Quantitative diffuse optical tomography for small animals using an ultrafast gated image intensifier

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1 Introduction

Optical approaches to small animal in vivo molecular imaging provide high sensitivity, stable nonradioactive probes, and an extensive array of functional reporting strategies. However, quantitative whole body assays remain elusive. While planar fluorescence reflectance imagers (FRI) provide quick assessment of probe concentrations, quantitative localization is lacking due to a rapid decrease in sensitivity with increased depth, masking of buried targets by superficial tissues, and poor resolution. Recently, the feasibility of in vivo continuous-wave fluorescence tomography (CWFT) has been demonstrated in small animals. CWFT addresses some of the limitations of FRI and can provide even sensitivity versus depth throughout live, intact mice.

A critical aspect of fluorescence tomography has been the use of differential measurements, often a ratio of the fluorescence emission intensity over the reemitted excitation light intensity. Emission/excitation ratio measurements reduce the influence of heterogeneous optical properties and path lengths. Thus far, small animal fluorescence tomography studies have therefore generally assumed homogeneous optical properties for image reconstruction. Previous studies that examined the influence of heterogeneous optical properties upon linear reconstructions of ratio metric data have come to different conclusions. Some studies have found the influence of moderately absorbing backgrounds on linear reconstructions to be weak, while other studies have found a more significant effect. The different conclusions are due in large part to differences in the optical property scenarios evaluated. For our application interest—whole body mouse imaging—it is reasonable to expect optical properties in mice to vary by an order of magnitude. Given these variations in optical properties, the incorrect assumption of homogeneous tissue properties can create significant errors in the reconstructed fluorescence concentrations.

Abstract. The quantitative accuracy of fluorescence and bioluminescence imaging of small animals can be improved by knowledge of the in situ optical properties of each animal. Obtaining in situ optical property maps is challenging, however, due to short propagation distances, requirements for high dynamic range, and the need for dense spatial, temporal, and spectral sampling. Using an ultrafast gated image intensifier and a pulsed laser source, we have developed a small animal diffuse optical tomography system with multiple synthetic modulation frequencies up to >1 GHz. We show that amplitude and phase measurements with useful contrast-to-noise ratios can be obtained for modulation frequencies over the range of ~250 to 1250 MHz. Experiments with tissue simulating phantoms demonstrate the feasibility of reconstructing the absorption and scattering optical properties in a small animal imaging system. © 2008 Society of Photo-Optical Instrumentation Engineers. DOI: 10.1117/1.2830656

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We present herein a time-resolved, ICCD-based small animal tomography system for obtaining quantitative absorption and scattering optical property maps. Using Fourier-transformed time-domain data, we show that measurements with useful contrast-to-noise ratios can be obtained even for modulation frequencies >1 GHz. Experiments in tissue simulating phantoms are used to demonstrate the feasibility of reconstructing volumetric optical properties.

2 Methods

2.1 Time-Resolved Diffuse Optical Tomography System

The design strategy for the current time-domain diffuse optical tomography (TD-DOT) system was to extend a previous fast scanning CWFT platform through addition of time-resolved measurement capabilities. The system uses a planar transmission imaging geometry and provides dense spatial sampling with a large field-of-view and short scan times (Fig. 1). A pulsed source laser is raster-scanned across a “source” window and illuminates the imaging cassette. Light transmission through the imaging volume is profiled on the opposing “detector” window using a lens coupled gated image intensifier. The pulsed laser source (Becker & Hickel, BHLP-700, 780 nm, \( T_p = 100 \) ps, 1 mW) provides a continuous train of light pulses (repetition rate 50 MHz). For time-resolved detection, an ultrafast image intensifier (PicoStar HR-12, LaVision, \( T_{gate} < 300 \) ps) relays time-gated images of the light levels on the detection plane to an electron multiplying charged coupled device (EMCCD; iXon 877f, Andor Technologies).

