Within-subject consistency of paired associative stimulation as assessed by linear mixed models.

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Abstract

Objective

Paired associative stimulation (PAS) is a TMS paradigm used to induce long-term potentiation in the human cortex. Little is known about the within-subject consistency of PAS-induced effects. We determined PAS-induced effects and their consistency in healthy volunteers between two PAS sessions. Additionally, we assessed the benefit of applying linear mixed models (LMMs) to PAS data.

Methods

Thirty-eight healthy volunteers underwent two identical PAS sessions with a >1 week interval. During each session, motor evoked potentials (MEPs) were assessed once before PAS induction and 3 times after at 30 min intervals.

Results

We did not detect any significant potentiation of MEP size after PAS induction. However, MEP size during PAS induction showed significant potentiation over time in both sessions ($LR(1) = 13.36$, $p<0.001$). Nevertheless, there was poor within-subject consistency of PAS-induced effects both during (ICC=0.15) and after induction (ICC=0.03-0.08). Additionally, statistical model selection procedures demonstrate that LMMs are more appropriate than conventional longitudinal models for estimating PAS-induced effects ($LR(34) = 214.73$, $p < 0.001$).

Conclusion

PAS-induced effects are more pronounced during than after induction, have a low within-subject consistency in any phase of the measurement, and are best estimated with LMMs. The implication of our study is that PAS is an inappropriate method to assess the evolution of brain plasticity over time periods longer than the PAS measurement itself.

Introduction

Synaptic plasticity is a fundamental process in our central nervous system, because it is essential for learning and memory [1, 2]. In addition, plasticity deficits are important in the etiology of neurocognitive disorders [3, 4]. Synaptic plasticity is conventionally measured with invasive techniques, such as field electrophysiology, which cannot readily be performed in human subjects. Non-invasive extracranial brain stimulation techniques, such as transcranial magnetic stimulation (TMS), have emerged in the last two decades, enabling measuring plasticity-like effects in human subjects. Specific TMS paradigms have been shown to produce long-term potentiation (LTP) as well as long-term depression (LTD) like effects, resembling the plasticity effects seen using intraparenchymal electrophysiological studies in animal models.

Paired associative stimulation (PAS) is one of the TMS paradigms used to induce long-term plasticity [5]. PAS is typically applied by pairing median nerve stimulation (MNS) with magnetic stimulation of the contralateral hand area of the primary motor cortex (M1). Consistent with the fundamental properties of spike-timing dependent plasticity (STDP) [7], when MNS precedes magnetic stimulations by 25ms, PAS stimulation induces a long-term increase in excitability of the M1 hand area that can be observed as an increase of motor-evoked potentials (MEPs) in the contralateral hand. In contrast, if the MNS precedes the magnetic stimulation by 10ms, the result is a long-term depression effect [6]. The resemblance to STDP is further strengthened by pharmacological studies showing that PAS is dependent on the function of the *N*-methyl-D-aspartate (NMDA) receptor, known to be essential for many forms of long-term synaptic plasticity [7].

Because of the temporal similarity of PAS results to classical STDP experiments in rodents, PAS has emerged as a potentially very useful proxy for studying long-term synaptic plasticity in human subjects. However, PAS produces highly variable results in humans [8, 9], which is often attributed to the challenge of achieving similar levels of standardization as for animal experiments. Environmental factors, lifestyle factors, experimental conditions and even genetic determinants have been suggested

to influence the magnitude of the PAS-induced plasticity (for review see Ridding et al. (2010) and

Wischnewski et al. (2016)) [10, 11]. However, these factors only explain between-subject variability,

whereas little is known about the within-subject consistency.

Besides inter- and intra-individual variability, PAS studies show variable effect sizes between

laboratories as well [9, 11]. In addition to optimizing experimental procedures, some types of

variability might be possible to account for by appropriate statistical modeling. PAS measurements

generate relatively complex data, combining both repeated measures as well as a nested data structure

(i.e. multiple MEP size assessments per time point). In the last decades, linear mixed models (LMMs)

have emerged as a statistical method that is specifically suited to handle such a data structure, reducing

the chance of both false-positive and false-negative results [12, 13]. Additionally, LMMs are excellent

for estimating reproducibility measures in the form of intra-class correlations. However, the utility of

LMMs for the analysis of PAS data has not yet been demonstrated.

