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Mapping breast cancer blood flow index, composition, and metabolism in a human subject using combined diffuse optical spectroscopic imaging and diffuse correlation spectroscopy

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Abstract. Diffuse optical spectroscopic imaging (DOSI) and diffuse correlation spectroscopy (DCS) are model-based near-infrared (NIR) methods that measure tissue optical properties (broadband absorption, μ_a , and reduced scattering, μ'_s) and blood flow (blood flow index, BFI), respectively. DOSI-derived μ_a values are used to determine composition by calculating the tissue concentration of oxy- and deoxyhemoglobin (HbO₂, HbR), water, and lipid. We developed and evaluated a combined, coregistered DOSI/DCS handheld probe for mapping and imaging these parameters. We show that uncertainties of 0.3 mm⁻¹ (37%) in μ'_s and 0.003 mm⁻¹ (33%) in μ_a lead to ~53% and 9% errors in BFI, respectively. DOSI/DCS imaging of a solid tissue-simulating flow phantom and a breast cancer patient reveals well-defined spatial distributions of BFI and composition that clearly delineates both the flow channel and the tumor. BFI reconstructed with DOSI-corrected μ_a and μ'_s values had a tumor/normal contrast of 2.7, 50% higher than the contrast using commonly assumed fixed optical properties. In conclusion, spatially coregistered imaging of DOSI and DCS enhances intrinsic tumor contrast and information content. This is particularly important for imaging diseased tissues where there are significant spatial variations in μ_a and μ'_s as well as potential uncoupling between flow and metabolism. © 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.JBO.22.4.045003](https://doi.org/10.1117/1.JBO.22.4.045003)]

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1 Introduction

Diffuse optical spectroscopic imaging (DOSI) is a functional imaging technology that measures hemodynamics and biochemical composition in centimeter-thick tissues.¹ We have developed a broadband handheld scanning probe for imaging breast tumors under clinically relevant conditions, such as detecting tumors in mammographically dense tissue,² distinguishing between malignant and benign tumors,³ monitoring tumor response⁴⁻⁶ to chemotherapy, and assessing breast density.⁷ In these studies, broadband NIR data (650 to 1000 nm) are used to calculate tissue absorption and scattering spectra and develop various metrics of endogenous contrast based primarily on tissue concentrations of oxy- and deoxyhemoglobin (HbO₂, HbR), water, and lipid. To enhance this information and provide a more complete picture of biochemical composition and oxygen consumption inside and in the vicinity of breast lesions, DOSI can be combined with diffuse correlation spectroscopy (DCS), a complementary method that measures deep tissue blood flow.

DCS utilizes the fluctuations of detected light intensity from the movement of particles to probe microvasculature blood flow

in deep tissue.⁸⁻¹⁰ The blood flow index (BFI) measured by DCS has been validated by arterial spin-labeled magnetic resonance imaging,¹¹ Doppler ultrasound,^{12,13} Xenon computed tomography,¹⁴ and laser Doppler flowmetry.^{15,16} Further, DCS systems have been used to monitor blood flow changes in clinical applications associated with brain,^{13,14,17} skeletal muscle,^{11,18,19} and tumors.^{10,20-23} When taken together, DCS and DOSI can provide a noninvasive measurement of the metabolic rate of oxygen consumption in tissue, and indeed, several studies have shown that the combination of DOSI and DCS techniques in breast lesion characterization provides a better understanding of tumor physiology and metabolism and may enhance treatment monitoring in the presurgical setting.^{20,23,24} However, these breast lesion studies were performed sequentially by using separate DCS and DOSI instruments. Serial measurements add additional time for data collection and can lead to uncertainties in the origins of the different signals.

This study aims to develop and validate a spatially coregistered, DOSI/DCS metabolic imaging system for enhancing breast tumor characterization. We combined DCS and DOSI techniques into a single handheld probe and developed integrated software that collects the data nearly simultaneously. Since the BFI depends on tissue optical properties, the BFI is typically calculated using fixed values from literature or using

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values measured with a separate instrument. DOSI is unique because it provides these optical properties (absorption, μ_a , and reduced scattering, μ'_s , parameters) over a broad spectral range from 650 to 1000 nm.²⁵ The impact of μ_a and μ'_s on BFI has been investigated in a previous study using a homogeneous liquid phantom where particles undergo Brownian motion.²⁶ We have extended this work to more realistic tissue simulating phantoms that employ a buried flow channel designed to simulate a subsurface flowing inhomogeneity. This allows us to analyze, for the first time, the combined impact of optical properties (μ_a , μ'_s), flow velocity, and channel depth on the estimation of the BFI.

Coregistered DOSI and DCS were then applied to construct two-dimensional (2-D) maps of subsurface particle flow in a solid tissue-simulating phantom with an embedded flow channel. Finally, images of tissue BFI, composition, and relative metabolism were obtained, for the first time, by coregistered DOSI/DCS mapping in a clinical measurement of a breast cancer patient. DOSI-derived optical properties at each measurement point were used to correct BFI in both clinical and tissue phantom images, improving accuracy and revealing unique and complementary contrast in breast cancer.

2 Material and Methods

2.1 DCS Theory

A typical DCS setup using optical fibers in tissue contact for light source and detector is shown in Fig. 1(a). The light source utilizes a long coherence length laser. The backscattered light intensity $I(t)$ is detected at a distance ρ from the source. The temporal speckle fluctuations of $I(t)$ is analyzed by calculating the normalized light intensity temporal autocorrelation function (g_2): $g_2(\tau) = \frac{\langle I(t)I(t+\tau) \rangle}{\langle I(t) \rangle^2}$, where τ represents the correlation delay time. Temporal speckle fluctuations cause variations in detected light intensity; thus, motions of red blood cells in microvascular blood vessels (arterioles, capillaries, and venules) provide the dominant contrast mechanism.^{8,11,27,28} Speckle fluctuations from a medium with higher flow rate produce a faster decay of g_2 . The Siegert equation²⁹ relates the normalized electric field autotemporal correlation function, $g_1(\tau)$, to $g_2(\tau)$

