Blood oxygen flux estimation with a combined photoacoustic and high-frequency ultrasound microscopy system: a phantom study

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Abstract. The metabolic rate of oxygen consumption, an important indicator of tissue metabolism, can be expressed as the change of net blood oxygen flux into and out of a tissue region per 100 g of tissue. In this work, we propose a photoacoustic and Doppler ultrasound method for imaging local blood oxygen flux of a single vessel. An imaging system for combined photoacoustic and high-frequency ultrasound microscopy is presented. This system uses a swept-scan 25-MHz ultrasound transducer with confocal dark-field laser illumination optics. A pulse-sequencer enables ultrasonic and laser pulses to be interlaced so that photoacoustic and Doppler ultrasound images are co-registered. Since the mean flow speed can be measured by color Doppler ultrasound, the vessel cross-sectional area can be measured by power Doppler or structural photoacoustic imaging, and multi-wavelength photoacoustic methods can be used to estimate oxygen saturation ($sO_2$) and total concentration of haemoglobin ($C_{Hb}$), all of the parameters necessary for oxygen flux estimation can be provided. The accuracy of the flow speed and $sO_2$ estimation has been investigated. In vivo sheep blood phantom experiments have been performed at different $sO_2$ levels and mean flow speeds. Blood oxygen flux has been estimated, and the uncertainty of the measurement has been quantified. © 2012 Society of Photo-Optical Instrumentation Engineers (SPIE). [DOI: 10.1117/1.JBO.17.3.036012]

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1 Introduction
The rate of oxygen consumption (MRO$_2$) is an important indicator of tissue metabolic activity. The ability to accurately image MRO$_2$ noninvasively will have an impact on research and clinical practice relating to cancer, diabetes, congestive heart failure and many other diseases relating to oxygen metabolism. MRO$_2$ is defined as the amount of oxygen consumed in a given tissue region per unit time per 100 g of tissue or the organ of interest. It can be estimated by quantifying the net oxygen flux of the blood vessels entering or exiting a defined tissue region. Blood oxygen saturation ($sO_2$), mean flow speed, and cross-sectional area of the blood vessel are the key parameters to estimate the local oxygen flux of a single vessel. Presently, techniques to obtain MRO$_2$ are limited. Near-infrared optical techniques have been developed to measure MRO$_2$ with the blood flow information provided by the diffuse correlation spectroscopy (DCS) and blood oxygenation information provided by diffuse reflectance spectroscopy (DRS). However, the spatial resolution is poor due to optical scattering. Positron emission tomography (PET) with $^{15}$O-oxygen has been used to measure oxygen consumption, but typical spatial resolution is ~1 cm for clinical systems, and ~1 mm for small animal systems. Magnetic resonance imaging (MRI) plays an important role in noninvasive measurement of MRO$_2$ by utilizing the blood oxygenation-level dependent (BOLD) effect. BOLD MRI can provide high spatial resolution, but it is only sensitive to deoxy-haemoglobin and has trouble distinguishing between changes in $sO_2$ and blood flow. Moreover, PET and MRI are not well-suited for many bedside clinical applications and can be cost-prohibitive.

Photoacoustic microscopy (PAM) is a novel technique that may provide a new way to image MRO$_2$. Photoacoustic imaging involves transmitting short laser pulses into biological tissues producing rapid heating in optically absorbing structures and detecting ultrasonic waves induced by thermo-elastic expansions. The strength of received PA signals is proportional to optical absorption, and the spatial resolution depends on ultrasonic detection. Maslov et al. developed the first dark-field confocal PAM system with a 50-MHz transducer, demonstrating a lateral resolution of ~50 μm. Recently, an alternative technique termed optical-resolution PAM (OR-PAM) utilizes the fine optical focusing to achieve high lateral resolution (~5 μm) with sacrifice of the imaging depth (~1 mm). The lateral resolution is determined by the size of a raster-scanned optical focal spot, and the axial resolution is still derived from time-resolved ultrasonic detection. Since oxy-haemoglobin (HbO$_2$) and deoxy-haemoglobin (HbR) are two of the major absorbing chromophores in human blood, and the optical absorption of blood is dependent on its oxygenation status, PAM is well suited to the imaging of vascular structures from macro vessels all the way down to individual capillaries and capable of functional imaging of $sO_2$ by multi-wavelength measurements. Zhang et al. have demonstrated in vivo imaging of microvasculature with high spatial resolution and $sO_2$ of single vessels in animals and total concentration of haemoglobin ($C_{Hb}$) in humans. Combined PAM and optical coherence tomography (OCT) method
has been proposed to estimate MRO$_2$ \emph{in vivo}, with sO$_2$ measured by OR-PAM and blood flow speed measured by spectral-domain OCT.

