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Abstract. It has been reported that a functional near-infrared spectroscopy (fNIRS) signal can be contaminated by extracerebral contributions. Many algorithms using multidistance separations to address this issue have been proposed, but their spatial separation performance has rarely been validated with simultaneous measurements of fNIRS and functional magnetic resonance imaging (fMRI). We previously proposed a method for discriminating between deep and shallow contributions in fNIRS signals, referred to as the multidistance independent component analysis (MD-ICA) method. In this study, to validate the MD-ICA method from the spatial aspect, multidistance fNIRS, fMRI, and laser-Doppler-flowmetry signals were simultaneously obtained for 12 healthy adult males during three tasks. The fNIRS signal was separated into deep and shallow signals by using the MD-ICA method, and the correlation between the waveforms of the separated fNIRS signals and the gray matter blood oxygenation level–dependent signals was analyzed. A three-way analysis of variance (signal depth $\times$ Hb kind $\times$ task) indicated that the main effect of fNIRS signal depth on the correlation is significant [$F(1,1286) = 5.34, p < 0.05$]. This result indicates that the MD-ICA method successfully separates fNIRS signals into spatially deep and shallow signals, and the accuracy and reliability of the fNIRS signal will be improved with the method. © The Authors. Published by SPIE under a Creative Commons Attribution 3.0 Unported License. Distribution or reproduction of this work in whole or in part requires full attribution of the original publication, including its DOI. [DOI: [10.1117/1.NPh.2.1.015003](https://doi.org/10.1117/1.NPh.2.1.015003)]

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1 Introduction

Functional near-infrared spectroscopy (fNIRS) measures the changes in cerebral hemodynamics and oxygenation by radiating weak visible or near-infrared light into the head and detecting the light reflected back (scattered) from another position.^{1–5} fNIRS has been applied to obtain two-dimensional topographical images of the changes in brain hemodynamics and oxygenation.^{6,7}

All over the world, fNIRS systems have been used in more and more situations,⁸ such as in neuroimaging research^{9,10} and medical purposes,^{11–14} especially for measuring the brain activity of infants and children^{15–20} and for creating wearable equipment,^{21,22} because they have a high level of safety^{23,24} and require few constraints.

One of the limitations of fNIRS is the potential effect of the extracerebral tissue on the signal. It was reported that an fNIRS signal can be contaminated by extracerebral signals.^{25–29} It has also been reported that the regional cerebral oxygen saturation is affected by extracranial contamination.^{30,31}

Another issue concerning extracerebral effects is the interference of systemic hemodynamics on fNIRS signals.^{32,33} This is

often referred to as broadly distributed signals caused by heart rate, blood pressure, and respiration. In other words, it is attributed to the effect of measuring systemically circulating blood. Systemic interference is included both in extracerebral and cerebral signals, so a signal originating from cerebral tissue may include a systemic contribution. Extracerebral veins have also been shown to affect fNIRS signals as a task-related systemic contribution.³⁴

To deal with the above-described interference issues, various methods have been proposed.¹⁰ The validity of these methods, however, was confirmed by making certain assumptions, namely (expected) waveforms,^{35,36} contrast-to-noise ratio,³⁷ and correlation with laser-Doppler signals.^{38,39} Few studies, however, have verified such methods by spatial analysis using simultaneous measurement by fNIRS and functional magnetic resonance imaging (fMRI). On the other hand, for providing higher spatial resolution, diffuse optical tomography using high-density probe arrangements has been proposed,^{40,41} and this technique was found to be consistent with nonsimultaneous fMRI. Although general consistency between fNIRS and fMRI has been reported,^{42–45} neither technique used multidistance optodes, and the purpose of these studies did not include the validation of methods for removing scalp effects. Through a concurrent multimodality study with fMRI and laser-Doppler flowmetry (LDF), a deep/shallow discrimination method can

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be validated from the spatial and temporal aspects. That kind of confirmation is highly valuable for the practical use of fNIRS with multidistance probes.

Accordingly, in the present study, we tried to validate a method with multiple-distance probes and independent component analysis (ICA)³⁸ to discriminate between the scalp and cerebral effects on the fNIRS signal using concurrently measured fMRI and LDF signals.

2 Materials and Methods

2.1 Participants

A total of 12 healthy adult males (mean age: 37.7 years; age range: 30 to 48 years) participated in measurements by simultaneous fNIRS with multidistance probes [with source-detector (S-D) distances of 15 (or 16) and 30 mm], fMRI, and LDF. All participants gave written informed consent to the study protocol, which was approved by the Ethical Committee of the Faculty of Medicine, the University of Tokyo [No. 3156-(2)]. None of the participants had a medical history of psychiatric or neurological illness or serious head injury, and none of them had a history of psychotropic drug use.

2.2 Data Acquisition

2.2.1 Functional near-infrared spectroscopy

An optical topography system (ETG-4000, Hitachi Medical Corporation, Japan) was used for the fNIRS measurements. The light sources consisted of continuous laser diodes with two wavelengths, 695 and 830 nm. The transmitted light (detected with avalanche photodiodes) was sampled every 100 ms.

A multidistance measurement (namely 15-, 16-, and 30-mm S-D distances) was conducted with 16 light sources and 16 detectors. Two probe holders were placed for covering the left prefrontal cortex and the left somatosensory or motor cortex. Ten-millimeter-thick low-elastic rubber sheets were used for holding optical-fiber probes. On the optical fiber probes for a 15-mm S-D distance, optical filters were used for attenuating optical intensity. In total, 22 channels and 15 channels were measured for S-D 30 and 15 (or 16) mm, respectively. The channel arrangement and appearance of the probe holders used for the fNIRS measurements are shown in Fig. 1. To mark the optical-probe positions, vitamin-E tablets were placed on the probe holders [Fig. 1(a)]. The left one of the probe holders described in Fig. 1(a) is placed on the left prefrontal position, and the right one is placed on the left parietal position. The positions of the (21) vitamin-E tablets used as markers are shown by yellow ellipses. White squares indicate the positions of S-D 30-mm measurement channels [Fig. 1(b)]. Red squares indicate the positions of S-D 15- and 16-mm measurement channels [Fig. 1(c)]. A photograph of the probe holder worn by a representative participant is shown in Fig. 1(d), left, and a corresponding T1-weighted image with vitamin markers is shown in Fig. 1(d), right.

