Calibration of diffuse correlation spectroscopy with a time-resolved near-infrared technique to yield absolute cerebral blood flow measurements

Mamadou Diop, Kyle Verdecchia, Ting-Yim Lee, and Keith St Lawrence

1Imaging Program, Lawson Health Research Institute, London, Ontario N6A 4V2, Canada
2Department of Medical Biophysics, University of Western Ontario, London, Ontario N6A 3K7, Canada
3Imaging Research Laboratories, Robarts Research Institute, London, Ontario N6G 2V4, Canada
*mdiop@lawsonimaging.ca

Abstract: A primary focus of neurointensive care is the prevention of secondary brain injury, mainly caused by ischemia. A noninvasive bedside technique for continuous monitoring of cerebral blood flow (CBF) could improve patient management by detecting ischemia before brain injury occurs. A promising technique for this purpose is diffuse correlation spectroscopy (DCS) since it can continuously monitor relative perfusion changes in deep tissue. In this study, DCS was combined with a time-resolved near-infrared technique (TR-NIR) that can directly measure CBF using indocyanine green as a flow tracer. With this combination, the TR-NIR technique can be used to convert DCS data into absolute CBF measurements. The agreement between the two techniques was assessed by concurrent measurements of CBF changes in piglets. A strong correlation between CBF changes measured by TR-NIR and changes in the scaled diffusion coefficient measured by DCS was observed ($R^2 = 0.93$) with a slope of 1.05 ± 0.06 and an intercept of 6.4 ± 4.3% (mean ± standard error). © 2011 Optical Society of America

OCIS codes: (170.1470) Blood or tissue constituent monitoring; (170.3660) Light propagation in tissues; (170.3890) Medical optics instrumentation

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I. Introduction

Diffuse correlation spectroscopy (DCS) [1,2]—also known as diffuse wave spectroscopy [3–10]—is emerging as a reliable noninvasive technique for continuous monitoring of perfusion changes in deep tissue [11–31]. The continuous measurement capability of DCS could be beneficial in areas such as neurointensive care to assist with the management of patients recovering from brain-injury [32]. In these patients, multiple factors including systemic hypotension, increased intracranial pressure, impaired autoregulation and spontaneous vessel constriction can cause cerebral ischemia [33,34]. Since the brain is extremely vulnerable to ischemic injury due to its high energy demands and limited substrate storage, continuous monitoring of cerebral blood flow (CBF) could improve outcome by providing the ability to detect ischemia before permanent tissue damage occurs. However, the capability of DCS to determine if CBF has fallen to ischemic levels is hampered by the current inability to interpret DCS measurements in terms of absolute blood flow.

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Our group has developed a dynamic contrast-enhanced near-infrared (NIR) technique for measuring absolute CBF using a light-absorbing dye, indocyanine green (ICG), as an intravascular contrast agent [35]. The accuracy of the technique has been demonstrated in multiple studies by comparison to CBF measurements acquired with perfusion computed tomography [35–37]. Initially developed for second-derivative NIR spectroscopy, it has also been extended to time-resolved (TR) NIR technology [38,39], which provides better depth sensitivity [40]. Although quantitative, a limitation with this technique is the need to inject a bolus of ICG every time CBF is measured. Since the clearance of ICG (mainly by liver) from the blood circulation takes 10-20 min, CBF measurements can only be obtained infrequently.

Considering that DCS and TR-NIR technologies are compatible, our goal was to combine the continuous monitoring ability of DCS with the quantitative, single time-point measurement capability of the dynamic contrast-enhanced technique in order to provide a method for continuous monitoring of absolute CBF. With this approach, quantitative CBF measurements obtained by the TR-NIR technique are used to convert DCS data into units of absolute blood flow. The effectiveness of this approach will depend upon the agreement between perfusion changes measured by the two techniques. To address this issue, concurrent DCS and TR-NIR measurements were acquired in newborn piglets in which CBF was altered by changing the arterial carbon dioxide tension (\(p_{a}CO_2\)) and by occluding the carotid arteries. This animal model was chosen to minimize signal contamination from extracerebral tissues. The agreement between the two techniques was assessed by converting TR-NIR CBF measurements into relative blood flow changes, which were compared to corresponding perfusion changes measured by DCS.

2. Materials and methods

2.1. Instrumentation

A detailed description of the time-resolved instrument and its characterization can be found in Ref. [39]. In brief, the light source of the time-resolved instrument was a picosecond diode laser (LDH-P-C-810, PicoQuant, Germany) emitting at 802 nm, the peak absorption wavelength of ICG in plasma. The laser output power and the pulse repetition rate were set to 1.4 mW and 80 MHz, respectively. The laser beam was attenuated by two variable neutral density filters (NDC-50-4M, Thorlabs, Newton, NJ) and coupled into a 1.5-m long multimode fiber (core = 400 µm and N.A. = 0.22; Fiberoptics Technology, Pomfret, CT). Photons emerging from tissue were collected by a similar optical fiber whose distal end was secured in front of an electromechanical shutter (SM05, Thorlabs). Light transmitted through the shutter was collected by a Peltier-cooled photomultiplier tube (PMT) (PMC-100, Becker & Hickl, Germany). The shutter was kept closed to protect the PMT from exposure to intense light and was only open during TR measurements. The output of the PMT and the synchronization signal from the laser driver (PDL 828, PicoQuant) were sent to a time-correlated single photon counting module (SPC-134, Becker & Hickl) that computed the temporal point spread function (TPSF). The shape of the TPSF depends on the optical properties (i.e., the absorption coefficient, \(\mu_a\), and the reduced scattering coefficient, \(\mu_s'\)) of the medium probed by the detected photons [41,42].