$$g_2(\tau) = 1 + \beta |g_1(\tau)|^2, \tag{1}$$

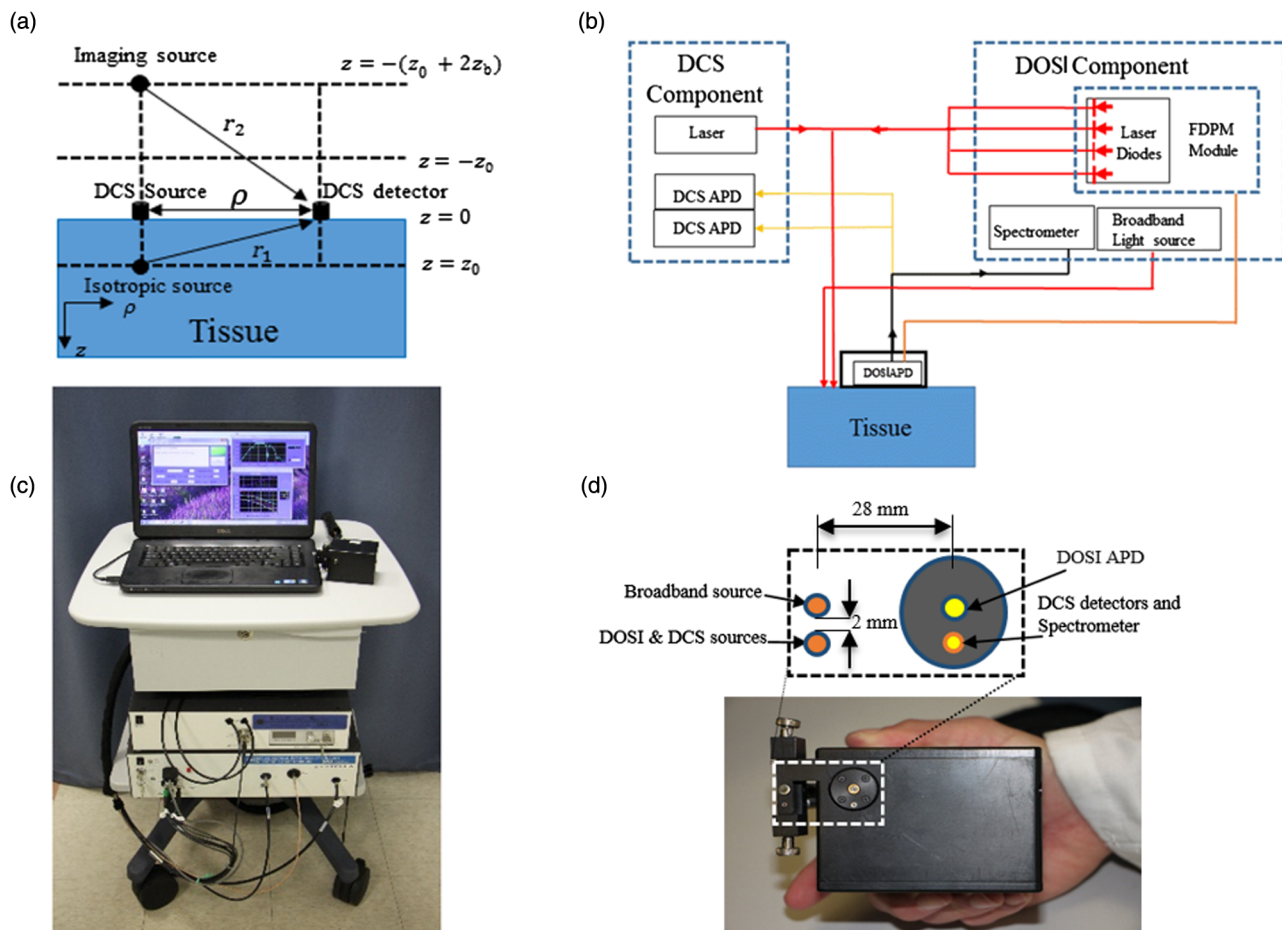


Fig. 1 (a) The semi-infinite measurement geometry for DCS including a positive isotropic light sources and a negative isotropic light (imaging) source for DCS theory; see Sec. 2.1 and Ref. 30 for details. (b) Block diagram of the DOSI/DCS instrument, red color represents the multimode source fibers. Collection fiber bundle is shown by black, which contains a DOSI spectrometer fiber and two DCS single mode fibers (shown with yellow), and orange is an RF cable. (c) Photograph of the entire system on a portable cart and (d) diagram and photograph of handheld probe.

where β depends on the optical setup and can be experimentally determined.³⁰

Similar to the photon diffusion equation, a correlation diffusion equation that describes the propagation of the unnormalized electric field temporal autocorrelation function $G_1(\tau)$ in a turbid media can be derived. For the setup in Fig. 1(a),^{8,9,15,30–33}

$$G_1(\rho, z, \tau) = \frac{3\mu_s'}{4\pi} \left[\frac{e^{-k(\tau)r_1}}{r_1} - \frac{e^{-k(\tau)r_2}}{r_2} \right], \quad (2)$$

where $r_1 = [\rho^2 + (z - z_0)^2]^{\frac{1}{2}}$, $r_2 = [\rho^2 + (z + z_0 + 2z_b)^2]^{\frac{1}{2}}$, $z_0 = \frac{1}{\mu_s'}$, $z_b = \frac{2(1+R_{\text{eff}})}{3\mu_s'(1-R_{\text{eff}})}$, and $R_{\text{eff}} = -1.440n^{-2} + 0.71n^{-1} + 0.668 + 0.0636n$ is the effective reflection coefficient, which is determined by the ratio of the refraction indices of two medium (e.g., $n = \frac{n_{\text{tissue}}}{n_{\text{air}}} \approx 1.33$). Equation (2) incorporates the extrapolated zero boundary condition, which approximates the isotropic light source located at $z = z_0$ and negative isotropic imaging source located at $z = -(z_0 + 2z_b)$. Schematic of the extrapolated boundary geometry is shown in Fig. 1(a).^{30,32–34} The effective decay rate $k(\tau)$ of $G_1(\rho, \tau)$ provides information about the motion of the scatterers in the underlying medium: $k^2(\tau) = 3\mu_s'\mu_a + \mu_s'^2 k_0^2 \alpha \langle \Delta r^2(\tau) \rangle$ where $\langle \Delta r^2(\tau) \rangle$ is the mean square displacement of the moving particles, k_0 is the wave-vector magnitude of the incident light field, and α is proportional to the fraction of scatterers that are moving (i.e., the fraction of tissue scatterers that are red blood cells). For diffusive motion $\langle \Delta r^2(\tau) \rangle = 6D_b\tau$, where D_b is the effective Brownian diffusion coefficient for the tissue scatterers.³⁰ The BFI (mm^2/s) is defined as $\text{BFI} = \alpha D_b$.^{15,35} We used the tissue optical properties measured with DOSI at each measurement iteration to calculate BFI. Assuming $\alpha = 1$, the factor $k(\tau)$ is fitted to the measured autocorrelation function, $g_2(\rho, \tau)$, via Eqs. (1) and (2) to derive the BFI.¹⁵

2.2 Instrumentation

A block diagram illustrating the combined DOSI and DCS system is shown in Fig. 1(b). The entire system and controlling PC fits on a cart for easy transport in a clinical setting [Fig. 1(c)].

