Assessing the narrowband to broadband conversion to estimate visible, near infrared and shortwave apparent albedo from airborne PolDER data

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Abstract – The airborne PolDER data acquired during the ReSeDA experiment were used to assess the accuracy of the narrowband to broadband conversion, commonly used to express the integrated value of albedo as a linear combination of waveband estimates. We focused on the estimation of shortwave, visible and near infrared apparent albedos from waveband hemispherical reflectances. The latter were computed using four classical linear BRDF kernel-driven models. The relative discrepancy over the estimates from the four models was about 10%. The narrowband to broadband conversion was performed considering several coefficient sets devoted to recent sensors. The validation against field measurements showed a relative accuracy of shortwave albedo about 9%, whereas it emphasized the quality of visible and near infrared albedos through an indirect procedure. However, the consequences of the theoretical uncertainties, made when calibrating the coefficient sets we used, remain unknown. Similarly, the calibration of coefficient sets devoted to the shortwave albedo estimation from PolDER data showed that it was not possible to discriminate the information provided by the blue, green and red PolDER channels. Indeed, the hemispherical reflectance data set we used was not accurate enough to address experimentally the validity of the spectral interpolation method. Finally, the calibration of coefficients showed that it might be possible to calculate shortwave albedo with a relative error of 8% using only red and near infrared channels.

remote sensing / multidirectional data / multispectral data / visible / near infrared / apparent albedo

Résumé – Évaluation de la méthode “narrowband to broadband conversion” pour estimer l’albedo apparent visible-proche infrarouge-courtes longueurs d’onde à partir de données aéroportées PolDER. Les données aéroportées PolDER acquises durant l’expérimentation ReSeDA ont été utilisées pour évaluer la précision de la méthode “narrowband to broadband conversion”, communément utilisée pour exprimer la valeur intégrée de l’albedo comme une combinaison linéaire de mesures spectrales. Nous sommes intéressés à l’estimation de l’albedo apparent visible-proche infrarouge-courtes longueurs d’onde à partir de la réflectance hémisphérique. Cette dernière a été calculée en utilisant quatre modèles classiques de FDRB à noyaux. La dispersion relative entre les estimations des quatre modèles était de l’ordre de 10 %. Nous avons calculé la valeur intégrée de l’albedo en utilisant des jeux de coefficients proposés dans la littérature pour des capteurs récents. La validation a montré que l’albedo courtes longueurs d’onde PolDER avait une précision relative de l’ordre de 9 %, alors qu’une procédure indirecte a mis en évidence la qualité des albedos visible et proche infrarouge. Cependant, les estimations de réflectance hémisphérique n’étaient pas assez précises pour évaluer les conséquences des incertitudes théoriques des jeux de coefficients. De même, la calibration de jeux de coefficients à l’aide des données ReSeDA a montré qu’il n’était pas possible de discriminer l’information fournie par les canaux PolIDER bleu, vert et rouge. Finalement, il était possible de calculer l’albedo avec une précision relative de 8 % en ne considérant que les canaux rouge et proche infrarouge.

télédétection / données multidirectionnelles / données multispectrales / albedo apparent / visible / proche infrarouge
1. INTRODUCTION

Instantaneous surface albedo is defined as the fraction of incident solar radiation reflected in all directions over a given spectral range of the solar spectrum. Subdefinitions are proposed according to the consideration of the spectral range and the consideration of the diffuse and/or the direct components of incoming solar radiation [16]. Shortwave, visible and near infrared albedos correspond respectively to the whole solar spectrum, the [400–700] nm and the [700–5000] nm spectral ranges. Inherent (or black-sky), white-sky and apparent albedo corresponds respectively to the direct component of incoming solar radiation, the diffuse component and both of them. Knowledge of these kinds of albedo in a spatially distributed manner is of prime interest for global climate modeling [7], meteorological studies [3, 5], and surface flux estimation [13, 22, 10]. The required accuracy varies from one application to another. Sellers [29] suggested a relative accuracy of ±2%, whereas GCM studies and weather forecast modeling require an absolute accuracy of ±5% that corresponds to a relative accuracy of ±10% [4].

Remote sensing is the most promising technique to estimate surface albedo in a spatially distributed manner. However, such a technique requires extrapolating the spectral and angular information collected from observations in a limited number of both viewing directions and wavebands. One possible approach is to express the spectrally integrated value of the albedo as a linear combination of top of canopy or top of atmosphere bidirectional reflectance, using coefficient sets calibrated over simulated or measured data sets [2, 15, 16, 30, 39]. Obviously, it is better to use top of canopy bidirectional reflectance to avoid the noise induced by the atmosphere variability. However, the accuracy of this method is limited by the representativeness of a single directional observation regarding to the angular variation of the reflected solar energy, so called BRDF for Bidirectional Reflectance Distribution Function [36]. Therefore, another approach consists of considering hemispherical reflectance, rather than bidirectional reflectance, when computing the integrated value of albedo as a linear combination of wavebands estimates [39]. This requires first characterizing the BRDF, which can be performed by processing the multidirectional information using either radiative transfer models [12, 24] or kernel-driven models [18, 38].