The light source of the DCS instrument was a continuous-wave laser (DL785-100-S, CrystaLaser, Reno, NV) emitting at 786 nm. The coherence length of the laser (> 5 m) was much longer than typical photon pathlengths and the maximum output power was 100 mW. The laser beam was attenuated with a variable neutral density filter (NDC-50-4M, Thorlabs) and coupled into an emission fiber similar to the one used in the time-resolved apparatus. Light emerging from the tissue was collected with a 4-m long single-mode fiber (SMF-28e+, N.A. = 0.14, core = 125 µm, single-mode cutoff wavelength at 1260 nm). Since the emission wavelength of the laser was much shorter than the single-mode cutoff wavelength of the detection fiber, the latter was expected to guide more than one propagating mode. The detection fiber was wrapped around into a 15-cm diameter cylinder to increase the losses in the higher-order modes and to convert them into non-propagating modes. Light collected by
the detection fiber was detected by a single photon detector (id100-MMF 50, id Quantique, Geneva, Switzerland) for the first three animals and by a single photon counting module (SPCM-AQR-15-FC, PerkinElmer Canada Inc, QC, Canada) for the remaining of the study. The output of the detector was sent to a photon correlator board (DPC-230, Beker & Hickl, Germany) that computed the normalized intensity autocorrelation function. This function contained information about the dynamics of the scatterers (mainly red blood cells) in the region probed by the detected photons [12].

2.2. CBF measurement with the time-resolved NIR technique

2.2.1. Tracer-kinetics model

The hemodynamics of an organ can be quantified by tracking the passage of a contrast agent through the vascular bed – an approach that has been adapted to imaging modalities including magnetic resonance imaging (MRI) and computed tomography (CT). With this method, the time-dependent concentration of contrast agent in tissue, \( Q(t) \), is related to its concentration in arterial blood, \( C_a(t) \), by the following expression [43]:

\[
Q(t) = C_a(t) \ast F \cdot R(t),
\]

where ** is the convolution operator, \( F \) is blood flow and \( R(t) \) is the impulse residue function, which represents the fraction of contrast agent remaining in the tissue at time \( t \). In our dynamic contrast-enhanced experiments, \( Q(t) \) was measured by the TR-NIR instrument and \( C_a(t) \) obtained noninvasively using a dye densitometer (DDG 2001, Nihon Kohden, Japan). The function \( F \cdot R(t) \) was extracted from the tissue and arterial ICG concentration curves using a stable deconvolution algorithm [44]. The initial height of the function is CBF since by definition \( R(0) = 1 \), and the area under \( F \cdot R(t) \) equals the cerebral blood volume (CBV) [35].

2.2.2. Measuring the time-dependent concentration of ICG in brain

The time-dependent tissue concentration of ICG was measured by continuously acquiring a series of TPSFs with a temporal resolution of 0.4 s, 10 s prior to and for 60 s after an intravenous bolus injection of ICG (0.5 ml, containing 0.05 mg of ICG per kg of animal weight). Tissue optical properties (\( \mu_a \) and \( \mu'_s \)) were obtained by fitting the measured TPSFs with the solution to the diffusion equation for semi-infinite homogeneous media using extrapolated-zero boundary conditions [42]. To account for the finite width of the laser pulse and its broadening by the fibers and the detection system, the instrument response function (IRF) of the apparatus was measured and convolved with the analytical solution [45]. A nonlinear optimization routine (MATLAB® function fminsearch) was used to fit the model convolved with the IRF to each measured TPSF. The fitting range was set to 80% of the peak value on the leading edge and 20% on the falling edge of the TPSFs. Determining the tissue ICG concentration first required measuring the baseline optical properties, \( \mu_a \) and \( \mu'_s \), which were determined from TPSFs acquired over the 10 s period prior to ICG injection. The fitting procedure included an additional scaling factor to account for the laser power, detection gain and coupling efficiency. Changes in light absorption caused by ICG were characterized using only \( \mu_a \) as a fitting parameter, with \( \mu'_s \) and the scaling factor fixed to their baseline values. Finally, the measured changes in the absorption coefficient were used to calculate ICG concentration by:

\[
Q(t) = \left[ \left( \mu_a(t) - \mu_a(0) \right) \right] / [\ln(10) \times \varepsilon_{\text{ICG}}],
\]

where \( \mu_a(0) \) represents the average baseline absorption coefficient determined prior to ICG injection and \( \varepsilon_{\text{ICG}} \) is the extinction coefficient of ICG at 802 nm (186 OD/mM/cm [46]).