2.2.2 Magnetic resonance imaging

MRI was performed with a Philips Achieva 3.0T TX system (Philips Medical Systems, The Netherlands) with a 32-channel SENSE head coil. A total of 130, 175, and 180 T2*-weighted gradient-echo echo-planar images (EPIs) were acquired while a

participant underwent a single session of the verbal-fluency task (VFT), working-memory task (WM), and finger-tapping task (TAP) (described below), respectively. The parameters used for acquiring EPIs are listed in Table 1. A single EPI volume consisted of 30 4-mm-thick (for VFT) or 35 3-mm-thick (for WM and TAP) axial slices interspaced by a 1-mm gap, covering the entire brain. Other imaging parameters included repetition time (TR) of 4000 ms (for VFT) or 2500 ms (for WM and TAP), echo time (TE) of 30 ms, flip angle (FA) of 80 deg, field-of-view (FOV) of 192×192 mm², and matrix size of 64×64 . The total measurement time (number of EPI scans $\times$ TR) was set to more than the total task duration (trial duration $\times$ repetition). For all the tasks employed, the initial four scans were discarded to allow for the T1-equilibration effects. Thus, the numbers of scans listed in Table 1 are those excluding the number of those “dummy” scans. Within the numbers of scans, the final eight scans in WM and the first one scan in TAP were not used to match the data size of blood oxygenation level-dependent (BOLD) signal to that of fNIRS signal. Following the functional imaging, a B0 field map was acquired by keeping the same head position (35 4-mm-thick axial slices, TR of 20 ms, and TEs of 2.3/4.6 ms). The B0 field map was later used to reduce the image distortion caused by inhomogeneity in the magnetic field. Further, for anatomically identifying activated regions in the brain, a T1-weighted structural image was obtained (FOV: 250×250 mm; in-place resolution: 1.1×1.1 mm; 301 contiguous sagittal slices with thickness of 0.6 mm; TR: 7.4 ms; TE: 3.4 ms; and FA: 8 deg).

2.2.3 Laser-Doppler flowmetry

Skin blood flow was measured with an LDF (MICROFLO DSP, Oxford Optronix Ltd., UK) equipped with two surface probes. One was attached to the skin, centered between the eyebrows (channel 1), and the other was attached to the left temple (channel 2). The LDF analog output was converted into a digital signal by an analog-to-digital converter (NR-2000, Keyence Corporation, Japan).

2.3 Tasks

The tasks performed in this study were a VFT,^{11,12} a verbal WM,^{46–48} and a TAP.⁴⁹ Target areas for each task were set as Brodmann areas (BAs) 9, 10, 44, 45, and 46 for VFT, BA 46 for WM, and BAs 1, 2, 3, 4, and 40 for TAP.

In the VFT, each trial consisted of a 40-s pretask control period, a 60-s task period, and a 70-s post-task control period. During each task period (60 s), the participants were requested to verbalize as many words as possible that began with a Japanese character enunciated through headphones every 20 s (three characters per trial). The characters, which were enunciated randomly, included /a/, /to/, /na/, /i/, /ki/, /se/, /o/, /ta/, and /ha/. During each control period, the participants were requested to repeatedly verbalize the five Japanese vowels (/a/, /i/, /u/, /e/, and /o/).¹¹ The sequence was repeated for three trials. Speech during fMRI scanning might cause movement artifacts in BOLD signals; therefore, in this study, we adopted a method to acquire all slices from the volume in the first period of the relatively longer TR and to make the remaining period a “no-sound” period.^{50,51} The acquisition time (TA) (for 30 slices) was set to 1205 ms and participants produced all speech (words and vowels) during the no-sound period of TRs, i.e., TR – TA = 2795 ms. We confirmed that this duration was

Fig. 1 Channel arrangement and appearance of probe holders for functional near-infrared spectroscopy (fNIRS) measurements: (a) Positions of sources, detectors, and vitamin-E markers. Left part is placed on left prefrontal position and right part is placed on left parietal position. Positions of 21 vitamin-E tablets used as markers are shown by yellow ellipses. (b) Positions of measurement channels with 30-mm source-detector (S-D) distance. (c) Positions of measurement channels with 15- and 16-mm S-D distances. The dotted ellipses indicate the channel groups for each S-D 30-mm channel. (d) Photograph of a probe holder worn by a representative participant (left) and the corresponding T1-weighted image with vitamin markers (right).

sufficient for all the participants to complete their articulation. This is the point that is different from the conventional VFT sequence. The temporal differences among slices exist within TA (1205 ms) and were not corrected in the present study. This is because the temporal change in the BOLD signal is several times longer than the time scale of the present TA, for which the benefit of correction can be minimal.

The WM and the TAP are described in our previous paper.⁵² Briefly, in the WM (which had an identical delayed-response paradigm), each trial started with a 1.5-s presentation of the target stimuli ("target" hereafter) on a PC display screen, which was followed by a delay of 7 s. A probe stimulus ("probe" hereafter) was then presented for 2.0 s or until the participant responded. The participant responded by pressing a button on a handheld pad connected to the PC. The button-pressing time was recorded. In the WM, one or four Japanese hiragana characters were presented as the target and a Japanese katakana

character was presented as the probe. The participants were instructed to judge whether the character presented as the probe corresponded to any of the target characters and then press the appropriate button. The intervals between the probe onset and the following target onset in the next trial were 24 s. Only a central fixation cross was presented during the interval and delay periods. In addition, a visual cue (changing the color of the fixation cross) was presented for 0.5 s prior to trial onset. Auditory cues (1000- and 800-Hz pure tones of 100-ms duration) were presented at the onset of the visual cue and probe, respectively. One-item and four-item conditions were presented in a pseudorandom order. The sequence was repeated for 16 trials (eight trials for the one-item condition and eight trials for the four-item condition).