2.2.1 DOSI hardware

DOSI estimates broadband near-infrared (NIR) (650 to 1000 nm) reduced scattering (μ_s') and absorption coefficients (μ_a) of turbid media through a combination of frequency domain photon migration (FDPM) and continuous wave (CW) NIR spectroscopy.²⁵ The principle of the measurement has been described,³⁶ the system utilized here is based on a custom 500-MHz frequency-domain module³⁷ and a high-speed 1-mm-diameter avalanche photodiode (APD) detector (#C5658 with S6045-03 APD, Hamamatsu Photonics, Shizuoka Pref., Japan). Four laser diodes (660, 690, 780, and 830) are sequentially swept from 50 to 500 MHz (each ~ 20 mW average power at the tissue) to acquire multifrequency phase and amplitude measurements of diffusely propagating photon density waves for each wavelength. The CW component of the DOSI system consists of a tungsten-halogen white light source (Micropak HL2000-HP, Ocean Optics, Florida) and a spectrometer (B&W Tek, Inc. Model 611, Delaware) to measure the backscattered reflectance spectra in the wavelength range 650 to 1000 nm.

2.2.2 DCS hardware

The DCS optical source is a long coherence-length (> 10 m) CW 785-nm laser (Crystalaser, Nevada). Laser output was adjusted to 40 mW, which was determined to ensure a safe intensity with reasonable detected photon count rate (more than 60 kcounts/s) for all the measurements. The DCS detectors consisted of two photon-counting modules (SPCM-AQRH-15-FC, Excelitas, Canada). The digital output of each detector was connected to a four-channel hardware correlator board (Flex01LQ-5, correlator.com, New Jersey) to measure the auto-correlation function (g_2) of the detected signal. Dark counts due to ambient light leakage were reduced by placing the detector units in an optical isolation box inside the system chassis.

2.2.3 Integrated DOSI/DCS handheld probe

DOSI and DCS sources and detectors were integrated into a single handheld probe with variable source-detector (s-d) separations (22, 28, or 34 mm) [Fig. 1(d)]. The s-d separations were set to 28 mm for all data reported here. Light sources for the DCS measurements and the FDPM component of DOSI were delivered with a five-fiber bundle of 200 μm step-index multimode fibers. The CW light source component of DOSI was delivered with a 3-mm-diameter bundle of 50- μm multimode fibers. The detector fibers include two single mode fibers (DCS detectors) and a 1-mm-diameter solid, multimode fiber (DOSI broadband detector/spectrometer) (LEONI Fiber Optics Inc., Virginia). The temperature-controlled FDPM APD detector was built into the probe for direct contact with the tissue. Custom software controls the serial collection of DOSI and DCS data and allows the instrument operator to easily enable either or both techniques. This feature allowed us to select the combined approach in the region of interest to minimize impact on the overall data collection time.

2.3 Liquid Phantom Experiment

The system was characterized using an liquid intralipid (IL)-based system with a submerged flow tube (3 mm ID, 0.8 mm wall), which allowed us to dynamically vary flow tube IL and background IL optical properties, flow rate, and depth of the flow tube. The combined (DOSI/DCS) handheld probe was wrapped with thin plastic film and secured at the surface of the liquid. FDPM calibration was done with the plastic wrap to factor out the effect of the plastic on the measurement.

A mixture of IL and nigrosin ink was used as a moving liquid through the flow tube for all the experiments. In each case, the predicted μ_s' and μ_a of the moving liquid at 785 nm were 20 and 0.06 mm^{-1} , respectively. These values are determined based on DOSI measurements. Three individual experiments were performed to study the effect of depth, absorption, and scattering on BFI: (1) to study the effect of depth and flow rates on BFI, the liquid surrounding the tube was formulated to simulate breast tissue ($\mu_a = 0.005 \text{ mm}^{-1}$ and $\mu_s' = 0.8 \text{ mm}^{-1}$) and distance between top of the flow tube, and probe surface (depth) varied from 0 to 20 mm in five steps (5 mm increment in each step). The flow rate varied between 0 and 400 mL/min in each depth. g_2 data samples were taken at each flow rate, and DCS integration time was 5 s for measuring g_2 . The relative change of blood flow, rBF, is the ratio of measured BFI for each flow rate and BFI for zero flow ($\text{rBF} = \frac{\text{BFI}_{\text{flow}}}{\text{BFI}_{\text{zero flow}}}$). (2) To study the effect of μ_a

on BFI, the absorption of the background liquid was changed in 10 steps from 0.0035 to 0.015 mm⁻¹ with constant μ'_s presented as mean (M) \pm standard deviation (SD) was 0.74 \pm 0.031 mm⁻¹. The flow tube was set at 10 mm depth and a constant 25 mL/min flow rate. (3) To study the effect of μ'_s on BFI, the scattering of the background liquid was changed in nine steps from 0.43 to 1.67 mm⁻¹ at a constant μ_a (0.0041 \pm 0.00027 mm⁻¹). Similar to the previous experiment, DOSI and DCS measurements were taken at each step with a constant 25 mL/min pump flow rate.

2.4 Silicone Phantom with Flow Channel

To demonstrate the capability of the probe for providing coregistered DOSI and DCS imaging, a solid phantom was made with a 4.6-mm-diameter flow channel, the top of which was located 5 mm below the surface. The phantom was created with 2.7 mL nigrosin (0.015%), 1.05 mg titanium dioxide (0.88 g/L), and 1.2 L polydimethylsiloxane, according to Ref. 11. The previously described moving IL mixture was used inside the flow channel.

2.5 Human Subject Measurement

The integrated system was used to characterize a triple negative (ER-, PR-, Her2-), grade 3 infiltrating ductal carcinoma of a 28-year-old premenopausal patient. The patient was measured 3 days following her first neoadjuvant chemotherapy infusion. The 20 \times 30 mm tumor was located 16 mm below the surface, as measured by ultrasound. The subject provided written informed consent, and the study protocol was approved by the University of California Institutional Review Board (UCI HS# 1995-563).

Tissue chromophore concentrations HbR, HbO₂, total hemoglobin (THb), lipid, water, and the tissue optical index (TOI = $\frac{[HbR] \times [water]}{[lipid]}$) were calculated from the broadband absorption spectrum. The TOI is a composite index that reveals contrast in more metabolically active regions (i.e., tumor) and was developed to identify breast tumor location from DOSI data.^{7,36} In addition, the relative tumor-to-normal oxygen consumption in

mammary tissue (rMMRO_{2(T/N)}) was calculated by combining DOSI and DCS data using²³

$$rMMRO_{2(\frac{T}{N})} = \frac{\gamma_T}{\gamma_N} \cdot \left(\frac{HbR_T}{HbR_N}\right) \cdot \left(\frac{THb_T}{THb_N}\right)^{-1} \cdot \left(\frac{BFI_T}{BFI_N}\right), \quad (3)$$

where *T* and *N* represent values from tumor and normal breast tissue locations, respectively. In this model, $\gamma_{T(N)}$ for either tumor or normal is the ratio of HbR to THb concentration in venous vasculature compare to the ratio of HbR to THb in the total vasculature. The ratio of $\frac{\gamma_T}{\gamma_N}$ was assumed to be 1 for simplicity. Further details can be found in Ref. 23. Normal tissue values of each parameter (HbR, THb, and BFI) were calculated using an average of three points far from the tumor region.

