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Quantification of CMRO₂ without hypercapnia using simultaneous near-infrared spectroscopy and fMRI measurements

Sungho Tak, Jaeduck Jang, Kangjoo Lee and Jong Chul Ye

Bio Imaging and Signal Processing Lab., Department of Bio and Brain Engineering, KAIST, 335 Gwahak-ro, Yuseong-gu, Daejeon 305-701, Korea

E-mail: jong.ye@kaist.ac.kr

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Abstract

Estimation of the cerebral metabolic rate of oxygen (CMRO₂) and cerebral blood flow (CBF) is important to investigate the neurovascular coupling and physiological components in blood oxygenation level-dependent (BOLD) signals quantitatively. Although there are methods that can determine CMRO₂ changes using functional MRI (fMRI) or near-infrared spectroscopy (NIRS), current approaches require a separate hypercapnia calibration process and have the potential to incur bias in many assumed model parameters. In this paper, a novel method to estimate CMRO₂ without hypercapnia is described using simultaneous measurements of NIRS and fMRI. Specifically, an optimization framework is proposed that minimizes the differences between the two forms of the relative CMRO₂-CBF coupling ratio from BOLD and NIRS biophysical models, from which hypercapnia calibration and model parameters are readily estimated. Based on the new methods, we found that group average CBF, CMRO₂, cerebral blood volume (CBV), and BOLD changes within activation of the primary motor cortex during a finger tapping task increased by $39.5\pm21.4\%$, $18.4 \pm 8.7\%$, $12.9 \pm 6.7\%$, and $0.5 \pm 0.2\%$, respectively. The group average estimated flow-metabolism coupling ratio was 2.38 ± 0.65 and the hypercapnia parameter was $7.7 \pm 1.7\%$. These values are within the range of values reported from other literatures. Furthermore, the activation maps from CBF and CMRO₂ were well localized on the primary motor cortex, which is the main target region of the finger tapping task.

(Some figures in this article are in colour only in the electronic version)

¹ Author to whom any correspondence should be addressed.

1. Introduction

Elucidating the underlying link between the changes in neuronal activity and the cerebral blood flow (CBF) or cerebral metabolic rate of oxygen (CMRO₂) has been of great interest in neuroimaging (Mukamel et al 2005, Schummers et al 2008, Gordon et al 2008, Smith et al 2002). The physiological background of neurovascular and neurometabolic coupling during brain activation is as follows: evoked neuronal activity induces dendrites to consume oxygen (O₂) rapidly (Kasischke et al 2004), which causes oxygen pressure (Ances et al 2001) to decrease (Malonek et al 1997). In the astrocyte, glycolysis is then enhanced (Kasischke et al 2004) and lactate is released (Hu and Wilson 2002). In response to the lactate, vasodilation occurs (Hein et al 2006). A summary of these processes is available in Gordon et al (2008). Fox and Raichle (1986) initially found that with somatosensory stimulation the increase in CBF (29%) exceeded a concurrent increase in CMRO₂ (5%). Numerous studies have confirmed that the fractional change in CBF is approximately minimum twice and maximum four times as large as the fractional change in CMRO₂ (Boas et al 2003, Hoge et al 1999a, 2005, Durduran et al 2004, Sheth et al 2004, Kastrup et al 2002, Kim et al 1999, Davis et al 1998). However, the physiological processes of flow-metabolism coupling during dynamic changes are still unresolved. Explaining the underlying mechanism of the flow-metabolism coupling is crucial for understanding the fundamental physiological effect of blood oxygenation level-dependent (BOLD) response. More specifically, the change in BOLD signal is caused by the combined effects of CBF, CMRO₂, and cerebral blood volume (CBV). To analyze the contributions of these components to BOLD response quantitatively, the balloon model (Buxton et al 1998b), the windkessel model (Mandeville et al 1999b), and the model formulated by Davis et al (1998) have been proposed. The main idea of the balloon model and the windkessel model is that the post-stimulus undershoot of the BOLD response is caused by a delay in the return to baseline by the CBV response, compared to CBF response. However, other studies have argued for a prolonged elevation of CMRO₂ (Frahm et al 2008, Lu et al 2004), and CBF undershoot in the presence of a slow venous CBV response (Chen and Pike 2009) as the primary contributor to BOLD undershoot. Therefore, to assess the main contributor of a transient BOLD response and investigate the neurovascular coupling in a quantitative manner, a robust estimation method of CBF and CMRO₂ is essential.

Determination of relative CMRO₂ changes using functional MRI (fMRI) has been traditionally obtained by measuring the CBF and BOLD changes (Davis *et al* 1998, Hoge *et al* 1999a, Kim *et al* 1999). Assuming that changes in CMRO₂ are negligible during hypercapnia (Horvath *et al* 1994), BOLD signals are calibrated against known changes in CBF. Relative CMRO₂ changes are then estimated from the relative changes of CBF and BOLD. While the fMRI approach of CMRO₂ estimation promises a high spatial resolution (Davis *et al* 1998, Uludağ *et al* 2004), there are several drawbacks. Most importantly, recent studies show that CMRO₂ continues to vary during CO₂ inhalation (Kliefoth *et al* 1979, Jones *et al* 2005, Sicard and Duong 2005), which challenges the important assumption of hypercapnia calibration, as inaccurate estimates of the hypercapnia calibration parameter lead to significant bias regarding the fractional CBF–CMRO₂ coupling ratio (Chiarelli *et al* 2007b). Although breath holding (Kastrup *et al* 1999) and hyperoxia calibration (Chiarelli *et al* 2007c) techniques have been introduced to overcome the disadvantages of the hypercapnia process, an additional measurement step is nonetheless required, which is inconvenient in that it may stress the subjects.

Near-infrared spectroscopy (NIRS) can also estimate changes in CMRO₂ from direct measurements of functional contrasts such as oxy-hemoglobin (HbO), deoxy-hemoglobin (HbR), and total-hemoglobin (HbT) (Boas *et al* 2003, Mayhew *et al* 2001a, 2001b,

Vignal et al 2008). Due to the high temporal resolution, the NIRS approach for CMRO₂ estimation has advantages over fMRI in that it is possible to characterize the transient response of CMRO₂ and CBF more accurately. However, current CMRO₂ estimation techniques based on NIRS measurements rely on multiple assumptions and depend considerably on baseline hemoglobin concentrations, which are difficult to measure through optical methods alone. Hoge et al (2005) proposed a method to determine relative CMRO₂ changes from the diffuse optical imaging and arterial spin labeling (ASL) techniques. The relative changes in CMRO₂ can be calculated as multiplying the relative oxygen extraction fraction and CBF, which are estimated from the optical and ASL measurements, respectively. While this approach does not require any relationship between CBF and CBV, uncertainties introduced by the assumptions of baseline hemoglobin concentration and the unknown partial volume difference between the MRI and optical measurements often limit the accuracy of this approach.

