A quantitative spatial comparison of high-density diffuse optical tomography and fMRI cortical mapping

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Abstract

Functional neuroimaging commands a dominant role in current neuroscience research. However its use in bedside clinical and certain neuro-scientific studies has been limited because the current tools lack the combination of being non-invasive, non-ionizing and portable while maintaining moderate resolution and localization accuracy. Optical neuroimaging satisfies many of these requirements, but, until recent advances in high-density diffuse optical tomography (HD-DOT), has been hampered by limited resolution. While early results of HD-DOT have been promising, a quantitative voxel-wise comparison and validation of HD-DOT against the gold standard of functional magnetic resonance imaging (fMRI) has been lacking. Herein, we provide such an analysis within the visual cortex using matched visual stimulation protocols in a single group of subjects (n = 5) during separate HD-DOT and fMRI scanning sessions. To attain the needed voxel-to-voxel co-registration between HD-DOT and fMRI image spaces, we implemented subject-specific head modeling that incorporated MRI anatomy, detailed segmentation and alignment of source and detector positions. Comparisons of the visual responses found an average localization error between HD-DOT and fMRI of 4.4 ± 1 mm, significantly less than the average distance between cortical gyri. This specificity demonstrates that HD-DOT has sufficient image quality to be useful as a surrogate for fMRI.

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Introduction

Functional brain mapping has revolutionized neuroscience research, by providing noninvasive investigations into human brain activity. However, functional imaging of the brain has, so far, found only limited clinical application with early uses in pre-operative planning (Nelles et al., 2009; Shimony et al., 2009; Wengenroth et al., 2011; Zhang et al., 2009). Functional imaging has the potential to play a larger clinical role in diagnosis, prognosis and monitoring due to its ability to find subtle changes in function before disease progresses to large-scale structural change. However, traditional functional brain mapping methods, including functional MRI (fMRI) and positron emission tomography (PET) are limited in many settings by immobility, expense, and constraints on subjects. Additionally, they have contraindications for metallic implants (fMRI) or use ionizing radiation (PET), limiting the number of repeated studies. In contrast, optical methods provide non-ionizing functional neuroimaging with potentially portable and wearable technology that is well-suited for many of the subjects inaccessible by fMRI or PET. Early diffuse optical imaging (DOI) methods used sparse sets of source-detector pairs to generate two-dimensional, low-resolution images of cerebral hemodynamics (Maki et al., 1995; Obrig and Villringer, 2003; Villringer et al., 1993). A more advanced method, diffuse optical tomography (DOT), relies on a variety of measurement strategies to improve lateral and depth resolution. Time-resolved (TR) measurements use time-gating (Benaron et al., 2000; Gibson et al., 2006; Hebden et al., 2002; Kohl-Bareis et al., 2002; Selb et al., 2005; Steinbrink et al., 2001) or frequency-domain phase data (Kohl-Bareis et al., 2002) to profile different tissue depths. However, the complexity and cost of TR systems impose practical limits and require tradeoffs between channel count, source and detector density, coverage (field-of-view), and frame rate. Another strategy uses...
high-density DOT grids with overlapping continuous wave measurements at multiple SD-pair separations (Bluestone et al., 2001; Boas et al., 2004a; Joseph et al., 2006; Zeff et al., 2007). Relative to DOI, the newer HD-DOT methods achieve higher resolution and improved localization accuracy (Gibson and Dehghani, 2009; Gibson et al., 2005; Habermehl et al., 2011; Koch et al., 2010; White and Culver, 2010b; Zeff et al., 2007). While HD-DOT’s ability to decipher detail has been established in studies of retinotopy in visual cortex (White and Culver, 2010a; Zeff et al., 2007) and finger-topy in the motor cortex (Custo et al., 2009; Koch et al., 2010; White et al., 2009), the image quality of HD-DOT at the voxel level has not been compared directly to fMRI, the current gold standard in hemodynamic-based functional neuroimaging. Establishing the relationship between HD-DOT and fMRI functional maps could significantly strengthen the impact that HD-DOT might have when used as a surrogate for fMRI. The purpose of this study is to validate HD-DOT functional mapping accuracy through a quantitative voxel-wise comparison to fMRI in subject-matched datasets of visual cortex activity. Previous comparative studies of diffuse optical and fMRI signals have used either non-imaging systems or sparse measurement datasets and performed comparisons unrelated to image quality. For example, thorough comparisons have been made in the measurement space of the DOI instrument (Cui et al., 2011; Huppert et al., 2006a, b; Sassaroli et al., 2006; Strangman et al., 2002; Toronov et al., 2001). Additionally, there have been detailed temporal evaluations (e.g., comparing the time course of the DOI response within a similar volume as that displaying an MRI response (Okamoto et al., 2004; Sakatani et al., 2007)). Throughout these studies, correlations were found between the time courses of BOLD and optical data. These findings along with parallel studies in rodent models (Bouchard et al., 2009; Culver et al., 2003; Custo et al., 2009; Devor et al., 2003; Dunn et al., 2005; Siegel et al., 2003) and human neonates (Villringer and Chance, 1997) lay the foundation for optical measurements to be used in calculations of metabolic markers such as CMRO2 at the bedside. Additionally, image- and time-domain comparisons have been made with simultaneously acquired MRI and NIRS (Toronov et al., 2007; Zhang et al., 2005). However, these studies used point-like activations and have not investigated image quality throughout an extended cortical region.