There are many potential methods to estimate the flow velocity and blood volume fraction. Recently, photoacoustic Doppler flowmetry has been proposed to measure transverse flow velocity based on Doppler bandwidth broadening, and the structural information of the flow can be provided as well. Yao et al. have developed mPAM (metabolic photoacoustic microscopy) system for simultaneously imaging sO$_2$ and blood flow speed to quantify MRO$_2$ \emph{in vivo}. However, the detectable flow speeds (<12 mm/s) are limited by the low laser pulse repetition rate, and the imaging depth is shallow (<1 mm). In our work, we propose to implement high-frequency Doppler ultrasound for flow estimation. High-frequency ultrasound is considered to include frequencies above ~20 MHz. The small beam size and large bandwidth provide high spatial resolution, which is important for imaging small animal models in preclinical research. Pulsed color Doppler ultrasound transmits ensembles of ultrasound pulses and estimates flow velocities from detected phase shifts in received echoes. By choosing appropriate pulse-repetition frequency (PRF) and Doppler ensemble length, color Doppler ultrasound is capable of imaging a wide range of flow velocities. However, like other flow meters based on the Doppler frequency shifts, the blood flow is hard to identify when the Doppler angle approaches 90 deg. Moreover, the noise signal with a low power but wide frequency spread can be displayed in color Doppler images as random velocities in random directions, which may appear as a mosaic of colors masking real vascular signals. In contrast with color Doppler, the technique of power Doppler ultrasound displays the back scatter strength of moving blood instead of flow velocities and directions. The intensity of power Doppler images is proportional to the number of scatterers moving within a vascular volume, rather than the frequency shifts they generate. The increased sensitivity to blood flow detection, and the angle-independent feature of power Doppler ultrasound make it particularly useful for imaging blood volume fractions even for tortuous vessels. For small vessels with low flow speeds, which are hard to detect by Doppler ultrasound, our previous work shows that photoacoustic imaging is more effective than power Doppler for the detection of blood.

The long-term goal of our research is to develop photoacoustic and ultrasonic imaging methods for noninvasive estimation of MRO$_2$ \emph{in vivo}. We recently developed a combined photoacoustic and high-frequency ultrasound microscopy system with a swept-scan 25-MHz transducer and confocal dark-field laser illumination optics. Previously we demonstrated the use of this system for interlaced photoacoustic and power Doppler imaging modes. In this article, along with our recent conference publication, we add color Doppler flow imaging to our system’s capabilities. Since the mean flow speed can be estimated using color Doppler ultrasound, vessel cross-sectional area can be quantified from power Doppler or photoacoustic images, sO$_2$ and C$_{\text{IOP}}$ can be estimated using multi-wavelength photoacoustic methods, our combined system is able to provide all of the parameters necessary for blood oxygen flux estimation. The accuracy of the flow velocity and sO$_2$ estimation has been investigated. \emph{In vitro} sheep blood phantom experiments have been performed at different sO$_2$ levels and mean flow speeds. We also experimentally demonstrate for the first time the feasibility of imaging local blood oxygen flux of a single vessel by combining photoacoustic oximetry and color Doppler ultrasound.

