



Physical interpretation of the correlation between multi-angle spectral data and canopy height

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[1] Recent empirical studies have shown that multi-angle spectral data can be useful for predicting canopy height, but the physical reason for this correlation was not understood. We follow the concept of canopy spectral invariants, specifically escape probability, to gain insight into the observed correlation. Airborne Multi-Angle Imaging Spectrometer (AirMISR) and airborne Laser Vegetation Imaging Sensor (LVIS) data acquired during a NASA Terrestrial Ecology Program aircraft campaign underlie our analysis. Two multivariate linear regression models were developed to estimate LVIS height measures from 28 AirMISR multi-angle spectral reflectances and from the spectrally invariant escape probability at 7 AirMISR view angles. Both models achieved nearly the same accuracy, suggesting that canopy spectral invariant theory can explain the observed correlation. We hypothesize that the escape probability is sensitive to the aspect ratio (crown diameter to crown height). The multi-angle spectral data alone therefore may not provide enough information to retrieve canopy height globally. **Citation:** Schull, M. A., et al. (2007), Physical interpretation of the correlation between multi-angle spectral data and canopy height, *Geophys. Res. Lett.*, 34, L18405, doi:10.1029/2007GL031143.

1. Introduction

[2] Canopy structure determines the amount of carbon sequestered in the vegetation pool, and incomplete knowledge of this amount contributes to the current uncertainty on predictions of future atmospheric CO₂ concentrations [International Panel of Climate Change, 2001]. Recent efforts using ground truth and lidar data to train MISR retrievals of vegetation structure have made significant strides toward developing a passive satellite technology for vegetation monitoring. It has been shown that multi-angle spectral data conveys information about fractional-area distributions [Braswell et al., 2003; Heiskanen, 2006; Chopping et al., 2006a, 2006b], stand basal area [Jenkins et

al., 2004], tree height [Ranson et al., 2005; Kimes et al., 2006, Heiskanen, 2006], forest cover density [Nolin, 2004] and biomass [Baccini et al., 2004; Chopping et al., 2007]. These results have employed the use of empirically based techniques, which are site specific and may not be extendable to operational use, given different structural and climatic conditions. Development of physically based approaches to interpret remote sensing data is therefore required not only to take full advantage of available remotely sensed data but also to advance our understanding of requirements for future space based measurements of vegetation 3D structure. In this paper, we use the concept of canopy spectral invariants to explain physics behind observed correlation between multi-angle multi-spectral data and canopy structure.

2. Canopy Spectral Invariants

[3] Photons that have entered the vegetation canopy undergo several interactions with leaves before either being absorbed or exiting the medium through its upper or lower boundary. As a result of an interaction, photons can either be scattered or absorbed by a phytoelement. The probability of the scattering event, or leaf single scattering albedo, ω_λ , depends on the wavelength and is a function of the leaf biochemical constituents. If objects are large compared with the wavelength of the radiation, e.g., leaves, branches, etc., the photon free path between two successive interactions is independent of the wavelength [Ross, 1981]. The interaction probabilities for photons in vegetation media, therefore, are determined by the structure of the canopy rather than photon frequency or the optics of the canopy. To quantify this feature, Smolander and Stenberg [2005] introduced the notion of recollision probability, p , defined as the probability that a photon scattered from a phytoelement in the canopy will interact within the canopy again. This spectrally invariant parameter is purely a function of canopy structural arrangement [Knyazikhin et al., 2005; Huang et al., 2007b; Lewis and Disney, 2007; Möttus, 2007]. Additionally scattered photons can escape the vegetation canopy, either through the upper or lower boundary. Their angular distribution is given by the directional escape probability, $\rho(\Omega)$ [Huang et al., 2007b].