2.6 Image Analysis

The surface of the solid phantom was measured over a 90 mm \times 90 mm field of view in 5 mm increments, centered on the flow tube. The resulting 19 \times 19 grid consisted of a total of 361 measurement points. Coredgistered DOSI/DCS scans were obtained at each point, and images were generated using cubic interpolation between the measured points. The *in vivo* study was performed using a 90 mm \times 80 mm field of view in 10 mm increments that covered the tumor and surrounding normal breast tissue. The resulting 9 \times 8 grid consisted of 72 DOSI measurement points. To minimize overall measurement time, combined DOSI/DCS data were acquired over a smaller 30 \times 80 mm field of view using a 3 \times 8 grid of 24 measurement points in 10 mm increments centered over the tumor. Similar to the silicon phantom study, functional maps of tissue chromophore concentrations [HbR, HbO₂, total hemoglobin (THb), lipid, water, TOI] and BFI were reconstructed from each grid point location. Map points were interpolated using the 2-D cubic interpolation function (interp2) in MATLAB[®] to round out the discrete shapes.

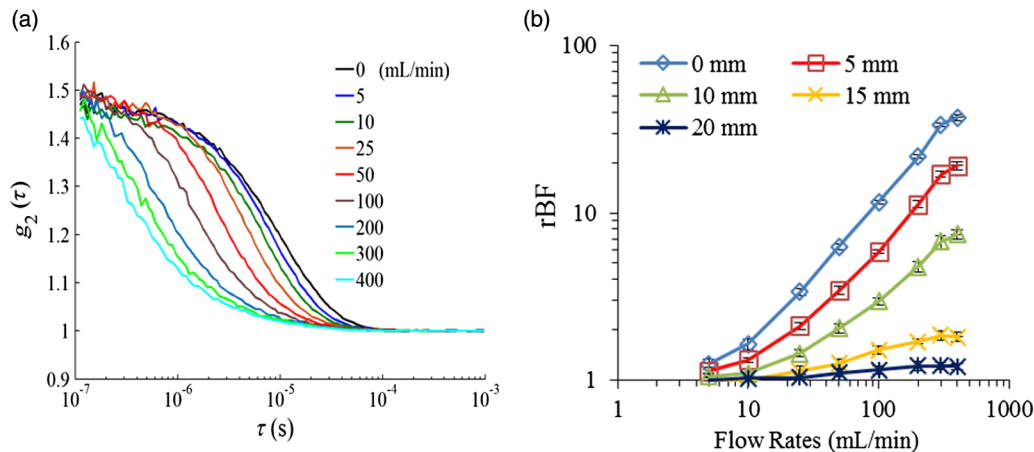


Fig. 2 DCS characterization. (a) Measured intensity autocorrelation functions g_2 , during liquid phantom depth/flow experiment for flow tube 5 mm below the surface. (b) The relative change of blood flow, rBF, versus pump flow rates, for various flow tube depths. Depth is the distance between the top of the flow tube and the handheld probe. Data are the average of 10 samples, and error bars indicate standard deviation. The DOS-measured optical properties of the liquid phantom were $\mu'_s = 0.75$ mm⁻¹ and $\mu_a = 0.005$ mm⁻¹ at 785 nm.

3 Results

3.1 Liquid Phantom Characterization

The measured intensity autocorrelation functions, g_2 , for different flow rates are shown in Fig. 2(a). The g_2 exhibits faster decay in the presence of higher flow rate.

Figure 2(b) shows the effect of flow tube depth on rBF for different flow rates. BFI was calculated using the corresponding μ_a and μ_s' values (0.0048 mm^{-1} and 0.78 mm^{-1} , respectively) from DOSI. Compared to a superficial flow tube, the BFI decreased to 52%, 24%, 10%, and 8% (mean ratio over all flow rates) when the depth of the tube was 5, 10, 15, and 20 mm, respectively.

To illustrate the effect of absorption error on the BFI, two groups of BFI were calculated while varying the background IL mixture μ_a with a constant flow rate. The first group was calculated by assuming constant $\mu_a = 0.0093 \text{ mm}^{-1}$ (median value), and the second group was calculated by using the extracted μ_a value from DOSI at 785 nm in each step. The average BFI for each group is shown in Fig. 3(a). BFI measured with constant μ_a value dropped slightly for a constant flow rate during the absorption titration. However, BFI recovered using DOSI-derived values remained nearly constant (with less than 5% variation). We observed that a 100% increase in μ_a (from 0.0076 to 0.015 mm^{-1}) induced almost 20% range of error on reconstructed BFI. The DOSI-measured μ_s' ($M \pm SD$) over the absorption titration was $0.74 \pm 0.031 \text{ mm}^{-1}$. Increased variation in BFI was noted at higher absorption cases due to the decreased DCS count rate and signal-to-noise ratio.

Next, the background IL concentration and thus μ_s' were varied with a constant flow tube depth and rate, with a constant absorption ($M \pm SD$) of $0.0041 \pm 0.00027 \text{ mm}^{-1}$. Similar to the absorption test, two groups of BFIs were measured either by assuming μ_s' was constant (0.8 mm^{-1}) or by using the DOSI-measured μ_s' . Illustrated in Fig. 3(b), a 60% increase of μ_s' (from 0.59 to 0.94 mm^{-1}) caused a 53% range of error in BFI—significantly more than the same proportional change in absorption. When using DOSI-correction, BFI was more consistent (with less than 12% variation). In general, we observed that

an uncertainty of 0.3 mm^{-1} in μ_s' leads to $\sim 53\%$ error in BFI, while an uncertainty of 0.003 mm^{-1} in μ_a yields only 9% error.

3.2 Silicone Phantom

A phantom with an embedded flow channel was made to demonstrate imaging with the combined probe. The measured optical properties of the phantom (outside of the flow channel) were $\mu_s' = 0.8 \text{ mm}^{-1}$ and $\mu_a = 0.008 \text{ mm}^{-1}$ at 785 nm. DOSI and DCS images calculated from two scans (no flow and 200 mL/min flow) are shown in Fig. 4. The values of μ_s' and μ_a ($M \pm SD$) for the three vertical lines (total of 19 before interpolation) centered in the tube were $0.88 \pm 0.0098 \text{ mm}^{-1}$ and $0.010 \pm 0.00097 \text{ mm}^{-1}$, respectively. BFI values ($M \pm SD$) in this area were $4.42 \times 10^{-6} \pm 2.52 \times 10^{-6} \text{ mm}^2/\text{s}$ for flow 200 mL/min and $3.22 \times 10^{-6} \pm 3.19 \times 10^{-7} \text{ mm}^2/\text{s}$ for flow = 0 mL/min.