## 2 Methods

### 2.1 Experimental System

Our customized imaging system incorporates a computer-controlled scanning, triggering, and data-acquisition system and a combined light delivery and ultrasound probe based on confocal dark-field laser illumination optics. Readers are referred to our previous work for specific details about the instrumentations and explanations on light delivery probe and imaging procedures. In brief, a digital-input-output (DIO) card (NI PCI-6542, National Instruments, Inc., USA) is used to generate a pulse sequence that is sent to the ultrasound pulser/receiver (5073PR, Panametrics; Waltham, Massachusetts) and/or laser to enable different imaging modes. Currently our system can provide ultrasound (US) B-mode, photoacoustic (PA) mode, and Doppler US mode. In PA mode, a tuneable optical parametric oscillator (Surelite OPO Plus, Continuum; Santa Clara, California) is pumped by a Q-switch Nd:YAG laser (SurelItie III, Continuum; Santa Clara, California) with a repetition rate of 10 Hz. The tuning range is from 410 to 710 nm. PA and Doppler US modes are combined by using the DIO card to interleave laser and US triggers so that images can be co-registered. A 25-MHz single element transducer (V324-SM, 12.7-mm focal length, Panametrics; Waltham, Massachusetts) is positioned inside the light delivery probe and can be adjusted vertically to match its focus point with the laser focus. Radio frequency (RF) data received by the US pulser/receiver is digitized by an eight-channel PCI data-acquisition card (CS8289, Gage Cobra, Gage Applied Systems, Inc., USA) with 12-bit dynamic range and sampling rates as high as 125 M Samples/s.

### 2.2 Principles of Multi-Wavelength PAM for Oxygen Saturation Estimation

Since HbR and HbO$_2$ are treated as the two dominant absorbing chromophores in blood, the blood absorption coefficient $\mu_a(\lambda)$ at wavelength $\lambda$ can be expressed as:

$$\mu_a(\lambda) = \epsilon_{\text{HbR}}(\lambda)C_{\text{HbR}} + \epsilon_{\text{HbO}_2}(\lambda)C_{\text{HbO}_2},$$  \hspace{1cm} (1)$$

Here $\epsilon_{\text{HbR}}(\lambda)$ and $\epsilon_{\text{HbO}_2}(\lambda)$ are the known molar extinction coefficients (cm$^{-1}$ M$^{-1}$) of HbR and HbO$_2$ at wavelength $\lambda$, respectively. $C_{\text{HbR}}$ and $C_{\text{HbO}_2}$ are the concentrations of these two forms of hemoglobin. The PA signal amplitude emitted from a subsurface location $x$ is proportional to the local optical absorption coefficient $\mu_a(x, \lambda)$, local optical fluence $\Phi(x, \lambda)$, and the Grueneisen parameter $\Gamma(x)$:

$$\rho(x, \lambda) = \Gamma(x)\Phi(x, \lambda)\mu_a(x, \lambda).$$  \hspace{1cm} (2)$$

Often the Grueneisen parameter is assumed to be a constant $\Gamma$. By interrogating the tissue with at least two optical wavelengths, we solve the following matrix equation to obtain $C_{\text{HbR}}(x)$ and $C_{\text{HbO}_2}(x)$:
Often the local optical fluence is unknown to us. In this paper we take two approaches to solve this equation. The first approach involves assuming that the fluence is a wavelength-independent constant. The second approach uses a separate PA experiment and some highly absorbing black dye to estimate a factor proportional to the subsurface wavelength-independent fluence. In either case, a least-squares approach using the Moore-Penrose pseudo-inverse can be used to solve for $C_{\text{HB}}(x)$ and $C_{\text{HBO}_2}(x)$. The oxygen saturation is then calculated as:

$$s_{O_2}(x) = \frac{C_{\text{HBO}_2}(x)}{C_{\text{HB}}(x) + C_{\text{HBO}_2}(x)} \times 100\%.$$  (4)