The main contribution of the present paper is to propose a novel method that estimates the changes in CMRO₂ and CBF with simultaneous NIRS and fMRI–BOLD measurements, in a manner that overcomes the limitations of earlier approaches. In the case of fMRI (Davis *et al* 1998), the relative coupling ratio between CMRO₂ and CBF can be represented as a nonlinear function of CBV and BOLD response. For NIRS (Mayhew *et al* 2001a, 2001b, Boas *et al* 2003), the relative coupling ratio between CMRO₂ and CBF arises from HbR and CBV in the venous compartment. Due to the simultaneous recording method, the relative coupling ratio separately derived from each modality should be identical. Therefore, by minimizing the difference between the coupling ratios from two biophysical models, the hypercapnia calibration parameter, the baseline hemoglobin concentrations, and other parameters can be accurately estimated without hypercapnia. Hence, the controversial assumption of CMRO₂ invariance during hypercapnia is not necessary in the proposed approach, which may enhance the reliability of this method in explaining the origin of BOLD.

2. Theory

2.1. Relative CMRO₂–CBF ratio from the BOLD biophysical model (Davis et al 1998, Hoge et al 1999a)

When the changes in the transverse relaxation rate caused by HbR, $\Delta R_{2,\text{HbR}}^*(t)$, are small enough, the fractional changes in the T_2^* -weighted BOLD signals are approximated by

$$\frac{\Delta \text{BOLD}(t)}{\text{BOLD}_0} \cong -\text{TE}\Delta R_{2,\text{HbR}}^*(t)$$

$$\cong H \left(1 - \text{rCBV}(t) \left(\frac{[\text{HbR}(t)]_v}{[\text{HbR}]_{v_0}}\right)^{\beta}\right), \tag{1}$$

where $\Delta R_{2,\mathrm{HbR}}^*(t) = c \left(\mathrm{CBV}(t) [\mathrm{HbR}(t)]_v^\beta - \mathrm{CBV_0} [\mathrm{HbR}]_{v_0}^\beta \right)$ (Boxerman *et al* 1995); the hypercapnia calibration parameter $H = \mathrm{TE} \cdot c \cdot \mathrm{CBV_0} [\mathrm{HbR}]_{v_0}^\beta$, where TE denotes the echo time, c denotes a constant dependent on the magnetic field strength and vascularity, and β is a constant that ranges from 1 to 2; $\mathrm{rCBV}(t)$ is the relative CBV to its baseline; $[\mathrm{HbR}(t)]_v = \mathrm{HbR}_v(t)/\mathrm{CBV}_v(t)$, $[\mathrm{HbR}]_{v_0} = \mathrm{HbR}_{v_0}/\mathrm{CBV}_{v_0}$, where the subscript '0' denotes the baseline value, and the subscript 'v' denotes the venous vasculature, respectively.

Assuming that the HbR concentration in arterial blood is negligible, Fick's law gives the following relationship (Hoge *et al* 1999a):

$$\frac{[\text{HbR}(t)]_v}{[\text{HbR}]_{v0}} = \frac{\text{rCMRO}_2(t)}{\text{rCBF}(t)},\tag{2}$$

where rCMRO₂(t) and rCBF(t) are relative CMRO₂, and CBF to their respective baselines. Substituting equation (2) for [HbR(t)] $_v$ /[HbR] $_{v0}$ in equation (1), the relative CMRO₂–CBF coupling ratio from the BOLD biophysical model is then derived as

$$\frac{\text{rCMRO}_2(t)}{\text{rCBF}(t)} = \text{rCBV}(t)^{-1/\beta} \left(1 - \frac{1}{H} \cdot \frac{\Delta \text{BOLD}(t)}{\text{BOLD}_0} \right)^{1/\beta}.$$
 (3)

2.2. Relative CMRO₂–CBF ratio from the NIRS biophysical model (Boas et al 2003, Mayhew et al 2001a, 2001b)

Assuming that the quantities in the venous vasculature are proportional to those in the total vascular compartment (Mayhew *et al* 2001a, 2001b), the relative coupling ratio from the NIRS biophysical model is derived from Fick's law (Hoge *et al* 1999a) as

$$\frac{\text{rCMRO}_{2}(t)}{\text{rCBF}(t)} = \frac{\text{rHbR}_{v}(t)}{\text{rCBV}_{v}(t)}$$

$$= \frac{1 + \gamma_{r} \left(\Delta \text{HbR}(t) / \text{HbR}_{0}\right)}{1 + \gamma_{v} \left(\Delta \text{CBV}(t) / \text{CBV}_{0}\right)},$$
(4)

where rHbR denotes relative HbR to its baseline, the venous HbR ratio γ_r and the venous HbT ratio γ_v are given by

$$\gamma_r = \frac{\Delta \text{HbR}_v(t)/\text{HbR}_{v0}}{\Delta \text{HbR}(t)/\text{HbR}_0} \qquad \gamma_v = \frac{\Delta \text{CBV}_v(t)/\text{CBV}_{v0}}{\Delta \text{CBV}(t)/\text{CBV}_0}. \tag{5}$$

2.3. Estimation of relative CMRO₂—CBF coupling ratio without hypercapnia calibration

Substituting equation (2) into equation (1), the fractional changes in BOLD are given by

$$\frac{\Delta BOLD(t)}{BOLD_0} = H \left(1 - rCBV(t) \left(\frac{rCMRO_2(t)}{rCBF(t)} \right)^{\beta} \right). \tag{6}$$

Assuming that there are no changes in CMRO₂ during hypercapnia (Horvath *et al* 1994), $rCMRO_2(t) \approx 1$, the hypercapnia calibration parameter *H* is derived from equation (6):

$$H = \frac{\Delta BOLD_H(t)/BOLD_{H0}}{1 - rCBV_H(t)rCBF_H(t)^{-\beta}},$$
(7)

where the subscript 'H' denotes the hypercapnia condition. The hypercapnia calibration has been extensively used for quantitative fMRI analyses (Davis *et al* 1998, Hoge *et al* 1999a, Leontiev *et al* 2007, Ances *et al* 2008, Chiarelli *et al* 2007a, 2007b, Wu *et al* 2002). However, there are many sources of bias associated with hypercapnia calibration. Since H is calibrated before the estimation of CMRO₂, possible changes in CMRO₂ cannot be incorporated (Jones *et al* 2002). Furthermore, the changes in CMRO₂ during CO₂ inhalation (Kliefoth *et al* 1979, Jones *et al* 2005, Sicard and Duong 2005) challenge the assumed iso-metabolic condition. This situation may lead to inaccurate estimates of the fractional CBF–CMRO₂ coupling ratio (Chiarelli *et al* 2007b).