3.3 Human Breast Tumor Imaging

DOSI and DCS images were acquired on a human breast tumor *in vivo*. The DOSI-measured optical properties over the tumor and surrounding normal tissue are shown in Figs. 5(a) and 5(b). DCS data were collected in a smaller region over the tumor, and images of BFI were calculated using DOSI measured optical properties [Fig. 5(c)] and using constant normal premenopausal breast optical property values ($\mu_a = 0.005 \text{ mm}^{-1}$ and $\mu_s' = 0.8 \text{ mm}^{-1}$) used in a prior DCS study²⁰ [Fig. 5(d)]. Peak contrasts in these maps [Figs. 5(c) and 5(d)] were measured by normalizing the max BFI on the tumor region to the BFI over a normal region (average of three BFI values far from the tumor). The BFI reconstructed with DOSI measured optical properties had a 50% higher peak contrast than BFI reconstructed using constant optical properties values (267% versus 217%). Figures 6(a)–6(e) show images of breast tissue chromophore concentrations (HbO₂, HbR, water, lipid, and TOI) calculated from DOSI-measured absorption. rMMRO_{2(T/N)} calculated from the DOSI-measured HbO₂ and HbR and the BFI is shown in Fig. 6(f).

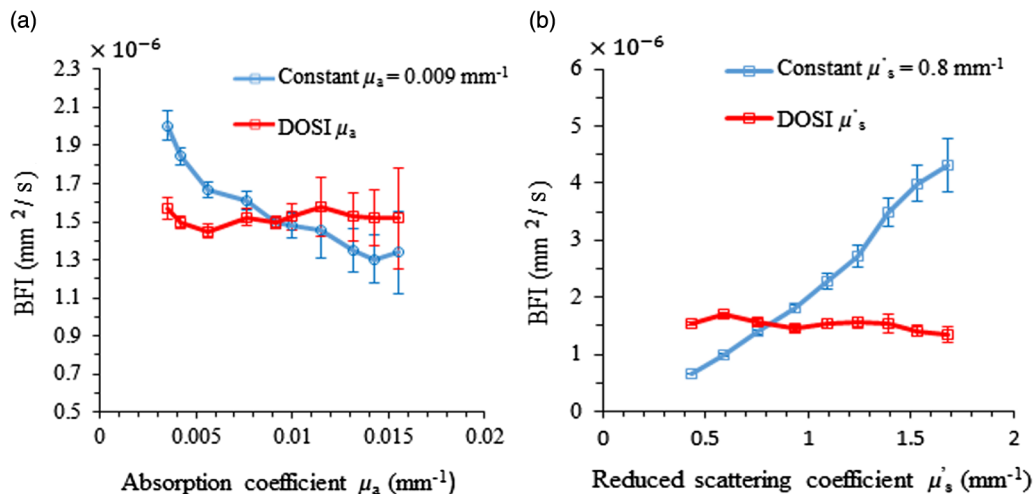


Fig. 3 Effect of absorption and scattering on BFI. BFI calculated with DOSI-corrected and constant optical properties are shown for (a) constant scattering and varied absorption and (b) constant absorption and varied scattering. The flow rate was constant during the measurements.

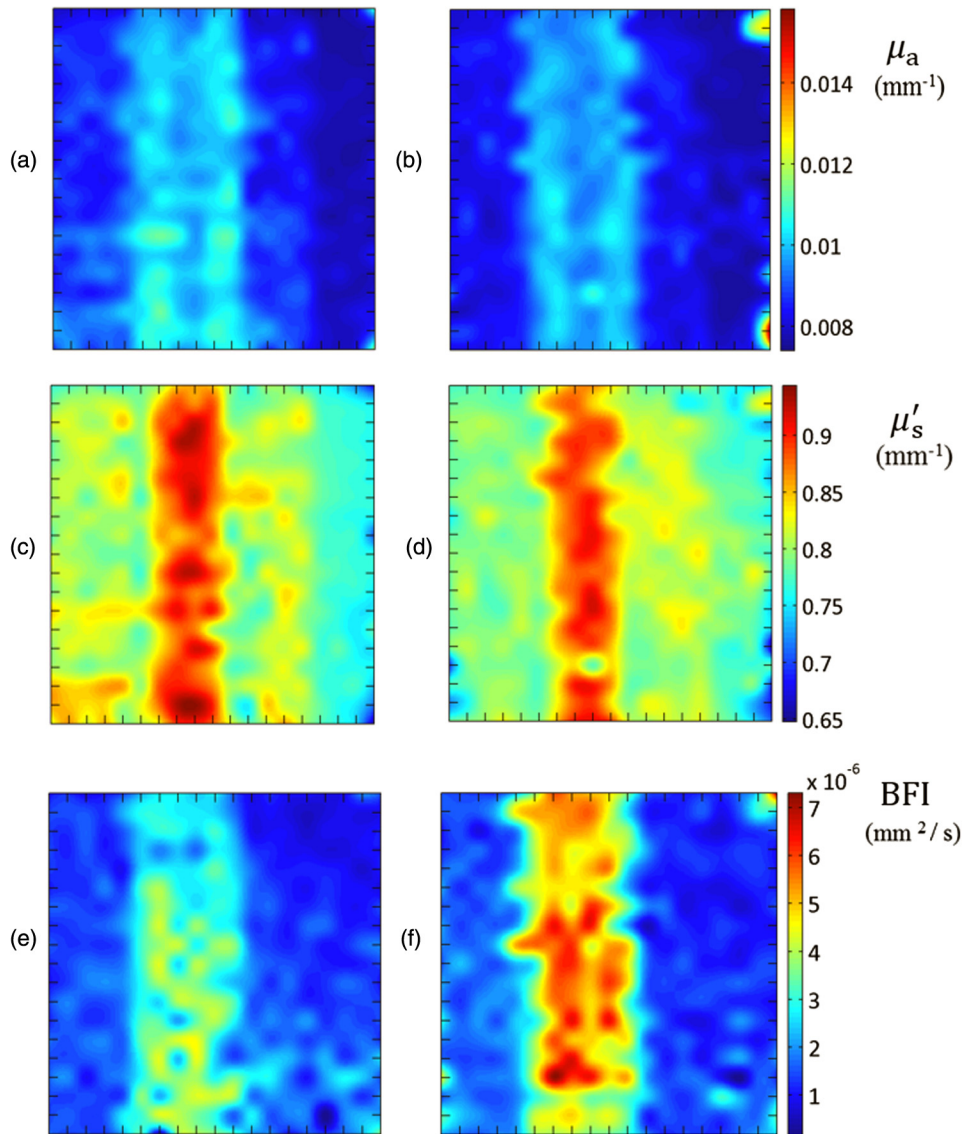


Fig. 4 DOSI and DCS images of silicone phantom with an embedded flow channel. (a, b) Absorption coefficient μ_a , (c, d) reduced scattering coefficient μ'_s , and (e, f) BFI. Images on left represent no flow (a, c, and e), where the images on the right represent, flow (b, d, and f). Tick mark separation equals 5 mm.