During the ReSeDA experiment, visible-near infrared remote sensing data were acquired using the airborne multangular PolIDER radiometer, simultaneously with field measurements of albedo. The PolIDER/ReSeDA data set provided an opportunity to assess the potential of the narrowband to broadband conversion. First, it covered the whole cycles of different crops including winter and summer ones. Second, the use of high spatial resolution remote sensing data reduced problems due to mixed pixels. Previous studies assessed the performances of classical kernel-driven models to both characterize the BRDF and estimate the hemispherical reflectance using this data set. Weiss et al. [40] analyzed the importance of the kernel number, as well as the impact of the directional sampling through both the number of data and their angular distribution regarding to the sun position. Jacob et al. [11] assessed the impact of the spatial variability, the atmospheric conditions and the waveband. They also evaluated the variability of the hemispherical reflectance estimates when considering several kernel-driven models, and its possible link with the fitting performances.

In this study, we focused on the estimation of the spectrally integrated value of visible, near infrared and shortwave albedo using top of canopy estimates of hemispherical reflectance in several wavebands. As mentioned previously, this narrowband to broadband conversion is generally performed using coefficient sets calibrated over simulated or measured data sets. We considered and compared several coefficient sets devoted to recent sensors, and calibrated over data sets that differed spectrally, directionally and experimentally. We also assessed the possibility to perform a coefficient calibration using the ReSeDA database.

2. DATA ACQUISITION AND PREPROCESSING

The ReSeDA experiment lasted from December 1996 to December 1997, in the South East of France (N 43°47', E 4°45'). The experimental site was a 5 × 5 km² agricultural region with maize, sunflower, wheat and alfalfa fields about 200 × 200 m² size [23, 26]. In this study, we used both the airborne PolIDER measurements and the field data of albedo. The following is an overview of the PolIDER and in situ data acquisition and preprocessing. More detailed descriptions are given by Leroy and Hautecœur [14], Olioso et al. [21], and François et al. [9].

The airborne PolIDER sensor [6] flew approximately one or two times per month during the year 1997, on clear sky days, at a 3000 m altitude resulting in a 20 m nadir spatial resolution. Four flight lines were parallel to the solar plane, and one perpendicular. The five lines were completed within 45 minutes centered on solar noon. The measurements were collected in four 40 nm width wavebands centered at 443 nm, 550 nm, 670 nm, and 865 nm. View zenith angle ranged from 0 to 50°. Data were radiometrically corrected in a similar way than that used by Leroy and Hautecœur [14]. The sensor calibration was performed by the Laboratoire d’Optique Atmosphérique (Lille, France) three times: before, during and after the experiment. The procedure accounted for ambient temperature, dark current, and the inter-calibration of the CCD matrix detectors. Calibration accuracy was about 5%. Atmospheric effects were corrected using the SMAC algorithm [27]. The input atmospheric variables were field measurements or climatological data of aerosol optical thickness,
water vapor content and ozone concentration. No accuracy was proposed for these corrections. Image registration was performed using data provided by both a Global Positioning System and a gyrosopic central unit. The images were geometrically matched according to a Lambert II projection that provided a 20 m sampling of the study area. This pre-processing provided a database of BRDF samplings, i.e., a pixel-by-pixel set of measurements of the reflected solar energy in a selection of viewing directions for each PoliDER waveband. Each measurement was the sum of the reflected fractions of incident direct and diffuse radiations, these fractions being weighted by the amount of the corresponding energy.

Ground based observations of albedo were estimated as the ratio of reflected to incident solar radiation measurements. The incident solar radiation was measured using a Kipp and Zonen CM6B pyranometer setup at the meteorological site located at the center of the experimental area. The reflected solar radiation was measured on seven locations corresponding to alfalfa, wheat, and sunflower crops using either Kipp and Zonen CM6B pyranometers or Skye Instruments SP1110 silicon sensors. These routine data were collected with both a 15 s time step and a 20 mn period averaging. The corresponding footprint ranged from 1000 to 3000 m². The intercalibration of the Kipp sensors provided a residual error about 10 W·m⁻², which yielded a relative error on computed albedo of 1.5%. The Kipp sensors were calibrated to provide estimates of the incoming solar radiation over the whole solar spectrum from measurements over the [300–3000] nm range. The Skye sensors measured the incoming solar radiation over the [350–1100] nm spectral band. The extrapolation of the Skye data to the whole solar spectrum required accounting for the spectral reflectance variations of the observed surfaces. For this, we calibrated a formulation between simulated Skye and Kipp albedo around solar noon. The simulations were performed by François et al. [9] using the radiative transfer model SAIL [33]. The SAIL inputs were soil/leaf optical properties, measurements of Leaf Area Index, and spectra of incident solar radiation computed from simulations of the atmospheric radiative transfer model 6S [34] that included numerous atmospheric situations. These simulations are described in details by François et al. [9]. The formulation we calibrated was linear with a residual error of 0.003:

$$\text{Albedo}_{\text{actual}} = 0.785 \times \text{Albedo}_{\text{Skye}} + 0.02.$$  (1)

