Enhancement of motor performance by a training method combining motor imagery and neurofeedback

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Abstract

Motor imagery and neurofeedback have been proposed as motor training approaches, but their effects on the enhancement of motor performance are still controversial. This study aimed to enhance motor performance using a training protocol combining motor imagery and neurofeedback. Seventeen participants were randomly assigned to the training or control group. The training group received real-time electroencephalography activity feedback relative to the motor imagery of the motor action. The control group reiterated the motor imagery of the intended motor action without feedback. The motor performance of the training group was enhanced significantly more than that of the control group. Subsequently, the proposed training protocol was tested in five elite male tennis players to investigate its applicability to sports activities. The motor performance was enhanced in three of these five professional tennis players. Using our training protocol, which combined motor imagery and neurofeedback training, we achieved enhanced motor performance. Furthermore, we could suggest the applicability to sports from the results with world-level tennis players.

Keywords: motor imagery, neurofeedback training, motor performance enhancement, sports, Event-related desynchronization

1 Introduction

Motor skills are essential for daily life, and they need to be learned and developed to

perform various activities. In sports, elite athletes must learn and develop specific motor

skills at extremely high levels that are beyond those required for ordinary activities of

daily living. To enhance motor performance, different training approaches have been

investigated in many fields. In neuroscience, to enhance motor performance efficiently and

effectively, motor imagery (MI) and neurofeedback (NFB) training have been proposed

and studied.

The motor performance of a given action can be enhanced by the MI method, which

consists of imagining the action without executing the movements involved (Decety and

Grèzes 1999). In cases of rehabilitation, it has been reported that MI therapy supplements

conventional physiotherapy or occupational therapy for patients with motor dysfunction

disorders (Caligiore et al. 2017; Zimmermann-Schlatter et al. 2008). These studies have

reported improvement of the movements of the upper and lower limbs among patients

following stroke. Conversely, in studies involving healthy participants, MI training can

yield effects that enhance motor performance (Gentili et al. 2006; Ruffino et al. 2017,

2019; Sobierajewicz et al. 2017). These studies have reported that the duration of an actual

movement is decreased by MI training. The above studies among patients and healthy

participants suggest that MI has the potential not only to improve the patients' motor

functions but also to enhance the motor performance of healthy individuals.

Among the many investigations of MI training, researchers noted that its effect seems to

be variable. A review analyzed 133 MI training studies with positive, unchanged, or

negative results (Schuster et al. 2011). The analysis suggested that successful MI training

could be achieved with an optimal protocol. However, the authors lacked a

neurophysiological viewpoint. Di Rienzo *et al.* have discussed the effects of MI training

on activity-dependent neuroplasticity (Di Rienzo et al. 2016). We agree that MI training

protocol needs to be optimized, but this end would be better served if we first understand

how MI can modulate neurophysiological activities.

The characteristic brain activity of MI is an event-related desynchronization (ERD) of mu-

band (8–13 Hz) oscillation in sensory-motor areas. This mu-band ERD has also been

observed during actual movement execution(Sobierajewicz et al. 2017). Many studies

have shown that mu-band ERD was induced in the contralateral hemisphere of the hand

during MI (Wolf et al. 2014; Wriessnegger et al. 2018). In sports, some studies have

reported that mu-band ERD of elite athletes was larger than that of amateurs or novices

during the MI of sports activity (Del Percio et al. 2007; Wolf et al. 2014) . The differences

of mu-ERD between elite and non-elite athletes have shown some correlation but not a

causal relationship.

In most MI studies, the mu-ERD has been used to monitor the mental states of subjects

during MI (Kim et al. 2017; Wriessnegger et al. 2018). It has been reported that the

vividness of MI correlates with mu-ERD amplitudes (Toriyama et al. 2018). The feedback

of mu-ERD will enable the trainees to confirm the vividness of MI by themselves, which

will lead to an enhancement of motor performance (Ruffino et al. 2017). For effective

training, the vividness of MI estimated by mu-ERD should be conveyed to a trainee rather

than a trainer.

To regulate motor-related neuronal activity, we overlaid NFB training on MI training.

NFB training is a method of self-regulation of neural activities to control specific mind or

brain states. In the NFB training paradigm, brain activities are transformed into external

sensory stimuli, which are fed back continuously. Individuals can learn to regulate their

brain states by receiving this continuous feedback. Several studies have reported the

enhancement or improvement of cognitive function (Hanslmayr et al. 2005; Kober et al.

2019; Nan et al. 2012) with feedback signals of sensory-motor rhythm (SMR, 12–15 Hz)

(Kober et al. 2019) and alpha band (individual alpha frequency [Nan et al. 2012] or

common frequency of 10–13 Hz [Hanslmayr et al. 2005]).

Although NFB training using SMR (13–15 Hz) and alpha-theta: 4–8 Hz (Rostami et al.

2012) has been used in many types of sports, the effects vary across sports and individuals

(reviewed in Gruzelier 2014; Jeunet et al. 2019). In an NFB-based study on archery,

focusing on brain activities peculiar to elite athletes, pre-elite athletes were trained using

the feedback of the EEG activity in the left temporal site (in which the elite athletes

showed higher activation during shooting) and had a higher enhancement of shooting

accuracy in the NFB group than in the sham or no feedback groups (Park et al. 2015;

Thompson et al. 2008). An NFB study of table tennis players evaluated SMR (12–15 Hz)

in 10 elite athletes and reported that some athletes showed an increase in accuracy that was

not observed in other athletes (Brown et al. 2012). Further, in an NFB study of golf

players a personalized EEG profile was determined during successful golf putts during a

3-day NFB training using individually assessed measures. The results showed that their

performance was significantly enhanced from day 1 to day 2; however, they could not find

significant enhancement on day 3 (Arns et al. 2007). To ensure stable effects, an effective

training protocol should be developed.