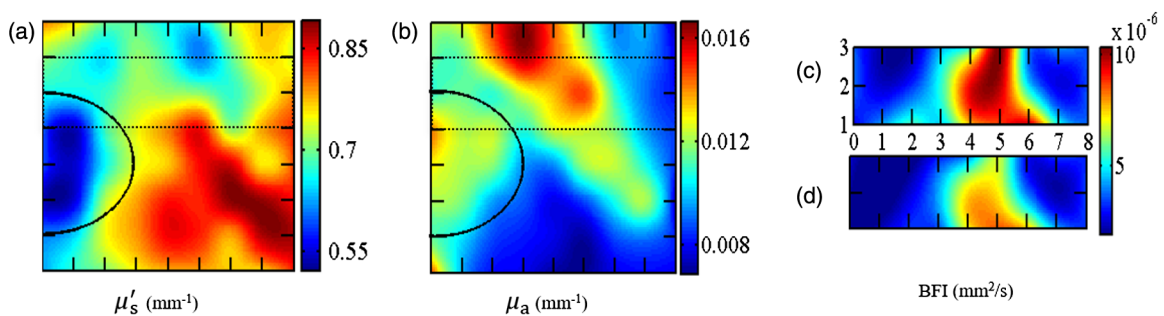


Fig. 5 Integrated DOSI and DCS imaging during the clinical measurement of a breast cancer patient (a) μ'_s and (b) μ_a mm^{-1} at 785 nm, (c) BFI (mm^2/s) reconstructed with μ'_s and μ_a values measured by DOSI at 785 nm in each, (d) BFI (mm^2/s) calculated with constant μ'_s and μ_a values (0.8 and 0.005 mm^{-1} , respectively). Dashed lines (a and b) represent the region where the lesion was located and DOSI/DCS data were taken. The areola region is outlined with a semicircle. Tick marks represent 10 mm separations.

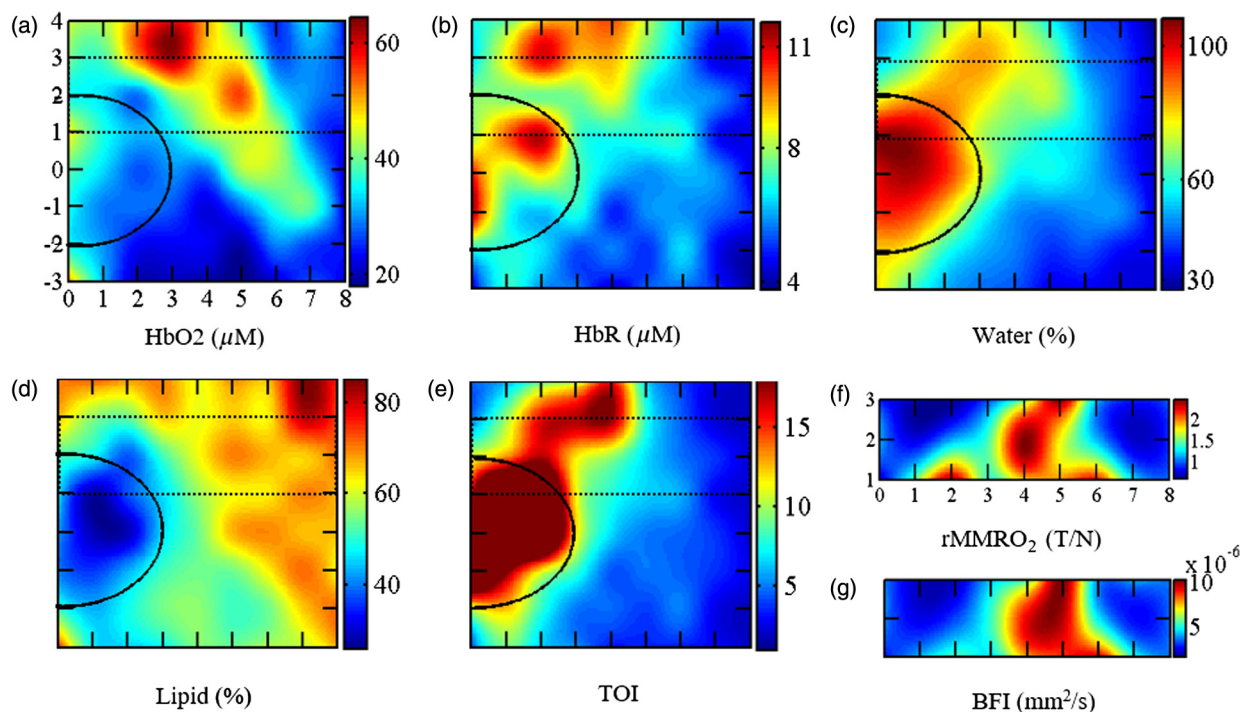


Fig. 6 Integrated DOSI and DCS imaging during the clinical measurement of a breast cancer patient. (a) HbO_2 (μM), (b) HbR (μM), (c) water (%), (d) lipid (%), (e) tissue optical index ($\text{TOI} = [\text{HbR}] \times [\text{water}] / [\text{lipid}]$), (f) rMMRO_2 (T/N), and (g) BFI (mm^2/s). Dashed lines (a–e) represent the region where the lesion was located and DOSI/DCS data were taken. The areola region is outlined with a semicircle. (f and g) The region that DOSI/DCS data were collected. Tick marks represent 10 mm separations.

For convenient comparison, the BFI from the same region is shown in Fig. 6(g). We observed that the spatial distributions of BFI and $\text{rMMRO}_2(T/N)$ around the tumor share similar features. However, the spatial distributions of TOI and $\text{rMMRO}_2(T/N)$ do not match, but the contrast from the areola is apparent in both images.

4 Discussion

In this study, we fabricated a combined DOSI/DCS metabolic imaging system designed specifically for breast tissue characterization, validated performance in several phantom studies, and performed breast cancer measurements in a human subject. A single integrated handheld probe was used to collect coregistered DOSI and DCS data from nearly identical tissue regions. Usability and data collection time were optimized by creating a custom software package that allowed the operator to enable either or both optical techniques.

The system was characterized and validated in several liquid and solid phantom experiments. First, a liquid phantom system with a submerged flow tube was developed to evaluate the impact of tube depth, flow speed, and background liquid optical properties on BFI. To simulate the DCS signals relevant to breast tissue, flow rates were selected to yield rBF (tumor/normal) values (from 1.90 to 2.70) at a tube located 15 mm below the surface based on previously measured flow changes of 32 patients.²⁰ Increasing the depth of the flow tube not only decreases the measured flow rate but also reduces flow rate sensitivity and dynamic range. We found that with 28 mm s-d separation, a linear flow signal was accessible up to 15 mm deep. However, while flow changes were accessible for a 20-mm deep flow tube, its linear dynamic range was greatly reduced. R^2

values derived from linear regression analysis of BFI at depths of 0, 5, 10, 15, and 20 mm were 0.98, 0.98, 0.98, 0.86, and 0.76, respectively.