In the TAP, the tip of the thumb was touched with the tip of each finger in serial order (forefinger, second finger, third finger, little finger, third finger, second finger, forefinger). On the

Table 1 Parameters used for acquisition of echo-planar image (EPI).

Task	VFT	WM	TAP
FOV	192 mm × 192 mm		
TR	4000 ms	2500 ms	
TE	30 ms		
FA	80 deg		
Number of pixels	64 × 64		
Number of slices	30	35	
Slice thickness	4 mm (gap: 1 mm)	3 mm (gap: 1 mm)	
Voxel size	3 × 3 × 5 mm	3 × 3 × 4 mm	
Number of scans	130	175	180

Note: VFT, verbal-fluency task; WM, working-memory task; TAP, finger-tapping task; FOV, field-of-view; TR, repetition time; TE, echo time; FA, flip angle.

computer screen, the color of the right or left arm of the fixation cross “+” changed alternately between black and yellow at 3.3 Hz (duration of each color: 150 ms). The participants were requested to tap the finger of the left/right hand when the direction of the yellow arm was left/right, synchronized with the presentation timing of the yellow arm. The task duration was 15 s, and there was a 25-s rest period between tasks. The right- and left-finger tapping tasks were repeated five times (10 trials in total).

After the initial dummy scans, analog pulse signals indicating fMRI scanning timings were sent from the MRI system to a PC, in which a software package (E-Prime, Psychology Software Tools, Inc., USA) was used to present visual and auditory stimuli to synchronize the stimuli presentation to the fMRI scanning and to send serial commands to the fNIRS system for recording the time of the stimuli presentation.

2.4 Data Analysis

MATLAB (The MathWorks, Inc., USA) was primarily used for the analysis. A flowchart of the data analysis is shown in Fig. 2.

Fig. 2 Flowchart of analysis.

2.4.1 Preprocessing of functional near-infrared spectroscopy and laser-Doppler-flowmetry signals

The oxy- and deoxy-Hb changes were calculated by using the optical density change of 695- and 830-nm light in accordance with the modified Beer-Lambert law.^{7,53}

As the preprocessing for the fNIRS data analysis, a low-pass filter (VFT: 1/8 Hz, WM: 1/5 Hz, TAP: 1/5 Hz) was applied for suppressing the pulse signals and a high-pass filter (VFT: 1/320 Hz, WM: 1/65 Hz, TAP: 1/80 Hz, inverse number of two times of each trial period) was applied for suppressing the low-frequency fluctuation. Low-pass and high-pass filters with the same cutoff frequencies described above were applied to the LDF signals.

2.4.2 Discrimination between deep- and shallow-layer functional near-infrared spectroscopy signals

A method for discriminating between deep and shallow signals included in original oxy- and deoxy-Hb fNIRS signals obtained with multidistance optodes by using the dependence of independent component amplitude (weight) on S-D distance, referred to as multidistance ICA (MD-ICA), was used.^{38,54} Briefly, a time-delayed decorrelation (TDD)-ICA^{55,56} was applied for obtaining independent components for each “channel group” that includes one S-D 30-mm channel and the nearest one to four S-D 15- or 16-mm channels. Channel groups for execution of TDD-ICA are listed in Table 2. They are also described in Fig. 1(c). For each independent component, the deep/shallow contribution ratios were calculated from the dependence of the signal amplitude (i.e., weight of component) on S-D distance. The deep and shallow subcomponents were then calculated by multiplying the independent components by deep/shallow contribution ratios. At this time, the original independent component is the sum of the deep and shallow subcomponents. Deep and shallow signals are then reconstructed using the sum of the subcomponents of all independent components. Delay times as a TDD-ICA parameter were set to 80, 16, and 21 s for VFT, WM, and TAP, respectively, which are about half the time of the block period [i.e., task plus control (rest) period].

As for the MD-ICA method, it is assumed that the partial optical path length of the deep layer linearly increases as the S-D distance increases, while that of the shallow layer does not change. This assumption is supported by several research works.^{57–60} Moreover, it was assumed that the fNIRS signals at each S-D distance can be expressed by the linear sum of hemoglobin change signals, which are proportional to the partial optical path length at the scalp and gray matter (GM).⁶¹

To apply the MD-ICA method to fNIRS data, at least two kinds of S-D distance ($> X_{i_{gr}}$) are necessary. $X_{i_{gr}}$ indicates the shortest S-D distance at which the detected light has sensitivity to absorption change in GM and is assumed to be 10.5 mm in adults.³⁸ Moreover, the channels in the same channel group should be close enough to each other. In this study, the center-to-center distance (center means midpoint between source and detector) between the long-distance (S-D 30 mm) and the short-distance (S-D 15 or 16 mm) channels was then set to be within 19 mm. The threshold of the center-to-center distance (19 mm) was set according to the previous study,³⁸ where we confirmed that the MD-ICA method successfully worked even when the center-to-center distance was 16.8 mm. The maximal center-to-center distance in the present case is 18.4 mm (e.g., between channels 15 and 24). The difference

Table 2 Channel groups for execution of time-delayed decorrelation independent component analysis (TDD-ICA).

Channel group no.	S-D 30-mm channel no.	S-D 15 or 16-mm channel no.
1	1	23, 26
2	2	24, 27
3	3	25, 28
4	4	26
5	5	26, 27
6	6	27, 28
7	7	28
8	8	26, 29
9	9	27, 30
10	10	28, 31
11	11	32, 36
12	12	33, 37
13	13	34, 36
14	14	35, 37
15	15	23, 24
16	16	24, 25
17	17	23, 24, 26, 27
18	18	24, 25, 27, 28
19	19	26, 27, 29, 30
20	20	27, 28, 30, 31
21	21	29, 30
22	22	30, 31

between 16.8 and 18.4 mm is only 1.6 mm, and we then assumed that 18.4 mm was also valid for execution of MD-ICA. The deep- and shallow-layer contributions' ratio for each channel was calculated by using the amplitude-weighted mean of contribution ratios.