The large variation in the effects on motor performance may be caused by two factors: the

training target (the kind of brain activity occurring during training) and the training

strategy (the strategy used to change the neuronal activity during NFB training). In

previous NFB studies on sports, training targets of EEG activities were not directly related

to the intended motor action, and training strategies were not standardized by

experimenters and were possibly unrelated to the intended action because strategies were

selected by the participants. In most NFB studies, including in those relevant to sports

training, the strategies have been freely determined by the trainee (Paret et al. 2019). To

enhance motor performance efficiently, the training target and training strategy should be

directly associated with the intended motor action.

Our main purpose is to determine whether the mu-band ERD enhancement can improve

motor performance. If the mu-band ERD differences between athletes and novices are the

result of actual motor training (Del Percio et al. 2007; Wolf et al. 2014), training of mu-

band activity will not provide any enhancement of motor performance. If motor

performance and mu-band activity can affect each other, motor performance will be

enhanced by mu-band training. In this study, we used the term "mu-ERD" to define the

mu-band activity induced by MI. Here, we hypothesized that mu-ERD feedback during MI

will regulate motor-related ERDs, and the enhancement can improve actual motor

performance. For mu-ERD training, real time feedback of motor-ERD during MI will be

effective using NFB methods.

In addition to mu-ERD, beta-ERD is another MI-related ERD (Paret et al. 2019). Many

studies have proposed that beta-ERD is also related with motor planning, execution, and

imagery (Formaggio et al. 2010; Kraeutner et al. 2014; Nasseroleslami et al. 2014; Neuper

et al. 2006; Gert Pfurtscheller and Neuper 1997; Wheaton et al. 2009). Some reports have

shown a correlation of beta-ERD with motor performance (Gehringer et al. 2019;

Jochumsen et al. 2017) and motor learning (Pollok et al. 2014), but others have shown a

negative correlation between beta-ERD and performance (Dal Maso et al. 2018; Gehringer

et al. 2018). To limit the effect of neurofeedback, we focused on the mu-ERD feedback

and not beta-ERD. We performed an offline analysis of the transfer effect on beta-ERD

enhancement in the proposed protocol.

Based on our hypothesis that motor-related ERD enhancement can enhance motor

performance, we developed a training protocol combining MI and NFB (MI-NFB). This

protocol was designed to enable the training of vividness of motor imagery and mu-ERD

regulation at the same time, and to provide more stable effects than either MI or NFB

alone. When trainees imagined a given motor action under the protocol, they received mu-

ERD feedback related to the action. The training strategy consisted of imagining the

intended motor action to enhance its performance.

MI-NFB approach has been demonstrated in patients and healthy persons (*e.g.*, Spychala

et al. 2020; Wriessnegger et al. 2018; Zich et al. 2017, reviewed in Jeunet et al. 2019). In

the previous studies of patients, their purpose was to restore the patients' motor functions

to normal or near-normal levels, and the target movements were designed for

rehabilitation (Spychala et al. 2020; Zich et al. 2017). In studies of healthy persons, the

MI-NFB approach has been investigated with several neuroimaging techniques such as

Electroencephalogram (EEG) (Bai et al. 2014; Wriessnegger et al. 2018; Zich, De Vos, et

al. 2015; Zich, Debener, et al. 2015), Magnetoencephalogram (MEG) (Boe et al. 2014),

functional Magnetic Resonance Imaging (fMRI) (Zich, Debener, et al. 2015), and Near-

infrared spectroscopy (NIRS) (Kober et al. 2014, 2018). Furthermore, some reports have

examined frequency band changes associated with MI (Kraeutner et al. 2014;

Wriessnegger et al. 2018).

It is worth noting that most MI-NFB studies have aimed to develop a brain computer

interface or to optimize MI parameters. Fewer investigations have aimed to evaluate the

enhancement of motor performance through training. The aims of this study were, first, to

enhance motor performance by MI-NFB training and, second, to enhance MI ability or

mu-ERD. Additionally, the proposed protocol was carried out in a limited number of elite

athletes to test the possibility of its application and to examine each case.

Our objective was to demonstrate that our novel MI-NFB training protocol can enhance

motor performance more efficiently than conventional MI training. First, we verified the

effectiveness of the training protocol in a laboratory with university students. The

unfamiliar motor action was designed originally as a pegboard game to evaluate the effect

of MI-NFB on motor performance itself superior to MI ability or motor-related ERDs.

Subsequently, we investigated its applicability to sports performance with a small sample

of world-class tennis players as a preliminary study. The tennis serve was the motor action

enhanced by the training protocol.

RESULTS

Experiment 1: Training protocol for healthy volunteers

Participants were assigned to two groups: the MI-NFB group and the control group. The

motor performance and related neural activities were analyzed according to the

performance of the dominant and non-dominant hand separately. Before starting the

training, their general MI ability was assessed using the Kinesthetic and Visual Imagery

Questionnaire (KVIQ) (Kober et al. 2018). Visual imaging score (VIS), kinesthetic

imaging score (KIS), and the sum of KIS and VIS (SS) were analyzed (Malouin et al.

2007). There was no significant difference between the groups in VIS (t $= -0.71$, $P =$

0.49), KIS (t $= -1.0$, $P = 0.34$), and SS (t $= -0.89$, $P = 0.39$) before the training.

