Operation of a P300-based brain-computer interface in patients with Duchenne muscular dystrophy

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A brain-computer interface (BCI) or brain-machine interface is a technology that enables the control of a computer and other external devices using signals from the brain. This technology has been tested in paralysed patients, such as those with cervical spinal cord injuries or amyotrophic lateral sclerosis, but it has not been tested systematically in Duchenne muscular dystrophy (DMD), which is a severe type of muscular dystrophy due to the loss of dystrophin and is often accompanied by progressive muscle weakness and wasting. Here, we investigated the efficacy of a P300-based BCI for patients with DMD. Eight bedridden patients with DMD and eight age- and gender-matched able-bodied controls were instructed to input hiragana characters. We used a region-based, two-step P300-based BCI with green/blue flicker stimuli. EEG data were recorded, and a linear discriminant analysis distinguished the target from other non-targets. The mean online accuracy of inputted characters (accuracy for the two-step procedure) was 71.6% for patients with DMD and 80.6% for controls, with no significant difference between the patients and controls. The P300-based BCI was operated successfully by individuals with DMD in an advanced stage and these findings suggest that this technology may be beneficial for patients with this disease.
conventional P300 speller. We then showed that the two-step procedure for the visual P300 BCI system provided significantly increased accuracy for ALS patients, compared with a conventional RC speller20.

McDonald showed that muscle weakness and muscle fatigue were important factors that adversely affected the quality of life in neuromuscular disease patients21. ALS is a well-known neuromuscular disease, but there are other neuromuscular diseases. Duchenne muscular dystrophy (DMD) is a neuromuscular diseases characterised by skeletal muscle atrophy of the whole body affecting proximal muscles and causing gait disturbance, due to the loss of the dystrophin-encoding gene, which encodes one of the most important proteins in myocyte membranes22. The DMD patients ultimately progress to a bedridden state requiring full assistance. In the past, their average life span was considered to be about 20 years, primarily because of respiratory failure, but due to advances in respiratory management, that life span has now increased to around 3023,24.

Patients with DMD show muscle weakness and muscle fatigue. After undergoing a tracheostomy, it becomes difficult for them to communicate verbally. These patients do not usually show deterioration in cognitive function and a worsening of consciousness level is associated with progression of the medical condition. Patients with DMD often use alternative augmentative communication (AAC) devices, but muscle weakness and muscle fatigue make it hard to use them continuously. Thus, it is desirable that patients with DMD also have the option of using communication devices that do not depend on muscle activity.

In this study, we investigated the efficacy of a P300-based BCI for patients with DMD. Bedridden patients with DMD and age- and gender-matched able-bodied controls were instructed to input hiragana characters using a region-based, two-step P300-based BCI. Here, we show that the P300-based BCI was operated by individuals with DMD at an advanced stage.

**Results**

We used a region-based two-step P300-BCI speller (Fig. 1)20 to compare the accuracy of hiragana character input in a DMD patient group and an age- and gender-matched able-bodied participant group. Eight patients with DMD (age: 23–38 years, mean: 30.3), diagnosed by genetic testing and proof of dystrophin deficiency, were recruited as participants (Table 1). These patients had experienced walking disabilities before reaching 5 years old, and their current level of activities of daily living (ADL) was at the total assistance level; their modified Rankin Scale score was 5. Online and offline performances in both groups were evaluated.

**Performance in BCI operation.** The mean online accuracy in region-based two-step P300-BCI operation, which indicated the accuracy of both steps in total, was 79.8% for patients with DMD and 83.4% for controls, which were not significantly different (two tailed t-test, p = 0.69, df = 7; Fig. 2(a)).

