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Abstract. Measurement of photon penetration in biological tissues is a central theme in optical imaging. A great number of endogenous tissue factors such as absorption, scattering, and anisotropy affect the path of photons in tissue, making it difficult to predict the penetration depth at different wavelengths. Traditional studies evaluating photon penetration at different wavelengths are focused on tissue spectroscopy that does not take into account the heterogeneity within the sample. This is especially critical in shortwave infrared where the individual vibration-based absorption properties of the tissue molecules are affected by nearby tissue components. We have explored the depth penetration in biological tissues from 900 to 1650 nm using Monte-Carlo simulation and a hyperspectral imaging system with Michelson spatial contrast as a metric of light penetration. Chromatic aberration-free hyperspectral images in transmission and reflection geometries were collected with a spectral resolution of 5.27 nm and a total acquisition time of 3 min. Relatively short recording time minimized artifacts from sample drying. Results from both transmission and reflection geometries consistently revealed that the highest spatial contrast in the wavelength range for deep tissue lies within 1300 to 1375 nm; however, in heavily pigmented tissue such as the liver, the range 1550 to 1600 nm is also prominent. © 2016 Society of Photo-Optical Instrumentation Engineers (SPIE) [DOI: [10.1117/1.JBO.21.12.126006](https://doi.org/10.1117/1.JBO.21.12.126006)]

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1 Introduction

Achieving deep photon penetration is one of the main objectives for *in vivo* optical imaging and treatment with light-based therapies. The use of visible light is convenient but only feasible for skin and other directly accessible tissues (i.e., esophagus and colon¹) and is not suitable for deep tissue penetration due to the presence of strong absorption from endogenous chromophores (blood)² and scattering from tissue components. Near-infrared (NIR) light is used to alleviate this scattering effect and is more suitable for deep tissue samples. Two optical windows exist within the NIR range: the first window ranges from 650 to 950 nm and is known as the traditional NIR or optical window I, and the second window ranges from 950 to 2200 nm and is referred to as shortwave infrared (SWIR), extended NIR, or optical windows II to IV.^{3,4} While photon penetration in the traditional NIR range is well studied^{5–7} due to the wide availability of silicon-type detectors, light sources, and optical elements, the field of SWIR is relatively less explored and demands different types of instrumentation.^{8,9} Recent rapid developments in SWIR optics, and especially efficient InGaAs sensors between 900 and 1650 nm and SWIR lasers, have made this spectral range attractive for a variety of deep tissue therapeutic applications¹⁰ and deep tissue imaging.^{11,12}

Recent studies have confirmed the feasibility of deep tissue imaging beyond 1000 nm¹¹ with different contrast agents that emit in this wavelength such as quantum dots,^{13,14} carbon

nanotubes,^{15–17} and other emitters, including small molecules.^{18–20}

A persistent challenge in SWIR imaging is identifying a set of wavelengths where tissue has the highest transparency. The answer for this question would allow the development of deeper imaging techniques with optimized light penetration and encourage the development of contrast agents that are optically active at the specifically identified (absorption and emission) wavelengths. The prediction of transparency at different wavelengths is complicated due to the contributions of several parameters, including Mie scattering,⁷ variations in absorption in accordance to the Beer-Lambert law, and anisotropy of tissue components.²¹ This complexity aggregates the demand for the experimental evidence of tissue transparency and the wavelength-depth relationship in SWIR.

Traditional studies evaluating photon penetration at different wavelengths have mostly focused on tissue spectroscopy that does not take into account the heterogeneity within the sample. Addressing heterogeneity is especially critical in SWIR where the individual vibration-based absorption properties of the tissue molecules such as water and lipids are affected by ions and nearby tissue components.²²

We have previously demonstrated with the use of tissue-mimicking phantoms (intralipid/India ink) that the wavelength-depth relationship can be evaluated via hyperspectral approaches using quantitative measurements of spatial Michelson contrast as a function of wavelength and depth.²³ We also demonstrated that at low thicknesses of the phantom a reasonable contrast of the objects can be achieved at any wavelength between

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700 and 1400 nm and between 1500 and 1600 nm in a transmission geometry.²³ Such contrast can be achieved mostly between 1300 and 1350 nm at larger thicknesses. This previous study relied on a point-to-point scanning process that was inherently slow (up to 30 min for a scan), which prevented the investigation of a freshly excised, unfixed biological tissue. In this study, we have modified the hyperspectral imaging (HSI) system to enhance image acquisition speed. With the new system, we were able to investigate the tissue in 3 min or less, thus preventing imaging from many artifacts such as drying while maintaining high spatial and spectral resolution.

The goal of this work was to identify the wavelengths in SWIR at which the tissue has the highest transparency; therefore, we selected Michelson spatial contrast as a metric of light penetration. First, we designed and calibrated an HSI microscope to generate images free of chromatic aberrations. Second, we optimized the imaging conditions to record the data within 3 min to minimize the destruction of the biological tissue. We then investigated contrast as a function of the wavelength in phantoms and several types of tissue in transmission and reflection geometry. We also performed simulation studies to validate our experimental data.

2 Materials and Methods

2.1 Instrument Design

The design of the HSI system shown in Fig. 1 featured an SWIR sensitive two-dimensional (2-D) charge-coupled device (CCD) InGaAs camera NIRvana 640 (Princeton Instruments) for image acquisition mounted on the top of an Olympus BX51 microscope equipped with a triaxis PC-controlled stage (Marzhauser Wetzlar). A standard 100-W halogen microscope lamp (Olympus) with the IR filter removed was placed in front of the entrance imaging spectrograph iHR320 (Horiba Inc.) as the light source. Continuous broadband light passed through a 600-groove/mm, blazed at 1000 nm grating, which dispersed light from 900 to 1650 nm into narrow bands. Narrow-band light was further

delivered to the sample through a liquid light guide (Newport Corp.) with high transmission in the 420 to 2000 nm range. The design also included a calibration setup composed from a xenon lamp, double grating monochromator, liquid nitrogen-cooled InGaAs diode array CCD camera Symphony II, and a cuvette holder with a standard made from Spectralon with high reflection (all components from Horiba Inc.).