In this study, we examined healthy volunteers using two identical PAS sessions with an interval of at

least 1 week, in order to determine the within-subject consistency between the two sessions. Using

linear mixed model (LMM) analyses, we estimated PAS-induced effects and PAS consistency.

Materials and Methods

Subjects

Thirty-eight out of 61 subjects were included in this study (reasons for exclusion are summarized in

Table S1 Table), who were recruited by advertising in the local community and on a Dutch research

subject-recruitment website. Subjects were included if aged 18-40, right-handed according to the

Edinburgh Handedness Inventory [14], in good health, medication free (excluding contraceptives) and

able and willing to give written informed consent. Subjects were excluded if they were women

lactating or pregnant, had a history of psychiatric illness and/or treatment, had a history of

neurological illness or did not meet the international safety guidelines considering TMS [15, 16]. All

subjects underwent the Wechsler Abbreviated Scale of Intelligence (WASI) [17] to determine their

intelligence quotient (IQ) [18] for descriptive purposes. This study was approved by the Medical

Ethical Review Board of the Erasmus MC Rotterdam in 2013, requiring study procedures to comply

with the latest version of the Declaration of Helsinki.

Fig 1. Schematic representation of the PAS paradigm.

(A) Schematic of one PAS session, in which the PAS induction is preceded by a baseline measurement

consisting of 20 TMS stimulations. To measure the change in MEP-size as of PAS induction, this

measurement is repeated 3 times with 30 min intervals. (B) Experimental setup during PAS induction,

where median nerve stimulation (MNS) at 3 times the sensory threshold (ST) precedes transcranial

magnetic stimulations (TMS) delivered at the stimulation intensity 1mV (SI1mV) by 25 ms. Motor

evoked potentials (MEPs) are measured using electromyography (EMG) of the abductis pollicis brevis

muscle. **C** Example traces of single MEPs before and after PAS induction.

Electromyography

Muscle activity was recorded from the left abductor pollicis brevis (ABP) muscle with

electromyography (EMG), using Ag-AgCl electrodes in a belly-tendon montage. EMG signals were

amplified using a universal amplifier (ANT Neuro, Enschede, The Netherlands) and digitalized at

5kHz for later offline analysis using Visor2 XT software (ANT Neuro, Enschede, The Netherlands).

During measurements, a continuous EMG signal and trigger related EMG epochs were plotted at real

time for online analysis, while applying a 50Hz notch filter and a 20-2000Hz bandpass filter.

Transcranial magnetic stimulation

Subjects were invited in the afternoon between 12 and 5.30 PM [19], and were asked to not perform

intense physical activities 24 hours prior to the measurement and to not smoke nicotine cigarettes or

drink coffee on the day of the measurement. They were seated in a comfortable chair with their left

arm resting on a pillow and were told to maximally relax their left hand during the measurement.

Magnetic stimulations were applied using a figure-of-eight coil with an inner diameter of 27mm and

outer diameter of 97mm, connected to a MagPro X100 with MagOption TMS device (MagVenture,

Farum, Denmark). The coil was held tangentially to the left primary cortex and diverging 45˚ from

midline. The electric field subsequently created in the cortex had a posterior to anterior direction.

To find the optimal position of the coil in order to maximally activate the ABP (the hotspot), TMS

stimulations were randomly placed around a predefined reference point. This reference point was

defined as the location at 10% of the ear-to-ear span lateral to Cz over the right hemisphere. Data on

coil location and position at every stimulation was collected using a neuronavigation system (ANT

Neuro, Enschede, The Netherlands), allowing a precise spatial definition of the hotspot and precise

determination of the angle and distance errors of every stimulation relative to the hotspot. All TMS

procedures hereafter described are performed at the hotspot.

The resting motor threshold (RMT) was determined using a maximum-likelihood threshold hunting

procedure [20]. For this procedure, a MEP with a peak-to-peak amplitude of $\geq 50\mu V$ with an onset

within a timeframe of 20-50ms post stimulation was considered a MEP. Subsequently, the stimulation

intensity 1mV (SI1mV) was determined, defined as the percentage of maximal stimulation output

(%MSO) of the TMS device that resulted in a mean MEP of 0.8 - 1.2 mV after 10 stimulations at

0.1Hz.