Rather than using hypercapnia calibration, the multi-modal approach is therefore proposed in this study to estimate the relative CMRO₂–CBF coupling ratio. Due to simultaneous recording, the coupling ratio derived from the BOLD and NIRS biophysical models should be identical. Hence, using equation (3) and equation (4) gives

$$rCBV(t)^{-1/\beta} \left(1 - \frac{1}{H} \cdot \frac{\Delta BOLD(t)}{BOLD_0} \right)^{1/\beta} = \frac{1 + \gamma_r (\Delta HbR(t)/HbR_0)}{1 + \gamma_v (\Delta CBV(t)/CBV_0)}.$$
 (8)

At this stage, the unknown parameters are H, β , γ_r , γ_v , CBV₀, and HbR₀. In order to estimate these values, the regression problem is converted as follows:

$$\min_{\mathbf{H}, \beta, \gamma_r, \gamma_v, CBV_0, HbR_0} \sum_{i=1}^{N} \left\| \left(\frac{\Delta BOLD(t_i)}{BOLD_0} \right) - H \left(1 - \left(1 + \gamma_0 \frac{\Delta CBV(t_i)}{CBV_0} \right) \left(\frac{1 + \gamma_r \left(\Delta HbR(t_i) / HbR_0 \right)}{1 + \gamma_v \left(\Delta CBV(t_i) / CBV_0 \right)} \right)^{\beta} \right) \right\|_{2}, \tag{9}$$

where $\{t_i\}_{i=1}^N$ denotes the sampling times for the BOLD time series, and $\gamma_0 = 1$ for the CBV from all vasculature. In this equation, $||\cdot||_2$ denotes the l_2 norm. The reason for converting the fitting problem equation (8) to equation (9) is to use NIRS measured oxygen species as covariates to fit the BOLD response.

It turns out that our nonlinear regression problem is severely ill-conditioned due to the redundant parameters. For example, an infinite number of (HbR_0, γ_r) pairs provides the same cost value; thus, HbR_0/γ_r should be estimated as one independent value $\eta_r = HbR_0/\gamma_r$. Furthermore, a previous study (Wu *et al* 2002) suggested that the CBV response in Davis's model comes from the venous compartment, i.e. $\gamma_0 = \gamma_v$ in equation (9). In this case, we can easily see that an infinite number of (CBV_0, γ_v) pairs with the same ratio provide the same cost values. Hence, we should estimate these values as one independent variable $\eta_v = CBV_0/\gamma_v$.

2.4. Estimation of CBF and CMRO₂

In the proposed fMRI–NIRS simultaneous recording process, CBF is not measured directly, but is instead estimated using the CBV–CBF relationship. This is done because the changes in CBF and CBV are closely related, as the increase of CBF is triggered by an increase in capillaries and veins, and the increase in the pressure expands the vessels. Grubb *et al* (1974) first described the steady-state relationship between relative CBF and relative CBV on whole brain by changing CBF with inhaled CO₂:

$$rCBV(t) = rCBF(t)^{\alpha}.$$
 (10)

Here, a value of $\alpha = 0.38$ is typically employed as the power of Grubb *et al* (1974) relationship. However, α has different values depending on the brain region (Wu *et al* 2002, Rapoport *et al* 1979). Furthermore, it is reported that during the transient state the CBF–CBV relationship is not simply governed by the power law.

To explain the transient response in which the CBV values return to their baseline more slowly than CBF after the stimulus is removed (Mandeville *et al* 1998), the balloon model (Buxton *et al* 1998b) and the windkessel model (Mandeville *et al* 1999b) have been proposed. The balloon model assumes that the changes in the blood volume mainly take place in the venous compartment and treats the venous compartment as a distensible balloon. The mass of blood and deoxy-hemoglobin is conserved through the venous balloon. Thus, the rate of volume change is represented as the difference between the inflow rCBF $_{in}(t)$ and the outflow rCBF $_{out}(t)$ with a time constant τ_{MTT} :

$$\frac{\mathrm{d}\left(\mathrm{rCBV}\right)}{\mathrm{d}t} = \frac{1}{\tau_{\mathrm{MTT}}}[\mathrm{rCBF}_{\mathrm{in}}(t) - \mathrm{rCBF}_{\mathrm{out}}(t)],\tag{11}$$

where $\tau_{\text{MTT}} = \text{CBV}_0/\text{CBF}_0$ denotes the mean transit time through the volume at rest, rCBF_{in}(t) corresponds to rCBF(t), and rCBF_{out}(t) is described as in an earlier study (Buxton et al 1998a):

$$rCBF_{out}(t) = rCBV(t)^{\frac{1}{\alpha}} + \tau_{\alpha} \frac{d(rCBV)}{dt}.$$
 (12)

Using equations (11) and (12), the relative CBF is given by

$$rCBF(t) = rCBV(t)^{1/\alpha} + (\tau_a + \tau_{MTT}) \frac{d (rCBV)}{dt},$$
(13)

where α is a constant, τ_{MTT} denotes the mean transit time, and τ_a is a viscoelastic time constant.

Using the estimated relative CBF values, the relative CMRO₂ can be calculated by multiplying the relative CMRO₂–CBF coupling ratio and rCBF:

$$rCMRO_{2}(t) = \frac{r\widehat{CMRO_{2}}(t)}{rCBF(t)} \cdot r\widehat{CBF(t)},$$
(14)

where $\frac{r\widehat{CMRO}(t)}{rCBF(t)}$ denotes the estimated relative coupling ratio and $r\widehat{CBF}(t)$ is the estimated relative CBF, respectively. Although a numerical model is used to estimate the CBF, our framework can be easily incorporated with direct blood flow measurements such as ASL, again without a hypercapnia calibration. In this case, the $\frac{r\widehat{CMRO}_2(t)}{rCBF(t)}$ values in equation (14) are measured by NIRS and BOLD simultaneous recording, whereas $r\widehat{CBF}(t)$ is obtained from ASL measurements.

2.5. Procedures for estimating CMRO₂ and CBF

The NIRS and BOLD measurements are prone to many noisy factors such as global drifts, vasomotion, blood pressure variation, or long-term physiological changes, etc. Since the amplitude of these bias is often comparable to that of the activated signals, a simple average of NIRS and BOLD response across many blocks often results in bias. This is especially problematic in our case where we need to estimate several calibration parameters by fitting two time traces from different modalities. In order to deal with this issue, we propose a wavelet-based adaptive averaging method for estimating HRF, which is described in the appendix.