### 3. HEMISPHERICAL REFLECTANCE ESTIMATION: AN OVERVIEW

A detailed discussion about the choice of the kernel-driven models used and the results obtained when assessing their performances is given by Jacob et al. [11]. We only give here an overview of the most important results, and some concluding remarks that were useful for the present study. The chosen models were four classical three-kernel-driven models: Li-Ross in the “Li-Sparse/Ross-Thick” variant [37], the semi-linearized version of MRPV [8], Roujean [28], and the non-reciprocal version of Walthall [35]. Model performances were assessed considering PoliDER data acquired over the in-situ data locations. It was shown that the fitting performances were not affected by the atmospheric conditions and were very close from one model to another regardless of waveband. These performances decreased systematically as the wavelength increased. This was explained by the aerosol diffusion, which induced at shorter wavelengths a lower signal to noise ratio, as well as a higher diffuse component not adequately described by the models. On the other hand, the hemispherical reflectance estimates were significantly different from one model to another. The most important differences occurred over the blue channel. Apart from this waveband, the agreement between the estimates was not systematically better as the wavelength increased. This showed that the fitting performances and the quality of the hemispherical reflectance estimates were not completely correlated, which emphasized the importance of the model interpolation performances. Finally, as noted previously by Privette et al. [25] and Lucht [17], the inter-comparison of the estimates showed that the hemispherical reflectance values could be significantly different from one model to another, with a relative discrepancy of about 10% (discrepancy expressed as the ARMSD and RRMSD, see Tab. I). A detailed discussion about these results is given by Jacob et al. [11].

Before using these hemispherical reflectance estimates to compute albedo, we rejected unrealistic values lower than 0 or higher than 1. These unrealistic values were ascribed to noise occurring on the data set of sampled BRDF. Table II displays the rejection rate over the whole study area apart from the mountain chain located at the south since the models were not designed for inclined areas. As a consequence of the

### Table I. Comparison between the hemispherical reflectance estimates from Li-Ross and MRPV according to the four PoliDER channels. Corr. Coef. means Correlation Coefficient. Results extracted from Jacob et al. [11].

<table>
<thead>
<tr>
<th>Channels</th>
<th>Blue (443 nm)</th>
<th>Green (550 nm)</th>
<th>Red (670 nm)</th>
<th>NIR (865 nm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>ARMSD</td>
<td>0.0078</td>
<td>0.0098</td>
<td>0.0124</td>
<td>0.0295</td>
</tr>
<tr>
<td>RRMSD</td>
<td>12.8%</td>
<td>8.1%</td>
<td>10.0%</td>
<td>8.3%</td>
</tr>
<tr>
<td>Corr. Coef.</td>
<td>0.9671</td>
<td>0.9891</td>
<td>0.9907</td>
<td>0.9851</td>
</tr>
</tbody>
</table>

### Table II. Percentage of pixels rejected over the study area to remove the unrealistic value of hemispherical reflectance provided by the kernel-driven models.

<table>
<thead>
<tr>
<th>Model</th>
<th>Blue</th>
<th>Green</th>
<th>Red</th>
<th>NIR</th>
</tr>
</thead>
<tbody>
<tr>
<td>Li-Ross</td>
<td>8.5%</td>
<td>1.2%</td>
<td>1.6%</td>
<td>0.2%</td>
</tr>
<tr>
<td>MRPV</td>
<td>6.1%</td>
<td>0.7%</td>
<td>0.7%</td>
<td>0.4%</td>
</tr>
<tr>
<td>Roujean</td>
<td>15.0%</td>
<td>3.4%</td>
<td>5.1%</td>
<td>0.5%</td>
</tr>
<tr>
<td>Walthall</td>
<td>12.0%</td>
<td>2.0%</td>
<td>2.1%</td>
<td>0.3%</td>
</tr>
</tbody>
</table>
perturbations we noted before, the rejection was systematically and significantly more important over the blue channel. Regardless of kernel-driven model, the lowest occurrence of unrealistic values corresponded to the near infrared channel, while this occurrence was similar over the green and red channels, and slightly higher at 670 nm. When considering the robustness to noise of each kernel-driven model, we noted that Roujean provided systematically the highest number of unrealistic values regardless of PolDER wavebands. The best robustness was observed for MRPV while the results we obtained with Li-Ross were also good as compared to Roujean and Walthall. These observations were important since they emphasized that the kernel-driven models we used could not provide systematically realistic values and could not therefore be used confidently in an operational manner. Furthermore, the recommendation of a kernel-driven model should be a compromise between the accuracy of the model and its robustness regarding to the noise occurring over the used data set. For further investigations, we selected PolIDER data acquired over the in-situ data locations.

4. ESTIMATING BROADBAND APPARENT ALBEDO FROM MULTISPECTRAL DATA

All the subdefinitions of albedo \( a(\theta_s, \varphi_s) \) given in Section 1 can be expressed as weighted integrations of the hemispherical reflectance \( \rho_{h,\lambda}(\theta_s, \varphi_s) \) over the considered spectral range, where the hemispherical reflectance is related to the diffuse component, and/or the direct component of the incoming solar radiation. As explained previously, the narrowband to broadband conversion consists of expressing the integrated value as a linear combination of the wavebands hemispherical reflectance \( \rho_{h,j}(\theta_s, \varphi_s) \) over the PolIDER channels \( j \).

\[
a(\theta_s, \varphi_s) = \beta_0 + \sum_{j=1}^{4} \beta_j \rho_{h,j}(\theta_s, \varphi_s)
\]

Since coefficient sets \( (\beta_j) \) devoted to the airborne PolIDER sensor did not exist, despite a study focused on the spaceborne version [15], we selected among the several sets suggested in the literature some proposed in recent studies: 3 by Weiss et al. [39] for generic sensors, and 6 by Liang et al. [16] and Liang [15] for the MISR sensor (see Tab. III). They were devoted to atmospherically corrected data.