To compare with the DCS measurements, changes in CBF relative to normocapnia were determined by:
\[ \Delta CBF = 100 \times \frac{CBF_{\text{cond}} - CBF_{\text{normo}}}{CBF_{\text{normo}}}, \]  

where \( CBF_{\text{cond}} \) is the average cerebral blood flow measured at either hypocapnia, hypercapnia or during carotid occlusion, and \( CBF_{\text{normo}} \) is the average CBF measured at normocapnia.

### 2.3. Diffuse correlation flowmetry

The dynamics of the scatterers in the region probed by the DCS instrument can be assessed by measuring the temporal dependence of the normalized intensity autocorrelation function:

\[ g_2(\rho, \tau) = \frac{\langle (I(\rho, t)I(\rho, t+\tau)) \rangle}{\langle (I(\rho, t))^2 \rangle}, \]  

which is related to the electric field autocorrelation function \( G_1(\rho, \tau) \) by the Siegert relation [15]:

\[ g_2(\rho, \tau) = 1 + \beta \frac{|G_1(\rho, \tau)|^2}{\langle (I(\rho, t))^2 \rangle}. \]  

Here \( \beta \) is the coherence factor and \( \langle (I(\rho, t))^2 \rangle \) represents the average detected intensity.

#### 2.3.1. Diffusion approximation model

It has been shown that in highly scattering, low absorbing media, \( G_1(\rho, \tau) \) is solution to the diffusion equation for large interoptode distances (>3/\( \mu_s' \)) [2,47]. In this study, the DCS data were analyzed using the analytical solution to the diffusion equation for a semi-infinite homogeneous medium with extrapolated-zero boundary conditions [12]:

\[ G_1(\rho, \tau) = \frac{3 \mu_s'}{4 \pi} \left( \frac{\exp(-k_\rho r_1) - \exp(-k_\rho r_2)}{r_1} \right). \]  

Here, \( k_\rho^2 = 3 \mu_s' \mu_a' + 6 \mu_s' k_\alpha^2 \alpha D_B \tau \) (for Brownian dynamics), \( r_i = \sqrt{\rho^2 + (z_0 + 2 z_{\text{eff}})^2} \), \( \rho \) is the interoptode distance measured on the surface of the piglet's head, \( z_0 \) is the distance above the surface of head at which the fluence vanishes, and \( z_{\text{eff}} \) is the effective depth of the source [20]. Each acquired \( g_2(\rho, \tau) \) function was fit with the analytical model using \( \mu_s' \) and \( \mu_a' \) measured with the TR-NIR system and the known interoptode distance. This procedure yielded estimates of \( \beta \) and the scaled diffusion coefficient (\( \alpha D_B \)). The integration range was between 0.5 and 10 \( \mu \)s. The lower limit was chosen to reduce the influence of noise at early correlation times, while the upper limit was set to suppress contributions from photons with short pathlength, which have mainly probed extracerebral tissues, and to not overweight the tail of the autocorrelation curve.

Changes in the scaled diffusion coefficient were evaluated by:

\[ \Delta (\alpha D_B) = 100 \times \frac{aD_B_{\text{cond}} - aD_B_{\text{normo}}}{aD_B_{\text{normo}}}. \]  

Here \( aD_B_{\text{cond}} \) is the average value of \( aD_B \) measured during a condition of blood flow change (hypocapnia, hypercapnia, or carotid occlusion) compared to normocapnia, and \( aD_B_{\text{normo}} \) is the average scaled diffusion coefficient obtained at normocapnia.
2.3.2. Model-independent method

We also investigated the correlation between CBF changes measured with the TR-NIR method and changes in the area under the intensity autocorrelation curves. In this model-independent approach, similar to the one adopted by Dietsche et al. [3], the area defined as

\[ A = \int_{t_1}^{t_2} (g_2(\tau) - 1) d\tau, \]  

was calculated for each intensity autocorrelation curve \((t_1 = 0.5 \mu s \text{ and } t_2 = 10 \mu s)\), and changes evaluated as follows:

\[ \Delta A = 100 \times \frac{A_{\text{normo}} - A_{\text{cond}}}{A_{\text{normo}}}. \]

Here \(A_{\text{normo}}\) and \(A_{\text{cond}}\) are the average areas under the autocorrelation curves measured at normocapnia and during a condition of blood flow change, respectively.

