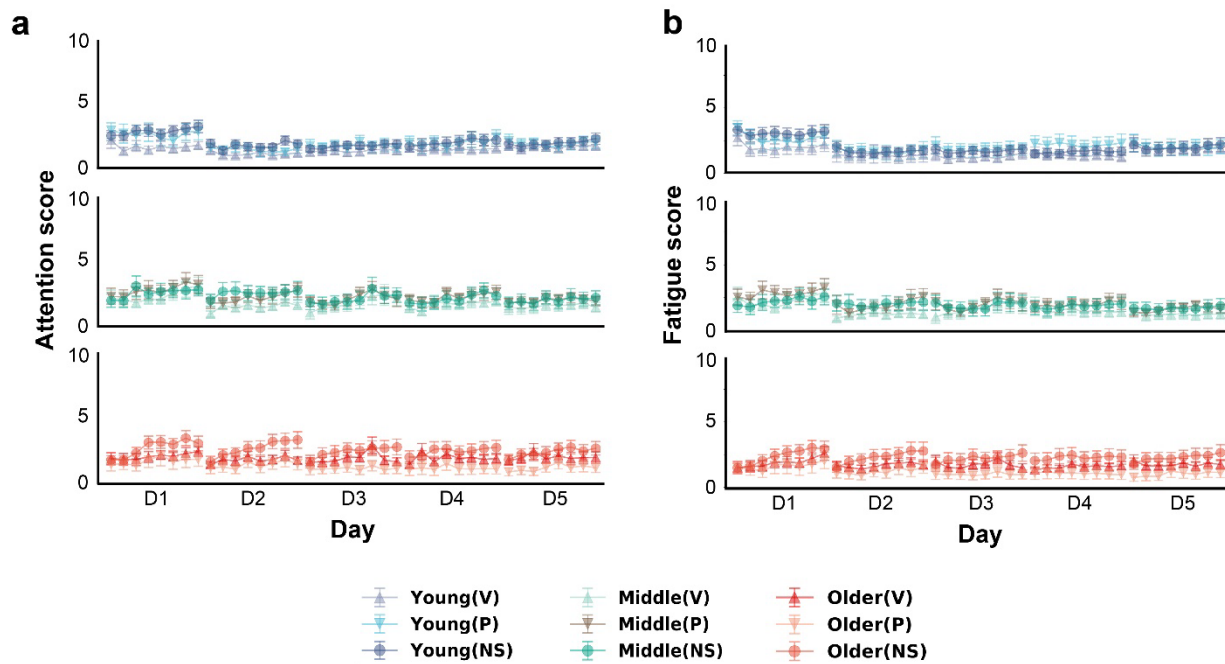


Figure S7. Self-reports on the state of attention and fatigue over the course of training. Average attention (a) and fatigue (b) scores provided by the participants of all groups of the first and the second experiments, with young adults at the top and older adults at the bottom panels. The error bars correspond to the standard error of the mean. V: Verum; P: Placebo; NS: No stimulation.

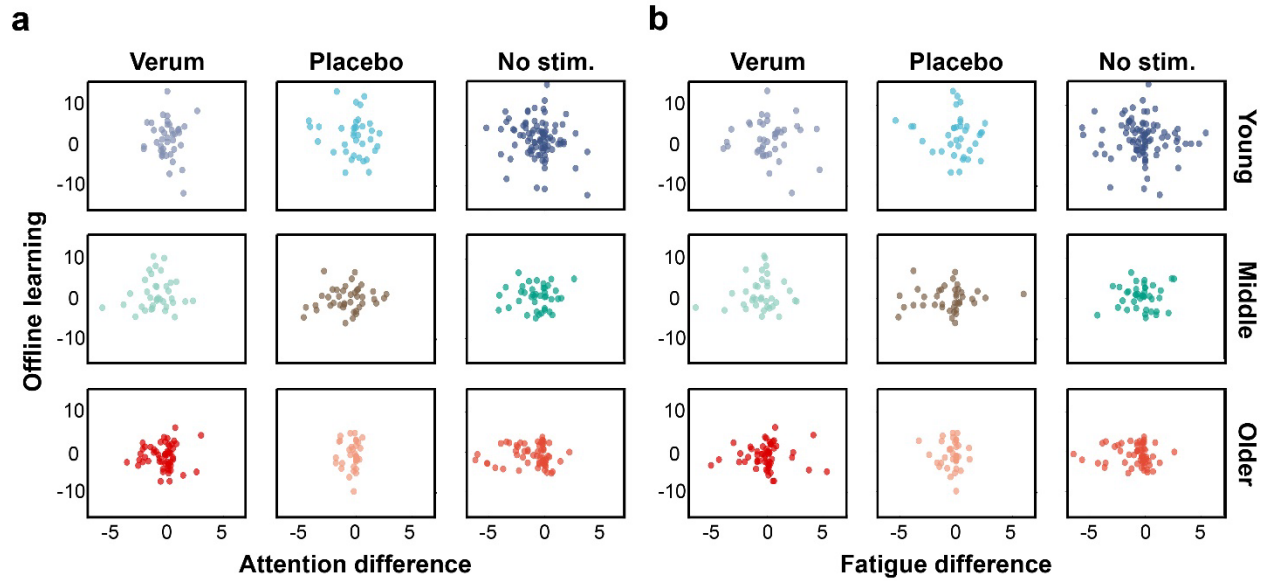


Figure S8. Offline learning as a function of attention and fatigue. The difference in attention and fatigue was calculated between the last block of a training session and the first block of the following training session. Please note that the figures depicting the relationship between offline learning and attention (a) and offline learning and fatigue (b) share the same y-axis. Please note as well that the top row shows the young adults and the bottom row shows the older adults for both attention and fatigue.

Statistical analysis of the main results

This subsection enlists all the statistical tests we ran for both experiments. In each table and for each aspect tested, we specify whether we fitted a linear model (LM) or a linear mixed-effect model (LMER) accounting for individual variation within each group; whenever between-subject variability warranted it, we used a linear mixed-effect model with random intercepts for individuals. The parameter $(1 + 1|ID)$ in the random column is used to specify a random effect of intercept between individuals, allowing the intercept of the fitted model to vary, as defined in the `lmer` package in R (72). We did not include a random effect of slope per day because inter-individual variability in each group did not justify it. We defined the factor “Day” as categorical, to fit individual lines per training day.