With the image intensifier operating in a “comb” mode, a sequence of gated transmissions accumulates on the EMCCD during each exposure burst. The pulse train is electronically shuttered on/off with switching times <100 μs to control the burst duration (i.e., exposure time) for each source position. The light transmitted through the tissue volume is temporally resolved by scanning the delay time \( T_{d} \) between the source laser pulse and the time-gated detection. The zero time delay, \( T_{d} = 0 \), is defined as the time delay at which the instrument response function is maximum. To obtain an instrument response function, a white sheet of paper is positioned in the imaging volume on the detector plane and a measurement set is taken. TD-DOT data sets consisted of measurements at 48 time-points \( (T_d \text{ from } -0.5 \text{ ns to } 4.2 \text{ ns in steps of } 0.1 \text{ ns}) \), with an image intensifier gate width of 0.4 ns for each source position. Temporally resolved scans are collected for each of 24 source positions, and data is extracted from the full CCD images for 24 detector positions symmetrically opposed to the source positions. The time-domain data are then converted from time-domain to frequency-domain using a Fourier transform for modulation frequencies of 208.3, 312.5, 416.7, 625, 833.3, and 1250 MHz. The frequency-domain data (\( \Phi_i \)) are indexed by \( i \), for a specified source position \( r_{s(i)} \), detector position \( r_{d(i)} \), and modulation frequency \( \omega_i \).

2.2 Diffuse Optical Tomography Algorithms

In this section, we detail the algorithms used to reconstruct the optical properties using light collected at the excitation wavelength. The data, \( \Phi_i \), are inverted using a first-order perturbation solution of the frequency-domain diffusion approximation. In the following paragraphs, we use both the diffusion coefficient \( D(\vec{r}) \) and the reduced scattering coefficient \( \mu_s^r(\vec{r}) \), which are related by the simplified expression \( D(\vec{r}) = \nu/3 \mu_s^r(\vec{r}) \), where \( \nu \) is the speed of light in the medium. The perturbation expansion separates the absorption \( (\mu_a(\vec{r})) \) and diffusion \( (D(\vec{r})) \) coefficients into background values \( (\mu_a^b \text{ and } D_a) \) and spatially dependent components \( (\Delta \mu_a(\vec{r}) \text{ and } \Delta D(\vec{r})) \). A scattered field, \( \mathbf{y} \), is defined as the negative log of the ratio of \( \Phi_{\text{target}}^i / \Phi_{\text{background}}^i \) with only background media, \( \gamma_i = -\log(\Phi_{\text{target}}^i / \Phi_{\text{background}}^i) \). The scattered field is then modeled using a linear Rytov approximation, \( \mathbf{y} = \mathbf{A} \mathbf{x} \), where \( \mathbf{x} = [x_a, D] \) is the optical property map, and \( \mathbf{A} = [W_a, W_D] \), where \( W_a \) and \( W_D \) represent the sensitivity functions for the absorption and diffusion coefficients, respectively. \( W_a \) and \( W_D \) are expressed in terms of Green’s functions:

\[
W_a(i,k) = \frac{-\hbar^3}{D_o} \frac{G(\vec{r}_{d(i)}, \vec{r}_{a(i)}, \omega)}{G(\vec{r}_{s(i)}, \vec{r}_{d(i)}, \omega)} \nabla^2 G(\vec{r}_{s(i)}, \vec{r}_{d(i)}, \omega),
\]

\[
W_D(i,k) = \frac{\hbar^3}{D_o} \frac{\nabla G(\vec{r}_{s(i)}, \vec{r}_{a(i)}, \omega)}{G(\vec{r}_{s(i)}, \vec{r}_{d(i)}, \omega)} \nabla G(\vec{r}_{a(i)}, \vec{r}_{d(i)}, \omega),
\]

where \( \hbar^3 \) represents the voxel volume. For the current imaging geometry, the Green’s functions are slab solutions to the diffusion approximation equation. Separating the real \( (R) \) and imaginary \( (I) \) components, the forward problem takes the following matrix form:

\[
\begin{bmatrix}
Y_R \\
Y_I
\end{bmatrix} =
\begin{bmatrix}
W_{aR} & W_{aI} \\
W_{dR} & W_{dI}
\end{bmatrix}
\begin{bmatrix}
x_a \\
x_D
\end{bmatrix}.
\]