### 2.3 Signal Processing Strategies of Doppler Ultrasound

Since tissues surrounding blood vessels often generate scattered signals much stronger than reflected signals from moving blood, clutter filtering is necessary before velocity estimation and power calculations. We use an eigen-filter, which is designed based on the statistics of the clutter, so that the response can be adapted to the tissue movement. The correlation matrix of the received RF data is decomposed into eigenvectors and eigenvalues by the discrete Karhunen-Loève transform (DKLT), and then eigenvalues are sorted in decreasing amplitude. The eigenvalue spectrum is a generalization of the Fourier power spectrum, which represents the power distribution of the frequency components. Since the power of clutter signals arising from stationary tissue is much higher than other components, we can partition the clutter subspace and eliminate clutter signals by choosing appropriate filter order. Appropriate filter order is determined experimentally by picking the filter order (typically two to three), which gives the best trade-off between clutter rejection and preservation of flow.

The combined system uses a swept-scan mode to continuously scan over the imaging target. This method can shorten the data-acquisition time, but the mechanical scan may cause changes of the amplitude and phase of the received signals, producing a gradual de-correlation. In this case, some clutter signals may pass through the filter, and the velocity estimator may interpret them as blood flow. Hence signal alignment is important before wall filtering. Within one line-of-sight, the peak magnitude of the cross-correlation between each adjacent pulse pair is found, and the alignment can be achieved by shifting each of the pulses by the determined amount.

The technique for velocity estimation is the auto-correlation technique introduced by Kasai et al. A sequence of A-mode pulses is obtained from transmissions in the same beam direction, and the flow velocity is calculated using the phase shift information. The Doppler ensemble length is determined by the scan speed and ultrasound PRF. The maximum and minimum detectable velocities are related to PRF and Doppler ensemble length. After signal alignment and wall filtering, the intensity of the signals from moving blood is accumulated within each Doppler ensemble to generate power Doppler images.

#### 2.4 Oxygen Flux and MRO$_2$ Estimation

The oxygen content in blood is the summed contribution of oxygen dissolved in plasma and chemically bound to haemoglobin (Hb) in red blood cells. The amount of oxygen carried in each form depends on the partial pressure of oxygen ($p_{O_2}$) to which each medium is exposed. Since the solubility of oxygen is limited in plasma, the specialized oxygen-binding mechanism of Hb is the major means of transporting oxygen in blood. As shown in Eq. (4), the gas-phase equivalent volume (in ml) of oxygen carried by Hb per ml of blood can be quantified as:

$$p_{O_2,\text{Hb}} = \zeta \times C_{\text{HB}} \times s_{O_2}.$$  (5)

Here $\zeta$ is the oxygen-binding capacity of Hb. From previous study when Hb is completely saturated with oxygen, $\zeta$ is 1.34 ml/g Hb, which means that each gram of Hb will bind with 1.34-ml oxygen. The unit of $C_{\text{HB}}$ is g/ml. Oxygen dissolved in plasma, despite its lower transport capacity, plays an important role for cells. In small capillaries, oxygen in plasma can diffuse through capillary walls and into cells for aerobic metabolism. According to Henry’s law, the gas-phase equivalent volume (in ml) of dissolved oxygen per 100 ml of blood can be quantified by the equation:

$$p_{O_2,\text{plasma}} = a_{O_2}(T) \times p_{O_2}.$$  (6)

Here $a_{O_2}$ with a unit of ml(oxygen)/100 ml(blood)/mmHg is the solubility coefficient of the oxygen in plasma dependent on temperature ($T$), and the unit of $p_{O_2}$ is mmHg.