[4] Under the assumption that the spectral invariants p and ρ remain constant in successive interactions, the bidirectional reflectance factor, $BRF_\lambda(\Omega)$, is an explicit function of spectrally invariant recollision and escape probabilities and spectrally varying single scattering albedo, ω_λ , i.e.,

$$BRF_\lambda(\Omega) = \rho(\Omega)\omega_\lambda i_0 + \rho(\Omega)\omega_\lambda^2 p i_0 + \dots + \rho(\Omega)\omega_\lambda^m p^{m-1} i_0 + \dots = \frac{\omega_\lambda}{1 - \omega_\lambda p} R_1(\Omega). \quad (1)$$

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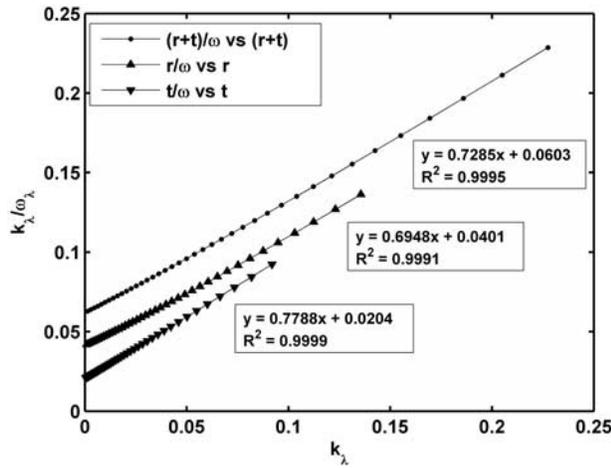


Figure 1. Canopy spectral invariants for canopy hemispherical reflectance, r_λ (upward triangle), transmittance of collided radiation, t_λ (downward triangle), and scattering coefficient $r_\lambda + t_\lambda$ (circle). By plotting values of the ratio $\kappa_\lambda/\omega_\lambda$ versus κ_λ a linear relationship is obtained. Here κ_λ represents either r_λ , t_λ , or $r_\lambda + t_\lambda$. The intercepts corresponding to r_λ and t_λ give the escape probabilities in upward (0.0401) and downward (0.0204) directions. The slope of the $(r_\lambda + t_\lambda)/\omega_\lambda$ versus $(r_\lambda + t_\lambda)$ line is an accurate estimate of the recollision probability (0.7285) [Huang et al., 2007b]. Its intercept (0.0603) is the sum of the upward and downward escape probabilities. The specification of the recollision probability from the spectral reflectance-only or transmittance-only underestimates (0.6948) or overestimates (0.7788) the true value. Calculations were performed using the stochastic radiative transfer equation for a vegetation canopy consisting of identical cylindrical “trees” uniformly distributed in the canopy layer [Huang et al., 2007a]. The aspect ratio, ground cover and plant leaf area index are set to 1, 0.16 and 10. The solar zenith angle and azimuth of the incident beam are 30° and 0° .

Here i_0 is the probability of initial collisions, or canopy interceptance defined as the portion of photons from the incident beam that are intercepted, i.e., collide with phytoelements for the first time. Canopy interceptance does not depend on the wavelength and is a function of the direction of incident beam and canopy structure. The term $R_1(\Omega) = \rho(\Omega)i_0$ is the escape probability expressed relative to the number of incident photons. Values of p and R_1 can be determined by fitting equation (1) to measured reflectance spectrum.

[5] In general case the recollision and escape probabilities vary with the scattering order m . For $m = 1$, the directional escape probability coincides with the bidirectional gap probability. This fundamental canopy structure parameter therefore is a special case of the directional escape probability. The probabilities, however, reach plateaus as the number of interactions m increases. Monte Carlo simulations of the radiation regime in 3D canopies suggest that the probabilities saturate after 2 to 3 photon-canopy interactions for low to moderate LAI canopies [Lewis and Disney, 1998]. Huang et al. [2007b] found that the relative error in the approximation (1) does not exceed

5% as long as the single scattering albedo is below 0.9. The use of this approximation minimally impacts the values of the escape probability. Violation of the above assumption, however, can result in a transformation of the recollision probability to its effective value as a result of fitting equation (1) to measured spectral reflectance [Huang et al., 2007b]. The reader is referred to Huang et al. [2007b] for the current state of understanding in this field.

