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Leproux, A.

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Supplement 2

Breast cancer spatial heterogeneity in near-infrared spectra and the prediction of neoadjuvant chemotherapy response

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Authors

Ylenia Santoro, Anaïs Leproux, Albert Cerussi, Bruce Tromberga, Enrico Gratton

Abstract

We describe an algorithm to calculate an index that characterizes spatial differences in broadband near-infrared (NIR, 650 to 1000 nm) absorption spectra of tumor-containing breast tissue. Patient-specific tumor spatial heterogeneities are visualized through a Heterogeneity Spectrum function (HS). HS is a biomarker that can be attributed to different molecular distributions within the tumor. To classify lesion heterogeneities, we built a Heterogeneity Index (HI) derived from the HS by weighing the HS in specific NIR absorption bands. It has been shown that neoadjuvant chemotherapy (NAC) response is potentially related to the tumor heterogeneity. Therefore, we correlate the heterogeneity index obtained prior to treatment with the final response to NAC. From a pilot study of 15 cancer patients treated with NAC, pathological complete responders (pCR) were separated from non-pCR according to their HI (-44±12 and 43±17, p=3·10⁻⁸, respectively). We conclude that the HS function is a biomarker that can be used to visualize spatial heterogeneities in lesions, and the baseline HI prior to therapy correlates with chemotherapy pathological response.
Breast cancer accounts for more than 20% of all newly diagnosed female cancers [1]. New treatment regimens and early detection have reduced mortality rates. Patients diagnosed with locally advanced breast cancer are considered to be at increased risk of disseminated disease. As a result, many of these patients now receive neoadjuvant chemotherapy (NAC), even if the tumor is primarily operable. Although there is no evidence to suggest that there is a survival benefit of neoadjuvant chemotherapy when compared with adjuvant treatment, some studies have shown an overall survival benefit for those patients who achieve a pathologic complete response (pCR) after neoadjuvant chemotherapy compared with patients without a complete response (non-pCR) [2]. Breast cancer is a heterogeneous disease that comes in several clinical and histological forms. Its clinical progression is difficult to predict using current prognostic factors and treatment is therefore not as effective as it should be [3-7]. The prediction of the efficacy of chemotherapy would potentially select good candidates who would respond well while excluding poor candidates who would not benefit from treatment. This would prevent many patients from experiencing unnecessary side effects. Several studies have been performed using conventional radiologic imaging techniques to investigate predictive factors for various chemotherapeutic modalities in breast cancer, but many of them have been found to be unsatisfactory [8-10]. A previous study [11] was able to predict chemotherapy response based on specific histological features of breast cancer. In this work we investigate the possibility of noninvasively predicting chemotherapy response prior to treatment based on biomarkers obtained from tumor spatial heterogeneities of spectral features measured using Diffuse Optical Spectroscopy (DOS).

DOS is a noninvasive, bedside technique that is commonly used to provide biochemical information on hemoglobin, bulk lipids and water concentration by measuring near-infrared (NIR) (650–1000 nm) tissue absorption and scattering [12]. DOS does not require exogenous contrast, and rapidly (e.g., tens of seconds) provides quantitative, functional information about tumor biochemical composition, making it desirable from a patient perspective. Multiple pilot studies have shown the utility of DOS in predicting early tumor response to neoadjuvant treatment [13, 14]. These studies correlated changes in tumor optical properties with tumor final pathological response; thus multiple time points were required to predict therapeutic response. There are inherent limitations in this approach as therapies are often multi-stage and the optimal timing of the post-treatment measurements is unknown. Furthermore, hormonal status and age introduce high interpatient variability in absorption spectra and might contribute additional functional changes independent of tumor response [15].

Using a self-referencing differential spectroscopy (SRDS) method that accounts for interpatient variability, metabolic differences were previously observed between malignant and normal tissues. These differences are assumed to be the result of subtle changes in molecular disposition, i.e. the location, concentration, and environment of a molecular species of NIR absorbers in tissues [16]. The SRDS method exploits the presence or absence of a unique endogenous spectral absorption fingerprint, called the specific tumor component (STC), which has been shown to separate malignant disease from benign tumors in a 40 breast lesion study [17]. The STCs are patient-specific signatures derived from scatter-corrected absorption spectra. We have
observed that these spectra can vary spatially within the same lesion and hypothesize that STCs may contain specific information on tumor heterogeneity. It has been shown that NAC response is potentially related to tumor cellular and structural heterogeneity [11]. These heterogeneous domains could be clustered by histological type or more homogeneously distributed throughout the tumor. Since heterogeneous clustering likely reduces therapeutic efficacy, our goal was to determine whether pre-therapeutic spatial variations of STCs could be correlated with NAC response.

We first introduce an algorithm for calculating the tumor spatial Heterogeneity Spectrum (HS). Then using the HS, a Heterogeneity Index (HI) is developed to quantify the amount of heterogeneity. Further the HI is correlated with NAC pathologic response in 15 patients. This suggests that noninvasive scatter-corrected absorption spectra could be used to predict neoadjuvant chemotherapy response in patients prior to therapy.

We used a DOS instrument to measure broadband NIR absorption and scattering spectra [18]. Absorption spectra were obtained at 1 cm spaced locations on grids over both tumor-containing tissue and normal tissue. The amount of points in a grid varied (from a minimum of 6 to a maximum of 15) depending on the dimension of the tumor. The lesion-free region in the contralateral breast served as patient-specific reference for the calculation of the STCs [16]. Tissue absorption and scattering spectra were calculated using custom software in MATLAB™. The resulting absorption spectra were analyzed by in-house Elantest software (Laboratory for Fluorescence Dynamics – University of California, Irvine) to calculate the STC spectra. Custom software for the calculation of the HS and HI was developed using MATLAB™.