Weiss et al. [39] proposed three sets obtained from linear regressions between simulated hemispherical reflectance and shortwave apparent albedo. The simulations were performed between 400 and 2500 nm with the radiative transfer model developed by Myneni et al. [20]. They were representative of several kinds of canopy at different latitudes and different seasons. These three sets differed by the number of used channels among the blue, green, red and near infrared spectral domains. Sets No. 2 and No. 3 were very similar. However, we used both of them since they could provide different results according to the hemispherical reflectance estimates in the blue and red channels.

Liang et al. [16] proposed sets dedicated to the shortwave, visible and near infrared inherent albedo using MISR data. The coefficients were computed from linear regressions over more than 100 observed reflectance spectra between 200 and 3000 nm that corresponded to vegetation, soil and snow. This study was extended by Liang [15] who calibrated sets to

<table>
<thead>
<tr>
<th>Author and Sensor and Albedo</th>
<th>Wavebands and Coefficient values</th>
<th>Offset</th>
</tr>
</thead>
<tbody>
<tr>
<td>Weiss et al. [39] Generic</td>
<td>445 560 665 855</td>
<td></td>
</tr>
<tr>
<td>SW/A Set No. 1</td>
<td>– – 0.57 0.46</td>
<td>–</td>
</tr>
<tr>
<td>SW/A Set No. 2</td>
<td>– 0.68 0.08 0.35</td>
<td>–</td>
</tr>
<tr>
<td>SW/A Set No. 3</td>
<td>0.06 0.69 0.001 0.35</td>
<td>–</td>
</tr>
<tr>
<td>Liang et al. [16] Vis/I Set No. 4</td>
<td>0.3511 0.3923 0.2603 –</td>
<td>–0.0030</td>
</tr>
<tr>
<td>NIR/I Set No. 5</td>
<td>– – – 0.6088 0.1442</td>
<td></td>
</tr>
<tr>
<td>SW/I Set No. 6</td>
<td>0.1587 –0.2463 0.5442 0.3748</td>
<td>0.0149</td>
</tr>
<tr>
<td>Liang [15] Vis/A Set No. 7</td>
<td>0.3810 0.3340 0.2870 –</td>
<td>–</td>
</tr>
<tr>
<td>NIR/A Set No. 8</td>
<td>−0.3870 –0.1960 0.5040 0.8300 0.0110</td>
<td></td>
</tr>
<tr>
<td>SW/A Set No. 9</td>
<td>– 0.1260 0.3430 0.4150 0.0037</td>
<td></td>
</tr>
<tr>
<td>A/C PolDER</td>
<td>423–463 530–570 650–690 845–885</td>
<td></td>
</tr>
</tbody>
</table>
compute shortwave, visible and near infrared apparent albedo, considering up to 256 spectra ranging between 250 and 2500 nm and derived from field and airborne measurements over numerous surfaces. We should note that Liang et al. [16] and Liang [15] used directional measurements by assuming the surfaces were Lambertian, which could be far from actual situations. Moreover, the coefficient sets proposed by Liang et al. [16] for the MISR sensor were dedicated to the estimation of inherent albedo whereas we focused in this study on apparent albedo. Nevertheless, we decided to test these sets in order to assess the possible discrimination of these two kinds of albedo regarding to the experimental context and the resulting inaccuracies.

Visible, near infrared and shortwave albedo calculations were performed considering the several combinations between the kernel-driven models and the coefficient sets. The shortwave albedo maps depicted values between 0.1 and 0.4. This variability was explained by the simultaneous presence of vegetative surfaces and bare soils over the site.

4.1. Intercomparison of albedo estimates

We first compared apparent and inherent albedos. Regardless of kernel-driven model, apparent and inherent albedos were very close over the visible domain (ARMSD between albedo from sets No. 4 and 7 of 0.0013 corresponding to a RRMSD of 1.4%) and significantly different over the near infrared domain (ARMSD between albedo from sets No. 5 and 8 of 0.0470 corresponding to a RRMSD of 14.4%). These results were consistent with the simulation study of Liang [15], and could be explained by a higher sensitivity of near infrared white sky and therefore apparent albedo to the atmospheric conditions. The sensitivity of the inherent and apparent albedo products to the choice of a kernel-driven model was more important over the visible domain (ARMSD between two estimates ranging from 0.0067 to 0.0137 corresponding to RRMSD ranging from 7.0 to 13.5%) than over the near infrared range (ARMSD between two estimates ranging from 0.0080 to 0.0240 corresponding to RRMSD ranging from 2.2 to 7.5%). Possible explanations were (1) a lower signal to noise ratio for visible albedo and (2) the contribution of the diffuse component that was more important over the visible domain and not well described by the kernel-driven models. Indeed, the contribution of the blue channel was significant when using sets No. 4 and 7, whereas we noted previously that the discrepancy between hemispherical reflectance estimates from the different kernel-driven models was important at 443 nm. Finally, we verified that apparent visible, near infrared and shortwave albedos were consistent by computing the ABias1 between each other. As expected for canopies and soils, visible albedos were lower than shortwave albedos (ABias about –0.1020) that were lower than near infrared albedos (ABias about –0.1044).