Clustering analysis

Many previous studies have reported that motor performance, even in skilled motor activities, is impaired by mental fatigue (Marcora et al. 2009; Schiphof-Godart et al. 2018). To verify the effects of the proposed training protocol without the influence of the harmful effects caused by mental fatigue, clustering analysis was performed for the fatigue reports and motor performance data. We used a data-driven method (a hierarchical cluster analysis of the mental fatigue during MI training in all participants) (see Fig. 1A and 1B for the detailed results). As a result, four participants who did not show a decrease in fatigue levels in the 3-day period were classified into another cluster (no-habituation group) (Fig. 1C), and they all underwent MI-NFB training. No one who performed a control task was classified as a no-habituation participant. The remaining data were separated according to the performed task into MI-NFB and control.

Dominant hand

We compared the working time of the motor performance test (MPT) using a pegboard before and after the training and analyzed by repeated measures ANOVA (pre-/post-training × groups). There was a significant main interaction effect ($F(2, 97) = 10.9$, $P = 0.0017$). The multiple comparison showed a significant time-series change in the MI-NFB group ($F(5,4) = 3.87$, $P = 0.013$, $\alpha = 0.017$), but not in the control or no-habituation groups ($F < 2.24$, $P > 0.10$). The multiple comparison showed a significant time-series change in the MI-NFB group ($t = 4.51$, $P = 0.011$, $\alpha = 0.017$), but not in the control or no-habituation groups ($P > 0.047$, $\alpha = 0.017$) (Fig. 2A).

To evaluate the training effect, we analyzed the changes of mu-ERD (8–13 Hz) during the

evaluation sessions using a repeated measures ANOVA (pre/post on the three training

days × groups). There was no significant main effect of groups ($F_{(2, 13)} = 2.78$, $P = 0.10$),

but a marginal main effect of the repetitions was observed ($F_{(5, 78)}$, $P = 0.041$). There

was no significant interaction effect ($F_{(10, 65)} = 1.35$, $P = 0.22$). (Fig. 3A). The mu-ERD

distributions are shown in Fig. 3B.

To investigate the transfer effect on beta band, we analyzed the beta-ERD changes (13–28

Hz). A repeated measures ANOVA (pre/post on the three training days × groups) showed

a significant main effect of group ($F_{(2,13)} = 6.27$, $P = 0.012$) and a marginal significant

interaction effect ($F_{(10,65)} = 1.87$, $P = 0.086$). The multiple comparison showed a

significant time-series change in the MI-NFB group ($F_{(5,4)} = 3.87$, $P = 0.013$, $\alpha = 0.017$)

but not in the control or no-habituation groups ($F < 2.24$, $P > 0.10$). Post-hoc test showed a

significant difference of beta-ERD between pre-training on day 1 and post-training on day

3 in the MI-NFB group ($t = 4.26$, $P = 0.013$) (Fig. 4A). The beta-ERD distributions were

indicated in Fig. 4B.

Non-dominant hand

The repeated measures ANOVA showed no significant main effect or interaction (pre-

/post-training × groups) in the working time of MPT using the non-dominant hand ($F <$

1.10, $P > 0.31$) (Fig. 2B).

To evaluate the training effect, we compared the mu-ERD during the evaluation session

using a repeated measures ANOVA (pre/post on the three training days × groups). There

was no significant main effect of group ($F(2, 13) = 0.37$, $P = 0.70$) and no significant

interaction effect ($F(10, 65) = 0.68$, $P = 0.74$) (Fig. 5).

The beta-ERD changes were analyzed in the dominant hand. A repeated measures

ANOVA (pre/post on the three training days × groups) showed no significant main effect

of group ($F(2, 13) = 2.48$, $P = 0.12$). There was no significant interaction effect ($F(10, 65)$

$= 0.88$, $P = 0.56$) (Fig. 6).

MI vividness

To evaluate the effect of MI-NFB on the MI subjective scale, the confidence rates were

measured. The self-confidence rate of the MI-NFB group was significantly lower than that

of the control group ($P = 0.012$).

To evaluate the effect of MI-NFB on MI ability and KVIQ scores were analyzed after the

training. These measurements showed no significant differences between the MI-NFB and

control groups, including No-habituation participants at post-measurement, VIS ($t = -0.59$,

$P = 0.57$), KIS ($t = -0.88$, $P = 0.40$), and SS ($t = -0.77$, $P = 0.46$).

Experiment 2: Effects of MI-NFB on elite tennis athletes

220   In the second experiment, we investigated the effect of the proposed training on elite

221   athletes. Five world-class male tennis players (T01–T05) were enrolled (Table 1). Because

222   of the small number of target participants, we analyzed the training effects individually

223   and did not perform any statistical analyses.

Motor performance test

Motor performance was evaluated individually based on the number of serves that hit the

target and the estimated service speed. Despite the lack of improvement in service speeds

(Fig. 7B), the number of hits of three players increased from pre- to post-MI-NFB training.

The number of hits of player T02 did not change, while the number of hits of player T03

decreased. (Fig. 7A). Motor performance was also evaluated subjectively based on the

self-evaluation of each player for each service by the visual analog scale method. The

subjective evaluations for services displayed an increase in players T02 and T04, while

there was a decrease in players T03 and T05, and no changes in player T01 (Fig. 7C).

The mu- and beta-ERD

236   To evaluate the training effect, relative mu-ERD and beta-ERD were analyzed by

237   subtraction of pre-training on day 1 from post-training on day 3 in all electrodes. The

238   topography indicated a grand average from right-handed athletes except for T04. The

239   topography of mu-ERD showed enhancement after the training, especially in the

240   ipsilateral side, though beta-ERD showed less enhancement than mu-ERD (Fig. 8).

DISCUSSION

In this study, we aimed to enhance motor performance by using MI-based NFB training.