The blood oxygen flux is defined as the volume of oxygen content transported per unit time. According to Eqs. (4) and (5), for a single vessel model, the oxygen flux can be expressed as:

$$\Phi_{O_2} = A \times \bar{v} \times \left[ \zeta \times C_{\text{HB}} \times s_{O_2} + a_{O_2}(T) \times \frac{p_{O_2}}{100} \right].$$  (7)

Here $\Phi_{O_2}$ is the oxygen flux (ml/min), $A$ is the cross-sectional area of the blood vessel, and $\bar{v}$ is the mean flow speed. If a region of interest is well defined, for $N$ feeding vessels and $M$ draining vessels, MRO$_2$ can be expressed as:

$$\text{MRO}_2 = \frac{\sum_i^N \Phi_{O_2}^\text{in}(i) - \sum_i^M \Phi_{O_2}^\text{out}(j)}{W},$$  (8)

where $W$ is the weight of the region of interest (in gram). The relative uncertainty of blood oxygen flux estimation can be determined by the relative uncertainties of $s_{O_2}$, $C_{\text{HB}}$, $\bar{v}$ and $A$. Simply, based on the assumption of error independence, for a single vessel model ($N = 1$):

$$\frac{\delta \Phi_{O_2}}{\Phi_{O_2}} = \sqrt{\left(\frac{\delta s_{O_2}}{s_{O_2}}\right)^2 + \left(\frac{\delta C_{\text{HB}}}{C_{\text{HB}}}\right)^2 + \left(\frac{\delta A}{A}\right)^2 + \left(\frac{\delta \bar{v}}{\bar{v}}\right)^2},$$  (9)

where $\delta$ denotes an error quantity. Uncertainty due to measurement variability can be quantified by measuring $s_{O_2}$, $C_{\text{HB}}$, $\bar{v}$ and $A$ repeatedly and estimating standard deviations.
2.5 Phantom Study

Experiments were performed on flow phantoms with the objective to show the feasibility of imaging blood oxygen flux using our combined PA and high-frequency US microscopy system. Transparent polymer tubing (Paradigm Optics; Vancouver, Washington) with inner diameter (ID) of 0.86 mm was used in construction of blood vessels. This ID is representative of small arteries and small veins. Six vessels were evenly positioned on an acrylic holder and embedded ~2-mm deep in a tissue-mimicking base. The tissue-mimicking base was composed of 10% cornstarch and 10% gelatin by mass, which provided an optical reduced scattering coefficient $\mu'_s = 9.2 \text{ cm}^{-1}$ at 532 nm, similar to human tissues. The ultrasonic and mechanical properties are also within the range of typical biological tissues.[6]

2.6 Double-Ink Studies

To determine the accuracy of flow velocity and $sO_2$ estimation, phantom studies were conducted using double-ink blood-mimicking fluid with both ultrasonic scattering and optical absorption. The double-ink technique provides a good first step toward oxygen flux estimation because all parameters regarding flow and mock-$sO_2$ can be easily controlled. The blood-mimicking fluid was prepared by dissolving 5% (by weight) cornstarch in water, which ensures the blood-to-tissue signal power consistent with in vivo signal levels. Red and blue ink samples (fiesta red and lake placid blue, Private Reserve Ink, Indiana) were diluted by the blood-mimicking fluid and mixed in various volumetric ratios to mimic different levels of $sO_2$. The fraction of the red ink concentration in the total ink concentration is termed mock-$sO_2$. Independent measurements of the optical absorption spectra of the ink samples were performed using a spectrometer (USB-4000 Miniature Fiber Optic Spectrometer, Ocean Optics Inc., USA). The double-ink blood-mimicking fluid was pumped through corresponding vessels at different flow rates using a calibrated syringe pump (NE-300 Syringe Pump, New Era Pump System Inc., USA) and imaged by our combined system.

2.7 Sheep Blood Studies

To demonstrate the capability of imaging blood oxygen flux, in vitro phantom experiments were performed using freshly collected arterial sheep blood with anticoagulant CPDA-1 (Citrate Phosphate Dextrose Adenine) solution. Figure 1 shows the experimental setup. The blood was loaded into a flask with a four-hole stopper. Pure oxygen and pure carbon dioxide flowed into the flask at different ratios of flow rates to obtain various $sO_2$ levels. The blood of certain $sO_2$ was withdrawn from one hole to a corresponding vessel of the flow phantom at different flow rates. Since the actual $sO_2$ is unknown, to validate $sO_2$ estimation, $pO_2$ was measured simultaneously when the blood flow was imaged by the combined system. An airtight acrylic mould was designed to hold the $pO_2$ electrode (DO_166MT-1 Micro Dissolved Oxygen Electrode, Lazar Research Laboratories, Inc., USA) as well as the temperature compensated pH electrode (HI 9126 pH/ORP meter, Hanna Instruments, USA) securely with the electrode tip completely immersed in blood.