2.4. Animal experiments

All animal experiments were conducted following the guidelines of the Canadian Council on Animal Care and approved by the Animal Use Subcommittee at the University of Western Ontario. Duroc-cross piglets were delivered from a local supplier on the morning of the experiment. Piglets were anaesthetized with 1.75 to 2.0% isoflurane (3.0% during surgery), tracheotomized and mechanically ventilated on an oxygen/medical-air mixture. Two incisions were made lateral to the trachea and vascular occluders (1.5 mm inner diameter; In Vivo Metric, Healdberg, CA) were placed around both carotid arteries proximal to the carotid bifurcation. The occluders were used to clamp the carotid arteries and create conditions of ischemia. An ear vein was catheterized for ICG injections (Sigma-Aldrich, St. Louis, MO). A femoral artery was catheterized to monitor heart rate and blood pressure and to intermittently collect blood samples for gas \((p_aCO_2, p_aO_2)\), pH and glucose analysis. If the blood glucose levels fell below 4.5 mmol/L, 1-2 ml infusion of a 25% glucose solution was administered intravenously. A heated water mattress was used to maintain rectal temperature between 37.5 and 38.5°C throughout the experiment.

After surgery, animals were placed in a prone position and a custom-made probe holder was strapped to the head to hold two sets (one for each instrument) of emission and detection probes, parasagittally, approximately 1.5 cm dorsal to the eyes. Once positioned in the holder, the probes were left in place throughout the entire experiment. The probe-holder was designed such that the emission and detection probes of each instrument were positioned at opposite ends (20 mm apart) of the diagonal of a rectangle, similar to the probe positions shown in Fig. 1b of Ref. [29]. This geometry ensured that the two instruments probed roughly the same brain region. Piglets were allowed to stabilize for 1 hour after surgery before any measurements were acquired.

Cerebral blood flow was increased by raising \(p_aCO_2\) and decreased by either occluding the carotid arteries or lowering \(p_aCO_2\). The \(p_aCO_2\) was changed by adjusting the breathing rate to create conditions of hypocapnia \((p_aCO_2 < 30 \text{ mmHg})\), normocapnia \((p_aCO_2 \approx 40 \text{ mmHg})\), moderate hypercapnia \((45 \text{ mmHg} < p_aCO_2 < 56 \text{ mmHg})\), and severe hypercapnia \((p_aCO_2 > 57 \text{ mmHg})\). Once the \(p_aCO_2\) had reached the target range, there was a 10-min delay before any measurements were acquired to ensure the physiological readings were stable. In addition, a final blood gas measurement was acquired before changing the condition of the animal to confirm the stability of the \(p_aCO_2\) during the measurements.

Sequential measurements with the DCS instrument, immediately followed by the TR-NIR technique, were made during the different blood flow conditions (Fig. 1). For each animal, three sets of DCS and TR-NIR data were acquired during normocapnia, two sets at severe hypercapnia, two sets at moderate hypercapnia, and a final two sets at either hypocapnia or...
during vascular occlusion. Each set of DCS and TR data included five intensity autocorrelation curves and one ICG blood flow measurement. Approximately 15 min were allowed between consecutive sets of measurements. All measurements were acquired while the animals were in steady-state as confirmed by the stability of the physiological parameters. Since the measurements were acquired sequentially, an electromechanical shutter placed in front of the PMT prevented any light from reaching the PMT during DCS measurements. The lasers were also sequentially blocked during the experiments. That is, the 786 nm CW laser was blocked during TR measurements and the 802 nm pulsed laser was blocked during DCS acquisitions. The initial count rate measured with the TR-NIR instrument was adjusted to 400 kHz (i.e., 0.5% of the laser repetition rate) to reduce dead-time effects [39]. The arterial ICG concentration curves were measured by attaching the dye densitometer probe to a forefoot. For the DCS measurements, the collection time was set to 30 s and the typical count rate was ~60 kHz and ~500 kHz for the first three and the last seven animals, respectively. The difference in count rate was a result of the difference in quantum efficiency between the two detectors (see section 2.1). Each experiment was completed in less than 7 hours.

2.5. Effects of errors in the optical properties on the recovered $a_D^B$

Simulations were conducted to investigate the effects of errors in the optical properties used in the fitting of the autocorrelation curves on the recovered $a_D^B$ values. Normalized intensity autocorrelation curves, $g_2(\rho, \tau)$, were generated using the diffusion approximation model (Eq. (5) and (6)) for three values of $a_D^B$: 25, 36 and 60 ($10^{-9} \text{cm}^2\text{s}^{-1}$). At each $a_D^B$, $g_2(\rho, \tau)$ data sets were generated by increasing $\mu_a$ in increments of 0.001 cm$^{-1}$ from 0.15 to 0.26 cm$^{-1}$ while keeping $\mu_s' = 8.0$ cm$^{-1}$. This procedure was repeated setting $\mu_a = 0.20$ cm$^{-1}$ and incrementing $\mu_s'$ from 7.0 to 10.0 cm$^{-1}$ in steps of 0.01 cm$^{-1}$. These values were chosen to mimic the range encountered in this study (see Tables 1 and 2). Each $g_2(\rho, \tau)$ curve generated in the aforementioned procedure was fitted using the analytical model and assuming fixed $\mu_a (0.2 \text{ cm}^{-1})$ and $\mu_s' (8.0 \text{ cm}^{-1})$, which are typical of optical properties measured at normocapnia (see Table 1). The percentage error in the recovered $a_D^B$ was evaluated as follows:

$$a_D^B_{\text{error}}(\%) = 100 \times \frac{a_D^B_{\text{fit}} - a_D^B_{\text{real}}}{a_D^B_{\text{real}}}.$$  \hspace{1cm} (10)

In this equation $a_D^B_{\text{fit}}$ is the value retrieved from the fitting procedure and $a_D^B_{\text{real}}$ is the input value (25, 36 or 60 $\times 10^{-9}\text{cm}^2\text{s}^{-1}$).