In this study we perform spatial voxel-wise comparisons between HD-DOT and fMRI data sets for cortical responses to visual stimulations throughout the visual field of view. The HD-DOT and fMRI datasets were co-registered on a subject specific basis by segmenting anatomical MRIs for each subject, locating and co-registering the HD-DOT cap placement on each subject’s head, and solving the forward light model within the subject-specific space. Visual activations were used because they have served extensively as a substrate for validation by other neuroimaging methodologies (Belliveau et al., 1991; Engel et al., 1997; Fox et al., 1986; Fox et al., 1985) and because the structure and function of the visual cortex has been comprehensively mapped via invasive anatomical and electrical studies in mammals (Pelleman and Van Essen, 1991; Gilbert and Wiesel, 1979; Rosa et al., 1993) and humans (Harding et al., 1991; Spalding, 1952). The comparison between HD-DOT and fMRI was quantified by calculating the center-of-mass of the imaged hemodynamic response to matched visual activations, and by a complete phase analysis of the responses to stimulations throughout the full visual field. The resulting HD-DOT image quality evaluation serves as a strong foundation and validation enabling further adoption of HD-DOT by both neuroscientists and clinicians.

Methods

Subjects and stimulus protocol

Five healthy adult right-handed subjects (aged 21–30) were recruited for this study. All subjects passed MR screening to ensure their safe participation. Informed consent was obtained for all subjects. The research was approved by the Human Research Protection Office at Washington University School of Medicine. All stimuli are angularly sweeping reversing black-and-white logarithmic checkerboard wedges (10 Hz reversal) on a 50% gray background (Engel et al., 1994; Warnking et al., 2002). The grids rotate around a white cross located at the center of the visual field, step 10° each second, and complete a full sweep every 36 s. For the HD-DOT stimuli, subjects are seated in an adjustable chair facing a 19” LCD display at a 90 cm viewing distance. For the fMRI stimuli, the stimulus is presented via a projector onto a screen that the subject could visualize from their position within the MR tube with a mirror attached to the head coil. The stimulus size is calibrated to be the same size on the retina as when presented in the HD-DOT setting: each wedge subtends a radial angle of 2.5°–10.5° and a polar angle of 60°. A set of stimuli consists of 10 repetitions of either clockwise- or counter-clockwise-rotating flickering wedges. Subjects are instructed to fixate on the central crosshair. Gray screens are presented for 30 s before and after each stimulus set.