After the HRFs are estimated, HbR, HbT, and BOLD time series can be estimated by convolving with the paradigm. The relative CMRO₂–CBF coupling ratio and model parameters including H, β , η_r , and η_v are then estimated from equation (9). This procedure makes the fitting problem more robust than the measurement noise.

The search ranges of model parameters are chosen within the physiological ranges which were estimated from other studies. Specifically, H ranges from 0 to 10% (Kastrup et~al~2002, Chiarelli et~al~2007a) and β ranges from 1 to 2 (Davis et~al~1998, Uludağ et~al~2004). The venous deoxy-hemoglobin ratio γ_r and venous total-hemoglobin ratio γ_v are both in the range of 0.5–1.5 (Boas et~al~2003, Mayhew et~al~2001b). The baseline oxygen saturation SO₂ and total baseline blood volume V_0 ranges from 55 to 80% and 40 to 140 μ M, respectively (Boas et~al~2003), Torricelli et~al~2001). In order to estimate the accurate parameters, we consider all possible values of unknown parameters by discretizing the parameter search range. All of the search ranges are summarized in table 1.

The optical and MR measurements reflect the changes over the different volume of tissue, which leads to different partial volume effect. Since the partial volume effect affects the accuracy of parameter estimation, it should be corrected. By changing the dynamic properties of the brain layer (scalp, skull, brain) and simulating the CBF in the bottom layer with a semi-infinite medium model, Durduran *et al* (2004) estimated the partial volume effect of NIRS. Since our probe geometry and the experimental protocol are very similar to their simulation condition, we employed the partial volume correction factor p of 6.2 from their work (Durduran *et al* 2004). In particular, we multiply the search range of CBV₀ by the partial

Table 1. Search range from model parameters. The four unknown parameters in our model are H, β , η_r , and η_v . The search space of the combined parameter η_r and η_v is calculated considering feasible parameter ranges of CBV₀, SO₂, γ_r , and γ_v . The partial volume factor p is 6.2 (Durduran *et al* 2004). Due to the space limit, a single representative reference is listed for each row.

Symbol	Description	Search range	References
Н	Hypercapnia calibration	[0-10] (%)	Chiarelli et al (2007a)
β	Davis's model parameter	[1–2]	Davis et al (1998)
η_r	$\eta_r = (1 - SO_2)CBV_0/\gamma_r$	$p \cdot [4-126]$	
η_v	$\eta_v = \mathrm{CBV}_0/\gamma_v$	$p \cdot [26-280]$	
CBV_0	Total baseline blood volume	$[40-140] (\mu M)$	Torricelli et al (2001)
SO_2	Baseline oxygen saturation	[55–80] (%)	Boas <i>et al</i> (2003)
γ_r	Venous deoxy-hemoglobin ratio	[0.5–1.5]	Boas <i>et al</i> (2003)
γ_v	Venous total hemoglobin ratio	[0.5–1.5]	Boas et al (2003)

volume factor p (see table 1). Finally, after the estimation of relative CMRO₂–CBF coupling and model parameters, the relative CBF and CMRO₂ are then calculated using a selected CBV–CBF relationship and equation (14).

3. Method

3.1. Behavior protocol

To evaluate the practicality of the proposed method, a right finger tapping experiment was performed. Subject instruction was visually presented using a beam projector. When the word 'go' appeared, the subjects were to perform a simple finger flexion and extension task repeatedly. When the word 'stop' appeared, the subjects were to stop moving and stare at a fixed point to avoid eye and head movements. As BOLD undershoot was observed for up to 60 s (Bandettini *et al* 1997, Mandeville *et al* 1999a), the finger tapping experiment with a long resting period was performed to investigate the BOLD undershoot. Specifically, a block design sequence consisted of 15 s of task and 72 s of rest in one cycle. The full experimental run consisted of 90 s of rest, followed by four task and rest cycles, followed by an additional 30 s of rest. The total recording time was 468 s. It was expected that the primary motor cortex, somatosensory cortex, and supplementary motor cortex would be mainly activated during finger tapping, based on experimental results from other literature (Witt *et al* 2008). Due to the limit of the penetration depth of the NIRS system, the basal ganglia and cerebellum were excluded from the region of interest (ROI).

3.2. Subject selection

For the finger tapping task, a total of three healthy subjects were examined (mean age $= 27 \pm 2.6$ years). No subject had a history of any neurological disorder. After all of the subjects were given instructions concerning the experimental environment and the operating mode of NIRS and fMRI, signed informed consent forms were obtained. This study was approved by the Institutional Review Board of the Korea Advanced Institute of Science and Technology (KAIST).

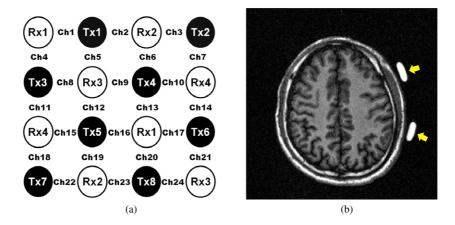


Figure 1. NIRS channel configuration and optode positions on the subject's head. (a) Schematic diagram of 24 NIRS channels configuration. Tx and Rx denote the illuminator and detector, respectively. Ch denotes the channel. (b) T1-weighted MR image showing the marker capsules highlighted by arrows in an axial section. NIRS channels are localized onto the cerebral cortex of the T1-weighted MR image using Horn's algorithm (Horn *et al* 1987) and the relationship between the MR coordinate and real coordinates is elicited in a 3D digitizer.

3.3. Data acquisition

Experimental data were simultaneously recorded using NIRS and fMRI. Detailed system specifications are described as follows. A continuous wave NIRS instrument (OXYMON MK III, Artinis, the Netherlands) was used to measure the changes in the optical density. The NIRS system emitted 781 nm and 856 nm laser lights at each source fiber. The fiber length was 10 m to connect the optodes in the MR scanner to the NIRS system in the MR control room. The sampling frequency was 10.0 Hz. The NIRS system had 24 channels with eight sources and four detectors, as shown in figure 1(a). The distance between the source and detectors was 3.5 cm. An optode holder cap was attached to the scalp around the left primary motor cortex, somatosensory cortex, and supplementary motor cortex. For spatial registration of the NIRS channels to the Montreal Neurological Institute (MNI) standard space, MR marker capsules were placed on the optode holder cap, as shown in figure 1(b). The marker capsule positions were used to elicit the relationship between the real coordinates and MNI coordinates using Horn's algorithm (Horn et al 1987). A 3.0T MRI system (ISOL, Republic of Korea) was used to measure the BOLD response. During the blocked task paradigm, the echo planar imaging (EPI) sequence was used with TR/TE equal to 3000/35 ms, a flip angle of 80°, and 35 slices with a 4 mm slice thickness. In the subsequent anatomical scanning session, T1-weighted structural images were acquired using the same scanner.