2.6. Statistical analysis

All data are presented as mean ± standard error (SE) unless otherwise noted. Statistical analyses were conducted using SPSS 16.0 (SPSS, Chicago, IL) and statistical significance was defined as $p < 0.05$. Physiological variables under the different capnic levels were compared using a two-tailed paired t-test including Bonferroni correction for multiple comparisons. Regression analyses were conducted to investigate correlations between CBF values from the TR-NIR technique and relative flow parameters from DCS.
3. Results

Measurements were acquired in ten newborn piglets (8 females and 2 males), less than 6 days old (average age = 1.5 day, range 0-5 days) with an average weight of 1.59 ± 0.40 kg (mean ± standard deviation, SD). Average optical properties and physiological parameters measured at hypocapnia, normocapnia, moderate and severe hypercapnia, and during carotid occlusion are summarized in Table 1.

Table 1. Optical properties and physiological parameters measured at hypocapnia, normocapnia, moderate and severe hypercapnia, and during carotid occlusion. Data are presented as mean ± SD.

| Condition              | \( \mu_a \) (cm\(^{-1}\)) | \( \mu_s' \) (cm\(^{-1}\)) | \( p_aCO_2 \) (mmHg) | pH        | HR (bpm) | MAP (mmHg) | \( p_aO_2 \) (mmHg) |
|------------------------|-----------------------------|-----------------------------|----------------------|-----------|-----------|-------------|---------------------|
| Hypocapnia (n = 3)     | 0.19 ± 0.03                 | 8.9 ± 1.5                   | 27 ± 4               | 7.59 ± 0.09 | 201 ± 53  | 43 ± 5      | 150 ± 45            |
| Normocapnia (n = 10)   | 0.20 ± 0.03                 | 7.9 ± 1.2                   | 39 ± 2               | 7.44 ± 0.08 | 167 ± 29  | 42 ± 5      | 121 ± 41            |
| Moderate Hypercapnia   | 0.23 ± 0.03*                | 8.6 ± 1.1                   | 52 ± 2*              | 7.36 ± 0.04* | 152 ± 29* | 39 ± 5      | 100 ± 32*           |
| (n = 7)                |                             |                             |                      |           |           |             |                     |
| Severe Hypercapnia     | 0.23 ± 0.04*                | 7.8 ± 1.0                   | 63 ± 7*              | 7.29 ± 0.03* | 176 ± 29  | 48 ± 5*     | 110 ± 28            |
| (n = 7)                |                             |                             |                      |           |           |             |                     |
| Carotid Occlusion      | 0.23 ± 0.09                 | 7.4 ± 2.1                   | 39 ± 2               | 7.46 ± 0.04 | 166 ± 7   | 38 ± 14     | 117 ± 48            |
| (n = 4)                |                             |                             |                      |           |           |             |                     |

In Table 1* indicates significant differences relative to the mean value at normocapnia. The statistical tests did not include data from the hypocapnic and vascular occlusion conditions due to the small sample size under these conditions. As expected, hypercapnia caused significant changes in the arterial carbon dioxide tension and arterial pH. Significant changes in heart rate (HR), mean arterial pressure (MAP) and arterial oxygen tension (paO2) were also found.

Figure 2 shows a typical intensity autocorrelation curve measured under normocapnia and the best fit to Eq. (5) and (6), which yielded estimates of \( \beta \) and \( aD_B \). The correlation factor obtained from the data shown in Fig. 2 (\( \beta = 0.158 \approx 1/6 \)) was in good agreement with the theoretical prediction based on the number of guided modes for the six-mode detection fiber used in this study [48]. This agreement shows that the higher waveguide modes (LP\(_{02}\) and LP\(_{21}\)) had been efficiently converted into non-propagating modes.

![Fig. 2. Typical intensity autocorrelation function measured at a source-detector distance of 2 cm on a piglet head and the best-fit curve obtained using the correlation diffusion model for a homogeneous semi-infinite medium assuming Brownian dynamics.](image)

Normalized intensity autocorrelation functions and the corresponding tissue ICG curves measured in one animal at hypocapnia, normocapnia and hypercapnia are displayed in Fig. 3. As expected when CBF increased, the autocorrelation function decayed faster and the derived scaled diffusion coefficient increased (i.e., \( aD_B^{\text{hyper}} > aD_B^{\text{normo}} > aD_B^{\text{hypo}} \)). A similar trend was observed in the corresponding TR-NIR measurements (Fig. 3B). As \( p_aCO_2 \) increased, the
maximum peak of the tissue tracer concentration curve increased, reflecting the changes in CBF between the different capnic levels. These qualitative observations are in good agreement with the measured values given in Table 2. Note, the coefficient of variation for the $\alpha_D$ values derived from a set of 5 temporal intensity autocorrelation curves was approximately 2%. This small value demonstrates that the signal-to-noise-ratio (SNR) of the DCS data was sufficient to obtain precise estimates of $\alpha_D$.