Below, we show tables with the results of all the statistical tests we ran. *Table S1* contains a key explaining the different aspects assessed in the execution of our motor task, while *Table S2* contains a key with the parameters reported for each statistical test. *Table S3* shows the results of the tests ran on the data from the first experiment, in which our groups of young, middle-aged and older adults trained without stimulation. *Table S4* contains the tests we ran including all groups of the second experiment and serves as a preamble, alongside the results reported on *Table S3*, justifying the testing of the effects of the intervention (i.e., atDCS) on each age group separately. The results shown in *Table S3* (and confirmed in *Table S5*) show clear age-related differences in the overall performance, as well as on the speed and on the accuracy, which alongside significant interactions found between the age groups and the intervention over the course of training (*Table S4*), suggest the mechanisms involved to be different, justifying a separate comparison for each age group. Our tests on the effect of stimulation for each age group are summarized on *Table S6*, while the subsequent tests done in the groups of older adults taking their generated chunking patterns into account are summarized in *Table S7*. *Table S8* and *Table S9* outline the results of the statistical tests we ran on the reported scores for attention and fatigue in the first and in the second experiment, respectively.

Table S1. Key describing the aspects tested for in the assessment of the behavioral and electrophysiological data of both experiments.

Aspect	Definition
Baseline	First exercise block presented to all subjects, lasting 90 seconds and containing a different sequence to that presented during the training or the catch blocks. It served as a measure of initial skill level.
Training	Set of blocks containing the main training sequence. On each training day, the training sequence was present in six out of seven blocks.
Online learning	Measure of performance change taking place over the course of a training session, quantified as the difference between the last and the first training block of each session.
Online slope	Slope of the straight lines fitted to the scores of each participant on each training day, reflecting the rate of change in performance on each day.
Offline learning	Measure of performance change occurring overnight (i.e., between training sessions), quantified as the difference in score between the last block of a training session and the first training block of the following session.
Speed	Speed of execution of the sequences, quantified as the total number of sequences generated per block.
Accuracy	Measure of precision in the execution of the sequence, quantified as the ratio of correct to total number of sequences.
FU scores	Scores obtained during the follow-up (FU) sessions.
Catch block scores	Scores obtained in the “catch” blocks presented halfway through each training session, containing a sequence different to the training sequence and different on each session.
D1 / D2-D5	Specification of whether an aspect concerns the first training day (D1) or the rest of the training week (D2-D5).
SICRest	Short-interval intracortical inhibition (SICI) measured when individuals were at rest, quantified as the ratio of the magnitude of the conditioning to the test pulse commonly used in the SICI TMS paradigm (please refer to the Methods). The specification “change Tr. D1” refers to the change in SICRest after the first training session, while “change Tr. Week” refers to the change after the training week.
Attention	Subjective measure of attention reported by each participant before and after each training block, captured using a visual analog scale in the form of an ungraded 10 cm straight line ranging from “completely attentive” (0) to “completely inattentive” (10).
Fatigue	Subjective measure of fatigue reported by each participant before and after each training block, captured using a visual analog scale in the form of an ungraded 10 cm straight line ranging from “awake” (0) to “tired” (10).
Label	The “label” parameter in <i>Table S7</i> corresponds to a classification we did within older adults receiving verum stimulation, grouping them according to whether they generated “young-like” or “old-like” chunking patterns on day 1.

Table S2. Key describing the parameters reported for each statistical test.

ANOVA parameters	Description
DF num.	Numerator degrees of freedom.
DF den.	Denominator degrees of freedom. Only applicable when fitting mixed-effects models (i.e., LMER).
F	F-value in an ANOVA, calculated as the ratio of the variance explained by a model to the unexplained variance.
p	Classical p-value, representing the likelihood of observing the current or more extreme data given that the null hypothesis is true (i.e., the true means of the compared groups are not different).

Post-hoc test parameters	Description
Estimate	Estimated difference between the models fitted to each group.
DF	Degrees of freedom.
t	T ratio, calculated as the estimate divided by the standard error of the estimate.
p	Classical p-value, as described above.
d	Effect size, calculated as the difference of the means divided by the standard deviation.
CI	Confidence level, calculated as the estimate \pm the value of the inverse cumulative density function of the Student t distribution given the degrees of freedom.

Table S3. Statistical tests run on behavioral data of the first experiment.

