Intercomparison of clumping index estimates from POLDER, MODIS, and MISR satellite data over reference sites

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\section*{A B S T R A C T}

Clumping index is the measure of foliage grouping relative to a random distribution of leaves in space. It is a key structural parameter of plant canopies that influences canopy radiation regimes and controls canopy photosynthesis and other land-atmosphere interactions. The Normalized Difference between Hotspot and Darkspot (NDHD) index has been previously used to retrieve global clumping index maps from POLarization and Directionality of the Earth’s Reflectances (POLDER) data at ∼6 km resolution and the Bidirectional Reflectance Distribution Function (BRDF) product from Moderate Resolution Imaging Spectroradiometer (MODIS) at 500 m resolution. Most recently the algorithm was also applied with Multi-angle Imaging SpectroRadiometer (MISR) data at 275 m resolution over selected areas. In this study for the first time we characterized and compared the three products over a set of sites representing diverse biomes and different canopy structures. The products were also directly validated with both in-situ vertical profiles and available seasonal trajectories of clumping index over several sites. We demonstrated that the vertical distribution of foliage and especially the effect of understory need to be taken into account while validating foliage clumping products from remote sensing products with values measured in the field. Satellite measurements responded to the structural effects near the top of canopies, while ground measurements may be biased by the lower vegetation layers. Additionally, caution should be taken regarding the misclassification in land cover maps as their errors can propagate into the foliage clumping maps. Our results indicate that MODIS data and MISR data, with 275 m in particular, can provide good quality clumping index estimates at spatial scales pertinent for modeling local carbon and energy fluxes.

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\section*{1. Introduction}

Leaves in canopies are generally grouped into various sub-canopy structures such as tree crowns, branches and shoots. These structures make the leaf spatial distribution non-random. The foliage clumping index (CI) is used to quantify the degree of deviation of this distribution from the random case (Chen and Black, 1992; Nilson, 1971). A CI value >1 implies that the foliage is regularly distributed; a CI = 1 a random distribution and if CI < 1 the foliage is more clumped than random (Chen et al., 2005; or see e.g. Chen, 1996 for in-depth discussion). The CI is an important parameter for the correct assessment of true leaf area index (L_t) from usually measured effective leaf area index (L_e; L_t = L_e/CI; Chen, 1996), and is also required for estimation of sunlit and shaded leaf fractions in the canopy (Govind et al., 2013) and for accurate modeling of the canopy-level gross primary production (GPP) (Baldocchi and Harley, 1995; Ryu et al., 2011). Chen et al. (2012) recently showed that global GPP can be overestimated by as much as 12% even when accurate L_t is available but clumping is ignored.
The CI can vary considerably even within a particular land cover type (Pisek et al., 2011a), and it is highly desirable to map the spatial distribution of this index using remote sensing data (Chen et al., 2003). Previous studies (Chen et al., 1999; Lacaze et al., 2002) have shown that directional reflectance has a potential to estimate CI. Sensors such as POLarization and Directionality of the Earth’s Reflectances (POLDER; Deschamps et al., 1994; Lier and Bach, 2008) and Multi-angle Imaging Spectroradiometer (MISR; Diner et al., 2002) are especially well suited for this purpose because they can acquire surface reflectances at multiple view angles of the same ground position on one orbit. POLDER is best used at the global scale because of its ~6 km nadir resolution and high angular resolution while MISR can be used at both global and regional scales with its 275 m resolution. In turn, Moderate Resolution Imaging Spectroradiometer (MODIS) at 500 m resolution currently provides the finest pseudo multi-angular data for the global land surface (Schaefer et al., 2002).

In this study we compared the most recent CI products from the respective sensors (POLDER – Pisek et al., 2010a; MODIS – He et al., 2012; MISR – Pisek et al., 2013). The evaluation of temporal and spatial consistency between the products was assessed by using (a) selection of sites representing the global distribution of biomes, (b) a set of sites with vertical profiles of clumping, and (c) sites with available CI estimates covering a complete seasonal cycle.