3.4. Data analysis

The concentration changes of HbO, HbR, and HbT were calculated from the measured optical density variations using the modified Beer–Lambert law (Cope and Delpy 1988). In order to remove the noise from the NIRS and BOLD time series, Gaussian smoothing with the full width at half-maximum of 2 s was applied. The wavelet–MDL based detrending algorithm (Jang *et al* 2009) was used to eliminate unknown global trends including breathing, subject movement, or instrumental instability.

According to the proposed method, the noise in the measurement was further reduced. HbR, HbT, and BOLD responses during one block of experiment were then estimated. Now,

by solving the regression problem in equation (9), the relative CMRO₂–CBF coupling ratio and model parameters were estimated. We selected the channels within the HbT activated region (p < 0.01, tube formula correction (Ye *et al* 2009, Sun 1993, Cao and Worsley 1999)) which overlapped with the primary motor cortex, and then estimated model parameters. Finally, the relative CBF and CMRO₂ were calculated using equation (10) and equation (14), respectively.

Statistical analyses of estimated CBF and CMRO₂ were performed using the software package NIRS-SPM (Ye *et al* 2009, Jang *et al* 2009). Specifically, the general linear model (GLM) (Worsley and Friston 1995), which explains measurements as a linear combination of explanatory variables, provided the Student's *t*-statistics on the interpolated channel positions. Given a specific *p*-value, Sun's tube formula (Sun 1993, Cao and Worsley 1999) was used to calculate an accurate threshold for CBF and CMRO₂ activation in an inhomogeneous Gaussian random field. To correct the *p*-value for the BOLD signal, random field correction was applied using a statistical parametric mapping (SPM) package (Wellcome Department of Cognitive Neurology, London, UK) (Friston *et al* 2006). Activated regions were rendered on the brain surface generated from an individual, which is included in the SPM package. Figures 2(a) and (b) show the schematics describing the proposed method of CBF and CMRO₂ estimation and corresponding statistical analysis method using NIRS-SPM, respectively. The proposed CBF and CMRO₂ quantification method will be included in the NIRS-SPM software package for download (http://bisp.kaist.ac.kr/NIRS-SPM).

4. Results

In order to confirm how well the proposed method estimates the model parameters, we separately generated the relative CMRO₂–CBF coupling ratio from BOLD and NIRS models, after solving the fitting problem in equation (9). Figure 3 shows time courses of the estimated CMRO₂–CBF coupling ratio on the channels included in the ROI for (a) the first subject, (b) the second subject, and (c) the third subject, respectively. The ROI was selected as the HbT activation area (p < 0.01, tube formula correction) which overlapped with the primary motor cortex. Due to variation in channel positions of individuals, the number of channels within the ROI was different between individuals. The solid lines indicate relative CMRO₂–CBF coupling ratios from the NIRS biophysical model, whereas the dotted lines indicate those from the BOLD biophysical model. The patterns of the relative CMRO₂–CBF coupling ratio from both models were mostly consistent. The coefficient of determination (R^2) from most of the channels is more than 0.9.

Figure 4(a) shows the group average of the estimated HbO, HbR, HbT concentration changes, and fractional BOLD changes. The solid black line denotes the task period. The error bar indicates the standard error of the mean (SEM) across the subjects at each time point. The ROI was selected as the HbT activation area which overlapped with the primary motor cortex. The group average HbO, HbR, and HbT changes during stimulus were $0.65 \pm 0.35 \,\mu\text{M}$, $-0.06 \pm 0.05 \,\mu\text{M}$, and $0.56 \pm 0.29 \,\mu\text{M}$, respectively. The group average of fractional BOLD changes was $0.5 \pm 0.2\%$. Note that post-stimulus undershoot of BOLD and HbT were observed.

Figure 4(b) shows group average time series of the relative CMRO₂–CBF coupling ratio. During the post-stimulus, overshoot was estimated. The average hypercapnia parameter H between the subjects was $7.7 \pm 1.7\%$, which is within the range of values estimated in the motor cortex (Kastrup *et al* 2002, Stefanovic *et al* 2004, Chiarelli *et al* 2007a). The average β between the subjects was 1.19 ± 0.23 . The β value is known to be 1.5 for the magnetic field strength of 1.5 tesla (Davis *et al* 1998). However, for higher field strengths, it is expected that the BOLD signal mainly comes from the extravascular space, corresponding to a β value

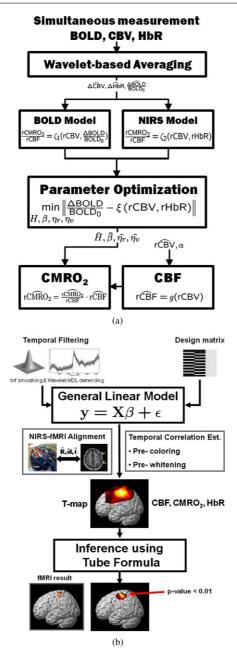


Figure 2. Schematics of cerebral blood flow (CBF)—cerebral metabolic rate of oxygen (CMRO₂) estimation method and corresponding statistical analysis method. (a) The proposed method to determine relative CMRO₂ (rCMRO₂) and relative CBF (rCBF). ζ_1 (rCBV, ΔBOLD/BOLD₀) and ζ_2 (rCBV, rHbR) denotes the relative CMRO₂/CBF ratio from the BOLD biophysical model (Davis *et al* 1998, Hoge *et al* 1999a) and the NIRS biophysical model (Boas *et al* 2003, Mayhew *et al* 2001b), respectively. In the parameter optimization step, ξ (rCBV, rHbR) denotes the biophysical model of fractional BOLD changes. (b) Statistical analysis framework using NIRS-SPM (Ye *et al* 2009, Jang *et al* 2009). Based on the general linear model (Worsley and Friston 1995) and Sun's tube formula (Sun 1993, Cao and Worsley 1999), activated regions from CBF and CMRO₂ are extracted.

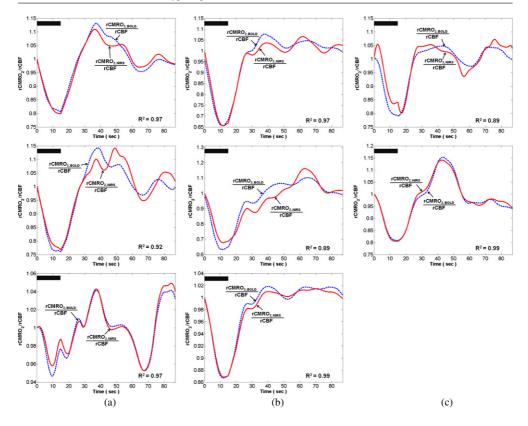


Figure 3. Time courses of the estimated CMRO₂–CBF coupling ratio on the channels included in the region of interest (ROI) for (a) the first subject, (b) the second subject and (c) the third subject, respectively. We define the ROI as the total hemoglobin activated region (p < 0.01, tube formula correction) which overlapped with the primary motor cortex. The solid lines indicate the relative CMRO₂–CBF coupling ratio from the NIRS biophysical model, whereas the dotted lines denote those from the BOLD biophysical model. The coefficient of determination (R^2) from most of the channels is more than 0.9. Due to the variation of channel positions for each individual, the number of channels within the ROI were three for the first and second subjects (figures 3(a) and (b)) and two for the third subject (figure 3(c)).

close to 1 (Buxton 2002). Therefore, in our 3.0T MRI system, the estimated β value is in agreement with the expected value.