HD-DOT imaging system and acquisition

The high-density imaging system has been described previously (Zeff et al., 2007). Briefly, our high-density DOT instrument uses light-emitting diode (LED) sources at 750 nm and 850 nm (750-03 AU and OPEST85, Roithner Lasertechnik) and avalanche photo diode (APD, Hamamatsu C5460-01) detectors (Zeff et al., 2007). Each detector has a dedicated 24-bit analog-to-digital converter (MOTU HD-192). Sources and detectors are coupled with fiber optic bundles to a flexible imaging cap held on to the back of the head with hook-and-loop strapping. After digitization, the APD measurements are written directly to hard-disk at 96 kHz. With temporal, frequency, and spatial encoding, the system works with a frame rate of 11 Hz in continuous wave mode (the time and frequency encoding is at ~10 kHz, much slower than the >10 MHz needed for measuring the time of flight for light propagation through tissue). The array has 24 source and 28 detector positions placed in two interlaced rectangular arrays with first- through fourth- nearest neighbor separations as follows: 1.3, 3.0, 3.9, and 4.7 cm. In order to ensure consistent pad placement between sessions, measurements of the distance between the optode array and the nasion and eyes are taken and recorded. Once the cap is placed comfortably with good signal-to-noise, the exact placement of the pad is found by measuring the locations of the outer four corner positions of the optode array using a commercially available 3D digitizer (Fastrak, Polhemus). Concurrently, we also measure the locations of anatomical landmarks on the head and face of the subject (e.g., the nasion) (Klem et al., 1999) to locate the pad relative to the subject’s head.

fMRI acquisition

fMRI scans are collected on a Siemens Trio (Erlagen, Germany) 3T scanner. Anatomical T1-weighted MPRAGE (echo time (TE) = 3.13 ms, repetition time (TR) = 2400 ms, flip angle = 8°, 1 x 1 x 1 mm isotropic voxels) and T2-weighted (TE = 84 ms, flip angle = 120°, 1 x 1 x 4 mm voxels) scans are taken at each session (for simplicity, we will subsequently refer to these scans as T1 and T2). Functional images are collected using a series of asymmetric gradient spin-echo planar (EPI) sequences (each brain volume had a TE = 27 ms, TR = 2000 ms, flip angle = 90°, 4 x 4 x 4 mm voxels) to measure the blood-oxygenation-level-dependent (BOLD) contrast. In keeping with standard methods for performing BOLD analysis, we transform the BOLD data to a 3 mm isotropic voxelated space.
Head modeling

To place the HD-DOT image space into the correct subject specific anatomic location, a full head model must be constructed. The head model incorporates the surface head shape, assumed optical properties of each voxel, and locations of the sources and detectors of the HD-DOT system. The basis for the head model is the subject-specific anatomical T1 (Fig. 1A) and T2. These volumes contribute complementary information that provides characteristic information for each tissue type. An iterative series of thresholding, region growing, and masking techniques are used to segment the head tissue into scalp, skull, cerebral spinal fluid (CSF), gray matter, and white matter regions (Fig. 1B).

To generate a space for the numerical light modeling, a high-density volumetric tetrahedral head mesh (Fig. 1C) is created from the segmented head using the Mimics software package (Materialize, Belgium). Each subject’s head mesh has approximately $5 \times 10^5$ nodes and $3 \times 10^6$ tetrahedral volume elements. To ensure proper resolution, the maximum inter-node distance both on the surface and within the mesh volume is set to 3 mm. Tissue-type-specific labeling of optical properties generates a more accurate light model than assuming homogeneous optical properties (Custo et al., 2006; Dehaes et al., 2011; Dehghani et al., 2000; Heiskala et al., 2009; Ripoll et al., 2000). Thus, each node is labeled by tissue type and assigned optical properties from the literature as summarized in Table 1 (Bevilacqua et al., 1999; Custo et al., 2006; Strangman et al., 2003). While errors in the assumed baseline optical properties will propagate through to the magnitude of the differential concentration changes (Strangman et al., 2003), this treatment of baseline optical properties is consistent with standard DOT processing (Blasi et al., 2002; Dehghani et al., 2003). The sensitivity matrix provides a mapping function that transforms the measured HD-DOT data to optical parameter changes within the model. The original tetrahedral finite-element based sensitivity function is then transformed into the 3 mm isotropic voxel space (standard in current BOLD processing practices) through a weighted spatial average to create the sensitivity matrix for the subject-specific space. As a result of these procedures the image spaces for the two modalities are co-aligned.