2.4.3 Functional near-infrared spectroscopy activation channel and effect size

The activation channel of the fNIRS signal was chosen from BAs 9, 10, 44, 45, and 46 for VFT, BA 46 for WM, and BAs 1, 2, 3, 4, and 40 for TAP. The BA number was determined⁶² for each projection point from the Montreal Neurological Institute (MNI) coordinates. The activation channel for each participant was determined by the effect size (Cohen's d) of the original fNIRS signal. The effect size is the amplitude difference between the mean of the task period [mean (task)] and that of the control period [mean (control)]

divided by the pooled standard deviation (σ). The equation for the effect size is expressed as

$$d = \frac{\text{mean}(\text{task}) - \text{mean}(\text{control})}{\sigma}, \quad (1)$$

$$\sigma = \sqrt{\frac{\sum_{i=1}^{n_1} (y_{1i} - \bar{y}_1)^2 + \sum_{i=1}^{n_2} (y_{2i} - \bar{y}_2)^2}{n_1 + n_2 - 2}}, \quad (2)$$

where y_1 and y_2 denote the vectors of raw Hb signals in the task and control periods, respectively; n_1 and n_2 denote the numbers of time points for y_1 and y_2 ; and $\bar{y}_1$ and $\bar{y}_2$ denote the temporal means of y_1 and y_2 , respectively.⁵² For the calculation of y_1 and y_2 , no temporal offset for the transient phase was set. For calculation of y_1 , the task periods of the four-item condition for WM and right-hand tapping condition for TAP of all repetitions were used. For calculation of y_2 , the periods of 30, 5, and 5 s before task onset were used for VFT, WM, and TAP, respectively.

In this study, the channels at which the effect size of oxygenated hemoglobin (oxy-Hb) is over 0.2 and that of deoxygenated hemoglobin (deoxy-Hb) is under -0.2 were first selected. After that, the channel at which the difference between the effect sizes of oxy- and deoxy-Hb in the target areas for each task is maximal was selected as an activation channel. The channels at which the absolute amplitude of the deep signal is over 0.6 mM-mm in the entire time span have been removed as noise channels.

2.4.4 Spatially weighted blood oxygenation level-dependent signal

The photon-diffusion region (sensitivity map) expressed in voxel coordinates for each channel of the fNIRS system was calculated for each participant. A gray matter (GM)-BOLD signal was calculated from a spatially weighted sum of BOLD signals at voxels in the photon-diffusion region that is included in the segmented GM region. The processing for obtaining a GM-BOLD signal is described in detail in our previous study.^{52,63} The photon-diffusion region and the GM-BOLD signal were calculated only for S-D 30-mm pairs. Short-distance channels were used for calculating deep and shallow signals of S-D 30-mm channels by the MD-ICA method.

2.4.5 Methods for evaluating discrimination performance

The MD-ICA method separates the fNIRS signal on the basis of signal depth (deep or shallow). As references for a shallow optical signal, skin blood flow (LDF signals) was measured. The following two methods were used to evaluate the performance of the MD-ICA method.

Correlation between fNIRS and LDF signals. The correlation coefficients of original, deep, and shallow signals versus the LDF signal were calculated by the way used in some literature,^{38,39,64} whereas Takahashi et al.²⁹ calculated the temporally integrated LDF signal (blood volume) to compare it with fNIRS signal because, in principle, the integrated LDF signal may relate more to the fNIRS signal than the direct LDF signal (blood flow) does. It was expected that the LDF signal had a higher correlation coefficient with the shallow signal than that with the deep one. For calculating means and standard

deviations of correlation coefficients between fNIRS and LDF signals, all S-D 30-mm channels (22 channels in total) of fNIRS and both LDF channels 1 and 2 were used. It has been reported that the fNIRS signal obtained with a short-distance probe (i.e., a surface fNIRS signal) is highly correlated with the LDF signal.^{38,64} The sign of the deoxy-Hb signal was inverted. While the total-Hb signal (oxy-Hb + deoxy-Hb) is more related to the blood flow signal than oxy- and deoxy-Hb signals in general, oxy- and deoxy-Hb signals were used for the correlation analysis with LDF signals because the present study focused on deep and shallow separation and the contribution ratio depends on Hb types (oxy/deoxy).³⁸ A two-sample *t*-test was used to compare the correlation coefficients among signal depths (original/deep/shallow) for each task and the Hb type with a Bonferroni correction for three comparisons.

Correlation between fNIRS and GM-BOLD signals. The correlation coefficients for the waveforms of the separated fNIRS signals and the GM-BOLD signals were analyzed, and the correlation coefficients for deep fNIRS and GM-BOLD signals were expected to be larger than those for shallow fNIRS and GM-BOLD signals. To investigate this expectation, a three-way (task $\times$ depth $\times$ kind) analysis of variance (ANOVA) was applied to the correlation coefficients for the fNIRS and GM-BOLD signals. The sign of the deoxy-Hb signal was inverted. The fNIRS data were down-sampled to match the fMRI data used for the correlation analysis. Some studies using concurrent fNIRS and fMRI measurements focused on the BOLD-significant (i.e., activation) area.^{43,65} In this analysis, on the other hand, the significance of the task-related change in GM-BOLD signal was not calculated for each channel and all the channels in the target areas for each task were used, because the significance of the change in GM-BOLD signal is not directly related to deep/shallow separation performance.

3 Results

3.1 Functional Near-Infrared Spectroscopy Channel Positions

The representative positions of each fNIRS channel were determined from the closest point on the brain surface to the midpoint of the source and detector positions. The BA number⁶² for each channel was determined for each participant in accordance with the determination of fNIRS channel positions in the MNI space (Table 3).

3.2 Grand-Average at Activation Channel

Grand-average continuous signals of fNIRS (original, deep, and shallow), GM-BOLD, and LDF signal changes (channel 2) at the activation channel obtained during VFT are shown in Fig. 3. Standard errors at each time point are displayed as translucent patches. Vertical solid and dashed lines indicate task onset and end timings, respectively. A task-related response during VFT was obtained for each signal.

3.3 Correlation with Laser-Doppler-Flowmetry Signals

Correlation coefficients (Fisher's *Z*-transformation converted from Pearson's correlation coefficient *r*) for oxy- and deoxy-

Table 3 Estimated location of each near-infrared spectroscopy (NIRS) channel on normalized brain image. Mean and standard deviation (SD) of Montreal Neurological Institute (MNI) coordinates across participants and corresponding Brodmann area (BA) numbers are shown for each channel. Percentage of participants by BA number is shown in parentheses.