Paired associative stimulation

Subjects underwent two identical paired associative stimulation (PAS) sessions at least one week

apart. First, baseline cortical excitability was assessed by applying a train of 20 magnetic stimulations

at the SI1mV at 0.1Hz. Second, the PAS induction phase followed, consisting of 200 paired

stimulations at 0.25Hz with each pair consisting of electric MNS followed by magnetic stimulation

with an inter stimulus interval of 25ms. Third, after this plasticity induction phase, the cortical

excitability measurement at baseline was repeated at three time points: immediately (Post 1), 30

minutes (Post 2), and 60 minutes (Post 3) after completing PAS induction (Fig 1B).

MNS during the PAS-induction was applied with a strength of three times the sensory threshold, using

a bipolar bar electrode connected to a constant current stimulator (model DS7A; Digitimer Ltd.,

Letchworth Garden City, UK) (Fig 1A). In case this stimulation intensity surpassed the pain threshold,

it was lowered to a painless but clearly noticeable level. The subject's attention level was standardized

as much as possible by applying four randomly timed electric stimuli during PAS induction to the

middle phalanx of the left thumb, and instructing participants upfront of PAS induction to focus their

attention on their left thumb, to count the number of thumb stimuli sensed and to report this number

after PAS induction [21]. These electric stimulations were administered at two times the sensory

threshold using a double ring electrode connected to a constant current stimulator (Micromed S.p.A,

Mogliano Veneto, Italy).

MEP-analysis

The EMG signal for every magnetic stimulation applied was stored for offline analysis as epochs of -

300ms to +300ms surrounding the TMS trigger (Fig 1C). Using software programmed in LabVIEW

(National Instruments, Austin, TX, US) pre-MEP noise, the maximal peak-to-peak amplitude and

MEP onset were determined using a seven-step data processing procedure:

1. Signals were linearly detrended.

2. The average amplitude value of the -300ms to -20ms before the TMS trigger was subtracted

to create a zero-baseline.

3. To prevent ringing due to filtering, the stimulation artefact was removed between -2ms to

+4ms surrounding the TMS trigger, followed by linear interpolation. For MEPs obtained

during PAS induction, the stimulation artefact of the MNS was removed similarly.

4. Signals were filtered using both a 20-2000Hz bandpass filter and a 50Hz-notch filter.

5.   Noise was determined on a time window of -25ms to +15ms surrounding the TMS trigger

that was depleted of baseline wandering by subtracting a 2nd-order polynomial fit. It was

defined as a peak-to-peak amplitude of >50μV or an SD of >15. Signals meeting these criteria

were discarded for further statistical analysis.

6.   The maximal peak-to-peak amplitude of every MEP was determined within a 20-48ms time

frame following the TMS trigger.

7.   The MEP onset was determined within an 18-30ms time frame following the TMS trigger.

Statistical analysis

Statistical analyses were performed using R version 3.3.3 [22] and the nlme package for mixed model

analysis specifically [23]. Session specific subject characteristics were compared using paired t-tests

for normally distributed data (RMT and SI1mV), a Wilcoxon Signed Rank test for non-normal

continuous data (starting time of TMS measurement), a Chi-square test for categorical data (attention

score) and LMMs for data related to individual MEPs (angle and distance error). Furthermore, we

used LMMs to estimate PAS-induced changes of MEP size, their correlations with baseline MEP size,

and intraclass correlations (ICCs). For these LMMs, the dependent variable was MEP size, which was

log2-transformed to better fit the assumption of normally distributed residuals. In addition, these

LMMs were adjusted for log2-transformed angle and distance error.

We built Model 1 to estimate PAS-induced effects on MEP size at each time point *after induction*

(Post 1, Post 2 and Post 3) within each session. This LMM included time point (categorical), session,

and their interaction. The random effects included subject specific random effects for each time point

in each session separately. An unstructured covariance matrix for the random effects was used (Model

1a) and was tested against the more restrictive compound symmetry structure (Model 1b), using a

likelihood ratio (LR) test. Subsequently, the LR test was also used to assess the main effects of fixed

effects.

Model 2 was built to estimate PAS-induced effects *during PAS induction*. This LMM included

stimulus number (continuous), session and their interaction. Stimulus number was regarded as

continuous time variable, as stimulations were equally spaced by 4 seconds in all PAS experiments.