Both transmission and reflection geometries were used for data acquisition. In the transmission geometry, the light delivered by the light guide passed through a microscope condenser and onto the sample. In the reflection geometry, the light delivered by the light guide projected onto the sample passed through a beamsplitter and a chromatic aberration-free 15× reflective objective (Beck) attached to the microscope through a set of adapters (Thorlabs) (Fig. 1). The transmitted or reflected light was collected with the same objective and channeled into the 2-D camera. Integration time, slit size, and wavelength increments were set in the FluorEssence software (Horiba). Image acquisition was collected through the LightField software (Princeton Instruments) and was synchronized with the wavelength selection.

2.2 Objective Selection for Chromatic Aberration-Free Imaging

The imaging system was corrected for chromatic aberrations, which is a known problem in HSI.²⁴ In the visible and NIR ranges, chromatically corrected objectives such as apochromats can mitigate the image distortion caused by chromatic aberrations. These objectives accommodate low dispersion glass elements to neutralize the difference in refracting indices at variable wavelengths; however, apochromatic objectives fully corrected in SWIR are not commercially available. One noticeable exception exists and is collectively known as a reflective objective. These types of objectives, first developed by Schwarzschild (circa 1904),²⁵ are based on a mirror system with no glass elements. Due to the all-reflecting optics, the objectives are chromatic aberration-free in the very broad spectral range from UV

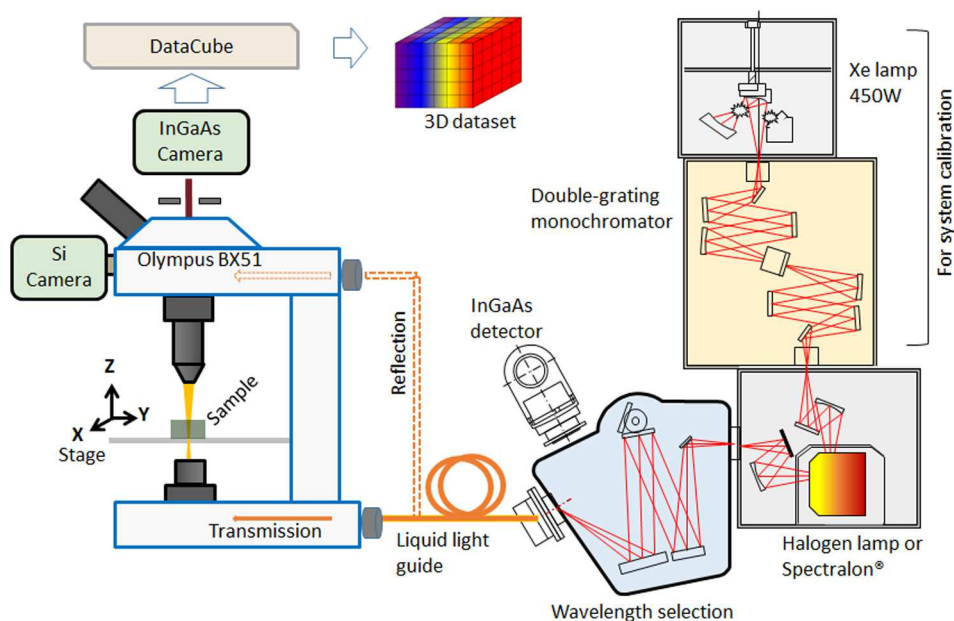


Fig. 1 Schematic of the HSI system for contrast measurement in phantoms in the spectral range 900 to 1650 nm.

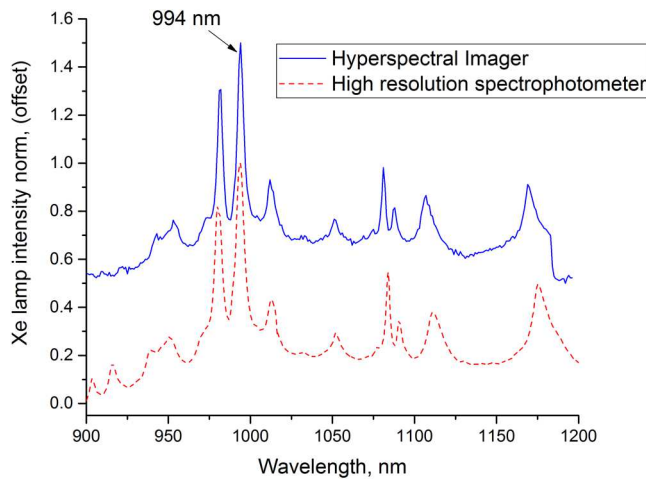


Fig. 2 Spectral profiles of the xenon lamp as measured by the high-resolution calibrated spectrophotometer and HSI. Similar appearance of the spectra collected from the imager and from the high-resolution spectrophotometer indicated spectrally calibrated HSI system and also demonstrate sufficient spectral resolution. All peaks are within ± 2 nm of each other.

(<300 nm) nm to IR (>10 μm). In the current work, we used a 15 \times reflective objective (working distance 23.5 mm and NA = 0.28, Beck) for all image acquisitions.