Since the absorption \( x_a \) and diffusion \( x_D \) variables differ by orders of magnitude, preconditioning of the Jacobian matrix is necessary to reduce parameter cross talk between \( x_a \) and \( x_D \). We precondition by normalizing the absorption and diffusion variables prior to inversion using a substitution of variables \( \bar{x} = L \mathbf{x} \), where

\[
L = \begin{bmatrix}
1 & 0 \\
0 & \frac{1}{D_a} \frac{\nabla^2 G(\vec{r}_{s(i)}, \vec{r}_{d(i)}, \omega)}{G(\vec{r}_{s(i)}, \vec{r}_{d(i)}, \omega)}
\end{bmatrix}.
\]
We also incorporate the measurement noise using a substitution of variables, \( \tilde{y} = Cy \), where the diagonal of \( C \) is a shot noise model. Both of these modifications are incorporated into the forward model, and we proceed with the inverse step using \( \tilde{y} = \tilde{A} x \), where \( \tilde{A} = C A L^{-1} \). To stabilize the inversion, we minimize an objective function that incorporates spatially variant Tikhonov regularization:\(^{23}\)

\[
\min \left\{ \| y - \tilde{A} x \|^2_2 + \alpha \| Q x \|^2_2 \right\},
\]

The penalty term for image variance is, \( \alpha \| Q x \|^2_2 \), where \( \alpha \) is a regularization constant, and \( Q \) is a spatially variant term where \( \text{diag}(Q) = \text{sqr}(\text{diag}(A^T A)) + \lambda \). Last, direct inversion of \( \tilde{A} \) is accomplished using a Moore-Penrose generalized inverse. After reconstruction, we convert maps of the diffusion coefficient to maps of the reduced scattering coefficient.

2.3 Heterogeneous Optical Properties and CW Fluorescence Tomography

To provide a context for phantom tests of our TD-DOT system, we evaluated the influence of heterogeneous optical properties on fluorescence tomography for the specific situation of small animal imaging. The distribution of effective attenuation coefficients in mice was evaluated using transmitted light measurements through mice. The TD-DOT system was reconfigured for use in the CWFT mode of operation by removing the gated image intensifier and using a continuous-wave laser diode (Hiachi, HL785G, \( P_{\text{in}} = 50 \text{ mW}, 785 \text{ nm} \)).\(^{1}\)

An anesthetized mouse was positioned in the imaging cassette. Transmission measurements (at the excitation wavelength without fluorescence emission filters) at opposing source-detector pairs were obtained over a 48 mm × 72 mm region with 2-mm-square spacing. To provide a consistent physical path length, the tank surrounding the mouse was filled with an optically matching solution, \( \mu_a = 0.2 \text{ cm}^{-1} \), and \( \mu'_s = 10 \text{ cm}^{-1} \), using Intralipid, water, and India ink.

In a second set of measurements, we created tissue-simulating phantoms for fluorescence tomography that matched the variation in optical properties found in the in vivo measurements. Three plastic tubes (3.5-mm OD) were placed in a background solution \( (\mu_a = 0.1 \text{ cm}^{-1} \text{ and } \mu'_s = 10 \text{ cm}^{-1}) \). Each tube had 1 mM concentration of Indocyanine Green (ICG) and each tube had variable concentrations of India ink. Emission and excitation measurements were obtained by scanning the phantom using a 12 × 64 source-detector grid with 1-mm-square spacing. Fluorescence tomography reconstructions were performed using a homogeneous tissue model following previous procedures.\(^{1}\)