The relationship between $sO_2$ and $pO_2$ is determined by the oxy-hemoglobin dissociation curve (OHDC), which demonstrates the nonlinear tendency for oxygen to bind to hemoglobin. Eq. (10) is the classic OHDC equation developed by Kelman:[10]

$$sO_2 = \frac{a_1 x + a_2 x^2 + a_3 x^3 + x^4}{a_4 + a_5 x + a_6 x^2 + a_7 x^3 + x^4}.$$  

Here, $x = pO_2 (\text{mmHg})$ in standard condition (temperature $T = 37^\circ C$, pH $= 7.4$, and carbon dioxide partial pressure $pCO_2 = 40 \text{ mmHg}$). Measurements conducted in other temperature or pH can be converted to standard condition by Eqs. (11) and (12):

$$x = f(T, pH, pCO_2) \times pO_2$$

$$f = 10^{[−0.024(T−37)+0.40(pH−7.4)−0.06(log_{10}pCO_2−log_{10}40)].}$$

Coefficients $a_1$ to $a_7$ in Eq. (10) are determined by fitting the equation in the least-squares sense to paired values of $sO_2$ and $pO_2$.
3 Results

3.1 Co-Registered Images Acquired from Double-Ink Flow Phantom

The flow phantom was immersed in water and imaged by cross-sectional scans at a speed of 0.5 mm/s, which provides 50 μm between each line-of-sight. The Doppler angle was 75 deg. Double-ink blood-mimicking fluid with four different mock-sO2 levels (20%, 40%, 60%, and 80%) was pumped through corresponding vessels at different flow rates. The mean flow speed is calculated from the volumetric flow rate divided by the cross-sectional area of the vessel. The mean flow speeds above 20 mm/s represent the blood flow in small arteries, and the speed range below 10 mm/s represents the flow in small veins down to large capillaries. Considering the slow scan speed, the Doppler ensemble length was chosen to be 10. Appropriate PRFs were set to ensure that the flow speed was within the detectable range. Optical wavelengths of 560 to 600 nm in steps of 10 nm were employed for sO2 estimation. A photodiode (DET36A/M, Thorlabs, USA) was used for laser pulse-energy normalization. Interlaced data were separated to form individual PA, B-mode, and Doppler US images. Then PA and Doppler images were co-registered and superimposed on the B-mode images with structural context. To better estimate the mock-sO2, M-mode measurements were repeated 100 times at the location with peak PA signal and the averaged amplitude of PA signals was used in Eq. (3). The mean flow speed and cross-sectional area of the vessel were estimated from ultrasound Doppler images.

Figure 3 shows representative co-registered power Doppler [Fig. 3(a)], color Doppler [Fig. 3(b)], PA [Fig. 3(c)], and mock-sO2 [Fig. 3(d)] images. The mock-sO2 image [Fig. 3(d)] was obtained by assigning the estimated sO2 value to a region obtained by thresholding the color Doppler image [Fig. 3(b)]. Appropriate regions of interest are selected...
so that the cross-sectional area of the vessel can be displayed in the center. The PA image [Fig.2(c)] was acquired at 560 nm and shows that only the top surface can be well visualized because of the minimal light penetration depth in the vessel. The power Doppler [Fig. 2(a)] and color Doppler [Fig. 2(b)] images both show good detection of the blood volume fraction. In practice, Doppler threshold levels will be important to accurately estimate the vessel area.