Aspect	Model	Dependent	Independent	Random	ANOVA					PostHoc tests								
					ANOVA param.	DF num.	DF den.	F	p	Level	Contrast	Estimate	DF	t	p	d	CI	
Baseline	LM	Score	AGE	NA	AGE	2		9.796524	0.000264	80	Young - Middle	1.642424	49	1.476524	0.310889	0.494407	-0.19425	1.183064
					DAY	2		4.909091	49		4.413228	0.000163	1.477748	0.674939	2.280557			
					AGE:DAY	2		3.266667	49		2.692989	0.025719	0.983341	0.19382	1.722861			
Baseline accuracy	LM	Per. Correct	AGE	NA	AGE	2		0.946997	0.394888									
Block 1 accuracy	LM	Per. Correct	AGE	NA	AGE	2		1.545493	0.223419									
Training	LMER	Score	AGE, DAY	(1 + 1 ID)	AGE	2	49	22.96936	9.17E-08	day 1	Young - Middle	4.49989	55.33501	2.170226	0.085361	1.092034	0.033397	2.15067
					DAY	4	1496	309.6215	3.95E-194		Young - Older	6.203192	55.33501	2.991701	0.011355	1.505391	0.403381	2.607401
					AGE:DAY	8	1496	30.80038	6.01E-45		Middle - Older	1.703302	55.33501	0.753302	0.732914	0.413357	-0.69293	1.519643
					day 2	Young - Middle	8.07949	55.33501	3.89661		0.000766	1.960731	0.797897	3.123566				
						Young - Older	11.78042	55.33501	5.681512		1.53E-06	2.858874	1.543574	4.171474				
						Middle - Older	3.700933	55.33501	1.636774		0.238972	0.898143	-0.23295	2.029232				
					day 3	Young - Middle	10.21638	55.33501	4.927198		2.34E-05	2.479312	1.233054	3.72557				
						Young - Older	14.85153	55.33501	7.16266		5.92E-09	3.604172	2.137729	5.070615				
						Middle - Older	4.635155	55.33501	2.049943		0.109952	1.12486	-0.02379	2.27351				
					day 4	Young - Middle	10.08712	55.33501	4.86486		2.92E-05	2.447944	1.20711	3.688778				
						Young - Older	15.63477	55.33501	7.540401		1.42E-09	3.794247	2.286533	5.30196				
						Middle - Older	5.547643	55.33501	2.4535		0.044959	1.346303	0.177046	2.515559				
					day 5	Young - Middle	10.77967	55.33501	5.198863		8.88E-06	2.616011	1.345595	3.886246				
						Young - Older	17.07833	55.33501	8.236606		1.08E-10	4.14457	2.558394	5.730745				
						Middle - Older	6.298658	55.33501	2.785644		0.019675	1.528559	0.339907	2.712111				
Online learning	LMER	ΔScore	AGE, DAY	(1 + 1 ID)	AGE	2	48.98015	0.816388	0.447949	day 1	Young - Middle	5.726304	244.9911	3.141517	0.005344	1.05351	0.325554	1.781465
					DAY	4	195.9789	8.045557	5.02E-06		Young - Older	9.151834	244.9911	5.020802	2.96E-06	1.683729	0.861894	2.505564
					AGE:DAY	8	195.9789	4.486743	5.10E-05		Middle - Older	3.42553	244.9911	1.723325	0.198532	0.630219	-0.11299	1.373425
					day 2	Young - Middle	-3.49969	244.9911	-1.91997		0.135206	-0.64386	-1.33036	0.042632				
						Young - Older	-1.24583	244.9911	-0.68348		0.773348	-0.22921	-0.89309	0.434679				
						Middle - Older	2.253854	244.9911	1.133875		0.494159	0.414658	-0.31566	1.144971				
					day 3	Young - Middle	-0.10921	244.9911	-0.05992		0.998023	-0.02009	-0.68066	0.640471				
						Young - Older	0.917291	244.9911	0.503236		0.869821	0.168761	-0.49359	0.831115				
						Middle - Older	1.026504	244.9911	0.516416		0.863415	0.188853	-0.53355	0.911255				
					day 4	Young - Middle	0.925097	244.9911	0.507519		0.867752	0.170197	-0.49219	0.832582				
						Young - Older	-0.51353	244.9911	-0.28173		0.957198	-0.09448	-0.75559	0.566629				
						Middle - Older	-1.43863	244.9911	-0.72375		0.74966	-0.26468	-0.89808	0.459371				
					day 5	Young - Middle	-3.75057	244.9911	-2.0576		0.10096	-0.69002	-1.35066	-0.02938				
						Young - Older	-3.85119	244.9911	-2.11281		0.089324	-0.70853	-1.36928	-0.04778				
						Middle - Older	-0.10062	244.9911	-0.05062		0.998588	-0.01851	-0.73885	0.701825				
Online slope	LM	Slope	AGE, DAY	NA	AGE	2		0.467166	0.627333	day 1	Young - Middle	0.88315	245	2.979489	0.008888	0.997667	0.277289	1.718045
					DAY	4		12.64101	2.29E-09		Young - Older	1.599765	245	5.397138	4.78E-07	1.807205	0.964321	2.650089
					AGE:DAY	8		4.776151	1.80E-05		Middle - Older	0.716615	245	2.217011	0.07031	0.809538	0.052857	1.566218
					day 2	Young - Middle	-0.4177	245	-1.40919		0.33772	-0.47186	-1.14549	0.201768				
						Young - Older	-0.29685	245	-1.0015		0.576616	-0.33535	-1.00204	0.331348				
						Middle - Older	0.120844	245	0.373859		0.925878	0.136514	-0.58381	0.856836				
					day 3	Young - Middle	0.227412	245	0.767223		0.723468	0.256901	-0.40685	0.92065				
						Young - Older	0.123662	245	0.417198		0.908566	0.139697	-0.52109	0.800486				
						Middle - Older	-0.10375	245	-0.32098		0.944806	-0.1172	-0.83724	0.602831				
					day 4	Young - Middle	0.123356	245	0.416167		0.908996	0.139351	-0.52143	0.800134				
						Young - Older	-0.09555	245	-0.32236		0.944342	-0.10794	-0.76823	0.552345				
						Middle - Older	-0.21891	245	-0.67724		0.776961	-0.24729	-0.9701	0.475514				
					day 5	Young - Middle	-0.45778	245	-1.54442		0.272071	-0.51714	-1.17794	0.14366				
						Young - Older	-0.7003	245	-2.36259		0.049438	-0.7911	-1.45178	-0.13042				
						Middle - Older	-0.24251	245	-0.75027		0.733748	-0.27396	-0.99758	0.449656				
					Young	day 1 - day 2	1.724773	245	6.462189		5.52E-09	1.948423	1.128125	2.768721				
						day 1 - day 3	1.459651	245	5.468858		1.11E-06	1.648923	0.886021	2.141824				
						day 1 - day 4	1.656072	245	6.204786		2.31E-08	1.870814	1.065898	2.675729				
						day 1 - day 5	2.18587	245	8.189772		1.93E-13	2.469309	1.675861	3.262757				
						day 2 - day 3	-0.26512	245	-0.99333		0.85819	-0.2995	-0.89972	0.30072				
						day 2 - day 4	-0.0687	245	-0.2574		0.999028	-0.07761	-0.67192	0.516703				
						day 2 - day 5	0.461096	245	1.727583		0.418962	0.520886	-0.07432	1.116095				
						day 3 - day 4	0.196421	245	0.735928		0.947895	0.221891	-0.37548	0.819262				
						day 3 - day 5	0.726218	245	2.720914		0.05377	0.820386	0.22462	1.416152				
						day 4 - day 5	0.529797	245	1.984986		0.276395	0.598496	0.004363	1.192628				
						day 1 - day 2	0.423926	245	1.31151		0.68433	0.478896	-0.25366	1.21145				
						day 1 - day 3	0.803914	245	2.487089		0.096898	0.908156	0.142093	1.67422				
						day 1 - day 4	0.896278	245	2.772839		0.046791	1.012498	0.235481	1.789514				
						day 1 - day 5	0.844939	245	2.614009		0.070924	0.954501	0.183701	1.725302				
						day 2 - day 3	0.379988	245	1.175579		0.765347	0.429261	-0.30069	1.159215				
						day 2 - day 4	0.472353	245	1.461329		0.588596	0.533602	-0.20213	1.269338				
						day 2 - day 5	0.421013	245	1.302499		0.689924	0.475606	-0.25677	1.207979				
						day 3 - day 4	0.092365	245	0.285751		0.998535	0.104341	-0.61553	0.82421				
						day 3 - day 5	0.041025	245	0.126921		0.999941	0.046345	-0.67301	0.765701				
						day 4 - day 5	-0.05134	245	-0.15883		0.999856	-0.058	-0.77742	0.661431				
						day 1 - day 2	-0.17185	245	-0.53164		0.984018	-0.19413	-0.91556	0.527309				
						day 1 - day 3	-0.01645	245	-0.0509		0.999998	-0.01859	-0.73784	0.700665				
						day 1 - day 4	-0.03924	245	-0.12141		0.999951	-0.04433	-0.76368	0.675012				
						day 1 - day 5	-0.11419	245	-0.35238		0.996646	-0.129	-0.8492	0.591207				
					day 2 - day 3	0.155393	245	0.480743	0.989055		0.175542	-0.54549	0.896578					
					day 2 - day 4	0.1326	245	0.410229	0.994024		0.149794	-0.57075	0.870339					
					day 2 - day 5	0.057654	245	0.178365	0.999772		0.06513	-0.65435	0.784609					
					day 3 - day 4	-0.02279	245	-0.07051	0.999994		-0.02575	-0.74502	0.693521					
					day 3 - day 5	-0.09774	245	-0.30238	0.998171		-0.11041	-0.83036	0.609532					
					day 4 - day 5	-0.07495	245	-0.23186	0.999356		-0.08466	-0.80432	0.634986					
Offline learning	LM	ΔScore	AGE, NIGHT	NA	AGE	2		8.27757	0.000354	AGE	Young - Middle	2.536574	196	3.342873	0.002843	0.559672	0.21987	0.899474
					NIGHT	3		0.079903	0.970857	AGE	Young - Older	2.637295	196	3.47561	0.001809	0.558185	0.243562	0.902229
					AGE:NIGHT	6		1.753384	0.110611	AGE	Middle - Older	0.100721	196	0.121721	0.991865	0.022223	-0.3379	0.382343
Speed	LMER	Seq. Number	AGE, DAY	(1 + 1 ID)	AGE	2	48.99997	28.30728	6.74E-09	day 1	Young - Middle	2.769697	53.16693	1.619623	0.246343	0.989968	-0.27039	2.250327
					DAY	4	1496	734.2932	0		Young - Older	4.869697	53.16693	2.84763	0.016918	1.740568	0.41101	3.070125
					AGE:DAY	8	1496	84.18678	3.80E-115		Middle - Older	2.1	53.16693	1.126096	0.502437	0.750599	-0.60453	2.105733
					day 2	Young - Middle	6.169192	53.16693	3.607529		0.001945	2.205404	0.816527	3.59356				
						Young - Older	10.72475	53.16693	6.27146		1.97E-07	3.833328	2.163674	5.502982				
						Middle - Older	4.555556	53.16693	2.442854		0.046412	1.628284	0.207383	3.049186				
					day 3	Young - Middle	8.530303	53.16693	4.988225		2.03E-05	3.048972	1.527227	4.570716				
						Young - Older	13.49697	53.16693	7.89256		4.85E-10	4.824199	2.943273	6.705125				
						Middle - Older	4.966667	53.16693	2.663306		0.027159	1.775227	0.339019	3.211436				
					day 4	Young - Middle	9.204545	53.16693	5.382499		5.01E-06	3.289965	1.724945	4.854986				
						Young - Older	15.21566	53.16693	8.897588		1.07E-11	5.438506	3.416337	7.460676				
						Middle - Older	6.011111	53.16693	3.223375		0.00604	2.148541	0.668412	3.62867				
					day 5	Young - Middle	10.33232	53.16693	6.041984		4.57E-07	3.693065	2.051288	5.33				

