Greater contribution of cerebral than extracerebral hemodynamics to near-infrared spectroscopy signals for functional activation and resting-state connectivity in infants

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1 Introduction

Cerebral blood volume change is measured by radiating weak visible or near-infrared light into the head and detecting the transmitted light from another position on the basis of near-infrared spectroscopy (NIRS). An NIRS imaging system, such as optical topography, that uses multiple light sources and detectors is used to obtain two-dimensional topographical images of the changes in brain hemodynamics and oxygenation. NIRS systems have been widely used for research and clinical purposes, especially for measuring the brain activity of infants, detecting the functional connectivity on the basis of spontaneous activity of the brain of infants while resting, and using wearable systems to conduct measurements in everyday situations. The general consistency between NIRS signals and blood oxygenation level dependent signals has been shown in studies on task-evoked brain activation and resting-state functional connectivity, validating NIRS as a neuroimaging tool. However, a major issue for NIRS is to what extent the deep tissue (such as cerebral tissue), as opposed to shallow-tissue (extracerebral tissue, such as the scalp), contributes to the measured signals of oxygenated and deoxygenated hemoglobin (oxy- and deoxy-Hb). The separation of deep- from shallow-tissue signals has also been performed by time-series analysis, such as adaptive filtering or independent component analysis (ICA). Most of the aforementioned studies assumed either the spatially homogeneous properties of systemic effects on the shallow tissues or the independence of temporal evolution of the deep and shallow signals. However, the assumptions might be violated if the systemic effects induced by tasks can locally affect the blood flow and oxygenation in both deep- and shallow tissues. Kirilina et al. used NIRS, functional magnetic resonance imaging (fMRI), MR-angiography, blood pressure, heart rate, skin conductance, and skin blood flow to study activation in the frontal cortex during a cognitive task and demonstrated that cognitive tasks strongly affect the blood volume of veins draining the scalp and that NIRS signals contain the task-evoked systemic

Abstract. While near-infrared spectroscopy (NIRS) has been increasingly applied to neuroimaging and functional connectivity studies in infants, it has not been quantitatively examined as to what extent the deep tissue (such as cerebral tissue) as opposed to shallow tissue (such as scalp), contributes to NIRS signals measured in infants. A method for separating the effects of deep- and shallow-tissue layers was applied to data of nine sleeping three-month-old infants who had been exposed to 3-s speech sounds or silence (i.e., resting state) and whose hemodynamic changes over their bilateral temporal cortices had been measured by using an NIRS system with multiple source-detector (S-D) distances. The deep-layer contribution was found to be large during resting (67% at S-D 20 mm, 78% at S-D 30 mm for oxygenated hemoglobin (oxy-Hb)) as well as during the speech condition (72% at S-D 20 mm, 82% at S-D 30 mm for oxy-Hb). A left-right connectivity analysis showed that correlation coefficients between left and right channels did not differ between original- and deep-layer signals under no-stimulus conditions and that of original- and deep-layer signals were larger than those of the shallow layer. These results suggest that NIRS signals obtained in infants with appropriate S-D distances largely reflected cerebral hemodynamic changes. 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.1.2.025003] Keywords: near-infrared spectroscopy; neuroimaging in infant; multidistance optodes; superficial tissue contribution; speech processing; functional connectivity.

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artifacts. Funane et al.\textsuperscript{40} proposed a method to discriminate between the deep- and shallow-tissue effects on the NIRS signal using multiple-distant (MD) probes and an ICA.\textsuperscript{41,42} Importantly, the MD-ICA method can be used to decompose NIRS signals even when the cerebral and extracerebral signals correlate.