To demonstrate the performance of color flow estimation, different mean flow speeds were imaged while the mock-sO2 stayed invariant at 40%. Co-registered color Doppler and mock-sO2 images at different mean flow speeds are shown in Fig. 6.

When the mean flow speed stayed at 20 mm/s, the mock-sO2 level was varied and imaged to show the performance of sO2 estimation. Figure 6 shows the co-registered color Doppler and mock-sO2 images at different sO2 levels.

### 3.2 Measurement Accuracy

Measurement accuracy can be elucidated by validating the mean of experimentally estimated quantities against known values. Mean flow speeds were quantified from color Doppler images [Fig. 3] and plotted in Fig. 4. Linear regression shows a good linear relationship between the measured and preset values with a correlation coefficient of $R^2 = 0.99$, a slope of $1.01 \pm 0.03$ (with 95% confidence bounds), and a zero-offset of $-0.91 \pm 1.38$ mm/s.

We used black ink (Private Reserve Ink, Indiana) as a wavelength-independent optical absorber to calibrate for the influence of wavelength-dependent local optical fluence. This black ink was pumped through each vessel as a separate experiment, and multi-wavelength PA signals were acquired. Because light is completely absorbed within one acoustic resolution voxel at the surface of the black ink (regardless of wavelength), these PA signals were proportional to the local wavelength-dependent fluence and used to normalize the existing data for improved estimation of sO2. Table 1 shows the mean values and standard errors of estimated mock-sO2 before and after black ink correction. A student’s t-test was carried out to evaluate if the mean values of mock-sO2 are significantly different from these two groups. Since the p-values are all within 0.05, the mean values of mock-sO2 in these two groups are significantly different.

Figure 5 shows the relation between measured and preset mock-sO2 values. Linear regression indicates a good linear relationship with a correlation coefficient of $R^2 = 0.99$, a slope of 0.97 ± 0.08, and a zero-offset of 0.01 ± 0.04.

### 3.3 In Vitro Phantom Study of Oxygen Flux Estimation

As illustrated in Fig. 4, the fresh sheep blood with different sO2 levels flows through the corresponding vessels embedded in tissue-mimicking base at different flow rates, and imaged by the combined system. Mean flow speeds were quantified from color Doppler images and plotted against preset values (10, 20, 40, 60, and 80 mm/s) with linear least-square fitting in Fig. 7. The correlation coefficient is 0.95 with a slope of 0.94 ± 0.11 and a zero-offset of 2.98 ± 3.93 mm/s.

Four wavelengths (558, 564, 570, and 576 nm) are used to estimate sO2. Since HbR and HbO2 have the same optical absorption coefficient at isosbestic wavelength (570 nm), the PA signal acquired at 570 nm is proportional to $C_{1\text{Hb}}$ regardless of the sO2 level. In previous study, $C_{1\text{Hb}}$ of the arterial sheep blood was measured to be 127 g/L. We set this value as the $C_{1\text{Hb}}$.

### Table 1 Mean values and standard deviations of mock-sO2 of double-ink phantom with/without correction.

<table>
<thead>
<tr>
<th>Preset mock-sO2 (%)</th>
<th>20</th>
<th>40</th>
<th>60</th>
<th>80</th>
</tr>
</thead>
<tbody>
<tr>
<td>Measured mock-sO2 (%)</td>
<td>18.8 ± 0.8</td>
<td>37.0 ± 1.1</td>
<td>56.3 ± 1.1</td>
<td>72.3 ± 1.4</td>
</tr>
<tr>
<td>Corrected mock-sO2 (%)</td>
<td>72.3 ± 1.4</td>
<td>39.7 ± 1.2</td>
<td>59.8 ± 1.2</td>
<td>74.4 ± 1.3</td>
</tr>
<tr>
<td>Student's t-test p-value</td>
<td>0.03</td>
<td>0.02</td>
<td>0.02</td>
<td>0.03</td>
</tr>
</tbody>
</table>

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Fig. 5 Measured mean flow speeds against preset values (double-ink flow phantom).