The group average time series of the relative CBF and CMRO₂ estimates are shown in figure 4(c). The rCBF_{GM} and rCBF_{BM} denote CBF estimated from the Grubb *et al* (1974) relationship ($\alpha=0.38$) and the balloon model ($\tau=5$ s) (Buxton *et al* 1998b), respectively. The rCMRO_{2,GM} and rCMRO_{2,BM} denote the relative CMRO₂, which are calculated by multiplying the relative flow-metabolism coupling ratio with rCBF_{GM} and rCBF_{BM}, respectively. Whereas the balloon model dynamically explains the transient response of CBF and corresponding CMRO₂ compared with Grubb's model, the large τ value more than 10 s introduces the pronounced fluctuations. In both models, at the onset of stimulus, CBF rapidly increased, and then persisted over larger fractional changes than CMRO₂. Upon the cessation of stimulus, CBF decreased faster than CMRO₂. The significant post-stimulus undershoot of CBF was observed. During stimulus, the CBF_{GM} increase was 39.5 \pm 21.4%.

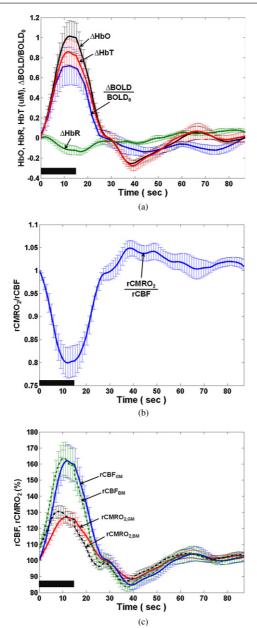


Figure 4. Group average time series of (a) the estimated HbO, HbR, HbT concentration changes, and fractional BOLD changes, (b) relative CMRO2–CBF coupling ratio, and (c) relative CBF (rCBF) and relative CMRO2 (rCMRO2) estimates from the finger tapping experiment. The rCBF_{GM} and rCBF_{BM} denote rCBF estimated from the Grubb *et al* (1974) relationship ($\alpha=0.38$) and the balloon model ($\tau=5$ s) (Buxton *et al* 1998b), respectively. The rCMRO2,GM and rCMRO2,BM denotes the rCMRO2 which was calculated as multiplying the relative flow-metabolism coupling ratio and rCBF_{GM} and rCBF_{BM}, respectively. The solid black line denotes the task period. The error bar indicates the standard error of the mean (SEM) across the subjects at each time point. The region of interest is the total hemoglobin activation region (p<0.01, tube formula correction) which overlapped with the primary motor cortex. During stimulus, the change in rCBF_{GM} was $39.5\pm21.4\%$. The change in rCMRO2,GM was $18.4\pm8.7\%$.

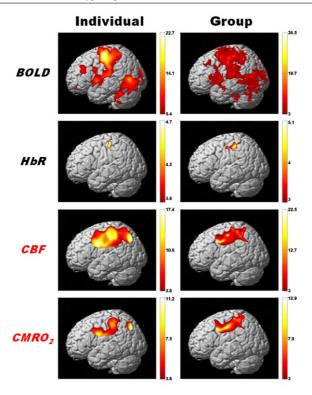


Figure 5. Activation areas from BOLD, HbR, CBF, and CMRO₂ during the finger tapping task. The first column shows the individual activation patterns (p < 0.01, BOLD: random field correction, NIRS: tube formula correction). The second column shows the group activation patterns (p < 0.05, uncorrected) during the finger tapping task. The activated region of CBF corresponds well with that of CMRO₂ and the target region of the finger tapping task. CMRO₂ activation was more tightly localized on the primary motor cortex, compared with CBF activation.

The change in CMRO_{2,GM} was $18.4 \pm 8.7\%$. The fractional CBF–CMRO₂ coupling ratio n(t) is then calculated as follows:

$$n(t) = \frac{\text{rCBF}(t) - 1}{\text{rCMRO}_2(t) - 1}.$$
(15)

Note that this is different from the *relative* CMRO₂–CBF coupling ratio, rCMRO₂(t)/rCBF(t), which results from the proposed fitting procedure. The group average of fractional CMRO₂–CBF coupling ratio, n, was 2.38 ± 0.65 . This coupling ratio is within the range of 2–4 reported from many other studies (Boas *et al* 2003, Hoge *et al* 1999a, 2005, Durduran *et al* 2004, Sheth *et al* 2004, Kastrup *et al* 2002, Kim *et al* 1999, Davis *et al* 1998).

Individual activation patterns from the simultaneously measured BOLD signal (p < 0.01, random-field correction), HbR, CBF, and CMRO₂ (p < 0.01, tube formula correction) are shown in the first column of figure 5. Group activation patterns (p < 0.05, uncorrected) are shown in the second column of figure 5. BOLD activations were observed in the premotor area, primary motor cortex and somatosensory cortex, which are the main target region of finger tapping task (Witt *et al* 2008). HbR activation was tightly localized on the primary motor cortex. Activated spots of CBF corresponded well with that of CMRO₂ and ROI of

finger tapping task. Note that CMRO₂ activation was more localized to the primary motor cortex, compared with CBF activation.