While NIRS studies have provided unique opportunities to reveal the developmental processes of the functional activation in the brains of infants\textsuperscript{12-17} and functional connectivity,\textsuperscript{18-20} very few have examined how deep- (e.g., cerebral) and shallow-tissue (scalp) layers of the head contribute to NIRS signals. Since the hemodynamic responses to visual or auditory stimuli in young infants show global properties over diverse cortical regions,\textsuperscript{43,44} it is crucial to elucidate the systemic effects on the hemodynamic responses for both shallow and deep tissues. Moreover, the dissociation of deep and shallow signals is important in the study of the functional connectivity of the cortex in the resting state, because a false-positive connectivity might be estimated if measured NIRS signals have a high contribution from the shallow tissue that exhibits temporally correlated hemodynamic changes over different locations of the scalp. It is also important to note that infants’ heads differ from those of adults not only in size, but also for optical properties regarding NIR light.\textsuperscript{45} Thus, an extrapolation from adult studies may not be valid. Taga et al.\textsuperscript{46} investigated the effect of the source-detector (S-D) distance on signal amplitude using MD probes with infants, but no method has been applied for separating shallow and deep signals. The MD-ICA method\textsuperscript{40} can be applied to the infant data obtained using MD probes if infant-specific parameters are determined, such as optical path length in each head tissue\textsuperscript{45,47} and brain-scalp distance.\textsuperscript{48}

This study is intended to investigate the cerebral and extracerebral contributions to NIRS signals during functional activation in infants. Furthermore, we also investigate those during a resting state, which has not been shown in the previous study.\textsuperscript{40} To do this, we develop a model to determine infant-specific parameters for optical properties and use the MD-ICA method,\textsuperscript{40} which was verified for adult participants, to quantify the effect of deep and shallow tissue on NIRS signals measured with infants’ heads in the existing data.\textsuperscript{46}

2 Method

2.1 Participants

The data used in this study were originally obtained for a previous study\textsuperscript{46} in which nine three-month-old (mean age, 113 days) infants (six girls and three boys) participated. All infants were studied while they were quietly sleeping. Informed consent was obtained from the parents of the infants prior to the initiation of the experiments. All experiments were approved by the Office for Life Science Research Ethics and Safety, The University of Tokyo.

2.2 NIRS Measurement

An NIRS system (ETG-100, Hitachi Medical Corporation, Tokyo, Japan)\textsuperscript{49} was used to measure the hemodynamic responses of infants. The optical fibers for the source and detectors were placed over the temporal cortices on both hemispheres. On the left hemisphere, the source was set on the T3 position on the basis of the international 10–20 system of electrode placement,\textsuperscript{50} and four detectors were mounted in a line at 10, 20, 30, and 40 mm from the source. On the right hemisphere, the source and four detectors were placed on homologous positions. Figure 1 shows the arrangement of the source and detectors in NIRS measurements for hemodynamic responses to auditory stimuli in infants. Thus all the measurement channels in the present study were likely located on the temporal cortex.\textsuperscript{16,50} These regions have also been reported to respond to speech sounds while infants are quietly sleeping.\textsuperscript{51}

Although in the previously obtained data\textsuperscript{46} two different intensities of NIR light were used, i.e., 0.6 mW (0.3 mW for each of the two laser diodes; 780 and 830 nm) and 1.2 mW (0.6 mW for each), in this study only the dataset measured under 1.2 mW was used because the signal-to-noise ratio of the detected light intensity is higher. The data obtained at 40-mm S-D distance were not used for the analysis because signal-to-noise ratios for several participants were not sufficient for applying the MD-ICA method.

2.3 Stimulation

Details of the stimulation have been described in previous papers.\textsuperscript{46,52} The test auditory stimuli consisted of speech sounds of Japanese words provided by a female Japanese speaker. Ten different sets of stimuli (mean duration: 2.68 s; range: 2.54 to 2.81 s) were used.

The experiments were conducted in a sound-attenuated room. The infants sat on the lap of an experimenter and were studied while they were quietly sleeping. The speech sounds were presented at a 73-dB maximum amplitude. We used an event-related paradigm under the speech condition in which the test auditory stimuli were presented every 15 s (i.e., stimulus onset asynchrony = 15 s). During interstimulus intervals, we did not present any sound. The measurements lasted 150 s. Under the no-stimulus (control) condition, the infants were given no sound and were measured for 150 s.

![Fig. 1](https://www.spiedigitallibrary.org/journals/Neurophotonics/images/000025003-2_001.png)
2.4 Preprocess for NIRS Data

For the preprocessing for the NIRS data analysis, the oxy- and deoxy-Hb changes were calculated by using the optical density changes of 780- and 830-nm light in accordance with the modified Beer-Lambert law.\(^{6,55}\)

The analysis flow of this study is shown in Fig. 2. We applied a 0.009 to a 0.2 Hz band-pass filter for suppressing the pulse signals and low-frequency fluctuation. The blocks were not used where the nonfiltered (raw) oxy-Hb signal changed very quickly (over 0.8 mM - mm change in 1 s) at either the S-D 10- or 20-mm channel because such blocks probably contain some artifacts. For the MD-ICA method, continuous data, including blocks, are necessary for applying ICA. If a block was eliminated, the blocks after or before the eliminated block all had to be eliminated, leaving longer continuous data that did not include the eliminated blocks. With this procedure, out of 180 blocks (9 participants × 2 conditions × 10 blocks), eight were not used.