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**Figure 1.** A region-based two-step P300-based hiragana speller. The subject was required to count the number of intensifications of the green/blue circle containing the target character to be input. Each circle flashed eight times, and the interval between two flashes was 175 ms, consisting of 100 ms of intensification (green) and 75 ms of rest (blue). We divided the 6 × 9 matrix into six circled regions including nine characters each. First, each region was intensified individually. When a region that included a target character was selected, the speller matrix moved to the second step. The second step used 3 × 3 regions with one character each, and each region was again intensified.
Figure 2(b) shows the results of the offline analyses for both groups in each flashing sequence. We used two-way repeated-measures ANOVA with group (DMD vs. control) and accuracy for each sequence (1–8). The main effect of group was not significant ($p = 0.0502, df = 1$; Fig. 2(b)), but the main effect of number of sequence was significant ($p = 3.71 \times 10^{-7}, df = 7$; Fig. 2(b)). The interaction effect was not significant ($p = 0.99, df = 7$).

Table 1. Demographic and clinical characteristics of the patients with DMD.

| No | Age | Gender | mRs | Tracheostomy | Communication                                   | Genetic deletion |
|----|-----|--------|-----|--------------|-----------------------------------------------|-----------------|
| 1  | 33  | male   | 5  | +            | lip-reading/visual scanning board exon 48–50   |                 |
| 2  | 28  | male   | 5  | +            | lip-reading/visual scanning board exon 18–44   |                 |
| 3  | 28  | male   | 5  | −            | speech exon 51                                  |                 |
| 4  | 27  | male   | 5  | −            | speech exon 48–50                               |                 |
| 5  | 23  | male   | 5  | −            | speech exon 48–52                               |                 |
| 6  | 38  | male   | 5  | +            | lip-reading/visual scanning board/ AAC device  | exon 7–9        |
| 7  | 36  | male   | 5  | +            | lip-reading/visual scanning board unspecified  |                 |
| 8  | 29  | male   | 5  | +            | lip-reading/visual scanning board unspecified  |                 |

Figure 2. Online and offline accuracy in the BCI operation. (a) Online accuracy in the region-based two-step P300-BCI operation, which indicated the accuracy of both steps in total, was 79.8% for patients with DMD and 83.4% for controls; no significant difference was observed between them ($t$-test, $p = 0.69$). Each triangle (DMD) and circle (controls) indicates each data point. (b) Offline accuracy for both groups in each flashing sequence was plotted (triangle: DMD, and circle: controls).

Figure 2(b) shows the results of the offline analyses for both groups in each flashing sequence. We used two-way repeated-measures ANOVA with group (DMD vs. control) and accuracy for each sequence (1–8). The main effect of group was not significant ($p = 0.0502, df = 1$; Fig. 2(b)), but the main effect of number of sequence was significant ($p = 3.71 \times 10^{-7}, df = 7$; Fig. 2(b)). The interaction effect was not significant ($p = 0.99, df = 7$).

Performance on character input. The mean online accuracy of inputted characters (accuracy for the two-step procedure) in the DMD patients was 71.6% (9.34 bit/min) and that in the controls was 80.6% (11.24 bit/min) (Fig. 3(a)), which were not significantly different (two tailed $t$-test, $p = 0.32, df = 7$).

Figure 3(b) shows the results of the offline analyses for both groups under the two-step procedure. We tested a two-way repeated measures ANOVA with group (DMD vs. control) and the accuracy of each sequence (1–8).
The main effects of group and the number of sequences were significant (group, \( p = 0.00081, \text{df} = 1 \); number of sequences, \( p = 9.9 \times 10^{-13}, \text{df} = 7 \); Fig. 3(b)). The interaction effect was not significant (\( p = 0.94, \text{df} = 7 \)).

Note that the two-step procedure was accomplished only when the responses were correct for the first and second steps. The accuracy of the first step was 85.1% for patients with DMD and 83.9% for controls, and that for the second step was 74.6% for patients with DMD and 82.8% for controls.

**EEG waveforms.** EEG waveforms obtained from P4 electrodes were averaged for patients with DMD and controls, as shown in Fig. 4, in which dashed lines indicate non-target ERPs, and solid lines indicate target ERPs. The P4 electrode was chosen specifically because, in previous studies, we showed the importance of electrode position in the operation of visual P300 BCIs with green/blue flicker stimuli\(^{25,26}\). The averaged waveforms of the target ERPs differed between the groups. In controls, two positive peaks were seen at approximately 243 and 535 ms after the onset of stimuli, which corresponded to the early and late components of P300, respectively, whereas in the DMD group, no clear peak corresponding to the early component was seen, but a blurred peak that may correspond to the late component was seen at approximately 450–550 ms.