We introduced an original performance test and developed a novel training protocol

unifying all MI and NFB tasks into the original test. To verify the effectiveness of the

proposed protocol, we compared motor performance following a training protocol with

that following MI training alone. In Experiment 1, motor performance and beta-ERD

amplitudes of participants who were trained under the protocol were more enhanced

compared with those of the participants trained with MI alone, though we could not

observe significant changes in mu-ERD. In Experiment 2, motor performances were

improved in three of five athletes, and mu-ERD amplitudes of the athletes trained by the

modified protocol for tennis players.

First, we discuss the different results between Experiment 1 and 2. Although both

protocols at least partly enhanced motor performances, ERDs in different frequency bands

were trained, mu or beta. This dissociated effectiveness in neuronal activity probably

could be attributed to the participants' familiarity levels of the motor action to learn. Both

of mu- and beta-ERD are well known MI- or motor-related activities (Formaggio et al.

2010; Kraeutner et al. 2014; Nasseroleslami et al. 2014; Neuper et al. 2006; Gert

Pfurtscheller and Neuper 1997; Sobierajewicz et al. 2017; Toriyama et al. 2018; Wheaton

et al. 2009; Wolf et al. 2014; Wriessnegger et al. 2018). With regard to mu-ERD, larger

mu-ERD has been observed more often in elite athletes than novices (Del Percio et al.

2007; Wolf et al. 2014). Beta-ERD has been discussed in the context of the motor learning

and efficiency after learning (Gehringer et al. 2019; Jochumsen et al. 2017; Pollok et al.

2014). Beta-ERD has been related to the maintenance of a current motor state (Engel and

Fries 2010), and, in one study, the short-term training was found to enhance beta-ERD,

while longer-term training did not result in changes of beta-ERD compared with pre-

training (Jochumsen et al. 2017). From these investigations, we can formulate the

hypothesis that beta-ERD can be enhanced when learning an unfamiliar motor action, but

it will be decreased after the action is familiarized. In line with this interpretation, a report

has shown a correlation between the beta-ERD suppression and motor skill retention

(Gehringer et al. 2018). In Experiment 1, we designed a novel motor task with short

training periods. To acquire the new motor action, beta-ERD was modulated by the MI-

related feedback. On the other hand, tennis services were very familiar among tennis elite

athletes, and beta-ERD did not changed. The protocol used in Experiment 2 would

enhance motor-related cortical activity by a different mechanism than that in Experiment

1, and it resulted in the enhancement of the motor performance and mu-ERD. The

different frequency band of enhanced ERD might reflect the different methods of learning

unfamiliar and familiar motor actions.

We hypothesized a mechanism for the proposed protocol, as described in Fig. 9C, before

we performed Experiment 1. According to our hypothesis, the enhanced motor

performance is derived from the mu-ERD feedback specified by the MI strategy of the

given motor action. The trainee can enhance their mu-ERD from MI with feedback using

own mu-ERD. This hypothesis (Fig. 9C) seems to be acceptable in the context of motor

learning of familiar actions like Experiment 2. The mu-ERD is believed to reflect motor

attention and multisensory integration, which are necessary to perform tasks more

efficiently (G Pfurtscheller and Lopes Da Silva 1999; Wolf et al. 2014). From these

reports, the effects of our training in Experiment 2 could be interpreted as the

enhancement of the mu-ERD, which means that enhancing motor attention and integrating

multisensory information results in a more efficient performance of an imagined motor

action, and that mu-ERD enhancement may have affected the high-performance of a

familiar action such as tennis service.

From the results of Experiment 1, we hypothesized another mechanism for unfamiliar

motor learning (Fig. 9A). The participants in MI-NFB were instructed to image the action

as clearly as possible, and the feedback circle was presented as a marker of proficiency in

the given motor action. Note that, they were not instructed to modulate the feedback

activity itself. Thus, our results that beta-ERD enhancement resulted in the protocol using

mu-ERD feedback would not contradict. The protocol used in Experiment 1 enabled

enhancement of MI proficiency of a given action that was unfamiliar to the participants,

and it provided enhancement of motor performance. We supposed that the beta-ERD

enhancement resulted from the learning of an unfamiliar action.

Next, we discuss the group comparisons among the MI-NFB, no-habituation, and control

groups, and subsequently, we interpret the results of the professional tennis athletes

individually.

With regard to the MI-NFB and control groups in Experiment 1, we found that the

performance of the dominant hand was enhanced after MI-NFB training. The participants

could enhance their beta-ERD from MI using their own mu-ERD as the feedback. Several

studies have reported that motor performance and ERDs of MI were positively correlated

(Babiloni et al. 2008; Gibson et al. 2014). We can speculate that the mu-ERD feedback

and strategy specification contributed to the success of the proposed training. In the

control group, the motor performance was not enhanced, which suggests that participants

were likely presented with feedback irrelevant to their own mu-ERD (Fig. 9B).

In conventional MI training studies, effects on motor performance are still controversial

and may be attributable to the lack of mu-ERD feedback. The degree of MI vividness

varied individually. Conventional studies did not involve any feedback of mu-ERD toward

the degree of vividness, and the lack of the feedback might have provided ambiguous

effect (Fig. 9D). In the conventional NFB-based studies on sports, motor performance

could not be significantly enhanced, and beta-ERD was not trained by mu-ERD feedback.

We speculate that there are two reasons for this: 1) the use of free strategies and 2) the

irrelevant feedback to the motor action to be trained (Fig. 9E).