Aspect	Model	Dependent	Independent	Random	ANOVA					PostHoc tests																																																		
					ANOVA param.	DF num.	DF den.	F	p	Level	Contrast	Estimate	DF	t	p	d	CI																																											
Accuracy	LMER	Per. Correct	AGE, DAY, BLOCK	(1 + 1 ID)	AGE	2	72.52015	5.837485	0.00446	day 1	Young - Middle	-0.00122	75.56009	-0.0468	0.998793	-0.01294	-0.56385	0.537963																																										
					DAY	4	1481	10.23586	3.55E-08		Young - Older	0.06209	75.56009	2.390027	0.050087	0.661015	0.076918	1.245112																																										
					BLOCK	1	1481	19.87458	8.89E-06		Middle - Older	0.063306	75.56009	2.234595	0.071861	0.673958	0.041443	1.306473																																										
					AGE:DAY	8	1481	2.510816	0.010367		Young - Middle	0.019482	75.69673	0.749573	0.734812	0.207406	-0.34708	0.76189																																										
					AGE:BLOCK	2	1481	3.748352	0.02378		Young - Older	0.077628	75.69673	2.986752	0.010492	0.82643	0.224228	1.248631																																										
					DAY:BLOCK	4	1481	1.115982	0.34729		Middle - Older	0.058146	75.69673	2.051518	0.107076	0.619024	-0.00887	1.246922																																										
					AGE:DAY:BLOCK	8	1481	3.131654	0.00162		Young - Middle	0.014007	75.8786	0.538613	0.852564	0.149124	-0.40405	0.702301																																										
					FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99998	27.97434	7.87E-09	day 5	Young - Middle	10.0237	57.78333	4.067446	0.000423	2.222246	0.946917	3.497575																																				
											DAY	2	410	10.36045	4.08E-05		Young - Older	15.71304	57.78333	6.376083	9.82E-08	3.483569	1.982418	4.98472																																				
											AGE:DAY	4	410	1.726828	0.143117		Middle - Older	5.689342	57.78333	2.117046	0.095219	1.261322	0.011863	2.510782																																				
											Catch block scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99995	16.57653	3.17E-06	day 10	Young - Middle	10.0269	57.78333	4.068742	0.000421	2.229555	0.947518	3.498391																														
																	DAY	5	245	1.233546	0.293915		Young - Older	17.54548	57.78333	7.119656	5.63E-09	3.88982	2.304128	5.475511																														
																	AGE:DAY	10	245	0.807684	0.621469		Middle - Older	7.518588	57.78333	2.797722	0.018852	1.666865	0.376669	2.95706																														
																	FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99998	27.97434	7.87E-09	day 60	Young - Middle	11.75296	57.78333	4.76915	3.82E-05	2.605622	1.268568	3.942676																								
																							DAY	2	410	10.36045	4.08E-05		Young - Older	18.45429	57.78333	7.488433	1.37E-09	4.0913	2.462034	5.720567																								
																							AGE:DAY	4	410	1.726828	0.143117		Middle - Older	6.701326	57.78333	2.493612	0.040554	1.485679	0.214914	2.756443																								
																							FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99995	16.57653	3.17E-06	All	Young - Older	17.2376	49	4.483804	0.000129	2.350274	1.087538	3.613011																		
																													DAY	5	245	1.233546	0.293915		Young - Middle	6.636418	49	2.573952	0.034474	1.471289	0.242662	2.699915																		
																													AGE:DAY	10	245	0.807684	0.621469		Middle - Older	17.2376	49	2.573952	0.034474	1.471289	0.242662	2.699915																		
																													FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99995	16.57653	3.17E-06	Young	day 5 - day 10	-1.92449	410	-2.51073	0.033209	-0.43706	-0.80866	-0.06547												
																																			DAY	5	245	1.233546	0.293915		day 5 - day 60	-0.50385	410	-0.65734	0.788305	-0.11443	-0.46454	0.235687												
																																			AGE:DAY	10	245	0.807684	0.621469		day 10 - day 60	1.42064	410	1.853397	0.153786	0.322635	-0.03862	0.683894												
																																			FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99995	16.57653	3.17E-06	Middle	day 5 - day 10	-1.9213	410	-2.06973	0.09745	-0.43634	-0.87758	0.004909						
																																									DAY	5	245	1.233546	0.293915		day 5 - day 60	1.225407	410	1.320077	0.38479	0.278297	-0.15166	0.708249						
																																									AGE:DAY	10	245	0.807684	0.621469		day 10 - day 60	3.146705	410	3.389806	0.00221	0.714634	0.242825	1.186443						
																																									FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)		AGE	2	48.99995	16.57653	3.17E-06	Older	day 5 - day 10	-0.09205	410	-0.09916	0.994594	-0.02091	-0.44298	0.401173
																																															DAY	5	245	1.233546	0.293915		day 5 - day 60	2.237392	410	2.410243	0.043171	0.508124	0.060234	0.956014
																																															AGE:DAY	10	245	0.807684	0.621469		day 10 - day 60	2.329443	410	2.509406	0.033326	0.529029	0.079036	0.979023

Table S4. Statistical tests run on behavioral data of the second experiment.