Channel	Mean MNI coordinates			SD	BA (%)		
	x	y	z				
1	-5.8	57.8	38.5	4.2	9 (83%)	10 (17%)	—
2	-24.7	50.7	37.7	4.7	9 (58%)	46 (42%)	—
3	-42.5	36.8	34.7	4.2	45 (42%)	46 (33%)	9 (25%)
4	5.5	65.2	25.7	4.8	10 (92%)	9 (8%)	—
5	-15.0	62.2	27.2	4.9	10 (67%)	9 (17%)	46 (17%)
6	-36.2	50.7	26.2	4.5	46 (83%)	45 (17%)	—
7	-50.7	34.3	21.8	4.4	45 (83%)	44 (8%)	46 (8%)
8	-7.2	68.7	14.5	4.8	10 (100%)	—	—
9	-27.3	62.5	15.0	4.0	10 (75%)	46 (25%)	—
10	-44.5	48.8	12.2	3.9	46 (50%)	45 (42%)	10 (8%)
11	-52.0	-23.3	57.0	5.5	3 (33%)	4 (33%)	1, 40 (17%)
12	-48.0	-37.8	59.2	3.9	40 (58%)	3 (33%)	1 (8%)
13	-61.5	-29.0	44.7	4.7	40 (42%)	2, 3 (25%)	1 (8%)
14	-56.5	-46.5	48.0	4.0	40 (83%)	39 (17%)	—
15	-13.0	51.5	43.8	2.8	9 (100%)	—	—
16	-31.7	42.3	40.8	4.8	9 (75%)	46 (25%)	—
17	-16.2	58.7	32.8	3.3	9 (33%)	10 (33%)	46 (33%)
18	-36.0	47.0	31.2	4.3	46 (75%)	45 (25%)	—
19	-18.8	63.8	21.2	3.6	10 (92%)	46 (8%)	—
20	-38.0	53.2	19.2	4.3	46 (83%)	10 (8%)	45 (8%)
21	-16.5	67.8	11.0	5.4	10 (100%)	—	—
22	-37.2	58.8	7.0	3.9	10 (50%)	46 (50%)	—

Hb signals (original, deep, and shallow) and LDF signals (including channels 1 and 2) during performance of VFT, WM, and TAP are shown in Fig. 4. Error bars indicate the standard deviations. Single (*) and double (**) asterisks denote the statistical significance at $p < 0.05$ and 0.01 (corrected for multiple comparisons), respectively.

3.4 Analysis of Variance of Correlation between Functional Near-Infrared Spectroscopy and Gray Matter Blood Oxygenation Level-Dependent Signals

Correlation coefficient Z (Fisher's Z) for the spatially weighted GM-BOLD signals and the fNIRS signals (deep and shallow signals of oxy- and deoxy-Hb) for the three tasks is shown in Fig. 5. All the channels in the target areas for each task were used. Figures 5(a) and 5(b) show the results for

oxy- and deoxy-Hb, respectively. Error bars indicate the standard deviations. A three-way ANOVA [signal depth (deep/shallow) $\times$ Hb kind (oxy/deoxy) $\times$ task (VFT/WM/TAP)] indicates that the main effect of the separated fNIRS signal depth on the signal correlation is significant [$F(1,1286) = 5.34$, $p < 0.05$] and that the interactions between the three effects are not significant. These results show that the mean of the correlation coefficients of the deep signal (mean: $Z = 0.149$) was significantly higher than that of the shallow signal (mean: $Z = 0.120$).

3.5 Deep-Layer Pooled Contribution Ratio Obtained by Multidistance Independent Component Analysis

Means and standard deviations of deep-layer pooled contribution ratio (%) for activation channels are listed in Table 4. Although the deep-layer pooled contribution ratios are a little

Fig. 3 Grand average of continuous data of fNIRS (original, deep, and shallow) and gray matter blood-oxygenation-level dependent (GM-BOLD) signals for activation channel, and laser-Doppler-flowmetry (LDF) signal changes (channel 2) obtained during verbal fluency task (VFT). Translucent patches indicate the standard error at each time point. Vertical solid and dashed lines indicate task onset and end timings, respectively. (a) Original signal; (b) deep signal; (c) shallow signal [oxy-Hb (solid line), deoxy-Hb (dashed line)]; (d) GM-BOLD signal change (%); and (e) LDF signal change (arb. unit).

Fig. 4 Correlation coefficients (Fisher's Z) between oxy- and deoxy-Hb signals (original, deep, and shallow) and LDF signal (including both channels 1 and 2) during a verbal fluency task (VFT), a working memory task (WM), and a finger tapping task (TAP). Error bars indicate the standard deviations. Single (*) and double (**) asterisks denote the statistical significance at $p < 0.05$ and 0.01 (corrected for multiple comparisons), respectively.

Fig. 5 Correlation coefficients (Fisher's Z) for (a) fNIRS oxy-Hb (deep and shallow) and spatially weighted GM-BOLD signals in target areas, and for (b) fNIRS deoxy-Hb (deep and shallow) and spatially weighted GM-BOLD signals in target areas during a VFT, a WM, and a TAP. Error bars indicate the standard deviations. The sign of deoxy-Hb signal is inverted. For any individual task, no significant differences were found between the correlations of deep and shallow fNIRS and GM-BOLD signals.

Table 4 Means and standard deviations of deep-layer pooled contribution ratio (%) at activation channels.

	oxy-Hb	deoxy-Hb
VFT	50.0 ± 17.1	55.1 ± 17.0
WM	56.2 ± 10.7	64.3 ± 9.7
TAP	60.9 ± 11.6	74.8 ± 5.3

lower than those reported in a previous study,³⁸ over half of the contribution of Hb signal to the fNIRS signals is originated from the deep layer, especially for oxy-Hb.