[6] Equation (1) is the basis for our data analyses. Specifically, we would like to show that the spectral BRF and $R_1(\Omega)$ convey comparable information about canopy structure. As such, the canopy spectral invariants, p and R_1 , can explain the physics behind the observed correlation between multi-angle spectral and lidar data, where the measurable directional escape probability is the variable that imbues canopy structure dependence to multi-angle spectral data.

3. Method Used and Its Limitations

[7] Equation (1) can be rearranged to a form which we will use to obtain the spectral invariants p and $R_1(\Omega)$ from multi-angle spectral data, namely,

$$\frac{BRF_\lambda(\Omega)}{\omega_\lambda} = pBRF_\lambda(\Omega) + R_1(\Omega). \quad (2)$$

[8] By plotting the ratio $BRF_\lambda/\omega_\lambda$, versus BRF_λ , a linear relationship is obtained (Figure 1), where the slope and intercept give the recollision and escape probabilities, respectively. We will use AirMISR data (section 4) at blue, green, red and near infrared spectral bands to specify p and $R_1(\Omega)$ using equation (2).

[9] The spectral invariant relationships are formulated for a vegetation canopy (1) bounded from below by a non-reflecting surface and (2) illuminated from above by a wavelength independent parallel beam. However, we will use measured multi-angle spectral data without correcting for canopy substrate effects (i.e. the fact that observed canopies do not have totally absorbing lower boundaries), which will impact the values of $R_1(\Omega)$ [Huang et al., 2007b]. The second assumption requires the use of the BRF. This assumption is not met in our research since the AirMISR BRF product was not available for all sites of the study area. We will use hemispherical-directional reflectance factor (HDRF) which characterizes surface reflective properties under ambient atmospheric condition [Martonchik et al., 2000]. At shorter wavelengths the diffuse component of the incident radiation is not negligible due to Rayleigh and aerosol scattering and exhibits strong variation with the wavelength. Since the interceptance, i_0 of the vegetation canopy is higher under diffuse illumination conditions [Tian et al., 2004; Min, 2005; Gu et al., 2002, Jenkins et al., 2007] the use of HDRF at blue and green spectral bands can result in a systematic overestimation of the escape probability (see auxiliary material¹), but since it is systematic it does not effect the outcome of our results.

[10] Equation (2) requires the spectral single scattering albedo ω_λ which depends on the scale at which this quantity

¹Auxiliary materials are available in the HTML. doi:10.1029/2007GL031143.

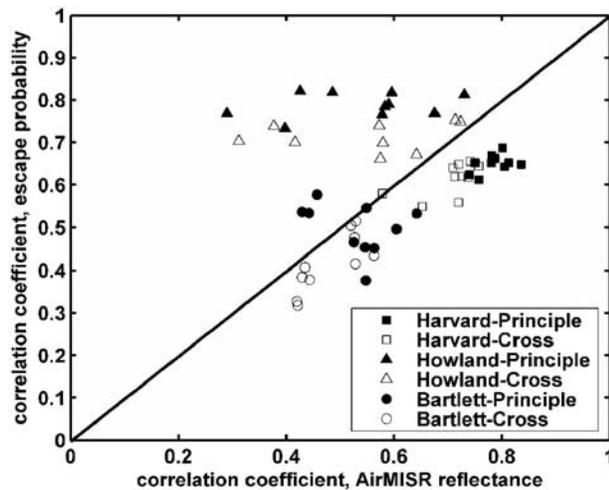


Figure 2. Correlation coefficients for escape probability predicted heights versus correlations for AirMISR predicted heights for three sites at two different observation geometries (Principle and Cross planes) and ten combinations of training and testing sets.

is defined. For example, the single scattering albedo of a needle, shoot, branch, tree crown, etc., are different. The choice of scale does not violate spectral invariant relationships [Smolander and Stenberg, 2005; Lewis and Disney, 2007] although it impacts the values of spectral invariants. We will use a leaf albedo spectrum measured during the Flakaliden field campaign [Huang et al., 2007b].

[11] Violation of the above assumptions and the use of one single scattering albedo pattern for all sites impact values of the spectral invariants and result in ambiguities in the scale at which they are defined. As we will see from our analyses, however, this will have a minimal impact on the estimation of information content of the escape probability.