Next, we separately varied the μ_a and μ'_s of the background liquid to demonstrate how bulk tissue optical properties can affect the DCS assessment of BFI. In previously published work,²⁶ a homogeneous liquid phantom was used to illustrate the effect of μ_a and μ'_s on BFI by comparing these values with Brownian motion of particles. In this work, we examined the impact of μ_a and μ'_s on BFI using a liquid phantom system with a submerged flow tube. Our experiments show that the BFI error is significantly higher for inaccuracies in μ'_s compared to μ_a , and it is independent of the flow rate. Our results are comparable to those from a previous study that showed an uncertainty of 0.3 mm^{-1} in reduced scattering leads to $\sim 50\%$ error in BFI, while an uncertainty of 0.003 mm^{-1} in absorption yields less than 10% error in BFI at 785 nm.²⁶ Although our absorption and scattering values were chosen to simulate breast tissue, they were varied over a large enough range to provide insight on DCS performance in other tissue types such as muscle and brain. Several phantom studies have investigated BFI error in cerebral blood flow measurements,^{38,39} and they have improved BFI accuracy through more complex modeling that accounts for irregular tissue geometries.^{40,41} It is possible that BFI accuracy in breast measurements can be further improved using models that account for arbitrary geometries⁴² and tissue heterogeneity.⁴³ These factors should be addressed in future studies.

A solid tissue-simulating silicone phantom with an embedded flow channel was fabricated for the first time to demonstrate the spatially coregistered information content of μ_a , μ'_s , and BFI. In the region over the flow tube, we observed the expected

increase in contrast in absorption, scattering, and BFI in both flow and no-flow conditions. BFI contrast in the no-flow case was likely due to the Brownian motion of the liquid scatterers. From these measurements, it is clear that the system is mainly sensitive to deep tissue lying within the field of view defined by the source–detector separation. Limited information is obtained from deep tissue more than 5 mm laterally outside the source–detector plane. Interestingly, the maximum absorption contrast was observed when either the source or detector was positioned directly over the phantom, while the maximum scattering contrast occurred at the midpoint of the source and detector, observations that were confirmed by Monte Carlo simulations (data not presented). Finally, even though the phantom was designed with a centered flow tube, we note that the flow channel appears shifted to the left in all the images. We verified that this is because the reference point used to position the probe was not located at the exact midpoint between the optical sources and detectors.

The combined DOSI/DCS system was used to characterize a high-grade breast tumor in a young patient. Two different images of BFI were calculated by using constant μ_a and μ'_s values and by extracting these values from the coregistered DOSI data. Although both BFI images showed increased blood flow in the tumor region (slightly different spatial variation), the BFI images reconstructed with DOSI μ_a and μ'_s exhibited higher BFI. This is due to overestimation of μ'_s (without DOSI) in some tissue regions leading to an underestimation of BFI. Overall, DOSI derived μ_a and μ'_s can vary up to 100% and 60% over different tissue regions, respectively, and the maximum error in the BFI reconstructed with constant μ_a and μ'_s values ranges from 20% to 53%. These findings underscore the importance of determining tissue optical and DCS properties from the same location. A prior study indicated that fixed optical properties did not significantly affect the relative change of BFI in tumor versus normal regions in a group of cancer patients. However, individual relative changes in BFI were clearly impacted when using DOSI-measured optical properties.²⁰ In addition, assuming constant optical properties can be particularly impactful when monitoring a subject during chemotherapy, since tumor optical properties are changing differently from the background.⁵

Our clinical data further underscore the importance of incorporating blood flow in characterizing breast tumors. Figure 6(f) shows the spatial distribution of oxygen metabolism in and around the tumor compared to the TOI, a high-contrast function used to localize tumors.²⁹ We observe colocalized contrast in the images due to the metabolically active tissue of the areola. However, both the peak locations and distributions of TOI and $rMMRO_{2(T/N)}$ around the tumor do not match. For example, one region of the tumor indicated by increased TOI contrast [$x, y = 2$ to 3 cm in Fig. 6(e)] corresponds to a relatively low BFI and $rMMRO_{2(T/N)}$. This is consistent with an increased vascular density but a lower blood flow that results in poor tissue perfusion in this region of the tumor. Consequently, there is a region adjacent to the peak TOI with a high BFI and $rMMRO_{2(T/N)}$, which may be due to the increased perfusion of peripheral tumor tissue. Overall, these data show how blood flow information augments and complements compositional data to provide a more complete picture of tumor perfusion and metabolism.

5 Conclusion

We describe a clinically compatible optical imaging system that simultaneously collects tissue composition and blood flow data

using DOSI and DCS, respectively. Specifically, the system measures tissue concentrations of oxy- and deoxyhemoglobin, water, and lipid, as well as blood flow. We validated instrument performance in both liquid and solid phantom systems and demonstrated coregistered imaging in phantoms. Measurements in a patient with breast cancer confirm the importance of accounting for spatial variations in tumor (T) and normal (N) tissue optical properties for optimizing BFI sensitivity and contrast. Finally, hemoglobin and flow parameters were combined to create images of the T/N relative metabolic rate of oxygen consumption [$rMMRO_2(T/N)$]. Together, quantitative DOSI and DCS imaging provides a more complete view of tumor composition and metabolism, which may aid in predicting, guiding, and personalizing cancer therapies.

Disclosures

B. J. Tromberg reports patents, which are owned by the University of California and licensed to commercial entities, that are related to the technology and analysis methods described in this study. This research was completed without their participation, knowledge, or financial support, and data were acquired and processed from patients by coauthors unaffiliated with any commercial entity.

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References

1. A. Yodh and B. Chance, "Spectroscopy and imaging with diffusing light," *Phys. Today* **48**(3), 34–40 (1995).
2. A. Leproux et al., "Assessing tumor contrast in radiographically dense breast tissue using diffuse optical spectroscopic imaging (DOSI)," *Breast Cancer Res.* **15**(5), R89 (2013).
3. S. Kukreti et al., "Characterization of metabolic differences between benign and malignant tumors: high-spectral-resolution diffuse optical spectroscopy," *Radiology* **254**(1), 277–284 (2010).
4. A. E. Cerussi et al., "Frequent optical imaging during breast cancer neoadjuvant chemotherapy reveals dynamic tumor physiology in an individual patient," *Acad. Radiol.* **17**(8), 1031–1039 (2010).
5. D. Roblyer et al., "Optical imaging of breast cancer oxyhemoglobin flare correlates with neoadjuvant chemotherapy response one day after starting treatment," *Proc. Natl. Acad. Sci. U. S. A.* **108**(35), 14626 (2011).
6. A. E. Cerussi et al., "Diffuse optical spectroscopic imaging correlates with final pathological response in breast cancer neoadjuvant chemotherapy," *Philos. Trans. R. Soc. London A Math. Phys. Eng. Sci.* **369**(1955), 4512–4530 (2011).
7. T. D. O'sullivan et al., "Optical imaging correlates with magnetic resonance imaging breast density and reveals composition changes during neoadjuvant chemotherapy," *Breast Cancer Res.* **15**(1), R14 (2013).
8. D. Boas, L. Campbell, and A. Yodh, "Scattering and imaging with diffusing temporal field correlations," *Phys. Rev. Lett.* **75**(9), 1855–1858 (1995).
9. D. Boas and A. Yodh, "Spatially varying dynamical properties of turbid media probed with diffusing temporal light correlation," *J. Opt. Soc. Am. A* **14**(1), 192–215 (1997).