4 Discussion

4.1 Correlations between Functional Near-Infrared Spectroscopy and Laser-Doppler-Flowmetry Signals

The correlations between the fNIRS and LDF signals (shown in Fig. 4) are very similar to those obtained in a previous study³⁸ that showed the correlation coefficients for deep fNIRS and LDF signals are significantly lower than those for shallow fNIRS and LDF signals. The correlation coefficient between deoxy-Hb and LDF signals under the TAP condition was extremely low. That was possibly because the deep-layer pooled contribution ratio of deoxy-Hb under the TAP condition was relatively high (Table 4) in the present study. The low correlation between deoxy-Hb and skin blood flow can be caused by the low contribution of the shallow signal. It should be noted that we did not temporally integrate the LDF signal, but the integrated LDF signal can be more correlated with fNIRS signal when a proper integration time is selected.²⁹

From the aspect of correlation between fNIRS and LDF signals, it was shown that the fNIRS signals were reasonably divided into signals with either higher or lower correlations with the LDF signal. It should be noted that during the finger-tapping task, the LDF signal had a higher correlation with the shallow fNIRS signal than that with the deep fNIRS signal, even if the target channels were located mainly in somatosensory or motor areas (BAs 1, 2, 3, 4, and 40) far from the LDF probes (attached on the forehead or temple). The result suggests that the LDF signal correlates with the shallow fNIRS signal in the broad area during the finger-tapping task.

4.2 Correlation between Deep Functional Near-Infrared Spectroscopy Signals and Gray Matter Blood Oxygenation Level-Dependent Signals

The mean of the correlation coefficients of the deep signal was significantly higher than that of the shallow signal. This is partly because deep (brain) and shallow (scalp) tissue layers are anatomically governed by different blood vessel systems (internal or external carotid artery). Different correlation coefficients for the fNIRS and the GM-BOLD signals would, therefore, be expected, i.e., the deep fNIRS signal should have stronger correlation with the GM-BOLD signal than that between the shallow fNIRS and the GM-BOLD signals.

From this point of view, the results of the correlation between deep fNIRS and GM-BOLD signals (Fig. 5) showed that the MD-ICA method successfully separates fNIRS signals into

deep and shallow signals that have higher and lower correlations, respectively, with spatially weighted GM-BOLD signals.

Deep and shallow signals can be similar as a result of MD-ICA method. The high correlation between deep and shallow signals was also obtained in previous studies.^{38,54} This can happen because the same independent components are commonly used for reconstructing deep and shallow signals, and the systemic signals did not be removed in order to quantify the contributions of deep and shallow signals. If the contributions of components that included both deep and shallow signals are almost the same, then the correlations of the shallow and deep signals with the GM-BOLD would be almost equivalent. In the present case, however, different correlations were obtained. This means that deep and shallow signals were different enough from each other to evaluate the separation performance. Although mean deep and shallow signals seem very similar, as shown in Fig. 3 for example, individual deep and shallow signals are different and have different correlations with LDF or GM-BOLD signals.

4.3 Deep-Layer Contribution Ratio Obtained by Multidistance Independent Component Analysis

It should be noted that the MD-ICA method quantifies the contribution ratios of both deep and shallow layers, but the ratios include the effect of systemic interference because the MD-ICA method discriminates fNIRS signals on the basis of signal depth only. Even if the deep-layer pooled contribution ratio is high, for example, it is possible that the systemic contribution in the deep layer is dominant.

It has been reported that there is interindividual variability in the correlation between the fNIRS signal and the scalp blood flow or mean blood pressure⁶⁶ and that the systemic changes that also affect extracranial signals can lead to false positives in fNIRS signals.³³ It should be noted that the effect of posture (sitting or supine) on the contribution of deep-layer tissue to fNIRS signal and its dependency on kind of task have not been investigated. Such an effect might cause the difference between the contribution ratios obtained in the current study and in a previous study.

4.4 Limitations

In regard to the proposed deep-shallow separation method (MD-ICA method), the structural parameter X'_{gr} was fixed for all participants and for all measurement channels. In this study, it was confirmed that the fixed parameter is effective, even in the case where the structural differences depending on individuals and positions within individuals are not considered and neither MRI structural data nor x-ray CT data are available. It should be noted that the deep- or shallow-tissue condition may be changed by changing the posture. The deep/shallow contribution ratio calculated in this study (i.e., supine posture) is not necessarily the same as that calculated for a sitting posture.

The measurement area was limited to only prefrontal, somatosensory, and motor cortices on the left side of the head. Other areas should be covered by the proposed method, so occipital and temporal areas should be further investigated. All participants in this study were male; it would, therefore, be more helpful to validate the proposed method by using female participants.

5 Conclusion

Though very few studies have validated a multidistance scalp-effect-removal method with concurrent fNIRS-fMRI measurement, this study shows that the previously proposed deep/shallow separation method (MD-ICA method) successfully separates fNIRS signals into “spatially” deep and shallow signals by comparing these signals with spatially weighed GM-BOLD and LDF signals. The result shows that the accuracy and reliability of the fNIRS signal will be greatly improved with the MD-ICA method. The correlation coefficients for shallow fNIRS and LDF signals were larger than those for deep fNIRS and LDF signals. This result is consistent with the results obtained in a previous study.³⁸ This method needs only small numbers of probes [at least two middle-distance (>10.5 mm) channels], so it will easily contribute to broad area (e.g., whole head) brain-imaging studies using cost-effective equipment.