5. Discussion

5.1. Interpretation of the fractional CBF-CMRO₂ coupling ratio

The determination of the fractional coupling ratio between CBF and CMRO₂, n, is important for the quantitative interpretation of the activated BOLD signal. A different n value for the same changes in CMRO2 induces a large difference in the BOLD changes. The fractional coupling ratio between CBF and CMRO2 was originally investigated by Fox and Raichle (1986). Numerous studies have reported the values of n in the range of 2-4 (Boas et al 2003, Hoge et al 1999a, 2005, Durduran et al 2004, Sheth et al 2004, Kastrup et al 2002, Kim et al 1999, Davis et al 1998). The different fractional changes in CBF and CMRO₂ can be explained by the oxygen limitation model (Buxton and Frank 1997, Gjedde et al. 1991), which asserts that a large increase of CBF accompanied by a small decrease of the oxygen-extraction factor can produce a fractional increase of CMRO₂. Physiologically, the decrease of the oxygen extraction factor can be explained based on two assumptions: (1) the increase in CBF is accomplished by increased capillary blood velocity rather than by capillary recruitment (Bereczki et al 1993); (2) oxygen delivery to tissue is limited at rest (Gjedde et al 1991) so that only a fraction of the oxygen delivered to the capillary bed is extracted from the blood and becomes available for metabolism. More specifically, oxygen diffusion from capillary plasma to the mitochondria is managed by the gradient of partial pressure of oxygen (pO₂) between the capillary and the tissue. To increase CMRO₂, the average capillary pO₂ should increase and consequently oxygen extraction should decrease, as the fractional saturation of oxy-hemoglobin, which is a function of the plasma partial pressure of oxygen, has a sigmoidal shape (Buxton and Frank 1997, Buxton 2002). At a steady-state condition, relative CMRO₂ is proportional to the ratio of relative CBF and relative oxygen-extraction fraction rOEF(t) (Buxton 2002): rCMRO₂(t) = rOEF(t)/rCBF(t). Due to the decrease in the oxygen-extraction fraction induced by the CBF increase, CBF should increase more compared to $CMRO_2$.

5.2. The relationship between CBF and CBV

The changes in CBF are closely related to the changes in CBV. Specifically, CBF increases are generated by drops in the arteriole resistance, that lead to increases in the pressure of capillaries and veins. The increase in the pressure expands the vessels (Grubb *et al* 1974), describing the steady-state relationship between CBF and CBV as the empirically derived power law shown in equation (10). However, since Grubb's law is based on the steady-state measurements of whole brain, it has limitations for explaining the transient relationship between CBF and CBV in a specific brain region. Previous studies found that the power factor α changes with different brain regions (Wu *et al* 2002, Rapoport *et al* 1979) or different vascular compartments (Zheng and Mayhew 2009). Furthermore, the α value is typically lower under stimulation than during hypercapnia and ranges from 0.18–0.36 depending on the simulation duration (Mandeville *et al* 1999b, Jones *et al* 2001, 2002). During transient phases of vessel dilation and contraction, dynamic relationships between CBF and CBV are not simply expressed as the power law, which suggests additional terms for explaining the transient behaviors.

The difference between the transient-state characteristics of CBF and CBV comes from the visco-elastic properties of vessels. Viscoelasticity of venous vessels leads the vessel volume to increase rapidly upon the onset of stimulus and slowly decrease upon the cessation of stimulus. The works in Buxton *et al* (1998a, 1998b) considered these viscoelastic effects in the CBF as the rate of CBV changes. According to equation (13), the balloon initially resists the changes in CBV and finally arrives at a steady state. The time constant τ controls the time required to adjust to a transient state. Similarly, Mandeville *et al* (1999b), Zheng and Mayhew (2009), and Huppert *et al* (2009) described capillary and venous compliance as windkessel theory passively responding to arterial pressure changes.

Figure 4(c) shows the group average CBF response estimated using the Grubb *et al* (1974) model (GM) and the balloon model (BM) (Buxton *et al* 1998b). The balloon model dynamically explains the transient response of CBF and corresponding CMRO₂, compared with Grubb's model. As the value of α decreases, the changes in CBF increase. In the balloon model, as the τ value increases, CBF increases more rapidly upon the onset of stimulus and more promptly decreases after cessation of stimulus.

Note that the limitation of our current estimation of CMRO₂ is dependent upon the accuracy of the CBF model. Hence, for the accurate estimation of rCMRO₂ a direct flow measurement will be useful. However, even with the separate flow measurement, our rCMRO₂/rCBF coupling ratio is still valid since these quantities are obtained from fMRI–NIRS simultaneous recording.

5.3. Post-stimulus undershoot of CBF and CMRO₂

In our experiments, the undershoot of the CBF response during post-stimulus was consistently observed using Grubb's model or the balloon model, etc. (Grubb *et al* 1974, Jones *et al* 2002, Buxton *et al* 1998a, 1998b, 2004). Therefore, it is likely that the post-stimulus undershoot of CBF, as shown in figure 4(c), did not stem from an inaccurate choice of parameters of the Grubb model or the balloon model, but rather from the inherent characteristics of the data. A number of studies have observed CBF undershoot (Hoge *et al* 1999a, 1999b, Uludağ *et al* 2004, Huppert *et al* 2009, 2006, Chen and Pike 2009, Obata *et al* 2004, Sheth *et al* 2004, Durduran *et al* 2004, Davis *et al* 1998). However, in many other studies, CBF undershoot was not observed (Lu *et al* 2004, Shen *et al* 2008, Kruger *et al* 1998, Hoge *et al* 2005, Kida *et al* 2007). Although the reasons for this difference remain controversial, the type of stimulus or the experimental paradigm (Hoge *et al* 1999b) and an autoregulatory CBF feedback mechanism of neuronal origin (Friston *et al* 2000, Uludağ *et al* 2004) are the possible sources of these inconsistency.

In our setup, the CBF response is not directly measured by the NIRS system. Hence, the CBF undershoot is numerically determined by the CBV undershoot. Many optical studies estimate the CBF undershoot corresponding to the CBV undershoot (Sheth *et al* 2004, Huppert *et al* 2009, 2006, Durduran *et al* 2004). Assuming that the hematocrit is constant, the changes in CBV are related to the changes in HbT. However, considering the evidence of the hematocrit changes in the vascular compartment (Mchedlishvili 1986), an elaborate relationship between CBV and HbT would provide more reliable observations regarding the CBV response in the NIRS system.

Figure 4(c) describes the group average CMRO₂ responses corresponding to CBF estimates from Grubb's model. Post-stimulus undershoot of CMRO₂ was also observed. In our method, the changes in CMRO₂ are determined by multiplying the relative CMRO₂–CBF coupling ratio by the estimated CBF changes. We observed that the dynamic range of the relative coupling ratios were not as large as that of CBF. Since the CMRO₂ is calculated using equation (14), the CBF undershoot therefore results in CMRO₂ undershoot. Although real flow measurements using ASL may be necessary to confirm whether the CMRO₂ undershoot really

comes from the neurovascular coupling mechanism, it is still interesting to note that several previous studies exhibit the post-stimulus CMRO₂ undershoot from optical measurements (Sheth *et al* 2004, Durduran *et al* 2004, Jones *et al* 2001) and MR measurements (Davis *et al* 1998, Mandeville *et al* 1999a).