4. Data Used

[12] The study area consists of three locations, which encompass a transition from evergreen needle leaf to deciduous broadleaf forest (Figure S2 in the auxiliary material). Howland Forest (45.2 N, 68 0.74 W) at International Paper's Northern Experimental Forest site is characterized as an evergreen needleleaf forest. Harvard Forest (42.54 N, 72.17 W) a Long Term Ecological Research (LTER) site is classified as a mixed deciduous broadleaf and evergreen needle leaf forest. Bartlett Experimental Forest (44.06 N, 71.29 W) a USDA Forest Service site is mostly a broadleaf deciduous forest with areas of conifers at the higher elevations. Lidar and multi-angle spectral data were acquired in the summer of 2003 as part of a NASA Terrestrial Ecology Program aircraft campaign and are publicly available at <https://lvis.gsfc.nasa.gov> and http://eosweb.larc.nasa.gov/PRODOCS/airmistr/table_airmistr.html, respectively.

4.1. Lidar Data

[13] Airborne Laser Vegetation Instrument System (LVIS) is a pulsed laser altimeter which measures range

by timing a short-10 ns duration pulse of laser light between the instrument and the target surface. LVIS point data collected from the H100 height measures (see auxiliary material) were sampled into a raster grid dataset at a 28 m nominal resolution using a window average scheme. A strong correlation between field measured and LVIS estimates of canopy height has been documented for the Bartlett and Howland sites [Kimes et al., 2006; Anderson et al., 2006] allowing for the use of LVIS as a surrogate for canopy height.

4.2. Multi-Angle Spectral Data

[14] The Airborne Multi-angle Imaging Spectrometer (AirMISR) instrument on NASA's ER-2 aircraft employs four channel spectral data at nine different viewing angles (See Figure S3 in the auxiliary material). Level 1 radiometrically and geometrically corrected radiance data product collected in the principle and cross-planes were converted to at-aircraft level HDRF. The data is provided in WGS 84 UTM projection and has a common ground resolution of 27.5 m with all the cameras co-registered to a common swath width of 11 km. The AirMISR dataset is resampled to 28 m resolution and co-registered to match the LVIS dataset. More information about the dataset characteristics and methods used can be found in the auxiliary material.

5. Multivariate Linear Regression Models

[15] For each study area and for each AirMISR sun-view geometry (Figure S3 in the auxiliary material), we first created a random subset (training set) from 1/3 of the AirMISR-LVIS dataset to derive two multivariate linear regression models for estimating LVIS canopy height measures. The independent variables for the first model were reflectance values of AirMISR HDRF for 4 spectral bands at 7 view directions, totaling 28 variables. To obtain the second model, we retrieved the escape probability at 7 AirMISR view angles using equation (2) first and then used the obtained values as the 7 independent variables in the model. The models were then used to predict LVIS height measure from the AirMISR HDRF and the escape probability in the remaining 2/3 of pixels (testing set). The correlation coefficients calculated using values of predicted and actual LVIS heights were employed as a measure of the information about canopy structure that AirMISR HDRF and escape probability convey. This approach was applied to AirMISR-LVIS data set degraded to 283 m resolution (see Figure S4 and associated text in the auxiliary material).

6. Results and Discussions

[16] Figure 2 shows the correlation coefficients for AirMISR predicted LVIS height estimates versus correlations for escape probability predicted LVIS heights for three sites at two different observation geometries and for ten combinations of training and testing sets. There are several important features noteworthy in the correlation coefficients. The correlation coefficient for the AirMISR predicted LVIS heights measure exhibits a higher sensitivity to the test data compared to its escape probability counterpart. The effect is most pronounced at Howland and less for Bartlett sites. This suggests that the test data should adequately represent variation in both structural and spectral

components in the case of multi-angle spectral data while the escape probability requires mainly structural information to train the model. As such, the latter is purely function of canopy structural arrangement. On average, the correlation coefficient is higher if the AirMISR data in the principal plane are used. Overall, the wavelength independent escape probability and multi-angle spectral data tend to provide a comparable amount of information about the LVIS height measure.