Functional data analysis

HD-DOT light measurement data were converted to log-ratio and high-pass filtered (0.02 Hz cutoff) to remove long term drift. An average of all 1st-nearest-neighbor measurements (sampling predominantly scalp and skull) was constructed as an estimate of systemic signals (Obrig and Villringer, 2003; White et al., 2009; Zeff et al., 2007). This signal was then regressed from all measurements. After a low-pass filter (0.5 Hz cutoff) removed residual pulse signals, the time traces were used for image reconstruction. Note that while a time trace of the average across channels of all 1st nearest-neighbor pairs was removed during the global signal regression, the individual 1st-nearest-neighbor channels retain variance after this regression and are used during the reconstruction. To manage system noise during cap fit, real-time data displays are used to adjust the cap (at the individual optode level if needed) to optimize the highest possible light level and lowest possible noise level. Further, the data is de-noised by removing noisy source-detector pair measurement channels that have signal standard deviations (across time) greater than 7.5% of their mean signal magnitude. Using this threshold, the following percentages of nearest neighbor (nn) measurements are typically retained: 100% 1st nn, 95% 2nd nn, 65% 3rd nn, and 19% placement have been described previously (Custo et al., 2009; Fuchs et al., 2002).

A sensitivity matrix for the subject-specific head model is calculated using a finite-element forward light model based on the diffusion approximation to the radiative transport equation using the NIRFAST software package (Dehghani et al., 2003). The sensitivity matrix provides a mapping function that transforms the measured HD-DOT data to optical parameter changes within the model. The original tetrahedral finite-element based sensitivity function is then transformed into the 3 mm isotropic voxel space (standard in current BOLD processing practices) through a weighted spatial average to create the sensitivity matrix for the subject-specific space. As a result of these procedures the image spaces for the two modalities are co-aligned.

<table>
<thead>
<tr>
<th>Optical properties of segmented head tissue for forward light model.</th>
<th>750 nm</th>
<th>850 nm</th>
<th>Index of refraction</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$\mu_a$ [mm$^{-1}$]</td>
<td>$\mu_s$ [mm$^{-1}$]</td>
<td>$\mu'_a$ [mm$^{-1}$]</td>
</tr>
<tr>
<td>Scalp</td>
<td>0.0170</td>
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<td>Skull</td>
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<td>0.94</td>
<td>0.0139</td>
</tr>
<tr>
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<td>0.004</td>
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<td>Gray matter</td>
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</tr>
<tr>
<td>White matter</td>
<td>0.0167</td>
<td>1.0908</td>
<td>0.0208</td>
</tr>
</tbody>
</table>

* Strangman et al., 2002.
** Custo et al., 2006.
*** Bevilacqua et al., 1999.
The sensitivity matrix is inverted as described previously (Dehghani et al., 2009; Zeff et al., 2007). Following the notation of Dehghani et al., 2009, we set the Tikhonov regularization constant $\lambda = 0.01$, and the spatially variant regularization parameter $\beta = 0.01$. Our reconstruction utilizes a spatial constraint to aid in estimating the orientation and resolution of the VOI of the HD-DOT system, as defined by the region of the imaging domain with high sensitivity (>50% of max sensitivity) determined through the forward light model (Fig. 1E). Sensitivity within a voxel is given by the sum over all measurements within the sensitivity matrix. To compare the image-volumes produced by the two modalities, we utilized two metrics. First, we quantified the localization error within four visual quadrants, areas within the FOV of the HD-DOT system that have high SNR. The center-of-mass (CoM) of an activation within a quadrant is calculated for each of the hemoglobin species concentrations as well as the BOLD data. The CoM error in location of an activation was computed as the Euclidian distance from the center-of-mass of the fMRI response to the center-of-mass of the HD-DOT response separately for each frame.