Contrary to our expectation, enhancement of motor performance while performing a motor

action using the non-dominant hand was not observed after MI-NFB training. Generally,

the efficiency of motor learning for a new action is lower in the non-dominant hand than

in the dominant hand. We can expect that the participants failed to learn the unfamiliar

action with the non-dominant hand using the training protocol. During the last evaluation

session, no significant difference in working time was observed when an action was

performed using the non-dominant hand. These results indicate that the training effect was

not sustained until the last evaluation session. If we could prolong the training period long

enough to sustain the beta-ERD enhancement until the last evaluation session,

performance with the non-dominant hand would also likely be enhanced. Several reports

have observed intermanual transfer, which is the ability to promote learning from a trained

limb to another untrained limb using MI, and they observed the transfer effect from the

trained non-dominant side to the untrained dominant side (Amemiya et al. 2010; Land et

al. 2016; Van Mier and Petersen 2006). While these reports observed the transfer in

keypad tapping or finger tracking, we employed an original peg-task including grasp,

move, and insertion of small pegs. The complexity of a given action might have

suppressed intermanual transfer from the dominant to the non-dominant hand. In the

present experiment, we could not investigate the intermanual transfer effect precisely

because both sides (dominant and non-dominant) underwent training. To analyze

intermanual transfer in the proposed protocol, it is necessary to train only one side and

examine the effect on the untrained side.

Self-confidence rates of MI were significantly higher in the control group than in the MI-

NFB group. These results indicate that there was a gap between subjective evaluation and

objective evaluation of MI. This cognitive gap was suggestive of a difficulty in estimating

the ability of MI by subjective assessment alone.

KVIQ was conducted to investigate the effects on the abilities of general MI for any motor

action by the training. There was no significant increase of KVIQ after the training. The

result indicated that the general MI abilities were not affected by the training in the

proposed protocol.

The participants in the no-habituation group who did not experience decreased mental

fatigue during the training did not show enhanced motor performance and mu-ERD.

Several previous studies have reported that high mental fatigue worsens motor

performance. The mu-ERD of the no-habituation group was not enhanced by training.

This result indicates that the four participants in the no-habituation group could not adapt

to the NFB training. The stress of not being able to adapt to the NFB training caused

mental fatigue, which may have worsened the motor performance. Therefore, controlling

the mental fatigue of trainees may be necessary for successful NFB training.

Experiment 2 with tennis athletes was performed to assess the extendibility of the

proposed training protocol. The training protocol was modified for tennis services. As a

result, we found some training effects in most of the players. It is worth noting that our

purpose was to enhance motor performance, not to improve MI vividness or MI-related

ERD. In this experiment, three of five athletes could enhance their motor performance of

services. Contrary to Experiment 1, mu-ERD was enhanced after the training, while beta-

ERD was not. As we discussed above, we speculated that the mu-ERD enhancement

reflected the proficiency of MI and performance of the familiar action.

We performed Experiment 2 as a case study, thus we discuss the results individually as
follows. Player T04 showed remarkable effects which included: 1) the number of services
that hit the target and 2) the values of subjective evaluation increased.

For players T01 and T05, the number of serves that hit the target increased after the
training. In contrast, a decline in performance and ball speed was observed in player T03.
Just before the post-performance test, player T03 could not do a warm-up exercise like the
other participants; thus, the reduced warming up may have been the cause of the decline in
performance. The service performance of player T02 did not differ from that of the pre-
performance test. The mu-ERD of player T02 was not induced during training on day 3,
likely because he could not train satisfactorily as a result of a system error. This failure to
train mu-ERD on day 3 may have inhibited the enhancement of motor performance.

Our purpose in Experiment 2 was to investigate the extendibility of the proposed training
protocol for elite athletes as a case report. This case report showed enhancement of motor
performance in three athletes, and we suggest a possibility to use the protocol for field
application, not only in laboratory experiments.

Wilson and Peper have hypothesized that elite athletes can benefit more from
neurofeedback than non-athletes (Wilson and Peper 2011). Athletes are highly motivated
to succeed in improvement and win. Feedback process is much easier for elite athletes
because it is a familiar concept during training and practice. They can concentrate on

many types of training with the belief that they will provide positive results. These

characteristics and attitudes suggest the effectiveness of the proposed training.

It is useful to note that the protocol needs further modifications and investigations with a

quantitative approach. More detailed measurements will provide a conclusion that the

proposed training is effective for sports training.

The limitations of the study should be considered. In Experiment 1, the working time of

the MI-NFB group decreased in the post-training more than that of the control group by

mu-ERD feedback. These results strongly supported that the enhancement of motor

performance was a result of MI-NFB training. However, we cannot reject the possibility

that the motor performance was enhanced by the placebo effect and not by MI-NFB

training itself. In our protocol of Experiment 1, we told the participants in the MI-NFB

group that the visual stimuli were associated with their MI proficiency. Conversely, the

participants of the control group were not told the meaning of the visual stimuli but were

simply told to ignore them. This procedure was adopted to let the participants in the

control group keep their motivation. Participants who received sham stimuli as feedback

can easily notice their irrelevance, which lowers their motivation, as they were instructed

to focus on the stimuli. However, the difference in instructions might have led to enhanced

performance as a result of the placebo effect.

An additional limitation was the small sample size. In the training protocol (Experiment

1), the MI-NFB group was separated further by the cluster analysis due to their fatigue

levels. This procedure was necessary to consider the relationship between training effect

and habituation, but it inevitably reduced the sample size. Though the sample size was not

sufficiently large, we observed some enhancement of motor performance in our protocol.

Regarding the effects of MI-NFB on tennis athletes (Experiment 2), the results of the

training in world-class tennis players indicated that the proposed training protocol could

easily be applied to sports and could contribute to the enhancement of motor performance.

We did not conduct group analysis in the tennis players experiment, because it was

difficult to recruit enough world-class players. It is necessary to measure data from more

elite players and to statistically compare the effect in a future study.

In both experiments, we did not measure electromyography, and we could not confirm

whether the participants performed MI without the actual motion. We cannot reject the

possibility that the observed mu-ERD was induced by an actual movement and not by

imaging. We removed noisy data containing amplitudes exceeding 100 µV at Fp1. Most

artifacts generated by body movements should have been rejected automatically, but it

may not be enough to reject motor artifacts.