**Discussion**

We used a region-based two-step P300-BCI speller to evaluate accuracy in BCI operation and character input in bedridden DMD patients and able-bodied age- and gender-matched control participants. In online analyses, the DMD group achieved reliable accuracies, comparable to those of the control group in both BCI operation (Fig. 2(a)) and character input (Fig. 3(a)).

DMD is a neuromuscular disease that shows skeletal muscle atrophy throughout the body, involving the proximal muscle and causing gait disturbances\(^{22}\). DMD patients eventually progress to a bedridden state requiring full assistance. Patients with DMD show muscle weakness and muscle fatigue. After performing a tracheostomy, it becomes difficult for such patients to communicate verbally. In a clinical context, these DMD patients can be considered as candidate BCI users. In this study, we showed systematically, for the first time, that bedridden DMD patients were able to use the BCI speller.

It is important for bedridden DMD patients to be able to use the BCI system because deterioration of attention has been reported in these patients. For example, Piccini et al. used visuospatial attention tasks to compare DMD
and control groups, showing that accuracy in the DMD group was significantly lower than in the controls, and that this was true not only in voluntary attention but also in automatic attention27.

Use of the region-based two-step system may have contributed to the successful operation of the BCI speller. In a previous study, we performed character-input experiments using a conventional row/column P300 speller for amyotrophic lateral sclerosis (ALS) patients under artificial ventilation control, but they could not control it well. However, when they used a region-based two-step BCI speller, the accuracy improved20. Similar to DMD patients, it has been reported that the ALS in later stages is often accompanied by deterioration in attention 28,29. Thus, the region-based two-step system, which has a larger flashing area than a conventional visual array, may help in achieving better accuracies in these patients at an advanced disease stage.

Although the online analyses showed that the DMD group was able to achieve reliable accuracies comparable to those of the control group in both BCI operation and character input, further evaluation in the offline analyses showed some differences between the DMD and control groups. In the offline analyses of BCI input, two-way repeated measures ANOVA was used with group (DMD vs. control) and the accuracy for each sequence (1–8). The main effect of group was not significant, but the main effect of number of sequences was (Fig. 2(b)). In the offline analyses of character input, we also used ANOVA; the main effects of group and number of sequences were significant (Fig. 3(b)). Thus, for DMD patients, character input was slightly more difficult than BCI input than it was in the control group.

Further evaluation of the EEG waveforms also showed some differences between the DMD patients and controls. We averaged EEG waveforms obtained from P4 electrodes in both groups (Fig. 4). In the controls, two positive peaks were seen at approximately 243 and 535 ms after the onset of stimuli, corresponding to the early and late components of P300, respectively. In the DMD group, no clear peak corresponding to the early component was evident, but a blurred peak was seen at approximately 450–550 ms.

P300 consists of the early component, occurring 150–300 ms after the presentation of a target stimulus, and the late component, occurring 300–500 ms after the presentation of the stimulus. It has been suggested that the early component corresponds to a stimulus-driven frontal attention mechanism and the late component is due to parietal activity51. Discrimination between a target stimulus and non-target stimuli at the time of task execution leads to the initiation of activity in the frontal lobe region, responding sensitively to attention efforts. The early component is generated when such stimulation is processed due to sufficient attention concentration, whereas the late component is generated when the activation of the resource for subsequent attention promotes a memory operation in the parietal region30–32. Reports have described that patients with frontal lobe lesions show a decrease in the amplitude of the early component and that frontal lobe integrity is necessary to generate the early component33,34. Thus, the lack of a clear peak corresponding to the early component in DMD patients may be related to their altered frontal lobe function.