2.4 Evaluations of Optical Property Maps Using Tissue Simulating Phantoms

The feasibility of reconstructing absorption and scattering optical property maps using the TD-DOT system was evaluated using tissue-simulating phantoms. Tubes filled with Intralipid, water, and India ink mixture to simulate objects with increased absorption \( (\mu_a = 0.44 \text{ cm}^{-1}, \mu'_s = 15 \text{ cm}^{-1}) \) and increased scattering \( (\mu_a = 0.12 \text{ cm}^{-1}, \mu'_s = 40 \text{ cm}^{-1}) \) were used for this purpose. The tubes were suspended in the imaging cassette filled with tissue matching fluid \( (\mu_a = 0.12 \text{ cm}^{-1}, \mu'_s = 15 \text{ cm}^{-1}) \) and were separated by a center-to-center distance of 2 cm. Target data (with tube phantoms) and background data (without tube phantoms) were collected using the TD-DOT system in TD mode operation (Sec. 2.1).

3 Results and Discussion

3.1 The Influence of Heterogeneous Optical Properties on CW Fluorescence Tomography

The heterogeneity of the optical properties in mice is illustrated by measurements of light transmission through a mouse (Fig. 2). The influence of heterogeneous optical properties...
concentration. With higher absorption incorrectly indicate less fluorophore concentration.

In vivo India ink concentration phantoms. The variation in the light transmission measured at opposing source-detector pairs across the three tube phantoms. The variation in the light transmission (due to variations in India ink concentration) between the three tubes is similar to that encountered in vivo (Fig. 1). (c) Fluorescence tomography reconstruction of data using a homogeneous tissue model. Images of the tubes with higher absorption incorrectly indicate less fluorophore concentration.

(Fig. 2) on fluorescence tomography reconstructions is illustrated using matching tissue phantom experiments (Fig. 3).

In Fig. 3 the transmitted light intensity varies by a factor of ~50 (brightest to darkest) across the body of a mouse. Using \( T \times \exp(-\mu_{eff} L) \), where \( L = 1 \text{ cm} \), and \( \mu_{eff} \) is the effective attenuation coefficient, a 50-fold variation in \( T \) corresponds to a ~fourfold variation in \( \mu_{eff} \). Since \( \mu_{eff} = (3 \mu_s \mu'_a)^{1/2} \), a ~fourfold variation in \( \mu_{eff} \) corresponds to approximately a 16-fold change in either the absorption or the reduced scattering coefficients.

The influence of this degree of heterogeneity on fluorescence tomography reconstructions was evaluated using tissue-simulating phantoms. Three tubes had identical fluorophore concentrations (1 \( \mu \text{M} \) ICG) and variable amounts of India ink so that the tubes left, center, and right had absorption coefficients \( 1 \times \), \( 5 \times \), and \( 10 \times \) the background, respectively (Fig. 3). Reconstructions using a homogeneous tissue mode produce images with values that incorrectly vary by up to a factor of 5. In such situations, \textit{in situ} optical property maps are needed to correctly reconstruct fluorophore distributions.

3.2 \textit{Evaluations of TD-DOT Optical Property Maps}

Figure 4 illustrates the concept of time-resolved measurements where the pulsed laser illuminates the imaging cassette and the gated intensifier time-resolves light emitted from the opposing window. Due to scattering and absorption, the detected pulse has decreased intensity and is delayed; its dependence on source-detector separation is also illustrated. Figure 4(a) shows the system response function with time \( t = 0 \text{ s} \) along with the response to a homogeneous medium as measured by three equally spaced detectors with increasing source-detector separation. The signal amplitude falls off exponentially with increasing source-detector separation, while the mean time-of-flight increases with distance, reflecting an increase in mean path length with increased source-detector distance. The data were converted using a Fourier transform from time-domain to frequency-domain. The signal amplitude [Fig. 4(b)] and phase [Fig. 4(c)] at three representative modulation frequencies: 312.5 MHz (blue), 625 MHz (green), and 1250 MHz (red), are shown for the middle (12th) source and all the detectors. (Color online only).