Aspect	Model	Dependent	Independent	Random	ANOVA				
					ANOVA param.	DF num.	DF den.	F	p
Baseline	LM	Score	AGE,STIM	NA	AGE	2		6.108802	0.00402
					STIM	1		0.001996	0.964523
					AGE:STIM	2		1.104202	0.33871
Training	LMER	Score	AGE,DAY,STIM	(1 + 1 ID)	AGE	2	54.99973	12.39233	3.61E-05
					DAY	4	1745	392.5667	2.70E-241
					STIM	1	54.99973	1.622449	0.208107
					AGE:DAY	8	1745	22.91452	1.39E-33
					AGE:STIM	2	54.99973	0.680037	0.510806
					DAY:STIM	4	1745	2.72677	0.027954
					AGE:DAY:STIM	8	1745	5.243177	1.67E-06
Speed	LMER	Seq. Number	AGE,DAY,STIM	(1 + 1 ID)	AGE	2	54.99969	26.35831	9.38E-09
					DAY	4	1745	1152.315	0
					STIM	1	54.99969	4.571259	0.036971
					AGE:DAY	8	1745	92.13703	9.04E-128
					AGE:STIM	2	54.99969	0.510373	0.603087
					DAY:STIM	4	1745	11.20811	5.58E-09
					AGE:DAY:STIM	8	1745	2.193037	0.025422
Accuracy	LMER	Per. Correct	AGE,DAY,STIM	(1 + 1 ID)	AGE	2	55	5.36158	0.00746
					DAY	4	1745	7.115792	1.11E-05
					STIM	1	55	0.029438	0.864402
					AGE:DAY	8	1745	6.245986	5.45E-08
					AGE:STIM	2	55	7.31263	0.001527
					DAY:STIM	4	1745	2.190772	0.067767
					AGE:DAY:STIM	8	1745	6.361158	3.66E-08

Aspect	Model	Dependent	Independent	Random	ANOVA					PostHoc tests												
					ANOVA param.	DF num.	DF den.	F	p	Level	Contrast	Estimate	DF	t	p	d	CI					
FU scores	LMER	Score	AGE, DAY	(1 + 1 ID)						day 60	Young - Older	11.73988	31.64578	4.399979	0.00033	2.921747	1.308741	4.534753				
												Middle - Older	1.433923	31.64578	0.55138	0.84652	0.356866	-0.96641	1.680145			
													day 5 - day 10	-0.60415	218	-0.55244	0.845334	-0.15036	-0.70682	0.406107		
													Young	day 5 - day 60	1.403806	218	1.283668	0.405902	0.34937	-0.2151	0.913849	
														day 10 - day 60	2.007953	218	1.836113	0.160307	0.499727	-0.07486	1.074318	
														day 5 - day 10	-0.27589	218	-0.26593	0.961775	-0.06866	-0.59523	0.457908	
														Middle	day 5 - day 60	4.028022	218	3.882544	0.000402	1.002468	0.396191	1.608746
														day 10 - day 60	4.303911	218	4.148469	0.000142	1.07113	0.454343	1.687917	
														Older	day 5 - day 10	-0.18914	218	-0.17295	0.983646	-0.04707	-0.60188	0.507735
															day 5 - day 60	1.331174	218	1.217252	0.444252	0.331294	-0.23219	0.894782
															day 10 - day 60	1.520315	218	1.390207	0.347788	0.378366	-0.18779	0.944524
					Catch blocks	LMER	Score	AGE, DAY	(1 + 1 ID)	AGE	2	25	8.715846	0.001343								
										DAY	5	125	2.02342	0.079779								
										AGE:DAY	10	125	0.783287	0.644731								
SICrest at Baseline	LM	SICI ratio (SICI/Test)	AGE	NA	AGE	2		0.097957	0.907064													
SICrest change Tr. D1	LM	ΔSICI ratio	AGE, TimePoint	NA	AGE	2		0.268557	0.765672													
					TimePoint	1		0.02577	0.873166													
					AGE:TimePoint	2		0.111119	0.895071													
SICrest change Tr. Week	LM	ΔSICI ratio	AGE, TimePoint	NA	AGE	2		0.034383	0.956226													
					TimePoint	1		0.803903	0.374499													
					AGE:TimePoint	2		0.244467	0.784112													

Table S6. Statistical tests run on data from the second experiment, comparing verum and placebo groups for each age group.