In individual datasets, we initially divided the continuous time course data into blocks. Each data block ranged from 1 s prior to stimulus onset to 15 s after onset. Then, we corrected the baseline using linear fitting to the mean signal of the first 1 s and that of the last 1 s in each data block. By averaging the signal changes over the data blocks for each subject under each condition, we obtained individual hemodynamic responses for each measurement channel. We finally obtained group-averaged hemodynamic responses at each measurement channel under each condition.

2.5 Deep-Shallow Signal Separation Method

There are several methods for removing the scalp effect from an NIRS signal.\(^{40,54,55}\) We previously proposed a method with MD probes to discriminate between the deep- (such as cerebral) and shallow- (scalp) tissue effects on the NIRS signals obtained from adult participants.\(^{40}\) We base the present study on the previous work and apply the previously proposed method to infant data.

In the method with MD-ICA, it is assumed that the partial optical path length of the deep layer linearly increases as the S-D distance from \(X_{lg}\) increases, whereas that of the shallow layer does not change. \(X_{lg}\) indicates the shortest S-D distance at which the detected light is sensitive to absorption change in gray matter. This is assumed to be 10.5 mm in adults (experimentally obtained).\(^{56}\) For infants, a different \(X_{lg}\) value, i.e., 8 mm, was used as mentioned above. Moreover, we assumed that the NIRS signals at each S-D distance can be expressed by the linear sum of the hemoglobin-change signals that are proportional to the partial optical path length at scalp and gray matter.\(^{56}\)

In the first step of this method, an ICA\(^{41,42}\) is used for obtaining signal sources (independent components) to acquire the deep and shallow subcomponents described in the following step. In this step of the analysis, three independent components were obtained for each hemisphere. If enough (e.g., four or more) channels are available, noise components can generally be truncated. However, we did not cut any components in the MD-ICA analysis because not enough channels were available. In the next step, the contribution ratios of the deep and shallow layers are calculated by the dependence of signal amplitudes (i.e., weight of component) on the S-D distance for each independent component. The deep and shallow subcomponents are then calculated by multiplying the independent components by deep/shallow contribution ratios. In the final step, deep and shallow signals are reconstructed using the linear sum of the subcomponents of all the independent components. This analysis supports a model that allows us to calculate an estimation of the relative contributions.

To explain in more detail, the deep and shallow signals are reconstructed by the following equations:

\[
\Delta C_{L_{dp}} = \begin{bmatrix}
P_{1} \cdots \ P_{m}
\end{bmatrix}
\begin{bmatrix}
C_{1} \\
\vdots \\
C_{n}
\end{bmatrix}
\begin{bmatrix}
r_{dp}^{C_{1}} u_{C_{1}} \\
\vdots \\
r_{dp}^{C_{n}} u_{C_{n}}
\end{bmatrix},
\]

where \( P_{1} \) to \( P_{m} \) indicate position numbers, \( C_{1} \) to \( C_{n} \) indicate component numbers, \( \Delta C_{L} \) indicates the hemoglobin change (product of concentration and optical path length), \( \Delta C_{L_{dp}} \) and \( \Delta C_{L_{sh}} \) indicate deep (dp) and shallow (sh) signals. \( u_{C_{i}} \) indicates the time course of the independent component \( (C_{n}) \). \( r_{dp}^{C_{n}} \) and \( r_{sh}^{C_{n}} \) indicate deep and shallow layer contribution ratios for component \( C_{n} \), and \( W_{dp}^{C_{n}} \) indicates the weight of component \( C_{n} \) as the output of the ICA at position \( P_{m} \). When all independent components are used, \( r_{sh}^{C_{n}} \) can be expressed as

\[
r_{sh}^{C_{n}} = 1 - r_{dp}^{C_{n}}.
\]