The utility of frequency domain data for distinguishing between absorbing and scattering objects can be observed directly in the measurement data. Two tubes with diameter 5 mm were imaged, one with increased absorption...
signals normalized by the phantom tube volume $V_T$ (or normalized volume integrated (NVI) signals) for the expected $NVI_e = \int S_{\text{expected}} \, dv / V_T$ and reconstructed $NVI_r = \int S_{\text{reconstruction}} \, dv / V_T$ images, where $S$ is the signal value in a voxel. The region of interest surrounding the tube phantoms is used in calculating $NVI_e$. The expected values for $NVI_e$ are 0.32 cm$^{-1}$ ($\Delta \mu_a$) and 25 cm$^{-1}$ ($\Delta \mu_s'$). The reconstructed $NVI_r$ values are 0.34 cm$^{-1}$ ($\Delta \mu_a$) and 22.58 cm$^{-1}$ ($\Delta \mu_s'$), both within an error <10% with respect to the $NVI_e$.

To provide a more comprehensive evaluation, we also tested two other situations: the ability of the system to reconstruct phantoms with both absorption and scattering perturbations, and the ability to image over a range of contrasts. For the first imaging task, we reconstructed three tubes (diam=3 mm), one with only absorption contrast ($\mu_a=0.44$ cm$^{-1}$, $\mu_s'=15$ cm$^{-1}$), one with both absorption and scattering ($\mu_a=0.44$ cm$^{-1}$, $\mu_s'=40$ cm$^{-1}$), and one with only a scattering contrast ($\mu_a=0.12$ cm$^{-1}$, $\mu_s'=40$ cm$^{-1}$) as shown in Figs. 8(a) to 8(c). In the reconstructed images [Figs. 8(b) and 8(c)], each tube shows up with the appropriate combination of contrast with minimal cross talk between absorption and scattering. For the second imaging task, we varied the absorption coefficient [Fig. 8(d), $\mu_a=0.44$ cm$^{-1}$, 0.34 cm$^{-1}$, 0.12 cm$^{-1}$, $\mu_s'=40$ cm$^{-1}$].

In Fig. 7, reconstructions of the absorption (a) and scattering (b) perturbations demonstrate the capability to separate the two optical properties using data at 625-MHz modulation frequency. In order to check the quantitative accuracy of the reconstructions, we compared volume-integrated 

\[ \text{NVI}_r = \frac{\int S \, dv}{V_T} \]

where $S$ is the signal value in a voxel. The region of interest surrounding the tube phantoms is used in calculating $\text{NVI}_r$. The expected values for $\Delta \mu_a$ are 0.32 cm$^{-1}$ and 25 cm$^{-1}$ ($\Delta \mu_s'$). The reconstructed $\text{NVI}_r$ values are 0.34 cm$^{-1}$ ($\Delta \mu_a$) and 22.58 cm$^{-1}$ ($\Delta \mu_s'$), both within an error <10% with respect to the $\text{NVI}_e$. 