In conclusion, in the present study, we succeeded to enhance motor performance with the

proposed training protocol using MI-NFB. We verified the hypothesis that the

combination of MI and NFB would provide stable effects. The novelty of the proposed

protocol was the specification of the strategy used for the MI of the intended motor action,

which enabled feedback of EEG related to the MI and enhanced its performance. Notably,

the protocol was effective for the trainees who did not experience increased mental fatigue

during the training. If mental fatigue can be controlled appropriately, the proposed

protocol can contribute to the development of more effective methods with applications in

therapy and sports.

MATERIALS AND METHODS

Experiment 1: Training protocol for healthy volunteers

Participants

Eighteen Kyushu University students (9 women, 9 men; mean ± standard deviation (SD)

aged 20.3 ± 1.7 years old; right-handed) were recruited for this study. One female

participant was removed from our analysis because she could not perform our MPT due to

an injury to her left thumb. Therefore, we analyzed the data from 17 participants (8

women; age, 20.4 ± 1.7 years old; right-handed). All participants performed the Flanders

handedness test in Japanese to assess individual handedness (Okubo et al. 2014). All

participants were healthy and had no previous history of neurologic disease or history of

participating in similar MI experiments. All participants provided written informed

consent. The experiments including any relevant details were approved by the local ethics

committee of the Faculty of Arts and Science, Kyushu University. All experiments were

performed in accordance with relevant guidelines and regulations.

Participants were divided into two groups, the MI-NFB group (5 women, 4 men; 20.3 ±

1.9 years old) and the control group (3 women, 5 men; 20.5 ± 1.6 years old), to avoid any

biases of age ($t(15) = -0.20$, $P=0.85$), sports experience ($t(15) = 0.68$, $P = 0.51$), and the

time period after retirement from sports ($t(15) = 6.39$, $P = 0.70$). One participant in control

group was removed in EEG analysis because of the measurement problem on day 1.

Experimental training procedure

The participants sat on a comfortable armchair in an air-conditioned soundproof room. A

monitor was placed at approximately 1.2 m in front of the participants (1920×1080

pixels).

The participants were asked to come to the experimental room for three days over a one-

week period. On the first day, all participants engaged in a pre-test session, to undergo the

assessment of motor performance; an evaluation session; a training session; and another

evaluation session. On the second day, they participated in the first evaluation session,

training session, and second evaluation session. On the third day, they first participated in

the evaluation session, and then in the training session, second evaluation session, and

post-test session for the assessment of motor performance.

Motor performance test (MPT)

We designed an MPT to control for unfamiliarity of the bilateral performance using a

pegboard, similar to that used in previous studies for rehabilitation scenarios (Braun et al.

2017; Opara et al. 2017). In the performance test, the participants were instructed to insert

four pegs into four different holes on the right or left side, as quickly as possible.

Immediately after the insertion, they were required to perform a 180° rotation of the pegs

and to reinsert them in the holes located on the opposite side. Finally, they removed all

pegs from the holes and pressed a key to measure the performance time (Fig. 10A). In the

right-hand trial, in which the participants were directed to use their right hand, they moved

the pegs from the right to the left side, and in the left-hand trial, they moved the pegs from

the left to the right side. Each trial started with the presentation of an instruction to 'Relax'

on the monitor for 5 s. An indication of 'Left' or 'Right' then appeared for 3 s to specify

the lateral trial. Next, a fixation cross was presented, and the participants started the

sequence of MPT. The fixation cross was presented while performing the task until a key

was pressed. Before the beginning of the next trial, a blank screen re-appeared. The

duration of the blank screen was randomly changed in the range of 2.1–2.5 s (Fig. 10C).

Five trials of the performance test were conducted with each hand. The performance time

was recorded from the time of presentation of the fixation cross to the moment of key

pressing, and this performance time was used to evaluate motor performance bilaterally. A

shorter performance time indicated high motor performance.

Training session

In the MI-NFB training session, the participants performed MI using the same motor

action as in the MPT. The displayed instructions and timings were designed to be the same

as in the MPT trials (Fig. 10C). After laterality was specified, a circle was presented

behind the fixation cross as a visual feedback, and the participants started MI. In the MI-

NFB group, the radius of the circle varied according to their mu-ERD (see EEG analysis

section below and equations 1 and 2). Participants were told that when they performed a

better MI, the feedback circle would be smaller. They were instructed to attempt to make

the circle smaller than the fixation cross, and as small as possible, by performing better

MI. When the length of the fixation cross was same as the circle radius, it indicated 10%

mu-ERD. In the control group, the radius of the feedback circle varied according to the

mu-ERD of another participant used as sham stimulation. They were instructed to ignore

the feedback circle and to concentrate on the MI.

A one-day training session consisted of five blocks, and one block included 5 MI trials for

each hand (Fig. 10B). Therefore, a total of 50 MI trials were conducted in a training

session. After each block, the participants took a 60-s break and answered a questionnaire

using a 7-point scale indicating the difficulty of the MI (1: very easy, 7: very difficult),

degree of mental fatigue (1: no fatigue, 7: high fatigue), and concentration (1: no

concentration, 7: high concentration). Then they provided a self-confidence rate in the

range of 0–100% for the performance of MI.

Evaluation session

In the evaluation session for the training effect, the participants performed MI as in the

MPT trial but without any visual feedback (Fig. 10B). The display instructions and timing

were the same as in the MPT trials (Fig. 10C). Participants in all groups were instructed to

concentrate on MI by focusing on the fixation cross in each trial. There were no

differences in instructions or display indications for all groups.