10. T. Durduran et al., "Diffuse optical measurement of blood flow in breast tumors," *Opt. Lett.* **30**(21), 2915–2917 (2005).
11. G. Yu et al., "Validation of diffuse correlation spectroscopy for muscle blood flow with concurrent arterial spin labeled perfusion MRI," *Opt. Exp.* **15**(3), 1064–1075 (2007).
12. E. M. Buckley et al., "Cerebral hemodynamics in preterm infants during positional intervention measured with diffuse correlation spectroscopy and transcranial Doppler ultrasound," *Opt. Express* **17**(15), 12571 (2009).
13. N. Roche-Labarbe et al., "Noninvasive optical measures of CBV, StO₂, CBF index, and rCMRO₂ in human premature neonates' brains in the first six weeks of life," *Hum. Brain Mapp.* **31**(3), 341–352 (2010).
14. M. N. Kim et al., "Noninvasive measurement of cerebral blood flow and blood oxygenation using near-infrared and diffuse correlation spectroscopies in critically brain-injured adults," *Neurocrit. Care* **12**(2), 173–180 (2010).
15. T. Durduran, "Non-invasive measurements of tissue hemodynamics with hybrid diffuse optical methods," *Med. Phys.* **31**(7), 2178 (2004).
16. Y. Shang et al., "Diffuse optical monitoring of repeated cerebral ischemia in mice," *Opt. Express* **19**(21), 20301 (2011).
17. R. Cheng et al., "Noninvasive optical evaluation of spontaneous low frequency oscillations in cerebral hemodynamics," *Neuroimage* **62**(3), 1445–1454 (2012).
18. G. Yu et al., "Time-dependent blood flow and oxygenation in human skeletal muscles measured with noninvasive near-infrared diffuse optical spectroscopies," *J. Biomed. Opt.* **10**(2), 024027 (2005).
19. G. Yu et al., "Intraoperative evaluation of revascularization effect on ischemic muscle hemodynamics using near-infrared diffuse optical spectroscopies," *J. Biomed. Opt.* **16**(2), 027004 (2011).
20. R. Choe et al., "Optically measured microvascular blood flow contrast of malignant breast tumors," *PLoS One* **9**(6), e99683 (2014).
21. L. Dong et al., "Noninvasive diffuse optical monitoring of head and neck tumor blood flow and oxygenation during radiation delivery," *Biomed. Opt. Express* **3**(2), 259–272 (2012).
22. U. Sunar et al., "Noninvasive diffuse optical measurement of blood flow and blood oxygenation for monitoring radiation therapy in patients with head and neck tumors: a pilot study," *J. Biomed. Opt.* **11**(6), 064021 (2006).
23. C. Zhou et al., "Diffuse optical monitoring of blood flow and oxygenation in human breast cancer during early stages of neoadjuvant chemotherapy," *J. Biomed. Opt.* **12**(5), 051903 (2007).
24. Y. Shang et al., "Portable optical tissue flow oximeter based on diffuse correlation spectroscopy," *Opt. Lett.* **34**(22), 3556–3558 (2009).
25. D. Jakubowski et al., "Quantitative absorption and scattering spectra in thick tissues using broadband diffuse optical spectroscopy," in *Biomedical Optical Imaging*, J. Fujimoto and D. Farkas, Eds., pp. 330–355, Oxford University Press, New York (2009).
26. D. Irwin et al., "Influences of tissue absorption and scattering on diffuse correlation spectroscopy blood flow measurements," *Biomed. Opt. Express* **2**(7), 1969–1985 (2011).
27. G. Maret and P. Wolf, "Multiple light scattering from disordered media. The effect of Brownian motion of scatterers," *Zeitschrift für Physik B Condens. Matter* **65**(4), 409–413 (1987).
28. D. Pine et al., "Diffusing wave spectroscopy," *Phys. Rev. Lett.* **60**(12), 1134–1137 (1988).
29. S. Rice, *Selected Topics in Noise and Stochastic Processes*, Dover, New York (1954).
30. D. A. Boas et al., "Near-infrared diffuse correlation spectroscopy for assessment of tissue blood flow," *Handbook of Biomedical Optics*, pp. 195–216, CRC Press, Boca Raton (2011).
31. D. A. Boas, "Diffuse photon probes of structural and dynamical properties of turbid media: theory and biomedical applications," PhD Dissertation, University of Pennsylvania, Philadelphia (1996).
32. X. Li, "Fluorescence and diffusive wave diffraction tomographic probes in turbid media," PhD Dissertation, University of Pennsylvania, Philadelphia (1998).
33. C. Zhou, "In-vivo optical imaging and spectroscopy of cerebral hemodynamics," PhD Dissertation, University of Pennsylvania, Philadelphia (2007).
34. S. A. Walker, S. Fantini, and E. Gratton, "Effect of index of refraction mismatch on the recovery of optical properties of cylindrical inhomogeneities in an infinite turbid medium," *Proc. SPIE* **2979**, 219 (1997).
35. C. Cheung et al., "In vivo cerebrovascular measurement combining diffuse near-infrared absorption and correlation spectroscopies," *Phys. Med. Biol.* **46**(8), 2053–2065 (2001).
36. T. D. O'Sullivan et al., "Diffuse optical imaging using spatially and temporally modulated light," *J. Biomed. Opt.* **17**(7), 071311 (2012).
37. K.-S. No et al., "Design and testing of a miniature broadband frequency domain photon migration instrument," *J. Biomed. Opt.* **13**(5), 050509 (2008).
38. J. Selb et al., "Sensitivity of near-infrared spectroscopy and diffuse correlation spectroscopy to brain hemodynamics: simulations and experimental findings during hypercapnia," *Neurophotonics* **1**(1), 015005 (2014).
39. K. Verdecchia et al., Eds., "Characterization of a hybrid diffuse correlation spectroscopy and time-resolved near-infrared spectroscopy system for real-time monitoring of cerebral blood flow and oxygenation," *Proc. SPIE* **9313**, 931310 (2015).
40. L. Gagnon et al., "Investigation of diffuse correlation spectroscopy in multi-layered media including the human head," *Opt. Express* **16**(20), 15514 (2008).
41. K. Verdecchia et al., "Assessment of a multi-layered diffuse correlation spectroscopy method for monitoring cerebral blood flow in adults," *Biomed. Opt. Express* **7**(9), 3659–3674 (2016).
42. Y. Shang et al., "Extraction of diffuse correlation spectroscopy flow index by integration of N th-order linear model with Monte Carlo simulation," *Appl. Phys. Lett.* **104**(19), 193703 (2014).
43. Y. Shang and G. Yu, "AN th-order linear algorithm for extracting diffuse correlation spectroscopy blood flow indices in heterogeneous tissues," *Appl. Phys. Lett.* **105**(13), 133702 (2014).

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