To quantify the localization error at every voxel within the FOV, we used phase analysis. Rather than using data from a small subset of the FOV as in the center-of-mass analysis, phase maps provide access to an error metric for every voxel within the FOV. We first define a physiologically-based constraint to the FOV for the phase analysis by calculating the spatially variant signal to noise ratio in the response defined as the ratio of the power of the Fourier transform at the stimulus frequency to the power at all other frequencies (excluding very low frequencies and harmonics of the stimulus frequency) (Saygin and Sereno, 2008; Sereno et al., 1995). In effect, this models the signal to noise as a ratio of two $\chi^2$ statistics, each with degrees of freedom equal to the number of time points. This ratio has an $F$-distribution from which we obtained $p$-values. Voxels with a $p$-value less than 0.001 were excluded in the analysis. Note that because the noise is not evenly distributed across frequencies, this is a conservative threshold estimate. The phase of the response at each voxel was calculated using methods described previously (White and Culver, 2010a). The hemodynamic delay is removed from the analysis by taking the vector average of the phase of the response to the counter-clockwise-moving stimuli with the response to the clockwise-moving stimuli after reflecting the imaginary part of the response to the clockwise stimuli about zero (Sereno et al., 1995). This way the stimulus phase is directly related to the response phase at each time point in the data. Each voxel's phase error $\Delta\theta_{\text{MRI}-\text{HD-DOT}} = \theta_{\text{MRI}} - \theta_{\text{HD-DOT}}$ is calculated as the difference between the phase measured with HD-DOT to that measured via fMRI. The norm of the gradient of the phase map $||\nabla\theta||$ at a given voxel is the magnitude of the change in retinal polar phase angle given a unit change in distance in the imaged volume. Therefore, dividing an error in phase $\Delta\theta_{\text{MRI}-\text{HD-DOT}}$ by this norm provides us with an estimate of the distance error for each voxel $\Delta d = \Delta\theta_{\text{MRI}-\text{HD-DOT}}/||\nabla\theta||$. All analyses for each subject are carried out within the subject-specific space as defined by the T1-weighted MRI.

**Results**

In response to flickering checkerboard wedges (Fig. 2) the location and spatial extent of the HD-DOT and fMRI activations within an individual subject are qualitatively similar (Fig. 3). All activations displayed are block averaged from a set of ten repetitions. For visual comparison, both the voxel and cortical surface representations were cropped to a threshold of 50% maximum response. In the parasagittal slice view of Fig. 3A, it can be seen that the responses in the
right cortex to a stimulation in the left visual field are detected in qualitatively similar locations. The activations lie within the opposite quadrant as the visual stimulus. While the HD-DOT signals lie on the same gyri as the fMRI signals, it is apparent that the activation is reconstructed towards the surface. Responses to the ventral stimuli are displayed in a dorsally located axial slice in Fig. 3B (Neurological orientation of the figures: subject left is figure left). Again, there is strong agreement between the fMRI BOLD signal and HD-DOT reconstructed hemoglobin concentrations. In the volumetric slice view, it can be challenging to visualize the full three-dimensional nature of the activation, especially the intersection with the convoluted cortical surface. Thus, Fig. 3C displays the response to stimulation within each of the four quadrants overlaid on the cortical surface as seen from behind the head. It is apparent that the spatial extents of the HD-DOT and fMRI activations are also qualitatively similar on the cortical surface. The spatial correspondence of HD-DOT with fMRI is more fully demonstrated in a movie of all the phase positions of the block-averaged periodic stimuli (Supplemental Movie 1). Qualitatively, the topography of the responses along the cortical anatomy agrees strongly throughout the entirety of the stimulus presentation; both modalities continually exhibit a response in the opposite quadrant of the visual cortex from the quadrant of the visual field.

With the HD-DOT and fMRI data sets co-registered we can examine the temporal hemodynamics within a single voxel of visual cortex. Both modalities present signals with a strong periodicity and high contrast to noise ratio (CNR) (Fig. 4A; red: oxygenated hemoglobin ($\Delta$HbO$_2$), blue: deoxygenated hemoglobin ($\Delta$HbR), and green: total hemoglobin ($\Delta$HbT), black: fMRI-BOLD). The high contrast to noise ratio is also evident in the Fourier transform of hemodynamic time course with a strong peak at the stimulation frequency, a smaller peak at the first harmonic and lower relative power at other frequencies (Fig. 4B). Even though the HD-DOT and fMRI recordings are taken during different sessions, the responses within the same single voxel for each modality, are highly correlated (Fig. 2b).