2.3 Spectral Calibration and Spectral Resolution

Spectral calibration was performed to ensure the correct spectral response of the HSI system. First, the detector Symphony II was calibrated using a neodymium doped glass that emits at 1052 nm when excited at 575 nm. Second, a xenon lamp emission was recorded using the calibrated Symphony II detector with a high spectral resolution (increment 0.25 nm, integration time 5 ms, emission slit 1 nm) to set the position of a reference peak at 994 nm (Fig. 2). Finally, the HSI was spectrally calibrated

with a 994-nm emission line from the xenon lamp (increment 1.0 nm, emission slit 1 nm).

Spectral resolution has an important effect on HSI performance, and determining optimal resolution is critical in HSI measurements. Low spectral resolution (>20 nm bandwidth) limits the ability of the method to resolve closely spaced spectral features, while high spectral resolution (<3 nm) requires significant increases in the integration time, which lowers output. Spectral resolution in the presented HSI system was adjusted by the width of the slits in the monochromator and the selection of the wavelength increment. The influence of both parameters on the broadening of the signal measured as full-width-at-half-maximum (FWHM) from a narrow linewidth 1064 nm laser is illustrated in Figs. 3 and 4. The results demonstrate a strong outcome of the slit width on FWHM, with fourfold FWHM increase from 1 nm slit to 8 nm (Fig. 3). The effect of the wavelength increment was somewhat less significant (1.5-fold increase FWHM from 1 nm increment to 4 nm, Fig. 4). The optimized image acquisition parameters were set to 5 nm slit and 5.27 nm increment.

2.4 Sample Preparation

The phantom's stock solution was composed of 18.75 mL intralipid (20% fat emulsion, Fresenius Kabi), 0.62 mL India ink (2% in water, Speedball), and 300 mL water.²⁶ Total amount of lipid was 1.1 w. %, which is similar to the amount typically seen in freshly excised chicken breast ($0.98 \pm 0.23\%$).²⁷ The absorption coefficients for individual components such as water and lipids in SWIR have been measured in our previous work.³ Based on the literature data, the μ_a values of intralipid phantoms are within 0.15 to 0.17 at 1100 nm and depend strongly on the wavelength following the shape of the absorption spectra.²⁸ The values μ_a' decrease monotonically on average from 4 cm^{-1} at 1100 nm to 1 cm^{-1} at 1700 nm.

The phantom assembly utilized a grid made from a stainless steel card (Fotofab Inc.) perforated with evenly spaced holes of about 230 μm diameter attached beneath a plastic polystyrene

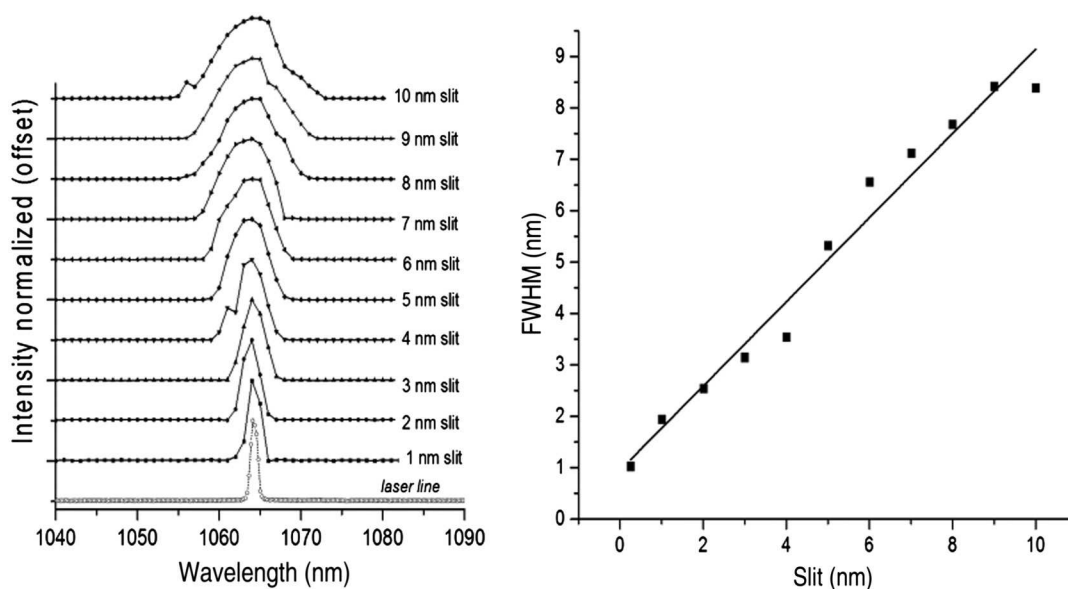


Fig. 3 Effect of slits on the HSI spectral resolution. FWHM of the recorded laser line at 1064 nm shows linear broadening of the signal at larger slits. Laser line (true line) was recorded with a high spectral resolution spectrophotometer. The spectra are normalized and plotted with an offset.