To provide a more comprehensive evaluation, we also tested two other situations: the ability of the system to reconstruct phantoms with both absorption and scattering perturbations, and the ability to image over a range of contrasts. For the first imaging task, we reconstructed three tubes (diam=3 mm), one with only absorption contrast ($\mu_a=0.44$ cm$^{-1}$, $\mu_s'=15$ cm$^{-1}$), one with both absorption and scattering ($\mu_a=0.44$ cm$^{-1}$, $\mu_s'=40$ cm$^{-1}$), and one with only a scattering contrast ($\mu_a=0.12$ cm$^{-1}$, $\mu_s'=40$ cm$^{-1}$) as shown in Figs. 8(a) to 8(c). In the reconstructed images [Figs. 8(b) and 8(c)], each tube shows up with the appropriate combination of contrast with minimal cross talk between absorption and scattering. For the second imaging task, we varied the absorption coefficient [Fig. 8(d), $\mu_a=0.44$ cm$^{-1}$, 0.34 cm$^{-1}$, 0.12 cm$^{-1}$, $\mu_s'=40$ cm$^{-1}$].
and 0.24 cm⁻¹, top to bottom, respectively] while holding the reduced scattering coefficient constant and then varied the reduced scattering coefficient while holding the absorption constant [Fig. 8(g), $\mu'_s = 20$ cm⁻¹, 30 cm⁻¹, and 40 cm⁻¹, top to bottom, respectively]. The reconstructions performed as expected for both the absorption titration [Figs. 8(e) and 8(f)] and the reduced scattering titration [Figs. 8(h) and 8(i)]. Targets showed graded contrasts and minimal cross talk between the absorption and reduced scattering coefficients.

Using single modulation frequency measurements in the reconstruction, the interparameter cross talk is less than 10%. Similar reconstructions were obtained for 208.3, 312.5, 416.7, 833.3, and 1250 MHz modulation frequency data sets. The resolution of the scattering object [e.g., full width half maximum (FWHM)] is better than the resolution for the absorbing object at all modulation frequencies. This is in agreement with previous theoretical and experimental results.²⁴

The interaction between modulation frequency and optical properties and their combined effect on image quality for single-frequency data sets has been examined. O’Leary et al. found in simulation studies that when background absorption is high ($\mu_a = 1$ cm⁻¹), an increase in modulation frequency from 50 MHz to 1 GHz has no effect on image quality.⁷ However, for low absorption (0.01 < $\mu_a < 0.1$ cm⁻¹), image quality improved. Jiang et al.⁵ using both simulated and measured data, found minimal effect of modulation frequency on reconstructions.

Reconstructions that combine measurements at several frequencies are expected to reduce the interparameter cross talk, provide better resolution, and increase the measurement information content. Intes et al.²³ have reported simulation studies with modulation frequencies from 20 to 500 MHz where different preconditioning techniques of the weight matrix were compared for image quality and cross talk. Their results indicate that increasing the number of frequencies gave a decrease in interparameter cross talk and the column scaling preconditioning scheme resulted in less cross talk and better estimates. Gulsen et al.²⁵ have compared multiple-frequency reconstructions with reconstructions from various individual frequencies. Multifrequency images had lowest measurement error and highest contrast-to-noise ratio (CNR), and also the noise near the source-detector positions was reduced. Future studies with more complex phantoms and a range of object contrasts are needed to evaluate and optimize the effects of modulation frequency and multiple-frequency reconstructions for our system.

### 4 Conclusions

The data presented demonstrate the feasibility of reconstructing the three-dimensional (3-D) optical property maps of small tissue volumes relevant for imaging mice from high-frequency time-resolved amplitude and phase measurement. Useful CNRs were obtained even for modulation frequencies in the range of 300 to 1250 MHz, and absorption and scattering components were successfully reconstructed and separated. The in situ heterogeneous optical property maps obtained with this diffuse optical tomography system can be used as a basis for quantitative fluorescence or bioluminescence¹,²,²⁷ tomography of small animals. In this paper, we reconstructed optical properties at a single wavelength. A more complete assessment, with mapping of the optical properties at both the excitation and emission wavelengths for a given fluorophore, could be achieved through a straightforward extension of the current system to a two or more wavelength system. Use of an ICCD permits integration of this approach into our existing CW small animal fluorescence tomography system. Compared to fluorescence reconstructions using homogeneous tissue models, volumetric images obtained using heterogeneous tissue models are expected to provide a more uniform mapping of fluorophore concentration and activity independent of local or nearby tissue optical properties.²⁸ We anticipate that as fluorescence tomography becomes more widely used, the development of methods to measure and incorporate heterogeneous optical property maps into fluorescence tomography reconstructions will become increasingly important.
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