The comparison of inherent or apparent shortwave albedos for a given coefficient set showed that the values were systematically close from one kernel-driven model to another (an example is given in Tab. IV). On the other hand, the estimates for a given kernel-driven model could be significantly different from one coefficient set to another when considering sets devoted to the apparent albedo (an example is given in Tab. V). This emphasized that the method we used to compute apparent shortwave albedo was more sensitive to the choice of a coefficient set than to the choice of a kernel-driven model. Moreover, the difference we obtained when comparing two different sets dedicated to apparent albedo (set No. 1 and 2 for example) could be more important than the difference between inherent and apparent estimates (set No. 6 and 9 for example). A possible explanation was the theoretical inaccuracy of the coefficient sets we used and/or both the theoretical and experimental inaccuracies of the hemispherical reflectance estimates. Besides, the difference between shortwave apparent and inherent albedos could be small since the PolIDER data were collected under clear sky conditions that might correspond to a low diffuse component of incoming solar radiation. Finally, we should note that sets No. 2 and 3 provided very close results, which indicated that it was possible to avoid the use of the blue channel over which significant perturbations occurred.

4.2. Validation of albedo estimates

The comparison of the airborne estimates against field measurements required accounting for the spatial, temporal and spectral features of both the PolIDER and in-situ data. Accounting for the spatial features consisted of assessing the influence of the spatial variability over the 400 m² size PolIDER estimates that were compared against 1000 to 3000 m² size field data. This was performed by computing the coefficient of variation (standard deviation/mean value) inside both 3 × 3
and 5 × 5 PolDER pixel windows. The coefficient of variation ranged between 1 and 2%, which emphasized the negligible effect of the spatial variability around field measurement locations. Therefore, we decided to perform the validation by extracting PolDER estimates through 3 × 3 pixel windows. Accounting for the temporal features consisted of averaging the field data over the period of the PolDER data acquisition that was about 45 minutes. Accounting for the spectral features when validating PolDER visible and near infrared albedos consisted of computing shortwave albedo $a_{\text{short}}$ from visible $a_{\text{vis}}$ and near infrared $a_{\text{NIR}}$ values in order to perform a comparison with the shortwave field estimates. This was performed by expressing $a_{\text{short}}$ as a weighted summation of $a_{\text{vis}}$ and $a_{\text{NIR}}$:

$$a_{\text{short}} = k \cdot a_{\text{vis}} + (1 - k) \cdot a_{\text{NIR}}$$

(3)

where $k$ is the ratio of the incident radiation over the [400–700] nm spectral domain to the whole solar spectrum. The proposed values in the literature range between 42 and 51% [32]. Accounting for ultraviolet radiation and assuming that surface albedos are quite similar over the ultraviolet and the visible domains increases $k$ by about 5%. Therefore, we performed the validation by considering $k$ values between 0.47 and 0.56.

Examples of validation using shortwave or visible and near infrared coefficient sets are given in Figures 1 and 2. They illustrate the best results that corresponded to the Li-Ross model with set No. 6, and the Walthall model with both sets No. 7 and 8 and a $k$ value about 0.55. These comparisons showed that there were no differences between Kipp and Skye estimates after the corrections of the latter (Sect. 2). We should note that the number of validation plots were not the same from one model to another. This was a consequence of the rejection filter applied to the unrealistic hemispherical reflectance values, which was discussed in the part of this study dealing with the directional issue.

For all the possibilities, we computed both the ABias/RBias and the ARMSE/RRMSE 2 between airborne and field estimates (an example is given in Tab. VI). Sets No. 1, 2, 3 and 9 overestimated the field measurements regardless of kernel-driven model. One could think in a first time that the overestimation provided by sets No. 1, 2, 3 could result from the spectral domain Weiss et al. [39] used to calibrate these sets, as suggested by Jacob et al. [11]. Nevertheless, the results we obtained with sets No. 6 and 9 invalidated this explanation. Indeed, these two sets were calibrated over similar spectral ranges that were different from that used by

$2$. The ARMSE (Absolute Root Mean Square Error) is the mean quadratic error between predicted and observed variables. The RRMSE (Relative Root Mean Square Error) is the ratio of the ARMSE to the average of the observed values.