First, we considered the STC spectra in the locations within the lesion for each patient [16]. We hypothesized that the variations of the lesion STC spectra are due to tumor heterogeneity. In order to mathematically characterize tumor spatial heterogeneity for each patient using the normalized STCs spectra, we defined the HS as follows:

$$HS(\lambda) = \frac{\sum_i (STC_i(\lambda) - \overline{STC}(\lambda))^2}{N_i \times N_\lambda}$$

Where $\lambda$ are the different wavelengths used during the measurements, $i$ is the location within the lesion, $N_i$ is the number of spatial points sampled within the lesion and $N_\lambda$ is the number of wavelengths.

The HS, computed for a single subject, shows a biomarker obtained from variations of the STC spectra at each wavelength (Figure 1). The HS provides information on spatial absorption variations, likely resulting from regional differences in lipid, hemoglobin, and water disposition. Specific wavelength domains contain information on the lesion heterogeneity in that particular range. For instance, it has been shown that the water absorption peak around 975 nm undergoes red shifting and broadening in vivo due to macromolecular binding [19]. The heterogeneity in the absorption spectra may provide additional insight regarding tissue pathophysiology. Further investigation of biochemical/histological origins of the heterogeneity spectra will be the subject of future work.
Hypothesizing that spatial STCs variations are associated with heterogeneous lesions, the amplitude of the HS would be correlated with treatment effectiveness. In other words, a highly variable HS would suggest that the subject will respond poorly to NAC. To verify this assumption, we investigated 15 patients, ranging in age from 31 to 71 years, seven pre- and eight post-menopausal, with pathologically confirmed diagnosis of invasive ductal carcinoma (IDC) (ranging from 1.1 cm to 9.8 cm). We performed pre-treatment absorption and scattering measurements. Final treatment pathological response was determined from standard pathology. Pathological complete response was defined as absence of invasive cancer and non-pCR defined as residual invasive tumor of any size, regardless of the presence of ductal carcinoma in situ (DCIS). Among the 15 patients 6 were pCR and 9 were non-pCR. Data were acquired in compliance with an institutionally approved human subjects research protocol (University of California, Irvine (UCI) 02-2306).

A weighted wavelength analysis method was used to exploit the HS to discriminate between pCR and non-pCR. The algorithm separates two types of spectra by using the ‘distance’ of a given spectrum from the average spectrum of each type. To account for spectral differences across the full wavelength region, the distance is calculated at
each wavelength point at which data were obtained, i.e. every half nanometer in the 650-1000 nm range. Furthermore, to maximize the differences, wavelength regions are weighted. Weighting factors for HS of non-pCR and HS of pCR are determined by an iterative process that calculates what combination of values for each wavelength region would best separate the coordinates of the pCR from the non-pCR patients. For every patient (or HS), the “similarity” between the HS and the average HS spectrum of a pCR and a non-pCR lesion was calculated and translated into an index, the Heterogeneity Index (HI). The HI quantifies the amount of heterogeneities in the HS. The HI ranges from -100 to 100, where negative and positive values would describe a homogeneous and heterogeneous lesion, respectively.

\[ HI = 2 \cdot \frac{D_{pCR}^* - D_{NR}^*}{D_{pCR}^* + D_{NR}^*} \]

Where \( D_{pCR}^* \) is the average distance of the considered HS from the average HS of all the pCR and \( D_{NR}^* \) is the average distance of the considered HS from the average HS of all the non-pCR (the factor of two is used to convert it from ±50).

We investigated a subset of 15 patients with IDC (6 pCR and 9 non-pCR) to optimize the weights for building the HI. The average HI values obtained for pCR and non-pCR tumors were -44±12 and 43±17 (\( p=3\times10^{-8} \)), respectively. As observed in Figure 2, the HI values for all patients clearly separate the two groups. The HI was positive for non-pCR tumors and negative for pCR tumors. A negative HI would describe a homogeneous lesion. The results confirm the hypothesis that spatially varying lesions, with a positive HI, result in a partially or non effective therapy.

![Figure 2. Distribution of the heterogeneity index. The pCR patients are clearly separated from the non-pCR patients.](image)

The data set was subjected to a round-robin analysis to determine the dependence of the classification of the patients’ HS on the particular group. Each patient was systematically removed from the set and for each of the reduced sets, the weighting factors were optimized. The HI of the omitted patient was calculated according to the
new weights as a test patient. We found that the HI changed slightly for each patient. However, the classification remained unchanged from the round-robin analysis. We were able to obtain two additional subjects, both non-pCR, whose measurements have been used to further test the classification algorithm. The lesions were correctly classified with a HI of 33 for one subject, and 40 for the other subject.

In conclusion, we have introduced a method for visualizing and analyzing the spatial heterogeneity of non-invasive, scatter-corrected tumor absorption spectra based on self-referencing differential spectroscopy (SRDS). We present an application of the method that, for the first time, relates tumor spatial heterogeneity to the pathological response of neoadjuvant chemotherapy. Preliminary results in 17 patients reveal that a spatial heterogeneity index (HI), calculated from the HS, can separate pathologic complete responders (pCR) from non-pCR patients. These results show that the SRDS technique is a promising approach for potentially predicting NAC outcome prior to therapy.

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References