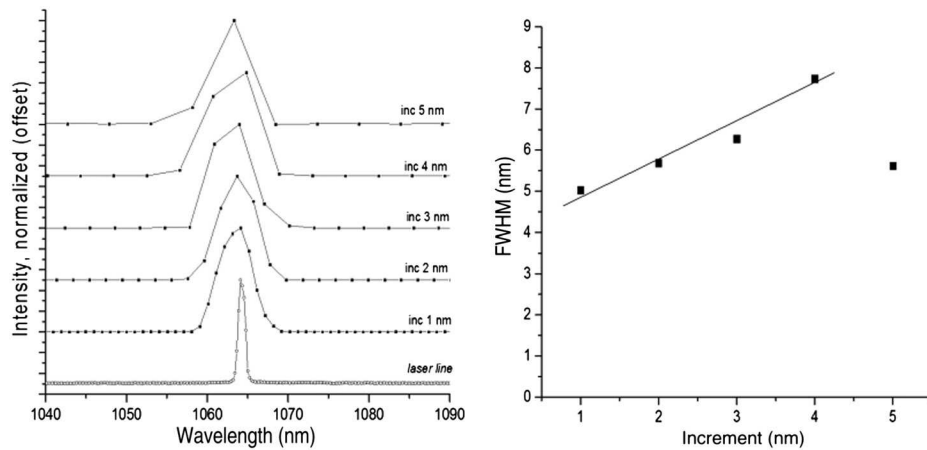


Fig. 4 Effect of the wavelength increment on the HSI spectral resolution. FWHM of the recorded laser line at 1064 nm shows broadening of the signal at larger increments. The outlier at 5 nm was due to low sampling rate below Nyquist value. The spectra are normalized and plotted with an offset.

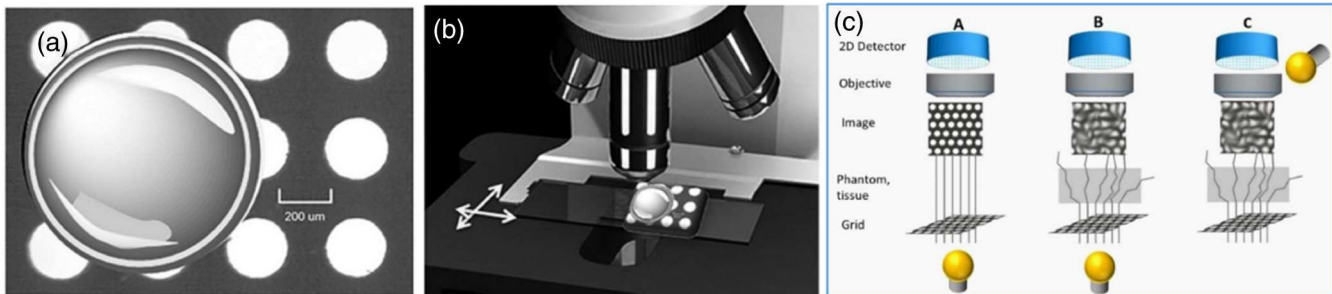


Fig. 5 (a) Phantom/tissue setup (closed-up view), (b) phantom/tissue placement, (c) principle of phantom and tissue setup for contrast measurements with hyperspectral imaging showing transmission (A and B) and reflection (C) mode.

petri-dish (35 mm × 10 mm) that was filled with the phantom stock solution (Fig. 5). The stock ranged from about 2 to 7 mL, corresponding to depths of 3.0 to 6.5 mm in the middle of the petri-dish, where the meniscus effect is minimal.

2.5 Tissue Preparation

Fresh chicken tissues including thigh, breast, and liver were obtained from a local supermarket (Whole Foods, St. Louis, Missouri), cut in small pieces and placed between two 25-mm-diameter cover glasses (Deckglaser) to ensure uniform distribution of thickness and prevent from drying. The thickness of the tissue was determined by measuring the distance between the two cover glasses using a digital caliper. The tissue was then placed on a 230-μm-diameter grid for contrast measurement.

2.6 Data Acquisition

The focus was adjusted manually prior to image sequence acquisition using a PC-controlled stage (Marzhauser Wetzlar) at 1300 nm to achieve the best visual contrast of the grid. A number of 180 collected frames corresponded to the spectral resolution of 5.27 nm. Dispersed light was coupled with a light guide with the distal end connected to either a transmission or a reflection port of the microscope. The power of the lamp, the slit width on the spectrograph, and the integration time of the camera for each image captured were optimized to achieve

an adequate signal strength (>1000 counts, with the noise level <100 counts). Using these optimized conditions, the integration time was set to 1 s, which brought the total time of recording to 3 min.

2.7 Data Processing

The set of collected frames were analyzed with ImageJ using the OpenSPE extension for direct export of frames. Regions of interest (ROIs) in the light and dark areas in the image (holes or the metal parts) were selected as shown in Fig. 6. Approximate maximum and minimum light intensities were obtained for each frame (wavelength) by averaging the light intensities in the ROIs selected with the plot Z-axis profile function. The Michelson contrast²⁹ was then calculated according

$$\text{contrast}(\lambda) = (I_{\text{light}} - I_{\text{dark}}) / (I_{\text{light}} + I_{\text{dark}}), \quad (1)$$

where I_{light} and I_{dark} correspond to the average light and dark intensities of the selected areas. A contrast value of 0 corresponds to no contrast and a contrast value of 1 corresponds to maximal contrast.

3 Results and Discussion

HSI-SWIR is a relatively established technique. Historically, satellite-based remote sensing was the first large-scale application of HSI-SWIR.³⁰ In the last decade, with the advances in