The model included subject specific random effects for stimulus number and session interaction and

session. The eventual model was selected in three steps using likelihood ratio tests. First, we started

out with a model using both natural cubic splines for stimulus number with three degrees of freedom

and an unstructured covariance matrix (Model 2a). Second, to investigate the correlation structure, we

tested Model 2a against a model with a compound symmetry structure (Model 2b). Last, to test

whether the relation between MEP size and stimulus number was non-linear, Model 2a was tested

against a model with a linear fit (Model 2c), using a LR test. After model selection, the LR test was

also used to determine the main effects of fixed effects.

As a measure of within-subject consistency we calculated ICCs from LMMs that included session as

an additional nesting level in the random effects. For the ICC of PAS-induced effects after induction,

fixed effects and subject specific random effects of time point (categorical) were used (Model 3). To

estimate the ICC of PAS-induced effects during PAS-induction over time, fixed effects as well as

subject specific slopes for stimulus number (continuous time variable) were included (Model 4). Since

the models used to calculate ICCs contained random effects for the respective time variables, the

variation partition method was used [24]. 95% confidence intervals (95%CIs) for each ICC were

estimated using 500 bootstrap samples.

Results

Session characteristics

Thirty-eight individuals (22 women; median age 23, range 19-38; mean IQ 107±10SD) underwent two

PAS sessions, which were spaced at least 1 week apart (median days between sessions was 14, IQR:

4). As displayed in Table 1, median starting time was significantly earlier in session 1 than in session

2, whereas both sessions did not differ in terms of baseline RMT, SI1mV or the level of attention

during PAS induction. Additionally, the angle error and distance error relative to the hotspot did not

differ between sessions either. The estimated means of both errors over each time point were minimal,

with the upper limit of the 95%CIs not exceeding 5° for the angle error and 2mm for the distance

error. There was, however, a significant effect of time point (including PAS induction) on both angle

error (LR(8) = 22.41, p = 0.004) and distance error (LR(8) = 18.55, p = 0.018).

Table 1. Session characteristics and comparisons.				
Characteristic	Session 1	Session 2	Statistic	P-value
RMT at baseline, mean (±SD), %MSO	48.5 (±10.3)	48.3 (±9.1)	t(37) = 0.17	0.87
SI1mV, %MSO	61.4 (±14.4)	60.1 (±14.7)	t(37) = 1.10	0.28
Start time, median (IQR), hh:mm	12:44 (12:28-13:08)	15:23 (15:00-15:41)	Z=719	<0.0001
Attention score, n (%)			LR(2) = 3.74	0.15
<4 stimuli reported	10 (29)	14 (38)		
4 stimuli reported	10 (29)	15 (41)		
>4 stimuli reported	15 (43)	8 (22)		
Angle error, estimated mean [95%CI]*, degrees			LR(5) = 5.26**	0.39
Baseline	2.69 [1.75, 4.16]	1.67 [1.11, 2.51]		
Induction	2.95 [1.99, 4.35]	2.11 [1.47, 3.01]		
Post1	2.07 [1.39, 3.08]	1.86 [1.12, 3.08]		
Post2	2.23 [1.41, 3.54]	1.29 [0.89, 1.87]		
Post3	1.83 [1.15, 2.91]	1.09 [0.81, 1.46]		
Distance error, estimated mean [95%CI]*, mm			LR(5) = 3.85**	0.57
Baseline	1.01 [0.84, 1.23]	0.87 [0.68, 1.12]		
Induction	1.11 [0.95, 1.29]	0.91 [0.79, 1.04]		
Post1	1.10 [0.97, 1.26]	0.91 [0.79, 1.06]		
Post2	1.20 [0.96, 1.50]	1.30 [1.02, 1.65]		

| Post3 | 1.31 [1.02, 1.68] | 1.34 [1.03, 1.74] |

PAS-induced effects post induction

We determined the PAS-induced effect on MEP size at each post-induction measurement in each

session. After filtering out MEPs with a noisy baseline, 5212 MEPs out of the originally 6080 MEPs

recorded could be used for statistical analysis, resulting in the exclusion of one complete session of

one subject. Plotting the individual trajectories of MEP size after induction (Fig 2A) indicates that

PAS-induced changes of MEP size were highly variable. It is, therefore, not surprising that a model

with an unstructured covariance matrix provided a better fit than one with a compound symmetry

matrix (LR(34) = 214.73, p < 0.001), and was therefore selected as the appropriate matrix to estimate

PAS-induced effects in our data.