One evaluation session had a block including 5 MI trials for each hand, and this session

was conducted before and after the one-day training session (Fig. 10B). After each

evaluation session was completed, the participants took a 60-s break.

EEG recording

EEG data were recorded using Cognionics Quick-20 headsets (CGX, San Diego, CA,

USA), consisting of 20 active-dry electrodes. These electrodes were placed according to

the international 10-20 system, and the recorded signals were digitized with a sampling

rate of 500 Hz. The reference electrode was placed on A1, at the left earlobe, and the

ground electrodes were placed near Fp1 and Fp2 at the forehead. An additional electrode,

A2 at the right earlobe was placed and measured (Fig. 10D). The active shield and active-

dry electrodes in the Quick-20 enable active noise cancellation with high sensor

impedances. The recommended sensor impedance of this device was below 2,500 kOhm,

but it can tolerate up to 5,000 kOhm. In this study, impedance levels were maintained

below 500 kOhm, which was 20% of the recommended level: 2500 kOhm. The impedance

levels were monitored during the EEG recordings. We recorded EEG data during the

training and evaluation sessions using a Data Acquisition Software (CGX, San Diego, CA,

USA).

Clustering analysis according to performance and fatigue levels

Previous reports have pointed out that motor performance is impaired by mental fatigue (Marcora et al. 2009; Schiphof-Godart et al. 2018). To evaluate the training effect without impairment from mental fatigue, the participants who showed different fatigue patterns should be discriminated. To classify the participants, the fatigue levels and motor performance data were subjected to Ward's hierarchical cluster analysis using Euclidean distances. The fatigue levels were recorded after each session every day, and we used the relative difference of fatigue data from day 1 to day 3 for clustering analysis, that is, five fatigue changes in every participant. The motor performance data were obtained by subtraction of working time on day 1 from day 3. The cluster analysis classified the data by fatigue changes and motor performances from day 1 to day 3.

EEG analysis

Real-time analysis

The decrease in real-time mu power from baseline (mu-ERD increased) was computed from the streaming EEG data. The recorded stream EEG data were stored for 1.024 s (512 points) before the onset of updating visual feedback, and the 1.024 s data were used to compute mu-ERD to define the radius of the feedback circle. We defined 8–13 Hz as the mu band. The baseline interval was defined as a relaxation period of 5 s just before a hand cue was presented (−8 s to −3 s before the hand cue onset), and the MI task interval was defined as the period from the presentation of the fixation cross to the moment of key

572   pressing (0 to key pressing). We computed the mu-ERD using the following equation:

$$\text{mu-ERD}(t) = \frac{A(t) - R}{R} \times 100, \tag{1}$$

574   where $A(t)$ is the power spectrum density of the time window during an MI trial at time t,

575   and R is the baseline power at the trial. The radius of the feedback circle was computed

576   from mu-ERD by using the following equation:

$$\text{radius}(t) = 5 \times \text{mu-ERD}(t) + \text{radius}_{BL}, \tag{2}$$

where radius_{BL} = 300 pixels. The visual angle when the radius_{BL} was presented was

approximately 10.5°. The analytic time window during an MI trial was a sliding window

with approximately 90% overlap. Accordingly, the feedback was updated approximately

every 100 ms, and the analytic time window during the relaxation periods (baseline

interval) was the sliding window without overlapping. The data in the baseline interval

was computed as mu power in each window and averaged. The averaged data was used for

the baseline power calculation (R in equation 1). The mu power spectrum density was

computed by using the 'psd_array_multitaper (array data, sampling rate)' function in the

Python MNE module. To remove eye-blinking artifacts, the time windows containing data

with absolute values exceeding 100 µV at the Fp1 electrode were rejected. If the 1-s

feedback was rejected because of the occurrence of artifacts during an MI trial, the data

before the last feedback was used to compute the radius of the visual circle. Therefore, the

size of the circle did not change when the data feedback contained artifacts. The data

recorded at C3 were used to compute the mu-ERD of the right-hand MI, and the data at C4

were used to compute the mu-ERD of the left-hand MI. These analyses and feedback

systems were developed using Python.

Offline analysis

Raw EEG data were band-pass filtered (3–50 Hz) and notch-filtered at 60 Hz. The recorded signals were re-referenced to the average of A1 and A2. The data were divided by −8 s to 15 s after the onset of MI start as one epoch. The mu-ERD was computed using equation (1). The EEG power from −6 s to −3 s was used as the baseline (R), and the power from 1 s to 5 s was used as an MI task activity (A). The remaining data from 5 s to 15 s after the MI onset was removed to exclude movement-related artifacts and EEG activity of the motor preparation for the button pressing.

Statistical analysis

In Experiment 1, we compared the data among the three groups, MI-NFB, control, and non-habituation. The motor performance were compared by a repeated measures ANOVA (pre/post all training days × group), and we used a t-test as post-hoc test. The mu-ERD and beta-ERD during the evaluation session was assessed by a repeated measures ANOVA (pre/post in each training day × group), and the t-test was used a post-hoc test. The significance α-value was set to 0.05, and Bonferroni correction was used for all statistical analyses. These tests were conducted in SciPy module (ver. 1.2.1) in Python (ver. 3.7.2).

Experiment 2: Effects of MI-NFB on tennis athletes

Participants

615   Five world-class male tennis players were enrolled, and they ranked in the top 200 of

616   Association of Tennis Professionals (ATP) men's tennis ranking on May 13, 2019. The

617   demographic data are summarized in Table 1.

618   Because of the inclusion of a small number of targeted participants, we analyzed the

619   obtained data individually and did not perform a group comparison.

620   Experimental procedure

Only a few modifications were made to the experimental procedure used for healthy

volunteers. The MPT was designed for tennis athletes as follows.