To reconstruct deep and shallow signals, weight \( W \) and contribution ratio \( (r) \) are used. \( W \) is obtained with the ICA process and \( r \) is obtained with the x-intercept using a product of weight and the effective amplitude of the component along
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Fig. 3 Geometric explanation for calculation of contribution ratios ($r$) of deep and shallow signals in original NIRS signal under two $x$-intercept conditions: (a) $X_{ex} < X_{gr}$ and (b) $X_{ex} \geq X_{gr}$. The deep-layer contribution ratios for component $Cl$ are expressed as $r_{dp} = B/A$ in (a) and $r_{dp} = B/(2B - A)$ in (b).

Fig. 4 Partial optical path lengths for scalp and gray matter in infant head model by Monte Carlo simulation.

8 mm for the infant head model. In the following, data are analyzed using 8 mm as the $X_{gr}$. The shortest S-D distance used in this study is 10 mm, which is longer than $X_{gr}$. Even if an S-D distance shorter than $X_{gr}$ is used (e.g., 5 mm) and a pure scalp signal is obtained, the contributions of shallow (scalp) and deep (cerebral) signals cannot be correctly quantified when the signals are similar (temporally correlated with each other). To overcome the limitation, we used the dependence of signal components on the S-D distance using multidistance optodes and ICA, where multiple S-D distances longer than the $X_{gr}$ are used.

2.6 Analysis After Deep-Shallow Separation

The following analyses were conducted after deep and shallow signals were reconstructed from the original NIRS signals (Fig. 2).

2.6.1 Deep-layer contribution ratio

Means and standard deviations of deep-layer contribution ratios for each S-D distance of left and right channels were calculated by the MD-ICA method. We performed a three-way analysis of variance (ANOVA) [S-D distance (10/20/30 mm)$\times$ hemisphere (L/R)$\times$ condition (no stimulus/speech)] for the deep-layer contribution ratios of oxy- and deoxy-Hb signals. The statistical threshold was set at $p < 0.025$ in order to take into account multiple comparisons between the two types of Hb signals.

2.6.2 Mean signal amplitude of blocks

We divided each continuous dataset into block data that included 1 s before the onset time to 12 s after the stimulation for each trial. After we divided the continuous data into blocks, the mean amplitude during a time window (4.1 to 7.0 s from onset for oxy-Hb and 5.0 to 7.6 s from onset for deoxy-Hb) was calculated. The time window was adopted from that of Taga et al.86 We performed a three-way ANOVA [S-D distance (10/20/30 mm)$\times$ hemisphere (L/R)$\times$ layer (original/deep/shallow)] for the mean signal amplitude under the no-stimulus and speech conditions. The statistical threshold was set at $p < 0.0125$ in order to take into account multiple comparisons between the two types of Hb signals under the two conditions.
2.6.3 Correlation coefficients between left and right channels

The functional connectivity among the cortical regions has been assessed by calculating the temporal correlations of the continuous oxy- and deoxy-Hb signals obtained at multichannels under no-stimulus conditions. To investigate the interhemispheric functional connectivity between the homologous regions of the temporal cortices and the effect of the superficial layer on the connectivity, the correlation coefficients between the left-right channels at each S-D distance were calculated for original, deep, and shallow signals. By using continuous data from original, deep, and shallow signals, correlation coefficients (Fisher’s Z) between the left-right channels were calculated at each S-D distance and task condition. The results were compared among all conditions. We performed a three-way ANOVA [S-D distance (10/20/30 mm) × layer (original/deep/shallow) × condition (no stimulus/speech)] for the correlation coefficients of oxy- and deoxy-Hb signals. The statistical threshold was set at $p < 0.025$ in order to take into account multiple comparisons between the two types of Hb signals.

3 Results

3.1 Block-Averaged NIRS Time Courses

Block-averaged NIRS time courses for each layer (original/deep/shallow) under the speech condition are shown in Fig. 5. Red solid and blue dashed lines indicate oxy- and deoxy-Hb, respectively. Vertical solid and dashed lines indicate the start (onset) and end of the stimulation period, respectively.

3.2 Deep-Layer Contribution

Table 1 lists means and standard deviations among participants of deep-layer contribution ratios for the left and right channels under the no-stimulus and speech conditions.