5.4. Post-stimulus undershoot of the BOLD response

In this study, the undershoot of the BOLD response during post-stimulus was observed, as shown in figure 4(a). BOLD undershoot is strongly influenced by the transient increase of the local deoxy-hemoglobin level. Currently, three theories to explain the post-stimulus undershoot of the BOLD response have been proposed. First, assuming that the CBF response rapidly returns to its baseline, the delayed recovery of CBV may accumulate the HbR content, which would then lead to BOLD undershoot (Buxton et al 1998b, Mandeville et al 1999b, Chen and Pike 2009). This observation has been embodied in biomechanical models such as the balloon model (Buxton et al 1998b) and the windkessel model (Mandeville et al 1999b). Second, the prolonged elevation of CMRO₂ has been postulated as the sole contributor of BOLD undershoot, in the presence of rapid CBF and CBV returns (Frahm et al 2008, 1996, Lu et al 2004, Kruger et al 1999). According to this theory, in order to replenish the ionic gradient that was decreased by action potential generation, CMRO2 is elevated upon the cessation of stimulus (Lu et al 2004). However, as recent research clearly shows that the magnitude and duration of BOLD undershoot do not depend on the intensity or duration of the stimulus (Chen and Pike 2009), it is uncertain that the elevated CMRO₂ is the sole contributor to BOLD undershoot. Third, CBF undershoot as well as the slow recovery of venous CBV has been postulated to contribute to the post-stimulus undershoot of the BOLD response (Chen and Pike 2009). Using a set of measures consisting of the amplitude, full width at half-maximum (FHWM), and time to fall of the post-stimulus undershoot, these studies have shown that the amplitude and duration of CBF undershoot were significantly correlated with those of BOLD undershoot during post-stimulus.

In the experimental results of this study (figure 4), the undershoot of CBV and the overshoot of the relative CMRO₂–CBF coupling ratio (rCMRO₂(t)/rCBF(t)) were consistently observed. According to Davis' model, in equation (6), the relative CBV and the CMRO₂–CBF coupling ratio are the two main factors that determine the BOLD undershoot. Although the CBV undershoot has a negative effect on the BOLD undershoot, larger overshoot of the relative CMRO₂–CBF coupling ratio results in the BOLD undershoot. In our data, the dynamic range of the relative coupling ratio was larger than that of CBV, suggesting that the relative coupling ratio was the main contributor of BOLD post-stimulus undershoot. This implies that the transient decoupling between the metabolism and the blood flow is the main source of BOLD undershoot.

6. Conclusion

A CBF and CMRO₂ estimation method was developed from simultaneous measurements of NIRS and BOLD. The proposed multimodal approach readily estimated CBF and CMRO₂ without any separate calibration under hypercapnia, hyperoxia, and breath-holding conditions. The relative CMRO₂–CBF coupling ratio was calculated by fitting biophysical models from NIRS and BOLD measurements. Then, CMRO₂ was estimated by multiplying the coupling ratio with CBF. Optimizing many unknown parameters within the physiological range greatly enhanced the accuracy of the proposed method. Experimental results from a finger tapping task showed that the estimated CBF and CMRO₂ time series were in good agreement with

physiological findings of other studies. The coupling ratio between CBF and CMRO₂ was also within the range of other studies. The CBF calculations from various models conducted in this study suggest that CBF exhibits post-stimulus undershoot whose magnitude variation is greater than that of CMRO₂, explaining the origin of the BOLD post-stimulus undershoot. Furthermore, the activated regions from CBF and CMRO₂ were localized on the primary motor cortex, which was the main target region of the finger tapping task.

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Appendix

We assume that the measurement at the *i*th channel is given by

$$y^{(i)}(t) = (\theta^{(i)} * s)(t) + n^{(i)}(t), \tag{A.1}$$

where $\theta^{(i)}$ denotes the hemodynamic response function (HRF), s is the input paradigm response, and $n^{(i)}$ is the noise contributor, respectively. In the proposed wavelet averaging, the HRF is represented as

$$\theta^{(i)}(t) = a\theta_J[0]\Phi(2^{-J}t) + \sum_{j=J_0}^J \sum_{k=0}^{2^{-j}M-1} d\theta_j[k]\psi(2^{-j}t - k), \tag{A.2}$$

where $\psi(t)$ is the wavelet and $\Phi(t)$ is the scaling function associated with a multiresolution analysis (Mallat 1999), J_0 denotes the finest scale that determines the smoothness of the trend, and the wavelet coefficients composed of approximation coefficients $\{a\theta_j[k]\}_{j,k}$ and detail coefficients $\{d\theta_j[k]\}_{j,k}$ are defined by the following recursions (Mallat 1999):

$$a\theta_0[k] = \theta[k], k = 0, ..., M - 1$$

$$a\theta_{j+1}[k] = \sum_n h[n - 2k]a\theta_j[n], k = 0, ..., 2^{-j-1}M - 1$$

$$d\theta_{j+1}[k] = \sum_n g[n - 2k]a\theta_j[n], k = 0, ..., 2^{-j-1}M - 1,$$

where j = 0, ..., J - 1, J is the maximum level of wavelet decomposition, $M \approx 2^J$, h is the low-pass filter, and g is the high-pass filter, respectively.

In equation (A.2), the paradigm related input s is known, and our goal is to estimate the hemodynamic response θ . In order to estimate the correct model order, our cost function is derived based on MDL principle for NIRS (Rissanen 1978, Jang *et al* 2009) and SIC principle for BOLD (Schwarz *et al* 1978, Meyer *et al* 2003):

$$c(n_0) = \frac{N}{2} \log \hat{\sigma}^2(n_0) + L(n_0), \tag{A.3}$$

where

$$L(n_0) = \begin{cases} \frac{3}{2}n_0 \log M : & \text{NIRS} \\ \frac{1}{2}n_0 \log M : & \text{BOLD,} \end{cases}$$

and

$$\hat{\sigma}^{2}(n_{0}) = \frac{1}{N} \left\| y^{(i)} - \left(a\theta_{J}[0](\Phi(2^{-J}t) * s) + \sum_{i=J_{0}}^{J} \sum_{k=0}^{2^{-j}M-1} d\theta_{j}[k](\psi(2^{-j}t - k) * s) \right) \right\|_{2}^{2}.$$

Here, n_0 denotes the number of non-zero coefficients in wavelet decomposition. The reason we use the different type of $L(n_0)$ for NIRS and BOLD is their different asymptotic behaviors depending on the length of time series (see details in Jang *et al* (2009)). In order to find the optimal parameter n_0 for equation (A.3), we perform the LASSO regression (Tibshirani *et al* 1996). More specifically, LASSO adds basis one by one along the homotopy path. In our approach, at each homotopy path, the cost in equation (A.3) is calculated, and the n_0 value that minimizes equation (A.3) is chosen as the optimal number of non-zero wavelet coefficients.

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