Aspect	Age Group	Model	Dependent	Independent	Random	ANOVA				PostHoc tests										
						ANOVA param.	DF num.	DF den.	F	p	Level	Contrast	Estimate	DF	t	p	d	CI		
Baseline	Young	LM	Score	STIM	NA	STIM	1		0.52376	0.479083	day 1	Verum vs. Placebo	0.582313	20.04746	0.196147	0.84647	0.116902	-1.12665	1.360452	
	DAY					4	543	1.849072	0.191054											
	STIM:DAY					4	543	0.015125	0.903289											
Block 1 accuracy	Young	LM	Per. Correct	STIM	NA	STIM	1		1.229327	0.282978	day 1	Verum vs. Placebo	2.201949	20.04746	0.741708	0.466866	0.442052	-0.80839	1.692493	
	DAY					4	543	2.597604	0.04746											
	STIM:DAY					4	543	1.285029	0.109615											
Training	Young	LMER	Score	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.090493	0.767199	day 1	Verum vs. Placebo	-2.68403	20.04746	-0.90409	0.376683	-0.53883	-1.79286	0.715193	
						DAY	4	543	0.027385	0.870515	day 2									
						STIM:DAY	4	543	140.4404	2.48E-82	day 3									
	Middle	LMER	Score	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.027385	0.870515	day 1	Verum vs. Placebo	1.051544	20.31419	0.517486	0.610406	0.297221	-0.90315	1.497596	
						DAY	4	543	0.228043	0.876449	day 2									
						STIM:DAY	4	543	3.380069	0.009564	day 4									
	Older	LMER	Score	STIM, DAY	(1 + 1 ID)	STIM	1	20.99994	4.997128	0.03638	day 1	Verum vs. Placebo	2.771352	24.74045	1.682165	0.105118	0.934891	-0.24504	2.114822	
						DAY	4	659	95.52916	4.39E-64	day 2									
						STIM:DAY	4	659	2.719338	0.028847	day 3									
	Online learning D1	Young	LM	ΔScore	STIM	NA	STIM	1		0.005869	0.93983	day 1	Verum vs. Placebo	3.619165	21	3.031921	0.006341	1.295378	0.322132	2.268624
							DAY	4	543	0.091144	0.766389									
							STIM:DAY	4	543	9.192547	0.006341									
Middle		LM	ΔScore	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.170196	0.685094	day 1	Verum vs. Placebo	0.948148	19.20977	0.61181	0.547837	0.421402	-1.02501	1.867819	
						DAY	3	51	1.493668	0.227352	day 2									
						STIM:DAY	3	51	0.228043	0.876449	day 3									
Older		LM	ΔScore	STIM, DAY	(1 + 1 ID)	STIM	1	21	0.007103	0.933821	day 1	Verum vs. Placebo	2.859259	19.20977	1.84499	0.080514	1.270791	-0.22213	2.763708	
						DAY	3	63	3.70962	0.015589	day 2									
						STIM:DAY	3	63	0.282582	0.837797	day 4									
Online learning D2-D5		Young	LMER	ΔScore	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.001108	0.973835	day 1	Verum vs. Placebo	0.00426	22.85555	0.129455	0.89813	0.058486	-0.8766	0.993576
							DAY	3	51	2.21858	0.096831	day 2								
							STIM:DAY	3	51	0.61681	0.607289	day 3								
	Middle	LM	ΔScore	STIM, DAY	(1 + 1 ID)	STIM	1	21.00008	0.229233	0.637041	day 1	Verum vs. Placebo	0.001119	22.85555	0.034016	0.97316	0.015368	-0.91956	0.950299	
						DAY	3	63	1.337383	0.270225	day 2									
						STIM:DAY	3	63	0.504896	0.6803	day 4									
	Older	LMER	ΔScore	STIM, DAY	(1 + 1 ID)	STIM	1	21.00008	0.229233	0.637041	day 1	Verum vs. Placebo	-0.0711	22.85555	-2.16077	0.01439	-0.97621	-1.95743	0.00501	
						DAY	3	62.99986	1.757451	0.16441	day 1 - day 2									
						STIM:DAY	3	62.99986	0.49405	0.687702	day 2 - day 3									
	Online slope D1	Young	LM	Slope	STIM, NIGHT	NA	STIM	1		0.0229	0.880166	day 1	Verum vs. Placebo	-0.03907	543	-2.93848	0.028267	-0.53649	-0.94824	-0.12474
							NIGHT	3		0.147828	0.930739	day 2 - day 4								
							STIM:NIGHT	3		0.447888	0.719593	day 3 - day 4								
Middle		LM	Slope	STIM, NIGHT	NA	STIM	1		0.210953	0.647487	day 1	Verum vs. Placebo	-0.01654	543	-1.24411	0.725431	-0.22714	-0.61126	1.56976	
						NIGHT	3		1.281362	0.2878	day 1 - day 5									
						STIM:NIGHT	3		0.125995	0.944954	day 4 - day 5									
Older		LM	Slope	STIM, NIGHT	NA	STIM	1		0.040191	0.841592	day 1	Verum vs. Placebo	0.005933	22.85555	0.180319	0.858491	0.081466	-0.85378	1.016716	
						NIGHT	3		1.314234	0.275203	day 1 - day 2									
						STIM:NIGHT	3		1.630893	0.188352	day 2 - day 3									
Online slope D2-D5		Young	LMER	Slope	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.001108	0.973835	day 1	Verum vs. Placebo	0.948148	19.20977	0.61181	0.547837	0.421402	-1.02501	1.867819
							DAY	3	51	2.21858	0.096831	day 2								
							STIM:DAY	3	51	0.61681	0.607289	day 3								
	Middle	LM	Slope	STIM, DAY	(1 + 1 ID)	STIM	1	1.038524	0.311771	day 1	Verum vs. Placebo	1.896299	19.20977	1.22362	0.235893	0.842805	-0.62102	2.306629		
						DAY	3	7.939899	0.000129	day 4										
						STIM:DAY	3	1.413525	0.246378	day 5										
	Older	LMER	Slope	STIM, DAY	(1 + 1 ID)	STIM	1	21.00008	0.229233	0.637041	day 1	Verum vs. Placebo	2.553704	19.20977	1.647825	0.115649	1.134988	-0.34749	2.617463	
						DAY	3	62.99986	1.757451	0.16441	day 2									
						STIM:DAY	3	62.99986	0.49405	0.687702	day 4									
	Offline learning	Young	LM	ΔScore	STIM, NIGHT	NA	STIM	1		0.0229	0.880166	day 1	Verum vs. Placebo	2.9	19.20977	1.871279	0.076617	1.288899	-0.20549	2.783291
							NIGHT	3		0.147828	0.930739	day 2								
							STIM:NIGHT	3		0.447888	0.719593	day 3								
Middle		LM	ΔScore	STIM, NIGHT	NA	STIM	1		0.210953	0.647487	day 1	Verum vs. Placebo	0.00426	22.85555	0.129455	0.89813	0.058486	-0.8766	0.993576	
						NIGHT	3		1.281362	0.2878	day 2 - day 4									
						STIM:NIGHT	3		0.125995	0.944954	day 3 - day 4									
Older		LM	ΔScore	STIM, NIGHT	NA	STIM	1		0.040191	0.841592	day 1	Verum vs. Placebo	0.005933	22.85555	0.180319	0.858491	0.081466	-0.85378	1.016716	
						NIGHT	3		1.314234	0.275203	day 1 - day 2									
						STIM:NIGHT	3		1.630893	0.188352	day 2 - day 3									
Speed		Young	LMER	Seq. Number	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.001108	0.973835	day 1	Verum vs. Placebo	0.948148	19.20977	0.61181	0.547837	0.421402	-1.02501	1.867819
							DAY	4	543	511.9617	1.21E-182	day 2								
							STIM:DAY	4	543	6.612859	3.35E-05	day 3								
	Middle	LMER	Seq. Number	STIM, DAY	(1 + 1 ID)	STIM	1	16.99999	2.2041	0.15595	day 1	Verum vs. Placebo	4.787037	18.36444	1.811946	0.08638	1.561754	-0.30992	3.433428	
						DAY	4	543	330.4279	6.51E-1447	day 2									
						STIM:DAY	4	543	3.795345	0.0047	day 3									
	Older	LMER	Seq. Number	STIM, DAY	(1 + 1 ID)	STIM	1	21	0.496911	0.488606	day 1	Verum vs. Placebo	4.851852	18.36444	1.836479	0.082529	1.5829	-0.29047	3.456273	
						DAY	4	659	264.3163	2.17E-135	day 4									
						STIM:DAY	4	659	0.703884	0.589453	day 5									
	Accuracy	Young	LM	ΔScore	STIM, NIGHT	NA	STIM	1		0.0229	0.880166	day 1	Verum vs. Placebo	0.00426	22.85555	0.129455	0.89813	0.058486	-0.8766	0.993576
							NIGHT	3		0.147828	0.930739	day 2								
							STIM:NIGHT	3		0.447888	0.719593	day 3								
Middle		LM	ΔScore	STIM, NIGHT	NA	STIM	1		0.210953	0.647487	day 1	Verum vs. Placebo	0.001119	22.85555	0.034016	0.97316	0.015368	-0.91956	0.950299	
						NIGHT	3		1.281362	0.2878	day 4									
						STIM:NIGHT	3		0.125995	0.944954	day 5									
Older		LM	ΔScore	STIM, NIGHT	NA	STIM	1		0.040191	0.841592	day 1	Verum vs. Placebo	-0.0711	22.85555	-2.16077	0.01439	-0.97621	-1.95743	0.00501	
						NIGHT	3		1.314234	0.275203	day 1 - day 2									
						STIM:NIGHT	3		1.630893	0.188352	day 2 - day 3									
FU scores		Young	LMER	Score	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.155273	0.698446	day 1	Verum vs. Placebo	-0.14785	21.71704	-3.92126	0.002655	-1.67475	-2.82044	-0.52907
							DAY	4	543	6.412699	4.78E-05	day 2								
							STIM:DAY	4	543	5.873906	0.000124	day 3								
	Middle	LMER	Per. Correct	STIM, DAY	(1 + 1 ID)	STIM	1	17	0.155273	0.698446	day 1	Verum vs. Placebo	-0.10713	21.71704	-2.45805	0.024432	-1.21357	-2.30346	-0.12368	
						DAY	4	543	13.05669	3.69E-10	day 2									
						STIM:DAY	4	543	0.705989	1.50E-05	day 3									
	Older	LMER	Per. Correct	STIM, DAY	(1 + 1 ID)	STIM	1	21	9.092071	0.006587	day 1	Verum vs. Placebo	-0.04822	21.71704	-1.10623	0.280733	0.54616	-1.58437	0.492054	
						DAY	4	659	3.554149	0.002959	day 4									
						STIM:DAY	4	659	0.054401	0.002959	day 5									
	SICI at Baseline	Young	LM	SICI ratio (SICI/Test)	STIM	NA	STIM	1		1.209097	0.287781	day 1	Verum vs. Placebo	0.113522	35.17942	2.841075	0.00743	0.927315	0.20904	1.64559
							DAY	4	543	3.554149	0.002959	day 2								
							STIM:DAY	4	543	0.054401	0.002959	day 3								
Middle		LM	SICI ratio (SICI/Test)	STIM	NA	STIM	1		1.625375	0.221739	day 1	Verum vs								