The estimated mean of baseline MEP size of session 1 (0.54 mV; 95%CI [0.43, 0.68]) did not differ

from that of session 2 (0.61 mV; 95%CI [0.53, 0.71]) (LR(4)=2.26, p=0.689). MEP size changed

significantly over time (LR(6) = 16.23; p = 0.013), which was mainly driven by a negative effect on

MEP size in Post 3 in session 2 (Table 2), instead of a positive effect on MEP size as one would

expect when performing PAS. PAS-induced effects did not differ between sessions, as the interaction

between time point and session was not significant (LR(3) = 1.93; p=0.586). The absence of this

interaction is reflected by the similar profile of MEP size over time illustrated in Fig 2B. There was a

moderate positive correlation between individual baseline MEP size and the PAS-induced effect at

each time point in session 1 ($r_{Post\ 1}$ = 0.68, $r_{Post\ 2}$ = 0.57, $r_{Post\ 3}$ = 0.52) and a poor positive correlation in

session 2 ($r_{Post\ 1}$ = 0.23, $r_{Post\ 2}$ = 0.21, $r_{Post\ 3}$ = 0.30).

Table 2. Fixed effects of PAS induction on MEP size per post-induction time point and session estimated by linear mixed effect modelling.

	Session 1	Session 2

Variable	β, %	95%CI, %	t (5165)	P-value	β, %	95%CI, %	t (5165)	P-value
Post 1	+16.83	[-9.05, 50.07]	1.22	0.22	-3.99	[-26.58, 25.55]	-0.30	0.77
Post 2	-10.00	[-30.92, 17.24]	-0.78	0.43	-23.92	[-42.44, 0.55]	-1.92	0.05
Post 3	-14.84	[-35.65, 12.69]	-1.12	0.26	-31.65	[-47.92, -10.30]	-2.74	0.006

Fig 2. PAS-induced effects for session 1 and 2 separately.

Subjects underwent two identical PAS sessions spaced >1 week apart, with session 1 displayed in red and session 2 in blue. (A) The mean change in MEP size over time per session (black line) plotted over the individual line plots (colored lines). (B) The mean change in MEP size over time for both sessions, where dots represent means of individual medians and bars represent their standard error. (C) Linear regression lines through all MEPs during PAS induction per session (black lines) plotted over the linear regression lines through MEPs per individual (colored lines). **(**D) The change of MEP size over time during the PAS induction, with every dot representing the mean MEP size over all participants for that stimulation number. Lines are fitted linear regression lines per session.

Potentiation during PAS induction

We, additionally, took a novel approach to determine the PAS-induced effect on MEP size by assessing this effect during PAS induction next to post induction alone. For this analysis, 9360 MEPs were available out of the 15200 MEPs recorded, due to filtering out MEPs with a noisy baseline. The exclusion of these MEPs resulted in the exclusion of one entire session for 9 subjects and the exclusion of both sessions for 4 subjects. Viewing the individual trajectories of MEP size development again indicates that there was high inter-individual variability (Fig 2C), which is reflected by the superior fit of the model with an unstructured covariance matrix to one with a compound symmetry covariance matrix (LR(8) = 525.31, p < 0.001). The development of MEP size over time appeared to be linear (Fig 2D), supported by the fact that a model with a cubic fit was not superior to one with a linear fit (LR(4) = 2.69, p = 0.612). Therefore, a model with an unstructured covariance matrix and a linear fit was selected as the appropriate model to estimate PAS-induced effects.

The estimated mean of MEP size at the start of PAS induction in session 1 (0.43 mV, 95%CI [0.27,

0.59]) did not differ from that in session 2 (0.44 mV, 95%CI [0.29, 0.66]) (LR(2) = 0.967, p = 0.617).

There was a main effect of time (LR(1) = 13.36, p < 0.001), as a result of a significant positive

increase of MEP size over time in both session 1 (+132%, 95%CI [+51%, +258%]) and session 2

(+79%, 95%CI [+19%, +169%]). However, there was no evidence of this time effect being different

between sessions (LR(1) = 0.87, p = 0.35), reflected by the similar slope of the MEP size development

in Fig 2D. There was a moderate negative correlation between MEP size at the start of PAS induction

and the change in MEP size over time for session 1 (r = -0.51) and a weak negative correlation for

session 2 (r = -0.41).