Relative CBF changes measured by the TR-NIR technique were compared to the corresponding changes in $\alpha_D$ (Fig. 4A). The apparent clustering of the data into three groups was due to the narrow $p_aCO_2$ range for the three capnic levels (see Table 1). This graph reveals a very strong linear correlation ($R^2 = 0.93$) between perfusion changes and changes in $\alpha_D$ with a slope of $1.05 \pm 0.06$ and an intercept of $6.4 \pm 4.3\%$. Statistical tests revealed that the slope was not significantly different from unity and the intercept was not significantly different from zero. The correlation between CBF changes measured by the TR-NIR

**Table 2. $\alpha_D$ and CBF values obtained from the experimental data presented in Fig. 3**

|               | Hypocapnia | Normocapnia | Hypercapnia |
|---------------|------------|-------------|-------------|
| $\alpha_D$ ($10^{-9}$ cm$^2$ s$^{-1}$) | 29.9       | 36.5        | 57.5        |
| CBF (ml/min/100g) | 35.2       | 45.6        | 65.5        |
Fig. 4. Perfusion changes measured with the TR-NIR technique compared to the corresponding changes in (A) the scaled diffusion coefficient ($aD_b$) and (B) area under the intensity autocorrelation curves ($\Delta A$).

technique and changes in the model-independent parameter $\Delta A$ extracted from the autocorrelation function is shown in Fig. 4B. The analysis revealed that changes in the area under the intensity autocorrelation curves were strongly correlated with blood flow changes ($R^2 = 0.86$), although the intercept was significantly different from zero ($p = 0.04$).

Figure 5 shows the relationship between absolute CBF measured with the TR-NIR technique and the corresponding $aD_b$ values. Since the apparatus used in this study was a hybrid TR-DCS instrument, the $aD_b$ data in Fig. 5A (a slope of $1.14 \pm 0.08$ (ml/min/100g)/(10$^{-9}$cm$^2$s$^{-1}$) and an intercept of $5.95 \pm 3.24$ ml/min/100g) were obtained using the measured $\mu_a$ (for each condition) and $\mu_s'$ (for each piglet) in the fitting of the autocorrelation curves. The benefit of using individual’s optical properties in the fitting of the DCS data was investigated by repeating the analysis using average values across all piglets (see Fig. 5B: a slope of $1.20 \pm 0.12$ (ml/min/100g)/(10$^{-9}$cm$^2$s$^{-1}$) and an intercept of $2.15 \pm 4.77$ ml/min/100g). In this case, the correlation between the two techniques was weaker, as evident by a reduction in $R^2$ (0.81 vs. 0.89), despite the relative homogeneity of the population. In both Figs. 5A and B, the intercept was not significantly different from zero.

Only data from the last seven animals were used in the comparison presented in Fig. 5 since the DCS measurements from these animals were acquired using a detector with high photon detection efficiency (~50%) to provide high SNRs. In contrast, the first three experiments were conducted using a detector with low photon detection efficiency (~7%). The lower SNR in these experiments caused a bias in the early time-points of the autocorrelation curves (i.e., the measured $g_2$ value was much greater than the predicted value of $1 + \beta$), which causes a subsequent bias in $aD_b$. Compared to the data in Fig. 5A, including data from the first three animals in the analysis resulted in a slope of $0.86 \pm 0.11$ (ml/min/100g)/(10$^{-9}$cm$^2$s$^{-1}$), an intercept of $11.45 \pm 4.63$ ml/min/100g and greater variability in the correlation ($R^2$ of 0.65 compared to 0.89).
Fig. 5. Correlation between CBF and corresponding values of absolute $\alpha D_B$: (A) $\alpha D_B$ was obtained using the measured optical properties for each animal and (B) $\alpha D_B$ was determined assuming $\mu_a = 0.19 \text{ cm}^{-1}$ and $\mu_s' = 7.4 \text{ cm}^{-1}$ across all animals. Data are presented from the seven experiments using the single photon counting module with high quantum efficiency.

Figure 6 displays the errors in the recovered $\alpha D_B$ values caused by errors in the optical properties used in the fitting of the autocorrelation curves. Since the percentage errors in the recovered $\alpha D_B$ were independent of the initial value, the results for only one $\alpha D_B$ are shown in the figure. Figure 6A indicates that if the $\mu_a$ value used in the fitting procedure is smaller than the true value, then the retrieved $\alpha D_B$ would be underestimated, while the opposite relationship is true for errors in $\mu_s'$ (Fig. 6B). As expected, the magnitude of the error in $\alpha D_B$ caused by the latter is much greater since $\mu_s' \gg \mu_a$. The error analysis also indicated that relative changes in $\alpha D_B$ are not affected by errors in the assumed $\mu_a$ and $\mu_s'$ values, provided these values do not change during the course of an experiment. However, a change in either $\mu_a$ or $\mu_s'$ will cause the same magnitude of error in $\Delta(\alpha D_B)$ as shown in Fig. 6 for absolute $\alpha D_B$.