The altered frontal lobe function in DMD patients has recently been discussed in light of their genetic background. DMD is accompanied by the loss of dystrophin, and dystrophin is known to play an important role...
in the development of the central nervous system in the foetal brain during pregnancy. In particular, it has a
growth-promoting effect on neurons; expression of dystrophin in the brain contributes to synapse maturation and
stabilisation25,26,27. Dystrophin is known to be highly expressed in the frontal lobe, together with the hippocampus
and the cerebellum25,27,28. Thus, attenuation of the early component in DMD patients may be due to altered frontal
lobe function, and the preserved late component is consistent with preserved parietal lobe function in DMD
patients.

In this BCI study, using a region-based two-step P300-BCI speller, although the online analyses and EEG
waveform analyses showed some differences between DMD patients and controls, likely due to altered frontal
lobe function in DMD patients, in the online analysis, the DMD group achieved reliable accuracies, comparable
to those of the control group in both BCI operation and character input. Thus, the P300-based BCI was operated
successfully by individuals with DMD in an advanced stage, and these findings suggest that a BCI may be benefi-
cial for patients with DMD.

Methods
Participants. Eight patients with DMD, aged 23–38 years (mean: 30.3), diagnosed by genetic testing and
proof of dystrophin deficiency, were recruited as participants (Table 1). These patients had experienced walking
disabilities before reaching 5 years old, and their current level of activities of daily living (ADL) was at a total
assistance level; their modified Rankin Scale score was 5. Of these eight patients, five had undergone tracheoe-
tomies, and were under mechanical ventilation control, with tracheotomy positive pressure ventilation (TPPV).
They communicated via lip reading, facial expressions, and with visual scanning boards. One of the patients
used an alternative augmentative communication (AAC) device, called the Den-no-shin (Hitachi KE Systems,
Chiba, Japan). Three patients had not undergone tracheotomies and were barely able to speak, due to muscle
weakness and tongue hypertrophy. No obvious neurocognitive deterioration was observed. As a control group,
eight able-bodied age- and gender-matched participants were recruited. Neither the patient group nor the control
group had any BMI training beforehand.

The present study was approved by the institutional ethics committee at the National Rehabilitation Center
for Persons with Disabilities. All participants provided written informed consent according to the institutional
guidelines. All experiments were carried out in accordance with the approved guidelines.

Experimental procedure. The P300 speller is a conventional method in BCI, which applies the oddball task
paradigm that elicits P300-like responses12. Icons arranged in a row/column matrix are intensified randomly, and
the subject is required to focus on one of the icons to be input, and to count the number of intensifications of the
target icon. Focusing on the intensification of the target as a rare stimulus enhances the P300-like responses. In
the original P300 speller13, 36 icons consisting of the English alphabet and 10 numerals or symbols were used and
arranged in a 6 × 6 matrix.

In this experiment, we used a region-based, two-step P300 speller that used green/blue intensification
(Fig. 1)20. This system uses a two-step paradigm, modified from previous research25–28,29. We divided the 6 × 9
matrix into six circled regions (i.e. 2 × 3 regions) including nine characters each. First, each region was inten-
sified individually. Each circle flashed eight times, and the interval between the two flashes was 175 ms, consis-
ting of 100 ms of intensification (green) and 75 ms of rest (blue). The traditional visual stimulus in the visual
P300 BCI was a luminance change in the icons12, but we applied the green/blue colour combination, because we
showed previously that adding a green/blue chromatic change to the luminance improved performance on the
P300 speller29. The subject was verbally instructed to attend to the next target character to be input, and also to
count the number of intensifications of the green/blue circle containing the target character. When a region that
included the target character was selected, the speller matrix moved to the second step. The second step used 3 × 3
regions with one character each, and each region was again intensified. The subject was again verbally instructed
to attend to the next target character to be input, and also to count the number of intensifications of the green/
blue circle with the target character.