Table 2 lists three-way ANOVA results for deep-layer contribution ratio ($F$ values). S-D distance had a significant main effect on deep-layer contribution ratios of both oxy- [$F(2,98) = 416.0, p < 0.0005$] and deoxy-Hb [$F(2,98) = 298.8, p < 0.0005$] signals. A post hoc test (Tukey-Kramer) was conducted to identify specific differences among the contribution ratios.
obtained at the three S-D distances. In the results for both oxy-
and deoxy-Hb signals, the mean of the contribution ratios at
each S-D distance was significantly different from those at
the other S-D distances (p < 0.0005).

The test conditions (no stimulus/speech) had significant main
effects on the deep-layer contribution ratios of both oxy-
Hb \( F(1.98) = 8.6, \ p < 0.005 \) and deoxy-Hb \( F(1.98) = 8.6, \ p < 0.005 \) signals. Interactions between hemispheres and
conditions were significant for both oxy-Hb (p < 0.025) and
dehoxy-Hb (p < 0.025) signals.

### 3.3 Mean Signal Amplitude

Mean signal amplitudes for each condition are shown in Fig. 6.
In the data obtained at S-D 20 and 30 mm, original- and deep-
layer signals showed similar changes in mean signal amplitudes.
In the data obtained at S-D 10 mm, which included relatively
higher scalp effects than other longer S-D distances, more sig-
nificant signal changes in oxy- and deoxy-Hb were obtained in
the deep signal than in the original and shallow signals. No sig-
nificant change in the mean signal amplitude was obtained under
the no-stimulus condition.

### Table 2 Three-way analysis of variance (ANOVA) results (F values)

|                | Oxy-Hb | Deoxy-Hb |
|----------------|--------|----------|
| (A) S-D distance (2, 98) | 416.0*** | 298.8*** |
| (B) Hemisphere (1, 98)     | 2.5     | 2.5      |
| (C) Condition (1, 98)      | 8.6**   | 8.6**    |
| A × B (2, 98)              | 0.1     | 0.1      |
| A × C (2, 98)              | 0.1     | 0.1      |
| B × C (1, 98)              | 5.6*    | 7.4*     |

The numbers in parentheses in the first column indicate the degrees
of freedom for each factor and interaction. Single (*), double (**), and
triple (***#) asterisks denote statistical significance at p < 0.025, 0.005,
and 0.0005, respectively.

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For both cases of oxy- and deoxy-Hb signals, post hoc test results showed that correlation coefficients of the shallow signals were significantly smaller than those of the other layer (original and deep) signals ($p < 0.0005$ for oxy-Hb, $p < 0.025$ for deoxy-Hb), whereas there was no difference between the correlation coefficients of original and deep signals ($p > 0.9$ for both oxy- and deoxy-Hb).

### 4 Discussion

#### 4.1 Deep-Layer Contribution Ratio

The present study analyzing data of three-month-old infants showed that the deep-layer contribution ratios (averaged between left and right channels) at a 30-mm S-D distance under the speech condition for oxy- and deoxy-Hb were 82 and 78%, respectively, as shown in Table 1. Note that Funane et al.\(^4\) reported that in adults, the mean deep-layer contribution ratios at the S-D 30-mm channels for oxy- and deoxy-Hb were 65 and 67% among several tasks, respectively. Thus, the deep-layer effect in infants was greater than that in adults. For 20-mm

![Mean signal amplitude](image)

**Table 3** Three-way ANOVA results ($F$ values) for mean signal change.

|                | No stimulus | Speech |
|----------------|-------------|--------|
| **Oxy-Hb**     |             |        |
| S-D distance   | 2.4         | 24.9***|
| Hemispheric    | 0.1         | 0.8    |
| Layer          | 0.4         | 18.7** |
| A×B (2, 148)   | 1.2         | 0.1    |
| A×C (4, 148)   | 0.3         | 5.8**  |
| B×C (2, 148)   | 0.0         | 0.1    |

The numbers in parentheses in the first column indicate the degrees of freedom for each factor and interaction. Triple (***) asterisks denote statistical significance at $p < 0.0005$. 