303 Consistency of PAS-induced effects

The within subject consistency of PAS-induced effects between the two sessions was small: Post 1 had

an ICC of 0.08 (95%CI [0.03, 0.16]); Post 2 had an ICC of 0.07 (95%CI [0.03, 0.14]); and Post 3 had

an ICC of 0.03 (95%CI [0.01, 0.09]) (Fig 3). Furthermore, the PAS-induced effects during induction

showed a similarly poor within-subject consistency (ICC = 0.15; 95%CI [0.05, 0.35]) (Fig 3), despite

their significant potentiation at group level. The ICC of baseline MEP size before induction was poor

(ICC = 0.02; 95%CI [<0.01, 0.04]), as well as at the start of PAS induction (ICC = 0.24; 95%CI =

[0.04, 0.42]). Nonetheless, the SI1mV did have a good within-subject consistency (ICC = 0.88; 95%CI

[0.83, 0.96]), as did the RMT at different time points ($ICC_{Baseline}$ = 0.85, 95%CI [0.77, 0.92]; $ICC_{Post 1}$

= 0.83, 95%CI [0.79, 0.90]; $ICC_{Post 2}$ = 0.85, 95%CI = [0.79, 0.92]; $ICC_{Post 3}$ = 0.85, 95%CI [0.78,

0.92]).

Fig 3. Intra-individual correlation of PAS-induced effects.

Scatterplots per phase during or after induction illustrating the correlation between individual PAS-

induced effects (dashed slopes) of session 1 against session 2. The slopes plotted are derived from the

models used to calculate the ICC of PAS-induced effects within subject and between session.

Discussion

To our knowledge, we performed the largest study reported to date aimed at quantifying the within-

subject consistency of PAS-induced effects. Additionally, this is the first PAS-TMS study that

quantifies the change in MEP size during PAS induction over time and the first using a linear mixed

model analysis approach. We performed two identical PAS sessions in one group of healthy

volunteers, resulting in pronounced potentiation over time during PAS induction, which was not

consistent within subjects. PAS-effects after induction did not show the expected potentiation, and

these effects were not consistent within subjects either. Additionally, we demonstrated that a linear

mixed model with an unstructured covariance matrix provides the best model fit for our PAS data.

PAS-induced effects during and after induction

The absence of post-induction potentiation in our study, and even the negative trend towards the last

post-induction measurement, is certainly not in line with most PAS studies (for review see [11]), for

which a few factors in our study could be responsible. First, our baseline MEP size is lower than the

baseline of around 1mV shown in most PAS studies. In addition, we found a poor to moderate positive

correlation between baseline MEP size and the PAS induced effects at each post-induction time point,

suggesting that individuals with a low baseline were prone to show lower PAS-induced effects.

Combined, these two observations could indicate that our stimulation intensity during induction was

relatively lower than most PAS studies, possibly causing our PAS-induced effects to be smaller.

Nonetheless, this hypothetical interaction between stimulation intensity and potentiation effect has

never been specifically studied and relatively lower stimulation intensities that are still supra-threshold

should still be able to potentiate MEP size, as double activation of the cortex by TMS and MNS is

conserved. Second, coil placement during experiments could have been inaccurate. However, as angle

and distance error of the coil position relative to the hotspot were minimal and carefully corrected for

in our model, this is unlikely the case in our study. Last, the known compromising effect of sleepiness

on MEP size could be a contributor [25]. As our subjects remained seated and were not allowed to

perform any type of physical activity or specific types of mental activity between post-induction time

points, it is plausible that subjects became increasingly sleepy during this phase of the experiment.

Unfortunately, we cannot support this speculation with actual measures of sleepiness during

experiments, as these were not assessed.

Nevertheless, we did find a significant increase of MEP size during PAS induction that shows striking

resemblance to the increase in excitatory post synaptic potentials seen in STDP experiments in rodents

[26]. From these animal studies, we know that the potentiation during plasticity induction correlates

with the potentiation after induction. However, whether our observed potentiation of MEP size during

PAS induction is indeed a proxy for NMDA-dependent LTP as well, should be confirmed by studies

using a sham-stimulation control and preferably also placebo-controlled studies with an NMDA-

receptor antagonist. It is noteworthy, however, that in our study MEP size at the start of PAS induction

showed a negative correlation with PAS-induced effects during PAS induction. As MNS during paired

stimulations has a known acute inhibitory effect on MEP size, also known as short-latency afferent

inhibition [27, 28], lower MEP size at the start of induction could indicate successful paired

stimulations and, therefore, be related to a more prominent PAS-induced potentiation.