Training effects were evaluated by the performance of tennis serves. All participants hit 10

services before and after the training the right or left side on a tennis court. After each

service, a participant evaluated the service subjectively using a visual analog scale ranging

from "worst performance" to "best performance." In the MPT, a service side was specified

for each participant, from which he was not good at shooting a serve. The service side to

shoot was determined according to the pre-training questionnaire of each participant. To

evaluate the service performances, an 81 cm × 81 cm target area was designated on the

tennis court (Fig. 11A). All players were instructed to hit the target area. The players'

motions and ball trajectories during services were recorded with three cameras from

different locations. Ball speeds were estimated from videos recorded in 480 fps using the

number of frames between the hit ball and grounded ball. We evaluated motor

performance using the number of hits on target and estimated ball speeds. After all tasks

were completed in the training protocol, we interviewed the participants.

MI-NFB training

The MI-NFB training procedure was the same as that used for healthy volunteers. For the MI task in the athletes, participants were instructed to imagine the motion of serving from a side that was specified in the MPT (Fig. 11B). In this task, hand laterality of imaging was the same as the hand used by each participant to grip a racket.

Online analysis

On day 1 of the training protocol for athletes, the size of the feedback circle was computed by equation 1 as in the case of healthy volunteers, and the coefficient of ERD in equation 1 was changed to 4 on day 2 and to 3 on day 3. This change was intended to increase the difficulty of MI-NFB training, because we hypothesized that elite athletes have superior motor imagery ability.

Declarations

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Conflict-of-interest statement

The authors declare that they have no conflicts of interests.

Availability of data and material

The datasets generated during and/or analysed during the current study are available from the corresponding author on reasonable request.

Code availability

None.

Author Contributions

C.S. and T.O. conceived the idea and designed the study. C.S. and K.T. collected the data and performed the analysis. C.S. and K.T. wrote the draft with critical comments from T.O., and all authors discussed and finished the manuscript.

Ethics approval

666 The experiments including any relevant details were approved by the local ethics

667 committee of the Faculty of Arts and Science, Kyushu University (#201905, #20).

Consent to participate

669 All participants provided written informed consent.

Consent for publication

671 All authors consent for publication.

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Figures captions

Fig. 1 A: Dendrogram of the hierarchical clustering analysis of mental fatigue and motor

performance test (MPT) working time. B: Scatter plots showing changes in mental fatigue

according to MPT working time. C: Mean and SEM of mental fatigue in each group

Fig. 2 Mean and standard error of mean (SEM) of the working time of the motor

performance test (MPT) with the dominant (A) and non-dominant hands (B) between pre

and post training. *: $P < 0.017$, with a Bonferroni correction. MI-NFB: mental imagery-

neurofeedback

Fig. 3 Results of mu-ERD with the dominant hand

A: Mean and SEM in the mu-ERD from the pre-training session on day 1 to post-training

on day 3. B: Topography of the grand averaged mu-ERD in each group. Color bars show

ERD [%]. White circles indicate channel location of C3, which was used as feedback

electrode. ERD: event-related desynchronization, MI-NFB: mental imagery-

neurofeedback

Fig. 4 Results of beta-ERD with the dominant hand

A: Mean and SEM in the mu-ERD from the pre-training session on day 1 to post-training

on day 3. B: Topography of the grand averaged mu-ERD in each group. Color bars show

ERD [%]. White circles indicate channel location of C3, which was used as feedback

electrode. ERD: event-related desynchronization, MI-NFB: mental imagery-

neurofeedback

Fig. 5 Results of mu-ERD with the non-dominant hand

Mean and SEM in the mu-ERD from pre-training on day 1 to post-training on day 3.

ERD: event-related desynchronization, MI-NFB: mental imagery-neurofeedback

Fig. 6 Results of beta-ERD with the non-dominant hand

Mean and SEM in the mu-ERD from the pre-training session on day 1 to post-training on

day 3. ERD: event-related desynchronization, MI-NFB: mental imagery-neurofeedback

Fig. 7 Results of experiments involving tennis athletes

A: Changes in the number of hits on target before and after the MI-NFB training. B:

Changes in the estimated speeds of serves before and after the MI-NFB training. C:

Changes in subjective evaluation before and after the MI-NFB training. MI-NFB: mental

imagery-neurofeedback

Fig. 8 Topography of subtraction from post-training on day 3 to pre-training on day 1

The colors showed grand averaged mu-ERD (left) and beta-ERD (right) of right-handed

tennis athletes. ERD: event-related desynchronization

Fig. 9 Conceptual scheme of the mechanism for the enhancement or non-enhancement of

motor performance

A: MI-NFB group with unfamiliar motor action. B: Control group. C: Tennis athletes with

familiar motor action. D: Conventional MI training. E: Conventional NFB training. ERD:

event-related desynchronization, MI: mental imagery, NFB: neurofeedback

Fig. 10 Experimental design of Experiment 1

A: Motor performance test (MPT) procedure. B: Procedure of the three training days. C:

Procedure of each session. D: Electrode setup

Fig. 11 Experimental design of Experiment 2

A: Location of target of service, which was performed as a motor performance test (MPT).

B: Procedure of the three training days

Table 1. Demographic information of the tennis players

Participant's ID	Age	Racket hand	Serve side	Best world ranking
T01	26	Right	Left	64
T02	31	Right	Left	36
T03	21	Right	Left	171
T04	24	Left	Right	58
T05	27	Right	Left	78

Figures

Figure 1

Figure 2

Figure 3

Figure 4

Figure 5

Figure 6

Figure 7

Figure 8

Tennis athletes
Post - Pre (Mu-ERD)

Tennis athletes
Post - Pre (beta-ERD)

Figure 9

Figure 10

Figure 11