Four of the eight DMD patients rested in a supine position on a bed, and a liquid crystal display (LCD), on
which visual stimuli were presented, was set ~100 cm from their eyes. The other four DMD patients and the
able-bodied control participants sat in a chair, and the LCD was also set ~100 cm from their eyes. The size of the
LCD was 21.7 cm in height and 38.5 cm in width.

EEG recordings and BCI analyses. EEG data were recorded from eight channels (Fz, Cz, Pz, P3, P4, Oz,
PO7, PO8) using an in-house cap with solid-gel electrodes30. All channels were grounded to AFz and refer-
enced to Fpz. We used an in-house amplifier, with which the sampling rate was 1024 Hz, the precision of the
analogue-to-digital converter was 24 bit, and the hardware notch filter was 50 Hz. The recorded signals were
band-pass filtered (0.1–50 Hz) and the data were down-sampled to 21 Hz for analyses.

To identify the region at which the subject was gazing, first, we applied a preparation session that derived a
feature vector to discriminate target and non-target for each individual. We used a specifically prepared panel
that included 3 × 3 regions, which include nine characters each, during the preparation session. We asked the
subjects to attend to nine regions each to derive the feature vector. Each region was intensified eight times, thus
72 segmented EEG data for target and 576 segmented EEG data for non-target were used. A total of 800 ms
EEG data were segmented according to the timing of the intensification onset to derive the feature vector. The
first 100 ms, just prior to flash onset, were used to correct the baseline, and the remaining 700 ms were used for
classification. Because the 700 ms baseline-corrected EEG corresponded to 15 data points, and the data were
collected from eight EEG channels, the feature vector had 120 dimensions. Fisher's linear discriminant analysis
was used to compute the feature vector. Target and non-target characters were classified using the feature vector
during the test session. The maximum of the summed scores, which was the result of this classification, was used
to determine the region to which the subjects were attending. The classification processes were applied to both online and offline analyses.

**EEG waveform analyses.** We evaluated the elicited EEG waveforms. EEG signals obtained from P4 electrodes were averaged for patients with DMD and controls. We specifically focused on the P4 electrode because, in previous studies, we showed the importance of this electrode position in the operation of visual P300 BCIs with green/blue flicker stimuli. The 800 ms waveforms extracted from the onset of intensification were averaged. The first 100 ms just prior to the onset of intensification was used for baseline correction.

**References**

1. Wolpaw, J. R., Birbaumer, N., McFarland, D. J., Pfurtscheller, G. & Vaughan, T. M. Brain-computer interfaces for communication and control. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 113, 767–791 (2002).

2. Birbaumer, N. & Cohen, L. G. Brain-computer interfaces: communication and restoration of movement in paralysis. *The Journal of Physiology* 579, 621–636, https://doi.org/10.1111/j.1469-8986.2006.525633 (2007).

3. Eric, C. L., Gervin, S., Jonathan, R. W., Jeffrey, G. O. & Daniel, W. M. A brain–computer interface using electrocorticographic signals in humans. *Journal of Neural Engineering* 1, 63 (2004).

4. Hochberg, L. R. et al. Neuronal ensemble control of prosthetic devices by a human with tetraplegia. *Nature* 442, 164–171, https://www.nature.com/nature/journal/v442/n7099/suppinfo/nature04970_S1.html (2006).

5. Birbaumer, N. et al. A spelling device for the paralysed. *Nature* 398, 297–298, https://doi.org/10.1038/18581 (1999).

6. Donchin, E., Spencer, K. M. & Wijesinghe, R. The mental prosthesis: assessing the speed of a P300-based brain-computer interface. *IEEE Transactions on Rehabilitation Engineering: A Publication of the IEEE Engineering in Medicine and Biology Society* 8, 174–179 (2000).

7. McFarland, D. J. & Wolpaw, J. R. Sensorimotor rhythm-based brain-computer interface (BCI): feature selection by regression improves performance. *IEEE Transactions on Neural Systems and Rehabilitation Engineering: A Publication of the IEEE Engineering in Medicine and Biology Society* 13, 372–379, https://doi.org/10.1109/TNSRE.2005.848627 (2005).