\[F(2, 148) = 10.5, p < 0.0005\] and deoxy-Hb $[F(2, 148) = 7.0, p < 0.005]$ signals. For both cases of oxy- and deoxy-Hb signals, post hoc test results showed that correlation coefficients of the shallow signals were significantly smaller than those of the other layer (original and deep) signals ($p < 0.0005$ for oxy-Hb, $p < 0.025$ for deoxy-Hb), whereas there was no difference between the correlation coefficients of original and deep signals ($p > 0.9$ for both oxy- and deoxy-Hb).
channels, deep-layer contribution ratios (averaged between left and right channels) under the speech condition for oxy- and deoxy-Hb were 72 and 66%, respectively, while for the 10-mm channels, deep-layer contribution ratios were <35%. Thus, a significant amount of signal originates from deep (such as cerebral) tissue at 20- and 30-mm S-D distances. This was also the case under the no-stimulus condition, suggesting that NIRS signals largely reflect spontaneous changes in cerebral blood oxygenation.

Because the deep-layer contribution ratio is high enough for both 20- and 30-mm channels, the optimal S-D distance depends on a trade-off between probe density and the area covered by a limited number of optodes. If the probe density needs to be high, the S-D distance should be 20 mm, whereas if the covered probe area needs to be larger, the S-D distance should be 30 mm, for example. According to Table 2, the interaction between hemisphere and condition was significant for both oxy- and deoxy-Hb. This was possibly because the contribution of the deep signal induced by stimulation may be larger in the channel for a large response than that in the channel for a smaller response in an infant’s head.

Fig. 7 Correlation coefficients (Z) between left and right channels for all participants. Bar graphs indicate means among participants.

The deep-layer contribution of the oxy-Hb signal is slightly more than that of the deoxy-Hb signal. Franceschini et al.,62 on the other hand, reported that oxy-Hb may be more altered by systemic changes than the deoxy-Hb in adult participants. If the systemic signal contribution is significant, the deep-layer contribution should decrease because such systemic signals are included in both cerebral and extracerebral tissues as reported by Scholkmann et al.9 Therefore, in infants, the extracerebral effect might be smaller than that in adults. Nonetheless, the task dependency of extracranial and systemic effects on NIRS signals in infants is an important issue30 to be further investigated.

### 4.2 Effect of Deep-Shallow Separation on Mean Signal Amplitude

As shown in Fig. 6, at S-D 20 and 30 mm, original and deep layer signals in oxy- and deoxy-Hb showed similar changes in mean signal amplitudes. For S-D 10-mm channel, where the most scalp effects are included among the three S-D distances, more significant signal changes in oxy-Hb were obtained in deep signals than in original signals. Significant signal changes were obtained only for the deep signal, especially for the left channel. From this result, MD-ICA effectively extracts a deep signal from infants’ NIRS signals when it is assumed that the deep signal (i.e., such as a cerebral signal) is induced during the stimulation. The amplitude of deep and shallow signals has the dependence on S-D distance that is expected from optical path length. For example, the shallow signal amplitude does not change much as the S-D distance is changed [Figs. 6(e) and 6(f)]. This result demonstrated that the assumed model for infants was valid enough to extract deep and shallow signals. In the ANOVA results, the significant interactions between the S-D distance and type of layer (original/deep/shallow) existed for the oxy-Hb signal because the signal obtained at a longer S-D distance includes a larger contribution from the deep layer based on optical partial path length.

| Table 4 Three-way ANOVA results (F values) for correlation coefficients between left and right channels. |
|---------------------------------|-----------------|-----------------|
|                                | Oxy-Hb          | Deoxy-Hb        |
| (A) S-D distance (2, 148)      | 33.3***         | 10.4***         |
| (B) Layer (2, 148)             | 10.5***         | 7.0**           |
| (C) Condition (1, 148)         | 0.4             | 0.0             |
| A × B (4, 148)                 | 1.3             | 1.8             |
| A × C (2, 148)                 | 0.1             | 0.1             |
| B × C (2, 148)                 | 0.9             | 0.3             |

The numbers in parentheses in the first column indicate the degrees of freedom for each factor and interaction. Double (**) and triple (***) asterisks denote statistical significance at \( p < 0.005 \) and 0.0005, respectively.
Under the speech condition, the deoxy-Hb signal at the S-D 10-mm channel did not show any significant change (only the left deep signal [Fig. 6(d)] is significant). This was not the case for the oxy-Hb signal. This result can be partly explained as a result of the signal amplitude of the deoxy-Hb signal being smaller than that of the oxy-Hb when comparing their variances, as shown in Fig. 5 (original, S-D 10-mm data). For both oxy- and deoxy-Hb signals, on the other hand, the significance of the changes was increased from original to deep signals (p values were decreased). This means that in deep signals, even though the signal amplitudes were lower than those of the original signals, the significance of the signal change increased, which was caused by a reduced signal variance. It has also been shown that the variance in the S-D 10-mm original signal during the set time window is largely caused by the shallow signal.