[17] An additional observation garnered from the research is that one cannot use single band multi-angle information to accurately measure canopy structure. The shapes of the HDRF and escape probability can be very similar (see Figure S6 in the auxiliary material). However, the use of single band multi-angle data results in a much weaker correlation as seen in a comparison of Figure 2 and Figure S7 in the auxiliary material. This is because the escape and recollision probabilities exhibit opposite tendencies in the sense that an increase in the escape probability is accompanied by a decrease in the recollision probability [Disney *et al.*, 2005]. The denominator in equation (1) therefore tends to suppress changes in R_1 , lowering the sensitivity of single band data to the canopy height. This result is consistent with findings reported by Kimes *et al.* [2006] and Heiskanen [2006] and suggests that single band multi-angle data are not sufficient to extract canopy horizontal structure parameters. Spectral information is required to extract the spectral invariants from the measured signal [Panferov *et al.*, 2001; Shabanov *et al.*, 2003; Knyazikhin *et al.*, 2005; Huang *et al.*, 2007b]. These variables imbue canopy structure dependence to multi-angle spectral data and, therefore, are plausible to explain the observed correlation between multi-angle spectral reflectance and canopy height.

[18] The correlation between the directional escape probability and LVIS canopy heights vary with site as can be seen in Figure 2 (see vertical axis). Howland Forest (needle leaf forest) has the highest correlation, Harvard Forest (mixed forest) shows intermediate results, and Bartlett Forest (mostly a broadleaf deciduous forest with areas of conifers at the higher elevations) exhibits the lowest correlation. The most obvious feature about each site is the difference in structural heterogeneity of the upper canopy: Howland Forest has the highest heterogeneity and gives the highest correlation while Bartlett is a dense forest with a closed canopy giving it a fairly smooth surface. As a result of this observation, we hypothesize that multi-angle spectral data are actually more sensitive to the aspect ratio (crown diameter to crown height ratio) and ground cover, and that this is responsible for the observed correlations. Indeed, a multi-angle sensor sees a tree from 3 sides, front (in relation to the sensor's direction), top, and back. At nadir the sensor sees the tree's crown diameter and at the front and back the sensor sees the tree crown's approximate height. If there is no visible height as in the case of Bartlett's closed forest, then there is no ability to measure height at all. The reason why we could predict height with higher accuracies at Howland is because the forest canopy has a conical structure allowing the sensor to see the sides of the trees and thus giving aspect ratio. A simple comparison of modeled and measured escape probabilities and visual inspection of sites do not reject our hypothesis (See Figure S5 in the auxiliary

material). Further research, however, is required to validate the speculative evaluation of multi-angle data imbuing information about canopy aspect ratio.

7. Conclusions

[19] The empirical analysis of AirMISR and LVIS data support results from previous studies [Ranson *et al.*, 2005; Kimes *et al.*, 2006; Heiskanen, 2006] that document a strong correlation of multi-angle multi-spectral data and canopy height. We found that the canopy spectral invariants can explain the observed correlation, where the wavelength independent directional escape probability is the variable that imbues the sensitivity of MISR data to canopy height. Spectral information is essential to extract the spectral invariant parameters from the measured signal. The significance of this result to the remote sensing of 3D canopy structure is two-fold. First, the canopy spectral invariants offer a simple, accurate and physically well-justified representation of canopy reflectance where structural and radiometric components of measured signal are naturally decoupled. This allows for retrievals of canopy structural parameters with fewer assumptions compared to traditional modeling approaches where structurally and spectrally varying parameters are functionally related in a complex manner. Second, many of existing approaches require models of the bi-directional gap probability to simulate canopy reflectance. Since this parameter is a special case of the directional escape probability, its estimation from multi-angle multi-spectral reflectance is virtually independent of any particular kind of surface model and thus enables a more direct link between canopy structure and optical remote sensing data. Finally, the information gained from canopy structure will lead to more accurate inputs for land surface-models as well as provide groundwork for more accurate biomass estimation.

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