<table>
<thead>
<tr>
<th>BRDF model and coefficient set</th>
<th>ARMSE</th>
<th>RRMSE</th>
<th>ABias</th>
<th>RBias</th>
</tr>
</thead>
<tbody>
<tr>
<td>Walthall and set No. 1</td>
<td>0.0417</td>
<td>21.4%</td>
<td>0.0360</td>
<td>18.5%</td>
</tr>
<tr>
<td>Walthall and set No. 2</td>
<td>0.0282</td>
<td>14.5%</td>
<td>0.0180</td>
<td>9.3%</td>
</tr>
<tr>
<td>Walthall and set No. 3</td>
<td>0.0255</td>
<td>13.2%</td>
<td>0.0187</td>
<td>9.7%</td>
</tr>
<tr>
<td>Walthall and set No. 4+5</td>
<td>0.0280</td>
<td>14.5%</td>
<td>0.0224</td>
<td>11.6%</td>
</tr>
<tr>
<td>Walthall and set No. 6</td>
<td>0.0160</td>
<td>8.3%</td>
<td>0.0003</td>
<td>0.2%</td>
</tr>
<tr>
<td>Walthall and set No. 7+8</td>
<td>0.0164</td>
<td>8.5%</td>
<td>0.0049</td>
<td>2.5%</td>
</tr>
<tr>
<td>Walthall and set No. 9</td>
<td>0.0248</td>
<td>12.8%</td>
<td>0.0145</td>
<td>7.4%</td>
</tr>
</tbody>
</table>
Weiss et al. [39]. However, set No. 9 induced a similar overestimation than sets No. 2/3 whereas set No. 6 provided very small over- or under-estimations. Besides, set No. 6 provided the best validation results regardless of kernel-driven model, whereas it was devoted to the estimation of inherent albedo. We should also note that sets No. 6/9 were calibrated using bidirectional reflectance whereas we expected best results when considering hemispherical reflectance such as sets No. 1 to 3. Moreover, we obtained similar results with set No. 3 and 9 whereas they were derived considering different sensor spectral wavebands. All these observations showed that the experimental inaccuracies of the hemispherical reflectance estimates (about 5% at least on the radiometric processing and 10% of relative discrepancy over the estimates from the four kernel-driven models) were too important to isolate the errors induced by the theoretical assumptions made when calibrating the coefficient sets dedicated to shortwave albedo. Moreover, the comparison was difficult since it was not possible to separate the different factors. For instance, set No. 1 to 3 could induce an overestimation because of the spectral domain chosen for the calibration as suggested by Jacob et al. [11]; whereas the overestimation from set No. 9 could result from the use of bidirectional reflectance data to perform the calibration, since bidirectional reflectance is often higher than hemispherical reflectance.

The validation of shortwave albedo estimates through the visible and near infrared values using both couples (sets No. 4 and 5) and (sets No. 7 and 8) showed that regardless of $k$ value between 0.47 and 0.56, the best estimates were provided by the coefficient sets dedicated to apparent albedo (i.e. sets No. 7 and 8). This might be explained by the necessity to well characterize the apparent albedo when distinguishing the visible and near infrared components. Indeed, near infrared white sky and therefore apparent albedo was more sensitive to the atmospheric conditions as mentioned previously. Furthermore, it was difficult to conclude about the quality of the visible and near infrared estimates since the validation was indirect and depended on $k$ values that varied according to the meteorological conditions. We could only mention that the shortwave estimates were close to the field measurements when considering a $k$ value about 0.55, which agreed with the value proposed by Varlet-Grancher et al. [32] when considering the French national scale. Besides, the validation results varied significantly according to this ratio, with a RMSE ranging from 8.5% ($k = 0.55$) to 15% ($k = 0.47$). Such a variation emphasized the importance of this parameter when deriving shortwave albedo from visible and near infrared ones.

When validating the shortwave albedo estimates for a given coefficient set, we noted that the best estimates corresponded systematically to both Li-Ross and Walthall models, with slightly better results for Walthall (an example is given in Tab. VII). The results we obtained with Walthall were consistent with previous studies, and were explained by the model robustness when considering several land use situations [1, 31]. The results we obtained with Li-Ross were also consistent with previous studies [17, 19, 25], and could be explained by the directional performances of the “Li-Sparse/Ross-Thick” variant over vegetative areas.

### 4.3. Calibrating coefficient sets for the narrowband to broadband conversion

The last step of this study aimed at calibrating some coefficient sets dedicated to the estimation of shortwave apparent albedo from the airborne PolIDER sensor. This was motivated by the following observations. First, no proposition existed in the literature from our knowledge regarding to the airborne PolIDER spectral configuration. Second, the airborne PolIDER sensor allows acquiring high spatial resolution data that can be used as reference for large-scale studies and satellite product validation. Third, resulting analysis from this study showed that albedo estimation was more sensitive to the choice of a coefficient set than to the choice of a kernel-driven model. This emphasized the importance of the narrowband to broadband conversion.

Before performing the coefficient set calibration, we assessed the information provided by each PolIDER waveband. For this, we evaluated possible redundant information by calculating the intercorrelation coefficients between the waveband estimates for 45 observations. The results we obtained when considering the Li-Ross model are displayed in Table VIII. We obtained similar results with the other kernel-driven models. It was shown that only the NIR channel was not correlated to the others. Therefore, we concluded that considering two channels should be sufficient to compute albedo, whereas the use of the NIR channel was absolutely necessary. The information provided by the blue, green and red channels was quite the same, or could not be discriminated, because of the noise over the hemispherical reflectance data set. Indeed, the correlation coefficient

<table>
<thead>
<tr>
<th>BRDF model</th>
<th>ARMSE</th>
<th>RMSE</th>
<th>ABias</th>
<th>RBias</th>
</tr>
</thead>
<tbody>
<tr>
<td>Li-Ross</td>
<td>0.0168</td>
<td>8.7%</td>
<td>−0.0012</td>
<td>−0.6%</td>
</tr>
<tr>
<td>MRPV</td>
<td>0.0227</td>
<td>11.6%</td>
<td>0.0118</td>
<td>6.1%</td>
</tr>
<tr>
<td>Roujean</td>
<td>0.0231</td>
<td>12.0%</td>
<td>0.0073</td>
<td>3.8%</td>
</tr>
<tr>
<td>Walthall</td>
<td>0.0164</td>
<td>8.5%</td>
<td>0.0049</td>
<td>2.5%</td>
</tr>
</tbody>
</table>

Table VII. Absolute and relative RMSE and Bias between airborne and field estimates of shortwave apparent albedo. We considered sets No. 7 and 8 along with a $k$ value about 0.55.