8. Sakurada, T., Kawase, T., Komatsu, T. & Kansaku, K. Use of high-frequency visual stimuli above the critical flicker frequency in a SSVEP-based BMI. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 126, 1972–1978, https://doi.org/10.1016/j.clinph.2014.12.010 (2015).

9. Edlinger, G., Allison, B. Z. & Guger, C. in *Clinical Systems Neuroscience* (eds K. Kansaku, L.G. Cohen, & N. Birbaumer) Ch. 3, 33–66 (Springer 2015).

10. Sutton, S., Braun, M., Zuhn, J. & John, E. R. Evoked-potential correlates of stimulus uncertainty. *Science* 150, 1187–1188 (1965).

11. Polich, J. Updating P300: an integrative theory of P3a and P3b. *Electroencephalography and Clinical Neurophysiology* 70, 510–523 (1988).

12. Iekami, S., Takano, K., Saeki, N. & Kansaku, K. Operation of a P300-based brain-computer interface by individuals with cervical spinal cord injury. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 122, 991–996, https://doi.org/10.1016/j.clinph.2010.08.021 (2011).

13. Sellers, E. W., Ryan, D. B. & Hauser, C. K. Noninvasive brain–computer interface enables communication after brainstem stroke. *Science Translational Medicine* 6, 257re257, https://doi.org/10.1126/scitranslmed.3007801 (2014).

14. Piccione, F. et al. P300-based brain computer interface: Reliability and performance in healthy and paralysed participants. *Clinical Neurophysiology* 117, 531–537, https://doi.org/10.1016/j.clinph.2005.07.024 (2006).

15. Sellers, E. W. & Donchin, E. A P300-based brain-computer interface: initial tests by ALS patients. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 117, 538–548, https://doi.org/10.1016/j.clinph.2005.06.027 (2006).

16. Hoffmann, U., Vesin, J.-M., Ebrahimi, T. & Diserens, K. An efficient P300-based brain–computer interface for disabled subjects. *Journal of Neuroscience Methods* 167, 115–125, https://doi.org/10.1016/j.jneumeth.2007.03.005 (2008).

17. Nijboer, F. et al. A P300-based brain-computer interface for people with amyotrophic lateral sclerosis. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 119, 1909–1916, https://doi.org/10.1016/j.clinph.2008.03.034 (2008).

18. McCane, L. M. et al. P300-based brain-computer interface (BCI) event-related potentials (ERPs): People with amyotrophic lateral sclerosis (ALS) vs. age-matched controls. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 126, 2124–2131, https://doi.org/10.1016/j.clinph.2015.01.013 (2015).

19. Iekami, S., Takano, K., Kondo, K., Saeki, N. & Kansaku, K. A region-based two-step P300-based brain-computer interface for patients with amyotrophic lateral sclerosis. *Clinical Neurophysiology: Official Journal of the International Federation of Clinical Neurophysiology* 125, 2305–2312, https://doi.org/10.1016/j.clinph.2014.03.013 (2014).

20. McDonald, C. M. Physical activity, health impairments, and disability in neuromuscular disease. *American Journal of Physical Medicine & Rehabilitation* 81, S108–120, https://doi.org/10.1097/PHM.0000000000000767.43578.3C (2002).

21. Bushby, K. et al. Diagnosis and management of Duchenne muscular dystrophy, part 1: diagnosis, and pharmacological and psychosocial management. *The Lancet Neurology* 9, 77–93, https://doi.org/10.1016/S1474-4422(09)70271-6 (2010).

22. Toussaint, M., Chatwin, M. & Soudon, P. Mechanical ventilation in Duchenne patients with chronic respiratory insufficiency: clinical implications of 20 years published experience. *Chronic Respiratory Disease* 4, 167–177, https://doi.org/10.1177/1479972307080697 (2007).

23. Tsuyoshi, M. Clinical importance of central nervous system dysfunction in myopathy. *Brain and Nerve* 68, 10, https://doi.org/10.11177/10.11477/mf.11405260 (2016).