The advantage of using linear mixed models for PAS data

Our results provide insight in the potential advantage of LMMs for analyzing PAS data over

conventional analysis methods. Most importantly, we show that using an unstructured covariance

matrix provides a better model fit than using a compound symmetry matrix. First, our finding at least

demonstrates that PAS data can greatly benefit from a model of which the covariance matrix can be

flexibly designed and suggests that the compound symmetry matrix is unsuited for PAS data. As the

compound symmetry assumes that the correlations of MEP size between each time point are equal,

this result is actually not surprising for data sampled in a dynamic biological system such as the

human brain. Since conventionally used RM-ANOVA is restricted to a compound symmetry matrix,

LMMs are therefore more appropriate for the analysis of PAS data and are likely to deliver more

reliable results. Consequently, future consistent implementation of LMMs for analyzing PAS data will

improve the reproducibility between PAS studies.

Consistency of PAS-induced effects

We show that PAS-induced effects, both during and after PAS induction, have poor intra-individual

consistency. However, one could argue that the lack of significant post-induction potentiation

compromises the validity of the consistency levels in this study. Nonetheless, we were able to show

significant potentiation during induction, which showed similar low consistency. Additionally, a lack

of significant potentiation at group level does not warrant that at individual level, as can be

appreciated from the individual traces in Fig 2A, and the low consistency found in this study indicates

that the subset of individuals showing potentiation were largely different in each session. It is,

therefore, unlikely that the consistency would be much higher in case of significant potentiation at

group level, which is in line with previous results reported by Fratello et al. (2006): in their study, they

also performed two identical PAS sessions within one group of healthy volunteers (n=18) and found

equally poor intra-individual consistency, despite finding significant PAS-induced potentiation both

PAS sessions [29].

The low consistency of PAS does not seem to hold for methods that measure brain plasticity in

humans in general, as intermittent theta burst stimulation (iTBS) was found to produce reasonably

consistent measures of brain plasticity with an ICC of 0.53 [30]. Another paradigm, anodal

transcranial direct current stimulation (aTDCS), showed moderate to no intra-individual consistency in

two separate studies [31, 32]. The contrast between the within-subject consistency between iTBS and

PAS can be understood from animal studies, suggesting that plasticity induction protocols based on

low frequency stimulations, such as STDP, are much more vulnerable to metaplasticity than high

frequency paradigms such as theta burst stimulation [33]. This is in line with findings in human TMS

studies, also reporting a vulnerability of PAS to metaplasticity [34, 35]. Altogether, this implies that

PAS is probably much more dependent on highly dynamic time-varying factors than iTBS, resulting in

a lower within-subject consistency of PAS.

In our view, the inconsistency of PAS-induced plasticity has major implications for its use. As our

results have revealed that PAS-induced effects cannot be seen as a trait, dividing subjects into PAS

responders and PAS non-responders or even selecting subjects based on previous PAS results seems

generally unsuitable. Moreover, our results suggest that PAS has limited value in longitudinal studies,

for example, in studies monitoring drug effects over longer periods of time or in patient follow-up

studies characterizing disease progression.

Conclusion

Our results demonstrate that PAS-induced effects have a high intra-individual variability and that

PAS-induced potentiation of MEP size seems to be most pronounced at the end of the induction phase

opposed to after induction. Additionally, our results support the use of LMMs for analyzing PAS data.

The main implication of our results is that, due to its low consistency, PAS is unsuitable to track the

development of human brain plasticity over longer periods of time.

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Supporting Information

S1 Table. Reasons for excluding recruited individuals before entering the study.

S1 Dataset. Full dataset of all variables reported. RDA-file containing all data on all variables

reported in this manuscript. This file can be opened and used with the open source R or RStudio

software [22].

S1 Syntax. Syntax for all linear mixed models reported. Syntax written in the R-language,

containing all linear mixed models that are used to generate results for this manuscript. This file can

be opened and used with the open source R or RStudio software [22].