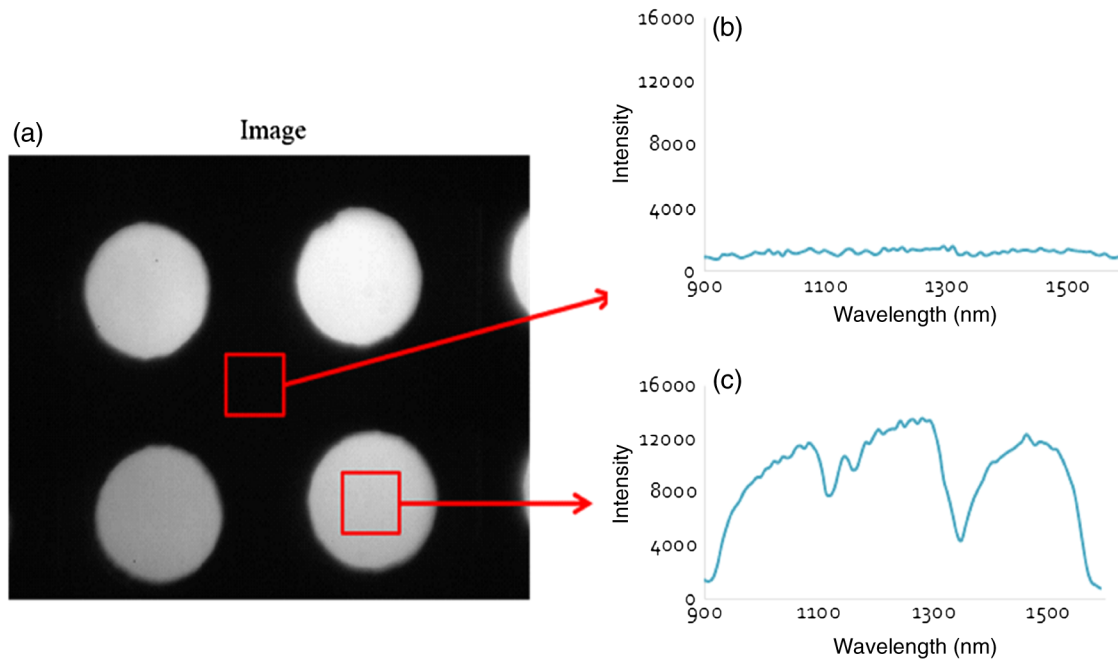


Fig. 6 Calculation of the Michelson contrast from the datacube and ROIs corresponding to the light and dark areas in the image. (a) Representative frame of a grid covered with a tissue-mimicking phantom. (b) Average intensity of the dark ROI over the spectral range from 900 to 1650 nm. (c) Average intensity of the ROI inside the hole over the spectral range from 900 to 1650 nm.

InGaAs technology, the applications of HSI-SWIR have been significantly broadened and currently cover a variety of research fields such as security,⁴ plant science,^{31,32} agriculture,³³ and food analysis.³⁴ The latter includes characterization of the food products,³⁵ detection of contaminants,³⁶ and quality control.³⁷ Given the better penetration of SWIR photons in scattering media, SWIR-HSI has also been recognized as a viable tool to monitor objects through thick smoke,³⁸ detect hidden damages in fruits,³⁹ and diagnose and demarcate deep bruises in humans.⁴⁰ In this work, we employ HSI-SWIR to measure the wavelength-dependent photon penetration in phantoms and different biological tissues.

Our method of measuring penetration depth was based on the quantitative evaluation of the spatial Michelson contrast introduced in our earlier publications.^{23,41} In this method, a sample of interest was placed on top of a grid, as shown schematically in Fig. 5, and the contrast was measured (as shown in the previous section and Fig. 6). Image contrast was independent of wavelength for a blank sample provided that the system was corrected for chromatic aberrations and the wavelength of detection was within detector sensitivity. The blank sample refers to the measurements from the Fotofab card without any of the intralipid/India stock solution. The contrast of the blank grid sample under transmission geometry in Fig. 7(a) shows this independence by exhibiting a steady contrast within the 0.88 to 0.91 range (1.0 under ideal conditions) from 935 to 1625 nm. The contrast substantially decreases at the margins due to the drop in sensitivity of the InGaAs sensor at these wavelengths, which makes the measurements below 535 and above 1625 nm no longer reliable. The value of the measured contrast was a little lower (less than 10%) than the expected value of 1.0 under ideal conditions.

The data for the transmission and reflection modes presented in Figs. 7(a) and 7(b) show general trends between contrast and

wavelength consistent for both geometries. Significantly lower depth penetration in the reflection geometry when compared to the transmission geometry was due in part to an average doubling in path length within the tissue. The increased photon path length of the reflection geometry leads to relatively more absorption events, which results in a lower I_{light} . Fresnel reflection also contributed to the lower depth penetration of the reflection geometry because of the involvement of a greater number of surfaces with different refractive indices.^{42,43} The lowest contrast within the studied spectral range was at 1400 to 1500 nm due to the strong water absorption at these wavelengths. The contrast at other wavelengths depended on sample thickness. At lower thickness, the highest contrasts were centered at two spectral bands: 1300 to 1375 nm and 1575 to 1625 nm. As the thickness of the phantoms increased, the contrast at the 1300 nm band became much more prominent than at 1600 nm [Fig. 7(c)].

These results reflect a compromise between the absorption properties of the chromophores and scattering. While the absorption coefficients vary greatly over the SWIR spectrum, the scattering coefficient of tissue decreases monotonically according to the Mie theory.⁴⁴ At 1600 nm, the absorption coefficient of water, the dominant chromophore, is an order of magnitude higher than that at 1300 nm 5.7 cm^{-1} versus 77.5 cm^{-1} ,⁴⁵ however, the scattering at 1600 nm is expected to be substantially lower than at 1300 nm. Under ideal Rayleigh scattering conditions, scattering is proportional to λ^{-4} .³³ With particles size similar to the wavelength of the incident electromagnetic radiation (Mie conditions), the scattering in SWIR is less wavelength-dependent and proportional from $\lambda^{-1.5}$ to $\lambda^{-0.6}$ depending on the type of tissue and often method of measurements.^{8,46} Based on this evaluation, thick tissue is expected to have the highest transparency at 1300 and 1600 nm, and this result is confirmed by our experimental data [Fig. 7(c)] and simulation (see Figs. 9 and 10).