24. Iekami, S., Takano, K., Wada, M., Saeki, N. & Kansaku, K. Effect of the green/blue flicker matrix for P300-based brain-computer interface. *Front Neurol* 3, 113, https://doi.org/10.3389/fneur.2012.00113 (2012).

25. Takano, K., Ora, H., Sekihara, K., Iwaki, S. & Kansaku, K. Coherent activity in bilateral parieto-occipital cortices during P300-BCI operation. *Front Neurol* 5, 74, https://doi.org/10.3389/fneur.2014.00074 (2014).

26. Piccini, G. et al. Developmental lag of visuospatial attention in Duchenne muscular dystrophy. *Research in Developmental Disabilities* 36, 53–61 (2015).

27. Phukan, J., Pender, N. P. & Hardiman, O. Cognitive impairment in amyotrophic lateral sclerosis. *The Lancet Neurology* 6, 994–1003 (2007).

28. Abe, K. et al. Cognitive function in amyotrophic lateral sclerosis. *Journal of the Neurological Sciences* 148, 95–100 (1997).

29. Brázdil, M., Rektor, I., Daniel, P., Dufek, M. & Jurák, P. Intracerebral event-related potentials to subthreshold target stimuli. *Clinical Neurophysiology* 112, 650–661, https://doi.org/10.1016/j.clinph.2012.03.013 (2012).

30. Brázdil, M., Roman, R., Daniel, P. & Rektor, I. Intracerebral somatosensory event-related potentials: effect of response type (button pressing versus mental counting) on P3-like potentials within the human brain. *Clinical Neurophysiology* 114, 1489–1496, https://doi.org/10.1016/j.clinph.2012.03.013 (2012).
32. Knight, R. T. Contribution of human hippocampal region to novelty detection. *Nature* **383**, 256 (1996).
33. Knight, R. T. Decreased response to novel stimuli after prefrontal lesions in man. *Electroencephalography and Clinical Neurophysiology/Evoked Potentials Section* **59**, 9–20 (1984).
34. Knight, R. T., Grabowecky, M. F. & Scabini, D. Role of human prefrontal cortex in attention control. *Advances in Neurology* **66**, 21–36 (1995).
35. Sogos, V., Curto, M., Reali, C. & Guerini, F. Developmentally regulated expression and localization of dystrophin and utrophin in the human fetal brain. *Mechanisms of Ageing and Development* **123**, 455–462 (2002).
36. De Moura, M. C. D. S., Valle, L. E. R., Resende, M. B. D., Reed, U. C. & Pinto, K. O. Visuospatial attention disturbance in Duchenne muscular dystrophy. *Developmental Medicine & Child Neurology* **52**, e10–e15 (2010).
37. Fazel-Rezai, R. Human error in P300 speller paradigm for brain-computer interface. In *2007 29th Annual International Conference of the IEEE Engineering in Medicine and Biology Society*. 2516–2519 (2007).
38. Treder, M. S. & Blankertz, B. (Co)vert attention and visual speller design in an ERP-based brain-computer interface. *Behavioral and Brain Functions* **6**, 28, https://doi.org/10.1186/1744-9081-6-28 (2010).
39. Takano, K., Komatsu, T., Hata, N., Nakajima, Y. & Kansaku, K. Visual stimuli for the P300 brain-computer interface: a comparison of white/gray and green/blue flicker matrices. *Clinical Neurophysiology* **120**, 1562–1566 (2009).
40. Toyama, S., Takano, K. & Kansaku, K. A non-adhesive solid-gel electrode for a non-invasive brain–machine interface. *Frontiers in Neurology* **X**, 114, https://doi.org/10.3389/fneur.2012.00114 (2012).

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**Author Contributions**

K.U., K.T. and K.K. designed the study; K.U., K.T., Y.O., T.K., O.O. and K.K. performed the study; K.U. and K.T. analysed the data, and K.U., K.T. and K.K. wrote the manuscript.

**Additional Information**

**Competing Interests:** The authors declare that they have no competing interests.

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