Fig. 6 Measured mock-sO2 against preset values (double-ink flow phantom).
Both $pO_2$ and pH were measured simultaneously when the blood flow was imaged by the combined system. All the measurements were conducted under room temperature ($T$) of 23 °C. Assuming that the CO$_2$ partial pressure $pCO_2 = 40$ mmHg, the measured $pO_2$ values are calibrated to standard condition according to Eq. (11). To investigate the relationship between measured $pO_2$ and $sO_2$, we applied a least-square fitting of the $pO_2$ and $sO_2$ values to Eq. (10). As shown in Fig. 9(a), our experimental results were in agreement with the sigmoidal shape of Kelman’s OHDC with a correlation coefficient of 0.97. The fitted curve rises steeply with increasing $pO_2$ and reaches 90% $sO_2$ at $pO_2$ of $\sim 40$ mmHg. In addition, the corresponding $sO_2$ was calculated from measured $pO_2$ according to Eq. (10). Figure 9(b) shows the relation between the $sO_2$ measured by PA method and calculated from measured $pO_2$. Linear regression shows that the correlation coefficient is 0.92 with a slope of 1.11 and a zero-offset of $-0.07$. From measurements of all these parameters, blood oxygen flux of a single vessel was estimated by Eq. (6), and the uncertainty was quantified by Eq. (9). Results are shown in Fig. 10. The mean standard error is 0.1 ml/ min.

4 Discussion

We have demonstrated for the first time a photoacoustic and ultrasound method for imaging local blood oxygen flux in a single vessel. Our combined photoacoustic and high-frequency ultrasound microscopy system shows good performance on both flow velocity and $sO_2$ estimation at penetration depths $>2$ to 3 mm. For the task of cross-sectional area estimation, a circular region is selected so that all the color pixels in the power Doppler image can be fitted into the circle. For different vessel sizes and flow speeds, it may be possible to improve our estimation of fractional blood volume by careful selection of photoacoustic or power Doppler parameters. On the other hand, color Doppler velocity estimation may be improved by only computing Doppler estimates in locations where photoacoustic or power Doppler signals exist.

One long-term goal of our research is to use this system to estimate MRO$_2$ in vivo. As shown in Eq. (3), the subsurface optical fluence is often unknown in photoacoustic imaging so that quantitative estimation of oxygen saturation is challenging for animal models. In this work, we used a black-ink correction of blood when quantifying oxygen flux. To determine the relative error in $C_{Hb}$ for Eq. (9), the amplitude of PA signals acquired at 570 nm (sheep blood with varying $sO_2$) was shown in Fig. 8.
method to reduce the influence of wave-independence of the local optical fluence. In future studies, we would like to couple PAM with a technique termed oblique incidence reflectometry, which can provide a good first-order estimate of local optical fluence to improve our SO2 estimation. This proposed method will require validation. Even if accurate fluence correction is not possible, relative changes of oxygen flux into and out of a given region, and within a given range of imaging depths may prove informative if not absolutely quantitative.

Moreover, the challenge of imaging depth and reduced imaging speed may be limitations of the high-frequency swept-scan system. A clinical array system may be incorporated to allow images at multi-cm depths and higher frame rates. Work in progress includes implementing the photoacoustic and Doppler ultrasound method for MRO2 estimation on a clinical array system.

5 Conclusion

We have developed a combined photoacoustic and high-frequency ultrasound imaging system with a 25-MHz single-element transducer and confocal dark-field laser illumination optics. Flow phantoms are constructed to validate the color flow estimation and multi-wavelength SO2 estimation. In vitro sheep blood phantom experiments have been performed at different SO2 levels and mean flow speeds. We also experimentally demonstrate for the first time the feasibility of imaging local blood oxygen flux of a single vessel by combining photoacoustic oximetry and color Doppler ultrasound. Uncertainties at all experimental conditions are quantified to show a good performance of blood oxygen flux estimation. The mean standard error is 0.1 ml/min, which means that we can distinguish oxygen fluxes that differ by this amount.

References