Similar spatial results were observed for all subjects recorded (Fig. 5, for clarity, only the HbR reconstructed response is shown for HD-DOT). For a comparison of image quality on every subject for each HD-DOT contrast, see Supplemental Fig. 1. In Subject 5, the lower quadrants are nearly missing in the HD-DOT image. This is due to low SNR on the ventral part of the pad for that subject. In general, dorsally located activations for HD-DOT had better contrast-to-noise and better agreement with the BOLD response than ventral activations.

To quantify the localization error of HD-DOT reconstructions relative to the fMRI BOLD signal, we calculated the CofM error as the Euclidian distance between the CofM of the activations. The average CofM errors across all subjects are: $\Delta$HbO$_2$ 4.9+/-1 mm, $\Delta$HbR 5.4+/-2 mm, $\Delta$HbT 4.7+/-1 mm, with an average across all hemoglobin contrasts of: 5.0+/-1 mm. The errors for the dorsal quadrants (4.2+/-1 mm) are lower than for the ventral quadrants (5.9+/-1 mm) (Fig. 3, Fig. 5, Supplementary Fig. 1).

Qualitatively, closer inspection shows that the HD-DOT signal diminishes with depth into the sulcal folds as compared to fMRI (Fig. 6); the HD-DOT response is contained almost entirely within...
the fMRI response area, and co-localized at the surface ridges of the cortical gyri.

To determine the image quality throughout the field-of-view of the imaging pad, we applied a more continuous metric of response activation using Fourier phase analysis. We calculated phase maps for both fMRI (Fig. 7a) and HD-DOT (Fig. 7c) that revealed the pinwheel structure of the stimulus (Fig. 7b). For clarity, only the HbO₂ contrast data is shown. The circular correlation coefficient (Lee and Fisher, 1983) CCC of the phases of the different modalities shows a strong correlation of 0.50 (Fig. 7d). This provides an assessment of how correlated HD-DOT and fMRI responses are for stimuli throughout the entire visual field of view.

We also used the phase analysis to evaluate the localization error between HD-DOT and fMRI. The phase error between the HD-DOT and fMRI was calculated for each voxel and converted into a distance error. After converting to a distance error, the map of errors of the imaging pad on Subject 2 is smooth across the field of view and not dependent on the location of the voxel relative to the stimulus (Fig. 7f). A map of the localization error across the whole field of view was created for each subject (Supplementary Fig. 2). After removing outliers and averaging over all voxels for all subjects, the average localization error for the HD-DOT imaging pad is ΔHbO₂ 4.2+/−1 mm, ΔHbR 4.2+/−2 mm.

ΔHbT 4.8+/−1 mm, giving an overall average of 4.4+/−1 mm. A summary of all localization errors is provided in Table 2.

Discussion

Our results provide an image quality benchmark test of HD-DOT via a voxel-to-voxel comparison with fMRI. We demonstrate that the location and spatial extent HD-DOT activations are qualitatively similar to fMRI activations throughout the accessible portions of the visual cortex (Fig. 3 and Fig. 5). Since the data sets were co-registered time traces of responses within individual voxels could be compared. Consistent with previous fMRI and NIRS studies, the signals are highly correlated between the two modalities (Fig. 4). Qualitatively, the cortical topography of the responses agree strongly throughout the phase cycle of the stimulus presentation (Supplemental Movie 1). Quantitatively, we find the correspondence of quadrant locations between the two modalities to have an average center-of-mass error of 5.0+/−1 mm. Using phase maps to calculate localization error in every voxel across the field of view we measured an average localization error in the HD-DOT of 4.4+/−1 mm. Through examination of the voxel overlap between HD-DOT and fMRI, we find excellent agreement along the cortical ridges (Fig. 6), while the concurrence falls off as the cortex folds away from the scalp surface and the sensitivity of the HD-DOT system diminishes.