Table S7. Statistical tests run on data from the second experiment, comparing young-like and old-like older adults in the verum group. Labels are either "Young-like" or "Old-Like".

Aspect	Model	Dependent	Independent	Random	ANOVA					PostHoc tests								
					ANOVA param.	DF num.	DF den.	F	p	Level	Contrast	Estimate	DF	t	p	d	CI	
Baseline speed	LM	Seq. Number	LABEL	NA	LABEL	1		6.0615	0.029928	BL	YoungLike - OldLike	4.208333	12	2.462011	0.029928	1.329638	0.077816	2.581461
Block 1 speed	LM	Seq. Number	LABEL	NA	LABEL	1		6.817824	0.022756	B1	YoungLike - OldLike	4.25	12	2.611096	0.022756	1.410153	0.149271	2.671036
Block 1 accuracy	LM	Per. Correct	LABEL	NA	LABEL	1		0.097421	0.760304									
Training	LMER	Score	LABEL, DAY	(1 + 1 ID)	LABEL	1	12	23.68295	0.000387	day 1	YoungLike - OldLike	3.896406	16.15486	2.569147	0.020478	1.361515	0.16111	2.56192
					DAY	4	398	84.58827	6.20E-52	day 2		6.578695	16.15486	4.337749	0.000499	2.298783	0.966261	3.631304
					LABEL:DAY	4	398	9.701391	1.70E-07	day 3		7.004135	16.15486	4.618268	0.000278	2.447443	1.089343	3.805544
									day 4	7.411037		16.15486	4.886565	0.00016	2.589626	1.206045	3.973207	
									day 5	9.362692		16.15486	6.173414	1.28E-05	3.27159	1.753634	4.789546	
Online slope D1	LM	Slope	LABEL, BLOCK	NA	LABEL	1		35.0624	7.62E-08									
					BLOCK	1		27.45138	1.28E-06									
					LABEL:BLOCK	1		2.214154	0.140681									
Online slope D2-D5	LM	Slope	LABEL, BLOCK	NA	LABEL	1		289.9642	3.54E-47									
					BLOCK	1		13.87193	0.00023									
					LABEL:BLOCK	1		0.083031	0.773411									
Speed	LMER	Seq. Number	LABEL, DAY	(1 + 1 ID)	DAY	4	398	224.2692	1.49E-100	day 1	YoungLike - OldLike	2.777778	13.2841	1.941788	0.073675	1.719401	-0.26609	3.704889
					LABEL	1	11.99999	16.75407	0.00149	day 2		5.590278	13.2841	3.907848	0.001731	3.460294	1.257319	5.663269
					DAY:LABEL	4	398	26.0044	3.62E-19	day 3		6.180556	13.2841	4.320478	0.000792	3.825667	1.562478	6.088856
									day 4	6.270833		13.2841	4.383586	0.000704	3.881547	1.608767	6.154328	
									day 5	7.722222		13.2841	5.39817	0.000113	4.779934	2.340342	7.219527	
Speed slope D1	LM	Seq. Number	LABEL, BLOCK	NA	LABEL	1		44.78938	2.75E-09	Slope D1	YoungLike - OldLike	0.547619	80	2.850178	0.005555	0.290895	0.070579	0.511211
					BLOCK	1		27.93586	1.06E-06									
					LABEL:BLOCK	1		8.123515	0.005555									
Speed slope D2-D5	LM	Seq. Number	LABEL, BLOCK	NA	LABEL	1		305.058	6.53E-49									
					BLOCK	1		13.34315	0.000301									
					LABEL:BLOCK	1		1.296887	0.255604									
Accuracy	LMER	Per. Correct	LABEL, Day	(1 + 1 ID)	LABEL	1	12	0.007673	0.931642	day 1 - day 2	Day	-0.08039	398	-4.57962	6.10E-05	-0.71397	-1.10482	-0.32313
					DAY	4	398	11.02619	1.72E-08	day 1 - day 3		-0.05978	398	-3.40553	0.0065	-0.53093	-0.89339	-0.16847
					LABEL:DAY	4	398	0.305363	0.874357	day 1 - day 4		-0.08503	398	-4.84412	1.80E-05	-0.75521	-1.15326	-0.35715
									day 1 - day 5	-0.10883		398	-6.19978	1.41E-08	-0.96656	-1.40542	-0.52777	
									day 2 - day 3	0.020609		398	1.17409	0.766217	0.183043	-0.14569	0.511781	
									day 2 - day 4	-0.00464		398	-0.2645	0.998922	-0.04124	-0.3654	0.28293	
									day 2 - day 5	-0.02844		398	-1.62016	0.485326	-0.25259	-0.58562	0.080446	
									day 3 - day 4	-0.02525		398	-1.43859	0.603055	-0.24228	-0.55541	0.106847	
									day 3 - day 5	-0.04905		398	-2.79425	0.043143	-0.43563	-0.78596	-0.0853	
									day 4 - day 5	-0.0238		398	-1.35566	0.656414	-0.21135	-0.54168	0.118976	