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References

1. B. Chance et al., “Cognition-activated low-frequency modulation of light absorption in human brain,” *Proc. Natl. Acad. Sci. U. S. A.* **90**(8), 3770–3774 (1993).
2. Y. Hoshi and M. Tamura, “Detection of dynamic changes in cerebral oxygenation coupled to neuronal function during mental work in man,” *Neurosci. Lett.* **150**(1), 5–8 (1993).
3. F. F. Jöbsis, “Noninvasive, infrared monitoring of cerebral and myocardial oxygen sufficiency and circulatory parameters,” *Science* **198**(4323), 1264–1267 (1977).
4. T. Kato et al., “Human visual cortical function during photic stimulation monitoring by means of near-infrared spectroscopy,” *J. Cereb. Blood Flow Metab.* **13**(3), 516–520 (1993).
5. A. Villringer et al., “Near infrared spectroscopy (NIRS): a new tool to study hemodynamic changes during activation of brain function in human adults,” *Neurosci. Lett.* **154**(1–2), 101–104 (1993).
6. H. Koizumi et al., “Higher-order brain function analysis by trans-cranial dynamic near-infrared spectroscopy imaging,” *J. Biomed. Opt.* **4**(4), 403–413 (1999).
7. A. Maki et al., “Spatial and temporal analysis of human motor activity using noninvasive NIR topography,” *Med. Phys.* **22**(12), 1997–2005 (1995).
8. D. A. Boas et al., “Twenty years of functional near-infrared spectroscopy: introduction for the special issue,” *NeuroImage* **85**(1), 1–5 (2014).
9. M. Ferrari and V. Quaresima, “A brief review on the history of human functional near-infrared spectroscopy (fNIRS) development and fields of application,” *NeuroImage* **63**(2), 921–935 (2012).
10. F. Scholkmann et al., “A review on continuous wave functional near-infrared spectroscopy and imaging instrumentation and methodology,” *NeuroImage* **85**(1), 6–7 (2014).
11. T. Suto et al., “Multichannel near-infrared spectroscopy in depression and schizophrenia: cognitive brain activation study,” *Biol. Psychiatry* **55**(5), 501–511 (2004).
12. R. Takizawa et al., “Neuroimaging-aided differential diagnosis of the depressive state,” *NeuroImage* **85**(1), 498–507 (2014).
13. E. Watanabe et al., “Non-invasive assessment of language dominance with near-infrared spectroscopic mapping,” *Neurosci. Lett.* **256**(1), 49–52 (1998).
14. S. Pu et al., “Association between cognitive insight and prefrontal function during a cognitive task in schizophrenia: a multichannel near-infrared spectroscopy study,” *Schizophr. Res.* **150**(1), 81–87 (2013).
15. T. Grossmann and M. H. Johnson, “Selective prefrontal cortex responses to joint attention in early infancy,” *Biol. Lett.* **6**(4), 540–543 (2010).
16. F. Homae et al., “Development of global cortical networks in early infancy,” *J. Neurosci.* **30**(14), 4877–4882 (2010).
17. F. Homae, “A brain of two halves: insights into interhemispheric organization provided by near-infrared spectroscopy,” *NeuroImage* **85**(1), 354–362 (2014).
18. Y. Minagawa-Kawai et al., “Prefrontal activation associated with social attachment: facial-emotion recognition in mothers and infants,” *Cereb. Cortex* **19**(2), 284–292 (2009).
19. G. Taga et al., “Spontaneous oscillation of oxy- and deoxy-hemoglobin changes with a phase difference throughout the occipital cortex of newborn infants observed using non-invasive optical topography,” *Neurosci. Lett.* **282**(1–2), 101–104 (2000).
20. G. Taga et al., “Brain imaging in awake infants by near-infrared optical topography,” *Proc. Natl. Acad. Sci. U. S. A.* **100**(19), 10722–10727 (2003).
21. H. Atsumori et al., “Development of wearable optical topography system for mapping the prefrontal cortex activation,” *Rev. Sci. Instrum.* **80**(4), 043704 (2009).
22. T. Funane et al., “Synchronous activity of two people’s prefrontal cortices during a cooperative task measured by simultaneous near-infrared spectroscopy,” *J. Biomed. Opt.* **16**(7), 077011 (2011).
23. Y. Ito et al., “Assessment of heating effects in skin during continuous wave near infrared spectroscopy,” *J. Biomed. Opt.* **5**(4), 383–390 (2000).
24. M. Kiguchi et al., “Comparison of light intensity on the brain surface due to laser exposure during optical topography and solar irradiation,” *J. Biomed. Opt.* **12**(6), 062108 (2007).
25. T. J. Germon et al., “Sensitivity of near infrared spectroscopy to cerebral and extra-cerebral oxygenation changes is determined by emitter-detector separation,” *J. Clin. Monit. Comput.* **14**(5), 353–360 (1998).
26. S. Kohri et al., “Quantitative evaluation of the relative contribution ratio of cerebral tissue to near-infrared signals in the adult human head: a preliminary study,” *Physiol. Meas.* **23**(2), 301–312 (2002).