2. Materials and methods

2.1. Foliage clumping estimates from remote sensing data

The main characteristics of the three most recent clumping products from the different multi-angle sensors investigated in this work are listed in Table 1. The CI can be estimated by using multi-angular remote sensing data because the variabilities in reflectances for different sun-sensor configurations contain information about the canopy structure (Goel, 1988; Li et al., 1995). The hotspot, where sun and view geometry coincide, and the darkspot, where the reflectance is at its minimum, form the basis of retrieving the clumping index from the angular signature in all three products. The Normalized Difference Hotspot–Darkspot (NDHD) index currently provides the finest pseudo multi-angular data for the global land surface (Schaefer et al., 2002).

Table 1

| Name | Spatial resolution | Algorithm | Parameterization | Global map | Temporal smoothing | References |
|------|-------------------|-----------|------------------|------------|-------------------|------------|
| POLDER | 6 km             | Model derived NDHD(NIR)–CI relationship | Vegetation type | Yes         | No                | Pisek et al. (2010a) |
| MODIS  | 300 m            | Model derived NDHD(red)–CI relationship | Vegetation type | Yes         | Yes               | He et al. (2012)    |
| MISR   | 275 m            | Model derived NDHD(red)–CI relationship | Vegetation type | Possible    | No                | Pisek et al. (2013) |

The CI can be estimated by using multi-angular remote sensing data because the variabilities in reflectances for different sun-sensor configurations contain information about the canopy structure (Chen et al., 2005). The hotspot acts as the normalizing factor that should minimize the dependence of the NDHD index on foliage optical properties that are important determinants of bidirectional reflectance (Asner, 1998).

2.2. POLDER data

The first global CI maps were derived using different generations of POLDER sensors that measure directional and polarized reflectance with an approximately 6 km nadir resolution (Chen et al., 2005; Pisek et al., 2010a). During a single satellite overpass, a surface target was scanned up to 14 (POLDER 1) or 16 times (POLDER 3) under different viewing angles, which also sampled the principle plane of the BRDF. It therefore provided a direct measure of the hotspot and darkspot for NDHD calculation. The final global CI map from POLDER observations (Pisek et al., 2010a) represents the mean annual clumping values calculated from the successful retrievals of POLDER-1 (October 1996 – June 1997) and POLDER-3 (the entire year 2005) (Fig. 1A). Gaps in the global coverage by POLDER observations (5% of vegetated areas, mainly in the tropics due to persistent cloud cover) were filled with mean CI values calculated from the successful retrievals over the same biomes for the dominant land cover types from the GLC2000 map. More details can be found in Pisek et al. (2010a).

2.3. MODIS data

He et al. (2012) derived a global CI map at 500 m resolution using the Bidirectional Reflectance Distribution Function (BRDF) product in 2006 from MODIS (Schaaf et al., 2002). While computing the NDHD using the red band (620–670 nm), He et al. (2012) reported that the hotspot calculated from the MODIS BRDF product was underestimated in comparison with POLDER measurements very near the hotspot. Without correcting the bias in the MODIS data, the MODIS-derived CI could be overestimated. He et al. (2012) developed an approach to correct the MODIS hotspot magnitude with co-registered POLDER-3 data acquired at about the same time. After the MODIS hotspot is corrected and the NDHD is calculated, the coefficients (A and B), calculated from the second-order polynomial fit of the tabulated relationship between CI and NDHD in Chen et al. (2005), were used to derive MODIS CI. He et al. (2012) assigned a single annual CI value, the median from its noisy seasonal trajectory, to each pixel in the map used in this study (Fig. 1B).