In this study a high density of overlapping measurements (Zeff et al., 2007) enabled a quantitative voxel-wise comparison of image quality of co-registered HD-DOT and fMRI over an extended region of cortex. Previous comparisons between fMRI and DOT have examined the co-localization between the two modalities in visual (Zhang et al., 2005) and motor (Joseph et al., 2006; Koch et al., 2010) cortices. However, these studies did not perform quantitative voxel-by-voxel comparisons. The motor papers used qualitative means to infer image agreement by projecting the fMRI data to individual DOI measurements at the scalp surface (Joseph et al., 2006), or used a generic head model for the forward light modeling rather than using the subject-specific anatomy to generate a realistic photon-propagation model (Koch et al., 2010) and then projected the data from the ‘model head’ onto the subject specific MRI anatomy. In contrast this visual study compared fMRI activations to DOT reconstructions in a subject-specific brain volume. The use of the subject-specific anatomy guarantees true alignment with the fMRI data set leading to the gyral specificity apparent in the reconstructions. The use of phase encoded retinotopy establishes the co-localization of HD-DOT and fMRI over contiguous regions of the visual cortex.

While topographic DOI methods remain in wide use (Cui et al., 2011; Franceschini et al., 2006; Khan et al., 2010; Lindauer et al., 2001; Ou et al., 2009; Sakatani et al., 2007; Tian et al., 2010), the image quality improvements of HD-DOT methods have been established by several research groups (Joseph et al., 2006; Koch et al.,...
errors greater than +/− each voxel within the scalp. Cortical depth of 0.75 cm, or about 1.75 cm below the surface of the cap. While elegant depth compensation algorithms have been proposed to increase the quality of depth localization of DOT (Niu et al., 2010), a full comparison of such methods is beyond the scope of the present work. The spatially-dependent regularization techniques herein used for human DOT have been detailed in references: (Dehghani et al., 2009; White and Culver, 2010a; White et al., 2009; Zeff et al., 2010; Zeff et al., 2007). Quantitative comparative studies, using both simulated and in vivo data, have demonstrated clear improvements in resolution of the tomography approach (Boas et al., 2001, 2004a, b; Koch et al., 2010; White and Culver, 2010b). While elegant depth compensation algorithms have been proposed to increase the quality of depth localization of DOT (Niu et al., 2010), a full comparison of such methods is beyond the scope of the present work. The spatially-dependent regularization techniques herein used for human DOT have been detailed in references: (Dehghani et al., 2009; White and Culver, 2010a; White et al., 2009; Zeff et al., 2007). Incorporation of additional overlapping mappings provides the most straightforward approach to further increases in image quality and extension of the imaging domain. Specifically, it has been shown (Dehghani et al., 2009) that the inclusion of larger source-detector pair separations leads to improved depth profiling and sensitivity. As the measurement pair distances increase, the SNR falls off exponentially (Boas et al., 2001). Thus, if DOT systems are developed with better sensitivity, they might sample deeper and possibly improve upon the resolution and localization accuracy of current systems. In these five subjects, the current system is capable of maintaining good localization (<5 mm error) down to a cortical depth of 0.75 cm, or about 1.75 cm below the surface of the scalp.

In this study, we have used NIRFAST, a finite-element solver of the diffusion equation (Dehghani et al., 2003) to construct the model solution to the forward light scattering problem. Alternatively, a solution to the full radiative transport equation (RTE) can be obtained via Monte Carlo (MC) methods. Recent comparative studies between RTE and diffusion modeling have shown that the diffusion equation is sufficient for general considerations of DOT in healthy adults (Custo et al., 2006). The favorable comparative fMRI-DOT results of this paper using diffusion modeling further support the use of diffusion modeling. However, there may be clinical situations, for example patients with cerebral edema, in which MC modeling will have significant advantages. While historically MC modeling methods were very time consuming, recent advances in GPU technologies hold promise for future studies to compare the image quality of reconstructions obtained via both RTE and diffusion modeling.

General Linear Models (GLMs), while standard in the fMRI literature, are not yet standard with DOT. The localization error of the centers of mass of the quadrants was calculated using the reconstructed hemoglobin concentrations for HD-DOT and the raw BOLD signal for fMRI. No noise-weighting correction was made for either modality. However, in the phase-map analysis, we did calculate the statistical significance for the measured response within each voxel with a conservative weighting by the noise (see Methods). The phase map analysis conducted herein provided a similar analysis as GLM by calculating both the correlation magnitude and phase of the response to a periodic stim. More generalized statistical parametric mapping (SPM) methods have been developed for NIRS data (Villringer et al., 1993; Ye et al., 2009), that address the uneven spatial sampling of NIRS topography. In principal, due to the improved spatial sampling of DOT compared to NIRS (White and Culver, 2010b), the extension of SPM methods used in fMRI for calculating roughness and addressing the multiple comparisons problem should be more straightforward for DOT than NIRS.