There are some differences between Figs. 6(e) and 6(f). Specifically, significant signal changes were obtained only for the oxy-Hb signal under the speech condition. According to Table 1, however, the deep-layer contribution of the oxy-Hb signal is higher than that of the deoxy-Hb signal. The signal amplitudes were calculated using only a task period (block data), whereas the deep-layer contribution reflects all measurement periods (continuous data). That result, therefore, suggests that low-frequency physiological fluctuations other than a task-related signal may have more effect on deep signals for the deoxy-Hb signal than those for the oxy-Hb signal. In other words, the task-related signal contributes more to the oxy-Hb signal than the deoxy-Hb signal. On the other hand, for deep signals [Figs. 6(c) and 6(d)], larger amplitudes and higher significant changes were obtained for oxy-Hb than for deoxy-Hb, which explains the higher deep-layer contribution for the oxy-Hb signal from another standpoint.

In principle, the shallow signal reconstructed by the MD-ICA method mainly reflects light absorption changes in the scalp, and the deep signal reflects those in tissue other than the scalp. From the results of the mean amplitude, the stimulation-related response cannot be explained only using the shallow signal. Therefore, the deep signal mainly originates from cerebral tissue when the blood change in the extracerebral layers in a relatively deeper area (e.g., blood vessels on the brain surface) is negligible.

Note that cerebral (originated from local neuronal activity) and systemic signals are not separated by the deep-shallow signal separation method (MD-ICA) because the method only uses signal depth on the basis of optical path length.

4.3 Interhemispheric Functional Connectivity

ANOVA and post hoc test results of layer (original/deep/shallow) effects showed that correlation coefficients of the original signals mainly reflect the correlation coefficients of the deep signals and that the left-right correlation cannot be accounted for only by the shallow layer. The significance of the S-D distance main effects also suggests that deep-layer signals largely contribute to the correlation coefficient between the left and right channels because the left-right correlation coefficients are significantly smaller at the S-D 10-mm channel than at other S-D distances. These results ensure that the functional connectivity obtained in infants in the previous study largely reflects correlations of deep (such as cerebral) hemodynamic changes.

4.4 Limitation

The infant head model was adopted from the literature and our original simulation, and $X_I$ was assumed to be 8 mm for all participants. The effects of the model error and linearity of the amplitude of the gray matter signal still need to be investigated, as does the case for the application of the MD-ICA method to adults.

In this study, one light source is shared by four detectors at one side; therefore, the measurement channel of each detector does not reflect the same position. That is, the channel of the further detector reflects changes observed in the closer detectors and those in the more posterior region in the head. In the MD-ICA method, the homogeneous absorption change in each layer (deep or shallow) is assumed, so the displacement of the channel positions may affect the analytical result and should be as small as possible.

Since this study focused only on the temporal cortex of the brain, the shallow-layer effect at other cortices of the brain should be further investigated. The shortest S-D distance at which the detected light is sensitive to absorption change in gray matter depends on the extracerebral tissue thickness, which has a variance around the head. It is, thus, necessary in the future to use structural data (extracerebral tissue thickness) for target areas.

5 Conclusion

The MD-ICA method was applied to dissociate deep- and shallow-layer contributions to NIRS signals obtained in infants and to investigate the left-right connectivity for original, deep, and shallow signals under no-stimulus (resting) and speech conditions. The results revealed that deep tissues contributed more than shallow ones to the NIRS signals when 20- and 30-mm S-D distances were chosen. The deep layers of the NIRS signals in infant heads contributed more than those in adult heads, the results of which were previously reported. Moreover, the original signal correlation between the left-right channels mainly originates from deep signal correlation. These results suggest that the cerebral contribution to NIRS signals is significant in functional activation and connectivity studies of infants.

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