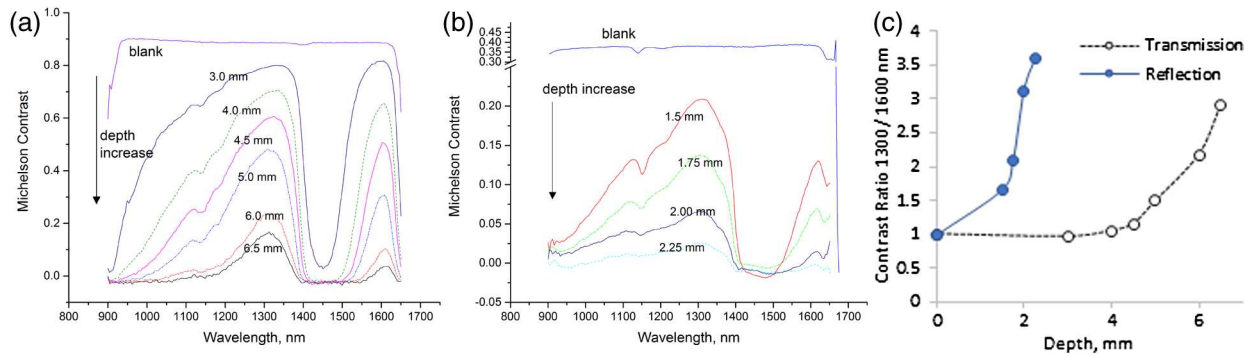


Fig. 7 Michelson contrast values as a function of wavelength from 900 to 1650 at different depth in intralipid/India ink phantoms from 0 to 6.5 mm for intralipid/India ink phantoms with 230- μ m-diameter grid. (a) Transmission geometry. (b) Reflection geometry. (c) Ratio of two Michelson contrast values (1300 to 1600 nm) at different depths.

Figure 8 summarizes the contrast values for representative tissues such as chicken thigh, breast, and liver in the transmission and reflection geometries at different thickness. Different organs have a distinct chemical composition (fat and proteins),

presence of pigments (i.e., hemoglobin), and connective tissue that contributes to scattering.

All tissues presented a loss of contrast at around 1400 to 1450 nm due to the high level of water in the tissue (>70%

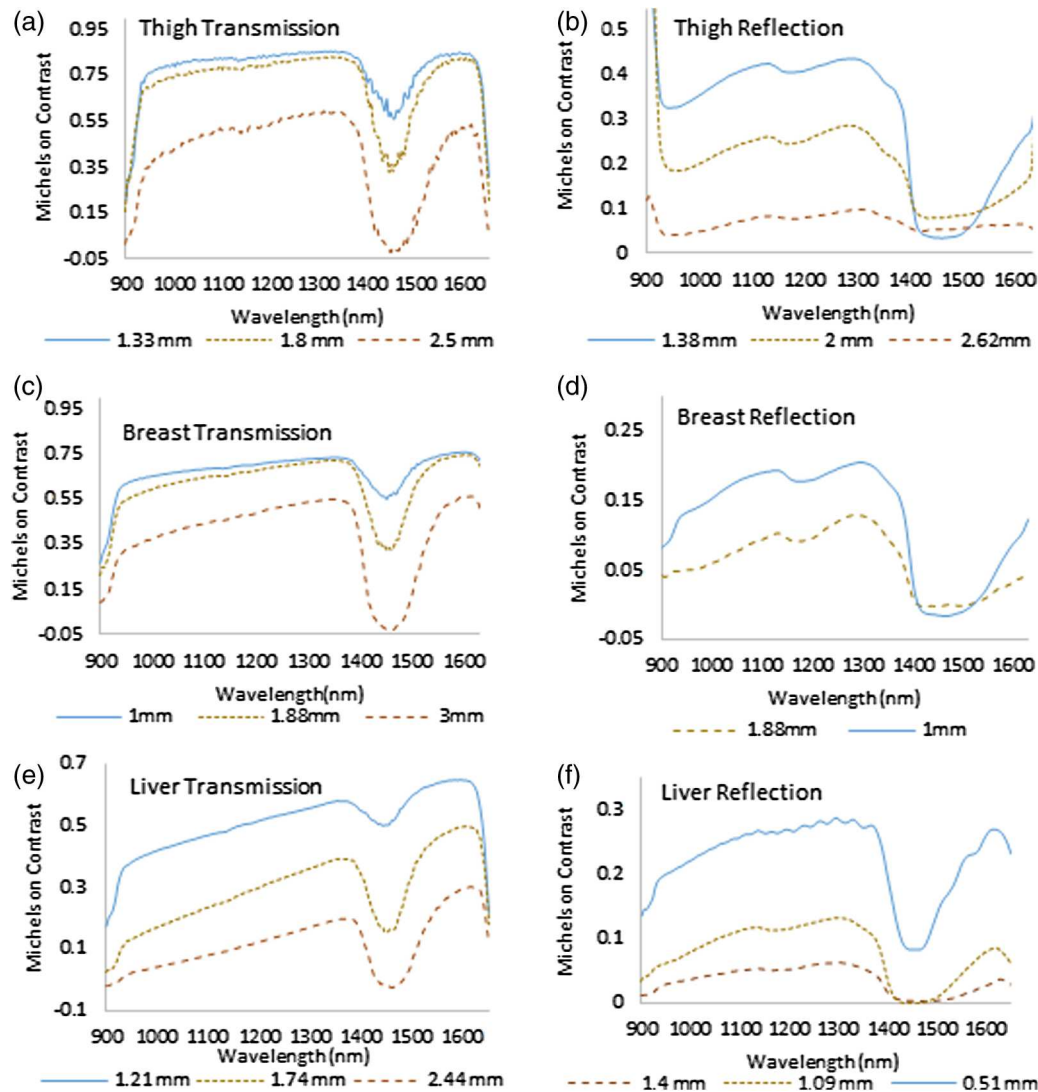


Fig. 8 Contrast data at different type of tissues and depth across wavelengths from 900 to 1650 nm. (a) Chicken thigh transmission, (b) chicken thigh reflection, (c) chicken breast transmission, (d) chicken breast reflection, (e) chicken liver transmission, and (f) chicken liver reflection. 230- μ m-diameter grid.