Table S8. Statistical tests run on scores given to rate attention and fatigue using a visual analog scale in the first experiment.

Aspect	AGE group	Model	Dependent	Independent	Random	ANOVA					Post-Hoc																																																								
						Parameters	DF num.	DF den.	F	p	Level	Contrast	DF	Estimate	t	p	d	CI																																																	
Attention	All	LMER	Att. Score	AGE, DAY, BLOCK	(1 + 1 ID)	AGE	2	45.68067	0.073668	0.92909	DAY	D1-D2	849	1.060739	8.743712	0	0.992083	0.573233	1.290933																																																
						DAY	4	1767	3.033341	0.016632																																																									
						BLOCK	1	1767	86.29617	4.40E-20																																																									
						AGE:DAY	8	1767	2.666987	0.006527																																																									
						AGE:BLOCK	2	1767	7.506309	0.000567																																																									
						DAY:BLOCK	4	1767	0.699577	0.592234																																																									
						AGE:DAY:BLOCK	8	1767	1.221993	0.281869																																																									
						DAY	4	849	6.482129	0.011811										DAY	D2-D3	849	0.044119	0.363677	0.996265	0.038768	-0.18149	0.259026																																							
	BLOCK	1	849	14.14725	0.000181																																																														
	DAY:BLOCK	4	849	0.290771	0.884045																																																														
	DAY	4	381	2.293873	0.058878	DAY:BLOCK	D3-D4	849	-0.2894	-2.38556	0.120209	-0.2543	-0.48745	-0.02115																																																					
	BLOCK	1	381	17.26319	4.02E-05																																																														
DAY:BLOCK	4	381	0.487373	0.745025																																																															
DAY	4	537	0.40211	0.807169	DAY:BLOCK										D4-D5	849	0.067699	0.558045	0.980936										0.059488	-0.1612	0.280173																																				
BLOCK	1	537	84.79408	7.99E-19																																																															
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	4	537	0.40211	0.807169																DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512																																							
DAY:BLOCK	4	537	84.79408	7.99E-19																												DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122																											
BLOCK	1	537	2.954083	0.019644		DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296																																																					
DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449																		
BLOCK	1	537	2.954083	0.019644																																														DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009									
DAY:BLOCK	4	537	2.954083	0.019644	DAY:BLOCK										D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09										1.430036	0.156587	0.297512																																				
BLOCK	1	537	84.79408	7.99E-19																																																							DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	1	537	2.954083	0.019644																DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449																																							
DAY:BLOCK	4	537	2.954083	0.019644																												DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009																											
BLOCK	1	537	2.954083	0.019644		DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512																																																					
DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122																		
BLOCK	1	537	2.954083	0.019644																																														DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296									
DAY:BLOCK	4	537	2.954083	0.019644	DAY:BLOCK										D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05										0.914232	0.047267	0.136449																																				
BLOCK	1	537	2.954083	0.019644																																																							DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	1	537	84.79408	7.99E-19																DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122																																							
DAY:BLOCK	4	537	2.954083	0.019644																												DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296																											
BLOCK	1	537	2.954083	0.019644		DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449																																																					
DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009																		
BLOCK	1	537	2.954083	0.019644																																														DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512									
DAY:BLOCK	4	537	2.954083	0.019644	DAY:BLOCK										D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10										1.570596	0.16788	0.300122																																				
BLOCK	1	537	2.954083	0.019644																																																							DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	1	537	2.954083	0.019644																DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009																																							
DAY:BLOCK	4	537	2.954083	0.019644																												DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512																											
BLOCK	1	537	84.79408	7.99E-19		DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122																																																					
DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296																		
BLOCK	1	537	2.954083	0.019644																																														DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449									
DAY:BLOCK	4	537	2.954083	0.019644	DAY:BLOCK										D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06										1.167333	0.069163	0.154009																																				
BLOCK	1	537	2.954083	0.019644																																																							DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	1	537	2.954083	0.019644																DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296																																							
DAY:BLOCK	4	537	2.954083	0.019644																												DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449																											
BLOCK	1	537	2.954083	0.019644		DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009																																																					
DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512																		
BLOCK	1	537	84.79408	7.99E-19																																														DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122									
DAY:BLOCK	4	537	2.954083	0.019644	DAY:BLOCK										D3 slope over BLOCKS	97	0.132602	3.776207	0.000275										0.844386	0.062908	0.202296																																				
BLOCK	1	537	2.954083	0.019644																																																							DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	1	537	2.954083	0.019644																DAY:BLOCK	D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09	1.430036	0.156587	0.297512																																							
DAY:BLOCK	4	537	2.954083	0.019644																												DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122																											
BLOCK	1	537	2.954083	0.019644		DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.202296																																																					
DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449																		
BLOCK	1	537	2.954083	0.019644																																														DAY:BLOCK	D5 slope over BLOCKS	97	0.111586	5.220472	1.02E-06	1.167333	0.069163	0.154009									
DAY:BLOCK	4	537	2.954083	0.019644	DAY:BLOCK										D1 slope over BLOCKS	97	0.227049	6.395316	5.64E-09										1.430036	0.156587	0.297512																																				
BLOCK	1	537	84.79408	7.99E-19																																																							DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122
DAY:BLOCK	4	537	2.954083	0.019644																																																															
BLOCK	1	537	2.954083	0.019644																DAY:BLOCK	D4 slope over BLOCKS	97	0.091858	4.088568	8.96E-05	0.914232	0.047267	0.136449																																							
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DAY:BLOCK	4	537	2.954083	0.019644																																					DAY:BLOCK	D2 slope over BLOCKS	97	0.234001	7.023921	2.98E-10	1.570596	0.16788	0.300122																		
BLOCK	1	537	2.954083	0.019644																																														DAY:BLOCK	D3 slope over BLOCKS	97	0.132602	3.776207	0.000275	0.844386	0.062908	0.2022									

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