<table>
<thead>
<tr>
<th>$\rho_{k,443}$</th>
<th>$\rho_{k,550}$</th>
<th>$\rho_{k,670}$</th>
<th>$\rho_{k,865}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1.0000</td>
<td>0.9202</td>
<td>0.8938</td>
<td>−0.5683</td>
</tr>
<tr>
<td>0.9202</td>
<td>1.0000</td>
<td>0.9443</td>
<td>−0.5549</td>
</tr>
<tr>
<td>0.8938</td>
<td>0.9443</td>
<td>1.0000</td>
<td>−0.7244</td>
</tr>
<tr>
<td>−0.5683</td>
<td>−0.5549</td>
<td>−0.7244</td>
<td>1.0000</td>
</tr>
</tbody>
</table>

Table VIII. Correlation coefficient matrix of the hemispherical reflectances over the four PolIDER channels provided by the Li-Ross model.
between these channels was about 0.9 whereas the relative discrepancy when comparing the four kernel-driven models for a given waveband was about 10%.

The calibration of the coefficient sets was performed using a pseudo-matrix inversion between the PolDER hemispherical reflectances and the field measurements of albedo:

\[
[a] = \begin{bmatrix} \beta \\ \rho_{\lambda,j} \end{bmatrix} \begin{bmatrix} 1 \\ \rho_{\lambda,j} \end{bmatrix} \Leftrightarrow \begin{bmatrix} \beta \\ \rho_{\lambda,j} \end{bmatrix} = \begin{bmatrix} 1 \\ \rho_{\lambda,j} \end{bmatrix} [a].
\] (4)

Moreover, we estimated the 90% confidence interval for the coefficients, as well as the residual error. The results we obtained are listed in Table IX. We only reported the results corresponding to a unique selection among the blue, green and red channels. Indeed, the results with the other selections were very similar, which was consistent with the correlation coefficients between these three bands. When considering only two channels, the coefficient sets were similar from one model to another, and the coefficient confidence interval ranged between 15 and 20% of the coefficient values. This emphasized once more the effect of noise occurring over the data set. Moreover, the results were physically consistent since the coefficients were positive and their sum was close to unity. Indeed, as suggested by Brest and Goward [2], the considered bands were assumed to be representative of given spectral intervals according to the surface reflectance variations, and the associated coefficients were assumed to correspond to the incoming solar radiation over these spectral ranges. When considering three or four PolDER wavebands, the results were worse despite a lower residual error. The confidence interval of the blue and green coefficients indicated that these coefficients could be equal to zero. Therefore, these channels could not be considered as noted previously. Moreover, adding the green and/or the blue channels increased significantly the uncertainty of the red coefficient. This showed that considering more than one band among the blue, green and red added more noise than information. As a consequence, the multilinear regression provided some physically unrealistic coefficients such as negative values whereas the coefficient sum was not anymore close to one, except for MRPV. Obviously, the residual error was better when considering all the bands. However, these results must be considered with many care since the fitting included the induced noise. From our results, it is recommended to consider only the red and NIR bands of the PolDER sensor. Indeed, two channels were sufficient and the red one was less noisy than the blue and green ones. However, these results depended on the observed surfaces and the accuracy of the database we used. Indeed, more accurate hemispherical reflectance estimates may emphasize the necessity to use more spectral information.

5. CONCLUSIONS

The objective of this study was to assess the accuracy of the narrowband to broadband conversion when using coefficient sets to calculate the integrated value of albedo as a linear function of hemispherical reflectances. This was performed considering shortwave, visible and near infrared albedos. The data set we used was acquired in the framework of the ReSeDA program, including remote sensing data collected with the airborne PolDER sensor and simultaneous field measurements of albedo. This multitemporal and high spatial resolution data set allowed performing comparison, validation and calibration of coefficient sets considering growth cycles of several crops while problems due to mixed pixels were reduced.

The hemispherical reflectance was estimated using four classical linear BRDF kernel-driven models: Li-Ross, MRPV, Roujean and Walthall. The intercomparison of the estimates showed a relative discrepancy of about 10%. The assessment of the model abilities to provided realistic hemispherical reflectance estimates despite the effect of noise occurring on the PolDER data showed that Li-Ross and MRPV were the most robust models.