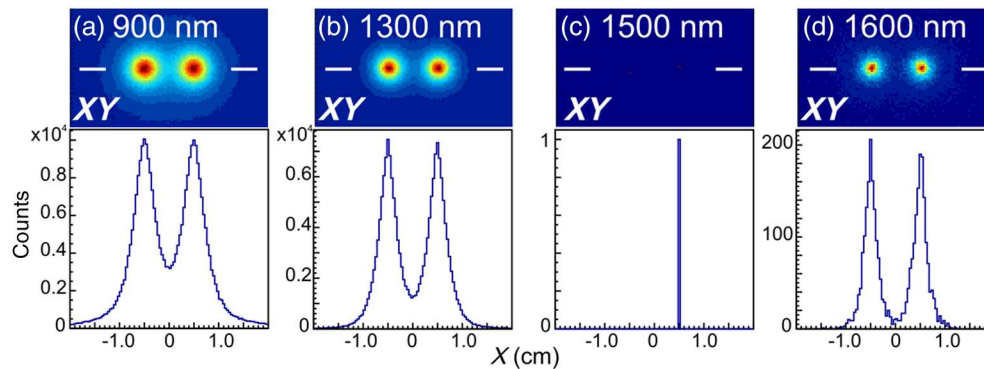


Fig. 9 Representative Monte-Carlo simulation of two parallel photon beam distributions in biological tissue in transmission geometry. (a) 900 nm: $\mu_a = 0.07 \text{ cm}^{-1}$, $\mu'_s = 1.3 \text{ cm}^{-1}$; (b) 1300 nm: $\mu_a = 0.8 \text{ cm}^{-1}$, $\mu'_s = 1.07 \text{ cm}^{-1}$; (c) 1500 nm: $\mu_a = 13 \text{ cm}^{-1}$, $\mu'_s = 0.95 \text{ cm}^{-1}$; (d) 1600 nm: $\mu_a = 5 \text{ cm}^{-1}$, $\mu'_s = 0.85 \text{ cm}^{-1}$.

in most of the soft tissues⁴⁷, which correlated with the trend observed with the intralipid/India ink phantoms. The contrast steadily increased with the wavelength, providing the largest contrast at 1300 to 1375 and 1575 to 1625 nm, which also correlated with the trend observed with the phantoms. These two bands showed similar contrast in the transmission geometry, which is important for light-based therapies. The contrast at 1300 to 1375 nm was consistently higher and particularly evident at larger depths in the reflection geometry, which is more practical for medical imaging. Freshly excised chicken tissue retains most of its interstitial fluid from the lymphatic system (>70% moisture²⁷) but, in general, contains a reduced amount of blood. We expect that the presence of blood will increase the absorption coefficient due to the increased level of oxygenated hemoglobin that is still significantly absorbing in this spectral range (although less than in the NIR and visible).^{3,48}

To explain the results, we performed a Monte-Carlo simulation. Because of a lack of optical data in the studied tissues (chicken breast, thigh, and liver), Monte-Carlo simulation was performed using literature data on absorption and reduced scattering coefficients in rat skin.⁴⁹ Scattering was modeled using the standard Henyey-Greenstein phase function.⁵⁰ In this simulation, two parallel beams of photons at different wavelengths (each beam = 1×10^7 photons) entered the tissue perpendicular to the surface. Reflection/transmission coefficients were calculated using the known Fresnel equation for nonpolarized photons through infinite XY surface of 10-mm thickness. The results of such simulations in the transmission geometry are

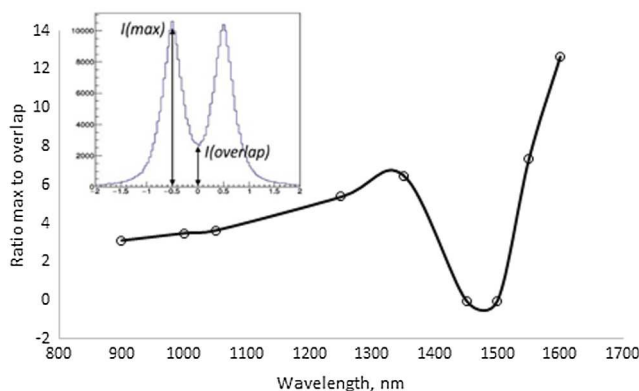


Fig. 10 Calculated resolution of two-peak histograms: $R = I_{\text{max}}/I_{\text{overlap}}$.

shown in Fig. 9. The resolution between two peaks was calculated as $R = I_{\text{max}}/I_{\text{overlap}}$ and presented in Fig. 10 as a function of wavelength. The trend of resolution versus wavelength was quite similar to what we measured experimentally in the transmission geometry with maximums at 1350 and 1550 nm. Beams were set 10 mm apart, 1×10^7 photons in each beam. Thickness of the tissue was set to 10 mm. For simplification, the values of the refractive index n and anisotropy g were considered to be independent from the wavelength in the 900 to 1600 nm range and equal to $n = 1.4$ (typical values of n for the tissue from 1.35 to 1.41⁸ and reviewed in Ref. 51), and anisotropy factor $g = 0.9$ (typical values 0.80 to 0.95⁵²). The upper panel shows the distribution of the photon from the bottom view (entrance) and top view. The lower panel shows the histogram of photons number from the middle position between the two beams.