![Graph](image)

Fig. 6. Effects of errors in the optical properties used in the fitting of the autocorrelation curves on the recovered $\alpha D_B$: $g(\rho, \tau)$ curves were generated by (A) incrementing $\mu_a$ from 0.15 to 0.26 cm$^{-1}$ while keeping $\mu_s' = 8 \text{ cm}^{-1}$, and (B) increasing $\mu_s'$ from 7 to 10 cm$^{-1}$ while keeping $\mu_a = 0.2 \text{ cm}^{-1}$. All autocorrelation curves were fitted with the same $\mu_a$ and $\mu_s'$ values (0.2 and 8 cm$^{-1}$, respectively).

4. Discussion

In this study, concurrent TR-NIR and DCS measurements were acquired in a neonatal animal model in which CBF was manipulated over a wide range from 17 to 90 ml/100g/min. The
analysis revealed a strong linear correlation between CBF changes measured with the TR-NIR technique and corresponding changes in $aD_B$. These results are in good agreement with Zhou and al [17], who also reported in the same animal model a strong correlation ($R^2 = 0.81$, slope $= 1.1$) between perfusion changes measured with fluorescence microspheres and changes in $aD_B$. Diffuse correlation flowmetry has also been validated against other blood flow techniques including the MRI-based arterial spin labeling method [14,18,49], transcranial Doppler (TCD) [11,29,50], laser Doppler flowmetry [30], and xenon-CT [51]. However, the current study is the first to report a comparison of DCS with another optical technique that measures CBF in a similar tissue volume to that probed by the DCS signal. Since the TR-NIR and DCS instruments use similar probe wavelengths and single-photon detectors, partial volume errors between the two techniques should be minimal compared to comparison studies involving imaging modalities or TCD. This difference may explain the stronger correlation reported here compared to previous findings [11,14,18,24,50,51].

The results of Fig. 4A suggest that a reference CBF measurement acquired with the TR-NIR technique can be used to convert relative perfusion changes obtained with DCS into absolute values. This approach requires injecting the contrast agent ICG in order to measure CBF. However, this is a minimally invasive procedure since ICG can be injected into a peripheral vein and the arterial ICG time-concentration curve can be measured noninvasively by a DDG. Furthermore, ICG is FDA-approved, has a strong safety record, and has very low acute toxicity (median lethal dose in mice is 650 mg/kg) [52]. The accuracy of this combined TR-NIR/DCS approach will depend on the precision of the CBF measurement acquired by the bolus-tracking technique. Using a repeat-measures ANOVA, our group previously determined the precision to be 9.7%, which is quite reasonable [35]. In addition, subsequent measurements of CBF could be acquired with the TR-NIR technique if relative blood flow abnormalities were suspected by the DCS measurements. A potential complication with combining DCS with the bolus-tracking method is the effects of ICG-induced changes in $\mu_a$ on the $aD_B$ measurements. In these validation experiments, each ICG injection was followed by a delay of 15 min to allow sufficient time for the dye to clear. However, delays of this magnitude are extremely conservative and DCS data collection could begin very shortly after the acquisition of the dynamic TR-NIR data with minimal effect on $aD_B$. This is because the change in $\mu_a$ following the first pass of ICG (see Fig. 3B) is approximately 5%, which would lead to an error in $aD_B$ of only 2%, according to Fig. 6.

The strong linear correlation between CBF and absolute $aD_B$ shown in Fig. 5A suggests that an alternative approach would be to use $aD_B$ directly as a measure of CBF. Since a theoretical model relating $aD_B$ to the underlying tissue hemodynamics does not currently exist, this approach would require generating a population-based calibration curve to convert $aD_B$ into CBF. Analogous to the piglet experiments, the general relationship between CBF and absolute $aD_B$ for a specific population, such as neonates, would have to be established. The advantage of this approach is that, once the calibration curve was generated, it would not be necessary to obtain a CBF measurement with the bolus-tracking technique in subsequent experiments. This approach would require accurate measurements of the tissue optical properties in every experiment, such as by time-resolved or frequency domain methods [30], since uncertainties in $\mu_a$ and $\mu_a'$ can lead to substantial errors in $aD_B$, as shown in Fig. 6. These simulations can also be used to predict the error in relative $aD_B$ if the optical properties changed during the course of the experiment. If the values of the tissue optical properties remained within the range of values measured in the piglet experiments, the errors would be relatively small. For example, the change in $\mu_a$ from normocapnia to hypercapnia was 15% (Table 1), which translates into an error in relative $aD_B$ of only 5%.