In our analysis we smoothed (Gaussian kernel with FWHM = 13 mm) the fMRI data to match the DOT imaging point spread function based on the rationale that we wanted to compare the differences in the modalities, not differences in the point spread functions. However an alternate approach would be to use standard fMRI processing which typically uses 8 mm Gaussian smoothing (Wenger et al., 2004). To evaluate the localization errors with normal fMRI processing we re-ran the quadrant localization error analysis with a set of fMRI data smoothed to 8 mm. The resulting average localization error of the quadrants was 4.5 +/− 2 mm, (within the error of the results from 13 mm smoothing). Thus the small localization errors between fMRI and HD-DOT hold even with standard fMRI processing.

<table>
<thead>
<tr>
<th>Subject</th>
<th>Center of mass error</th>
<th>Phase map error</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(mm, mean +/− SD)</td>
<td>(mm, mean +/− SD)</td>
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<tr>
<td>HB02</td>
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The fMRI and HD-DOT data reside within the same model space. Thus, the magnitude of an error in the location of activation or a phase map value can be directly compared. The purpose of the current paper is a within-subject voxel-to-voxel comparison of HD-DOT and fMRI. We used five subjects to obtain a sampling across head types. We provide group averaging of the within subject errors between fMRI and DOT (errors in CoM and in the phase plots). Group averaging of the DOT data on a voxel by voxel basis requires spatial normalization that, while standardized for fMRI, has not yet been established in the literature for DOT. However, spatial normalization of DOT data will be an important area of future work to enable voxel-wise group level comparisons, data averaging, and better alignment with fMRI processing.

A general limitation of DOT systems is that while the upper portions of the cortical hull can be sampled, deep brain structures and deep cortical tissue along midline (e.g. area V1 of the visual cortex) cannot be sampled. To address this issue we constrained our comparison of fMRI and DOT to the regions of the tissue that were well sampled by the DOT imaging array. This limited the potential confound of having displaced or mismatched sampling volumes. Retinotopic stimuli, even at a single focal location in the visual field of view, lead to multiple activations within the visual cortex due to the presence of multiple representations (i.e. V1, V2, V3 etc...). Qualitatively, the cortical topography of the responses are in agreement throughout the accessible visual areas (Supplemental Movie 1). The phase analysis confirms this quantitatively.

Conclusion

These co-registered retinotopic results establish that HD-DOT methods can map brain function with good (~5 mm localization error) voxel-to-voxel correspondence with fMRI. By using a phase-encoded visual paradigm this study not only compared point-activations but also full maps of visual cortex, a standard analysis in fMRI. By using a phase-encoded fMRI paradigm this study not only compared point-activations but also full maps of visual cortex, a standard analysis in fMRI. Encoded visual paradigm this study not only compared point-activations but also full maps of visual cortex, a standard analysis in fMRI. Encoded visual paradigm this study not only compared point-activations but also full maps of visual cortex, a standard analysis in fMRI.

Supplementary materials related to this article can be found online at doi:10.1016/j.neuroimage.2012.01.124

Role of the funding source

This work was supported in part by NIH grants R01-EB009233 (J.P.C.), T90-DA022871 (Imaging Science Fellowship, B.W.R.) and a Fulbright Science and Technology Ph.D. Award (S.L.F.). The funding source had no involvement in the study design, collection, analysis, interpretation of the data, writing of the paper, or decision to submit the paper for publication. J.P.C and Washington University have financial interests in Cephalogics LLC based on a license of related optical imaging technology by the University to Cephalogics LLC.

Acknowledgments

We thank Gavin Perry and Martin Olevitch for help with HD-DOT instrumentation and software; and Fran Miezin for developing the BOLD sequence we used; Donna Dierker for help and patience with Caret and FreeSurfer software; and Tracy Nolan and Linda Larson-Prior with some MRI data acquisition.

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