Analysis of the tissue-based data revealed the difference between the types of the tissue. While thigh and breast showed similar contrast apparently due to the similarity in their tissue composition, the liver presented a significantly lower contrast (Fig. 11). This difference became especially noticeable at larger thickness. The liver is highly vascularized compared to the thigh and the breast tissues and is therefore richer in red cells and hemoglobin. Hemoglobin is strongly absorbing in the visible but also absorbing at longer wavelengths with the absorption coefficient at 1300 nm comparable to that of water.³ Thus, despite the known low scattering coefficient of the liver compared to other tissues,⁶ the liver is substantially less transparent in the SWIR than other soft tissues such as breast and thigh.

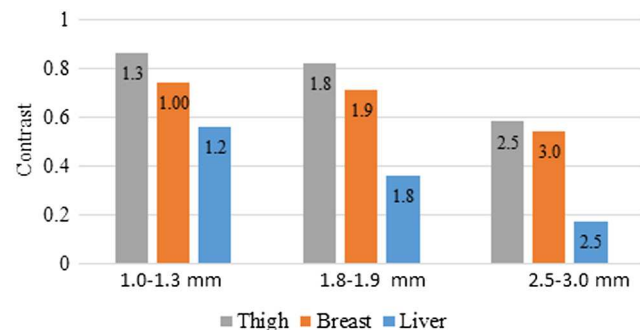


Fig. 11 Michelson contrast of biological tissues at different depth at 1300 nm. Values in the columns show the actual thickness of the tissue in mm.

There are a growing number of publications that explore the SWIR range as a new optical window II.^{13,53,54} This optical range has been explored quite recently for the diagnostics of middle ear pathologies⁵⁵ and arthritis.⁵⁶ In this regard, our study confirms that light penetration in certain spectral ranges within the SWIR range is higher than in the more traditional NIR range. In this work, we have experimentally demonstrated that the highest transparency lies in a narrow range of 1300 to 1375 nm and in some cases (i.e., transmission geometry) in the range of 1575 to 1625 nm.

Since tissue lipids have a distinct absorption profile, HSI-SWIR can be applied to identify a lipid-based plaque in the *in vivo* and *in vitro* models of atherosclerosis.^{57,58} As the detection of the coronary plaque is often identified with an optical fiberscope catheter, the outcome of the study will depend not only on the depth penetration but also on the contrast between the plaque and the blood vessel. This contrast can be significantly improved using our HSI-SWIR imaging approach as we have presented elsewhere.⁵⁹

The presented contrast versus wavelength data obtained from the HSI approach contain the combined information about scattering, absorption, and other optical parameters. In general, the observed trend follows the light penetration patterns reported in the literature.⁸ In the NIR (<900 nm) range, the scattering in general dominates the absorption $\mu'_s > \mu_a$, providing relatively low contrast. At longer wavelengths (<1000 nm), where the scattering values are comparable to those of absorption ($\mu'_s \approx \mu_a$), the contrast increases with the decrease of scattering. Higher values of absorption ($\mu_a \gg \mu'_s$) even in combination with relatively low scattering, as observed in the 1400- to 1500-nm range, block the light penetration and consequently lead to a weak contrast. Such variability in light penetration in SWIR has been observed in a number of studies.^{8,28,49,60}

4 Conclusions and Future Work

Using HSI, we investigated the transparency of the tissue at different wavelengths in SWIR. We focused on a spectral range from 900 to 1650 nm measuring the depth penetration in biological tissues using a Michelson spatial contrast as a metric of light penetration. Chromatic aberration-free hyperspectral images in transmission and reflection geometries were collected with a spectral resolution of 5.27 nm and a total acquisition time of 3 min. Relatively short recording time minimized artifacts from sample drying. Results from both transmission and reflection geometries in phantoms and biological tissues consistently revealed the highest spatial contrast within 1300 to 1375 nm in deep tissues. The most critical parameters affecting depth penetration in soft tissue were water and hemoglobin absorptions. In heavily pigmented tissue such as liver, the depth penetration was significantly lower than for other studied tissues such as the breast and the thigh. The presented data opens up the possibility for the rational design of fluorescent probes and light-based therapies for deeper penetration, which could have potential applications in clinical settings. We have recently identified QDs with the high quantum yield in the desired optical range at about 1300 nm.⁶¹

Our finding suggests that 1300 to 1375 nm is the best optical band for deep tissue imaging within the tested range; however, the poor sensitivity of the commercial InGaAs detector beyond 1650 nm did not allow us to measure the contrast at longer wavelengths. It has been recently reported that light penetration is increasing up to 1800 nm.^{4,11} The measurement of the

Michelson contrast in these longer wavelength would require a different type of detector with sufficient sensitivity.

As part of our long-term interest in SWIR imaging, we are currently developing a computational model of SWIR light penetration in a variety of biological media. This theoretical model requires reliable quantitative values of absorption and reduced scattering coefficients in SWIR, as well as other wavelength-related parameters, such as anisotropy and refractive index. Although these parameters are available for some tissues, such as skin, they cannot be directly applied to the studied type of tissues. We are currently developing a method to extract light-tissue penetration parameters in SWIR using the generated contrast-wavelength dependence. A related work is currently being conducted in our lab and will be reported in a separate paper.

Disclosures

The authors have nothing to disclose.

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