The calibration-curve approach also requires that the DCS data be acquired at high count-rates to provide sufficient stability in the retrieval of $aD_B$. The sensitivity of $aD_B$ to count rate was evident in this study by the substantial reduction in the correlation between CBF and $aD_B$ when data acquired with the low quantum efficiency detector were included in the comparison. Noise in the autocorrelation function introduces a bias that affects the accuracy of the absolute $aD_B$ values. However, the effects are minimal on relative $aD_B$ measurements.
provided the count rate at different flow conditions is kept roughly constant, which can be achieved by adjusting the laser power. Greater SNR can be obtained by using detectors with high quantum efficiency or multiple detectors in parallel [8]. The count rate can also be improved by using few-mode detection fibers instead of truly single-mode fibers [3]. However, the increased count rate due to the larger active area of the detection fiber comes at the expense of an increased number of guided modes and, consequently, reduced correlation factor. For the few-mode detection fiber used in this study, $\beta$ was expected to be \(-0.167\) (for an ideal six-mode fiber) [48], which was in good agreement with the measured value ($\beta = 0.158$). The small discrepancy between the predicted and experimental $\beta$ values had no effect on the perfusion measurements and was likely due to a more efficient coupling of the fundamental $LP_{01}$ mode compared to the two $LP_{11}$ modes.

A potential complication with both the reference-CBF measurement and calibration-curve approaches is tissue heterogeneity, due to the presence of a substantial extracerebral layer in the region probed by the detected photons. While the contribution of extracerebral tissues is not expected to be substantial in neonates, it cannot be ignored in juvenile and adult subjects. Extracerebral contamination has been shown to cause substantial underestimations of CBF by ICG bolus-tracking techniques [53] and affect the accuracy of brain optical properties measurements [41]. A study involving computer simulations and layered tissue-mimicking phantoms also suggested that the accuracy of DCS can be degraded by extracerebral contamination [26]. It may be possible to overcome this limitation by applying multi-layer diffusion models and using depth-sensitive optical techniques [10,37,40,41].

Besides investigating $aD_B$ as a relative flow marker, we also compared changes in the area under the intensity autocorrelation curves ($\Delta A$) to CBF changes. This model-independent approach is appealing due to its conceptual simplicity and does not require any knowledge about the optical properties of the medium. The results of Fig. 4B show a strong correlation between perfusion changes and changes in $\Delta A$, but unlike in the case of $aD_B$, the correlation between $\Delta A$ and relative CBF was better represented by a second-order polynomial ($R^2 = 0.89$) rather than a linear curve ($R^2 = 0.86$), suggesting a non-linear scaling of CBF with $\Delta A$.

A potential limitation with this study was the use of two different methods to create conditions of low CBF. In the first seven animals, vascular cuffs placed around the common carotids were used to lower CBF. However, in three of these animals neither technique (DCS or TR-NIR) detected a change in CBF after the vascular occluders were inflated, which was likely due to collateral flow through the vertebral arteries. Consequently, hypocapnia rather than carotid occlusion was used to reduce CBF in the remaining three animals. Since the purpose of the study was to investigate the agreement between the two techniques over a range of CBF values, altering the method used to reduce CBF should not affect the main conclusion of the study. A second potential limitation was the exclusion of data from the first three animals in the comparison between absolute $aD_B$ and CBF. It was necessary to exclude these data due to poor SNR caused by low count rates. In the remaining seven animals, DCS data were acquired using a detector with a high photon detection efficiency, which improved the SNR and reduced the variability in the retrieved $aD_B$ values. Another potential limitation was that the tissue optical properties used in the fitting of the autocorrelation curves were measured at 802 nm, and not at the emission wavelength of the DCS instrument (786 nm). However, it is reasonable to assume that the difference in optical properties between these two wavelengths is small and, consequently, the error in the DCS measurements is minimal. For example, we have measured the optical properties at 802 and 760 nm (data not presented) and the difference in $\mu_s$ was less than 5%, which would cause an error of less than 3% in $aD_B$.

A final consideration is the fact that $\alpha$ depends on the fraction of moving scatterers in the probed region, i.e., the cerebral blood volume (CBV). In this study, a good correlation was observed between the CBV measurements obtained from the bolus-tracking method and $aD_B$ ($R^2 = 0.66$, data not presented). However, this correlation is not surprising considering the well-known coupling of CBV and CBF during variations in arterial blood $CO_2$ tension [54], and a good correlation between CBV and CBF was also observed in this study ($R^2 = 0.77$). It
would be interesting to determine if the relationship between $aD_b$ and CBF changes under condition in which CBV and CBF become uncoupled, such as during ischemia [55].

5. Conclusion

In this study we demonstrated that DCS data could be converted into cerebral blood flow using an all-optical technique approach. An extension of this approach would be to combine DCS with a TR-NIR system equipped with multiple lasers. This combination would enable continuous monitoring of both CBF and cerebral blood oxygenation, from which the cerebral metabolic rate of oxygen could be determined. The ability to continuously measure absolute blood flow and oxygen utilization could help improve the outcome of patients at risk of brain injury by early identification of cerebral ischemia.

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