The narrowband to broadband conversion was first assessed comparing and validating several coefficient sets proposed recently in the literature. The intercomparison of visible, near infrared and shortwave albedos allowed verifying experimentally several observations previously reported in simulation studies. The intercomparison of the several shortwave albedo estimates emphasized that the method we

<p>| Table IX. Values of the calibrated coefficients (Coef) obtained from linear regressions between the field data of albedo and the PolDER hemispherical reflectance estimates. The associated 90% confidence intervals (Co. In.) are given below the coefficients with the ± sign. The corresponded residual errors as given through ARMSE and RRMSE. |
|-----------------|--------|--------|--------|--------|--------|--------|--------|</p>
<table>
<thead>
<tr>
<th>BRDF model</th>
<th>Blue Coef</th>
<th>Green Coef</th>
<th>Red Coef</th>
<th>NIR Coef</th>
<th>Offset</th>
<th>ARMSE</th>
<th>RRMSE</th>
</tr>
</thead>
<tbody>
<tr>
<td>LSRK</td>
<td>-9.2659</td>
<td>0.8118</td>
<td>0.3978</td>
<td>-0.0025</td>
<td>0.0153</td>
<td>7.63%</td>
<td>7.94%</td>
</tr>
<tr>
<td></td>
<td>±0.4158</td>
<td>±0.3651</td>
<td>±0.0652</td>
<td>±0.0292</td>
<td>9.4%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>LSRK</td>
<td>-0.3867</td>
<td>1.0044</td>
<td>0.4122</td>
<td>-0.014</td>
<td>0.0173</td>
<td>7.63%</td>
<td>7.94%</td>
</tr>
<tr>
<td></td>
<td>±0.3488</td>
<td>±0.3954</td>
<td>±0.0648</td>
<td>±0.0295</td>
<td>9.4%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MRPV</td>
<td>-0.5836</td>
<td>0.3483</td>
<td>-0.0037</td>
<td>0.0173</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.1349</td>
<td>±0.0574</td>
<td>±0.0346</td>
<td>8.91%</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>MRPV</td>
<td>0.1029</td>
<td>0.4950</td>
<td>0.3384</td>
<td>-0.0022</td>
<td>0.0173</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.3845</td>
<td>±0.3580</td>
<td>±0.0687</td>
<td>±0.0354</td>
<td>8.91%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>MRPV</td>
<td>-0.2247</td>
<td>0.1436</td>
<td>0.5766</td>
<td>-0.0017</td>
<td>0.0170</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.3211</td>
<td>±0.3873</td>
<td>±0.3752</td>
<td>±0.0690</td>
<td>8.76%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Roujean</td>
<td>-0.5794</td>
<td>0.3253</td>
<td>0.0110</td>
<td>0.0187</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.1488</td>
<td>±0.0591</td>
<td>±0.0350</td>
<td>9.6%</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Roujean</td>
<td>-0.1720</td>
<td>0.3145</td>
<td>0.0101</td>
<td>0.0186</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.4912</td>
<td>±0.4257</td>
<td>±0.0755</td>
<td>±0.0354</td>
<td>9.6%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Roujean</td>
<td>0.5609</td>
<td>0.9673</td>
<td>0.3527</td>
<td>0.0008</td>
<td>0.0177</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.4812</td>
<td>±0.4787</td>
<td>±0.4625</td>
<td>±0.0734</td>
<td>9.17%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Walthall</td>
<td>0.0024</td>
<td>0.4803</td>
<td>0.3565</td>
<td>0.0087</td>
<td>0.0152</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.4665</td>
<td>±0.3895</td>
<td>±0.0658</td>
<td>±0.0276</td>
<td>7.92%</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Walthall</td>
<td>-0.5934</td>
<td>0.2937</td>
<td>0.6170</td>
<td>-0.0076</td>
<td>0.0144</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>±0.4667</td>
<td>±0.5017</td>
<td>±0.3879</td>
<td>±0.0631</td>
<td>7.48%</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
used was more sensitive to the choice of a coefficient set than to the choice of a kernel-driven model. The validation against field measurements showed a relative accuracy of about 9% for the shortwave apparent albedo observed. This was an acceptable result regarding to the inaccuracy of the PoIDER data processing that was at least about 5%. However, the hemispherical reflectance estimates were not accurate enough to assess the uncertainties of the assumptions made when calibrating the coefficient sets we used. An indirect validation of the visible and near infrared albedos suggested that their quality was good, whereas it was difficult to conclude because of the procedure used that depended on the ratio of visible to shortwave radiation. Nevertheless, the ratio that provided the closest estimates to field data corresponded to the value proposed in literature when considering similar regions. Regardless of considered albedo, we noted that the validation results corresponded to both the Li-Ross and Walthall BRDF kernel-driven models. Since Li-Ross was also one of the most robust models when assessing their abilities to provide realistic hemispherical reflectances despite noisy data, we could conclude that this model was the most interesting among the four ones we used, for both its accuracy and its feasibility.

Finally, we evaluated the possibility to calibrate coefficient sets devoted to the estimation of shortwave albedo from the airborne PoIDER sensor. The calibration results showed that this was difficult when using hemispherical reflectance computed from measured data. The statistical tools and physical indicators we used to assess the quality of the calibrated coefficients emphasized the sensitivity of the method to the uncertainty of the hemispherical reflectance estimates. Indeed, the latter were not accurate enough to discriminate the information provided by the blue, green and red PoIDER channels. Further investigations are necessary to provide accurate reflectance estimates that allow addressing experimentally the validity of the narrowband to broadband conversion. Finally, it was possible to calculate shortwave albedo with a relative error of about 8% using only red and near infrared channels.

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