27. L. Minati et al., "Intra- and extra-cranial effects of transient blood pressure changes on brain near-infrared spectroscopy (NIRS) measurements," *J. Neurosci. Methods* **197**(2), 283–288 (2011).
28. P. Smielewski et al., "Clinical evaluation of near-infrared spectroscopy for testing cerebrovascular reactivity in patients with carotid artery disease," *Stroke* **28**(2), 331–338 (1997).
29. T. Takahashi et al., "Influence of skin blood flow on near-infrared spectroscopy signals measured on the forehead during a verbal fluency task," *NeuroImage* **57**(3), 991–1002 (2011).
30. S. N. Davie and H. P. Grocott, "Impact of extracranial contamination on regional cerebral oxygen saturation: a comparison of three cerebral oximetry technologies," *Anesthesiology* **116**(4), 834–840 (2012).
31. H. Sørensen et al., "Cutaneous vasoconstriction affects near-infrared spectroscopy determined cerebral oxygen saturation during administration of norepinephrine," *Anesthesiology* **117**(2), 263–270 (2012).
32. S. B. Erdoğan, M. A. Yücel, and A. Akin, "Analysis of task-evoked systemic interference in fNIRS measurements: insights from fMRI," *NeuroImage* **87**, 490–504 (2014).
33. I. Tachtsidis et al., "False positives in functional near-infrared topography," *Adv. Exp. Med. Biol.* **645**, 307–314 (2009).
34. E. Kirilina et al., "The physiological origin of task-evoked systemic artefacts in functional near infrared spectroscopy," *NeuroImage* **61**(1), 70–81 (2012).
35. L. Gagnon et al., "Improved recovery of the hemodynamic response in diffuse optical imaging using short optode separations and state-space modeling," *NeuroImage* **56**(3), 1362–1371 (2011).
36. R. B. Saager, N. L. Telleri, and A. J. Berger, "Two-detector corrected near infrared spectroscopy (C-NIRS) detects hemodynamic activation responses more robustly than single-detector NIRS," *NeuroImage* **55**(4), 1679–1685 (2011).
37. Q. Zhang, E. N. Brown, and G. E. Strangman, "Adaptive filtering to reduce global interference in evoked brain activity detection: a human subject case study," *J. Biomed. Opt.* **12**(6), 064009 (2007).
38. T. Funane et al., "Quantitative evaluation of deep and shallow tissue layers' contribution to fNIRS signal using multi-distance optodes and independent component analysis," *NeuroImage* **85**(1), 150–165 (2014).
39. S. Kohno et al., "Removal of the skin blood flow artifact in functional near-infrared spectroscopic imaging data through independent component analysis," *J. Biomed. Opt.* **12**(6), 062111 (2007).
40. A. T. Eggebrecht et al., "Mapping distributed brain function and networks with diffuse optical tomography," *Nat. Photonics* **8**(6), 448–454 (2014).
41. D. K. Joseph et al., "Diffuse optical tomography system to image brain activation with improved spatial resolution and validation with functional magnetic resonance imaging," *Appl. Opt.* **45**(31), 8142–8151 (2006).
42. D. A. Boas, A. M. Dale, and M. A. Franceschini, "Diffuse optical imaging of brain activation: approaches to optimizing image sensitivity, resolution, and accuracy," *NeuroImage* **23**(Suppl 1), S275–S288 (2004).
43. X. Cui et al., "A quantitative comparison of NIRS and fMRI across multiple cognitive tasks," *NeuroImage* **54**(4), 2808–2821 (2011).
44. F. B. Haeussinger et al., "Reconstructing functional near-infrared spectroscopy (fNIRS) signals impaired by extra-cranial confounds: an easy-to-use filter method," *NeuroImage* **95**, 69–79 (2014).
45. S. Heinzel et al., "Variability of (functional) hemodynamics as measured with simultaneous fNIRS and fMRI during intertemporal choice," *NeuroImage* **71**(1), 125–134 (2013).
46. R. Aoki et al., "Relationship of negative mood with prefrontal cortex activity during working memory tasks: an optical topography study," *Neurosci. Res.* **70**(2), 189–196 (2011).
47. R. Aoki et al., "Correlation between prefrontal cortex activity during working memory tasks and natural mood independent of personality effects: an optical topography study," *Psychiatry Res.: Neuroimaging* **212**(1), 79–87 (2013).
48. H. Sato et al., "Correlation of within-individual fluctuation of depressed mood with prefrontal cortex activity during verbal working memory task: optical topography study," *J. Biomed. Opt.* **16**(12), 126007 (2011).
49. H. Sato et al., "Intersubject variability of near-infrared spectroscopy signals during sensorimotor cortex activation," *J. Biomed. Opt.* **10**(4), 044001 (2005).
50. S. Abrahams et al., "Functional magnetic resonance imaging of verbal fluency and confrontation naming using compressed image acquisition to permit overt responses," *Hum. Brain Mapp.* **20**(1), 29–40 (2003).
51. W. B. Edmister et al., "Improved auditory cortex imaging using clustered volume acquisitions," *Hum. Brain Mapp.* **7**(2), 89–97 (1999).
52. H. Sato et al., "A NIRS-fMRI investigation of prefrontal cortex activity during a working memory task," *NeuroImage* **83**, 158–173 (2013).
53. D. T. Delpy et al., "Estimation of optical pathlength through tissue from direct time of flight measurement," *Phys. Med. Biol.* **33**(12), 1433–1442 (1988).
54. T. Funane et al., "Greater contribution of cerebral than extracerebral hemodynamics to near-infrared spectroscopy signals for functional activation and resting-state connectivity in infants," *Neurophotonics* **1**(2), 025003 (2014).
55. J. F. Cardoso, "Blind signal separation: statistical principles," *Proc. IEEE* **86**(10), 2009–2025 (1998).
56. T. Katura et al., "Extracting task-related activation components from optical topography measurement using independent components analysis," *J. Biomed. Opt.* **13**(5), 054008 (2008).
57. Y. Fukui, Y. Ajichi, and E. Okada, "Monte Carlo prediction of near-infrared light propagation in realistic adult and neonatal head models," *Appl. Opt.* **42**(16), 2881–2887 (2003).
58. A. Hirasawa et al., "Influence of skin blood flow and source-detector distance on near-infrared spectroscopy-determined cerebral oxygenation in humans," *Clin. Physiol. Funct. Imaging* (2014).
59. G. E. Strangman, Q. Zhang, and Z. Li, "Scalp and skull influence on near infrared photon propagation in the Colin27 brain template," *NeuroImage* **85**(1), 136–149 (2014).
60. T. Yamada, S. Umeyama, and K. Matsuda, "Separation of fNIRS signals into functional and systemic components based on differences in hemodynamic modalities," *PLoS One* **7**(11), e50271 (2012).
61. T. Funane et al., "Dynamic phantom with two stage-driven absorbers for mimicking hemoglobin changes in superficial and deep tissues," *J. Biomed. Opt.* **17**(4), 047001 (2012).
62. C. Rorden and M. Brett, "Stereotaxic display of brain lesions," *Behav. Neurol.* **12**(4), 191–200 (2000).
63. A. Sassaroli et al., "Spatially weighted BOLD signal for comparison of functional magnetic resonance imaging and near-infrared imaging of the brain," *NeuroImage* **33**(2), 505–514 (2006).
64. F. Aletti et al., "Deep and surface hemodynamic signal from functional time resolved transcranial near infrared spectroscopy compared to skin flowmotion," *Comput. Biol. Med.* **42**(3), 282–289 (2012).
65. A. Seiyama et al., "Circulatory basis of fMRI signals: relationship between changes in the hemodynamic parameters and BOLD signal intensity," *NeuroImage* **21**(4), 1204–1214 (2004).
66. I. Tachtsidis et al., "Investigation of frontal cortex, motor cortex and systemic haemodynamic changes during anagram solving," *Adv. Exp. Med. Biol.* **614**, 21–28 (2008).

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