2.4. MISR data

MISR consists of nine cameras arranged to view along track that acquire image data with nominal view zenith angles relative to the surface reference ellipsoid of 0.0°, ±26.1°, ±45.6°, ±60.0°, and ±70.5° (forward and aftward of the Terra satellite) in four spectral bands (446, 558, 672, and 866 nm). In the global mode, the 672 nm (red) band images are acquired with a nominal maximum cross-track ground spatial resolution of 275 m in all nine cameras and information from all bands is provided at this resolution in the nadir camera as well (Diner et al., 1998, 2002; Pisek et al., 2013).
used the 275 m resolution data to obtain surface bidirectional reflectance factors (BRFs). The Rahman–Pinty–Verstraete (RPV) model (Rahman et al., 1993) was inverted against these full-resolution BRF values in the red band. The initial hotspot (HS) and darkspot (DS) values in the principal plane were reconstructed for solar zenith angle of 60° using the four kernel coefficients for a given pixel from the RPV model inversion. Because the MISR-hotspot might still be underestimated due to original data acquisition away from the principal plane, the same hotspot correction of He et al. (2012) as in case of MODIS is applied for MISR data as well. Using the above described approach, seasonal time series of CI can be reconstructed from the available MISR observations for a given location.

3. Methodology

The intercomparison procedure was defined to comply as much as possible with the best practices proposed by the CEOS WGCV LPV subgroup (Garrigues et al., 2008; Baret et al., 2009). It corresponds to Stage 1 validation as defined by the CEOS

| Network | Site | Country | Lat (deg) | Lon (deg) | Land cover |
|---------|------|---------|-----------|-----------|------------|
| LPV     | Aek Loba | Indonesia | 2.63 | 90.58 | Palm tree plantation |
| LPV     | Albemal | NC, USA | 36 | –78 | ENF |
| LPV     | Alpilles2 | France | 43.81 | 4.71 | Crops |
| VALERI  | Barax | Spain | 39.07 | –2.1 | Crops |
| LPV     | Bondville | IL, USA | 40.01 | –86.29 | ENF |
| LPV     | BOREAS NSA | Canada | 55.87 | –98.48 | ENF |
| LPV     | BOREAS SSA BERMS | Canada | 53.65 | –105.32 | ENF |
| LPV     | Braschaaat (De Inslag) | Belgium | 51.3 | 4.51 | MF |
| VALERI  | Camerons | Australia | –32.6 | 116.25 | EBF |
| VALERI  | Chibolton | UK | 51.16 | –1.43 | Crops and forest |
| VALERI  | Concepcion | Chile | –37.47 | –71.47 | MF |
| LPV     | Counami | French Guiana | 5.35 | –53.24 | EBF |
| VALERI  | Counami | French Guiana | 5.35 | –53.24 | EBF |
| VALERI  | Demmin | Germany | 53.89 | 13.21 | Crops |
| VALERI  | Donga | Benin | 9.77 | 1.78 | Grassland |
| LPV     | Flakaliden | Sweden | 64.11 | 19.45 | ENF |
| LPV     | Funduca | Romania | 44.41 | 26.58 | Crops |
| VALERI  | Gilching | Germany | 48.08 | 11.32 | Crops and forest |
| VALERI  | Gngaara | Australia | –31.53 | 115.88 | EBF |
| LPV     | Gourma | Mali | 15.32 | –1.55 | Grassland |
| LPV     | Guanacaste | Costa Rica | 10.87 | –85.66 | Tropical dry forest |
| VALERI  | Haouz | Morocco | 31.66 | –7.6 | Crops |
| LPV     | Harvard Forest | MA, USA | 42.5393 | –72.1779 | DBF |
| VALERI  | Hirsikangas | Finland | 62.64 | 27.01 | ENF |
| VALERI  | Honombi | Mali | 15.33 | 1.48 | Grassland |
| VALERI  | Hyytiälä | Finland | 61.85 | 24.29 | ENF |
| LPV     | Järvesjä | Estonia | 58.31 | 27.3 | ENF |
| LPV     | Kejimkujik NP | Canada | 44.45 | –65.28 | MF |
| LPV     | Konza Prairie | KS, USA | 39.09 | –96.57 | Crops |
| LPV     | Krasnoyarsk | Russia | 57.27 | 91.6 | ENF |
| LPV     | Laprida | Argentina | –36.99 | –60.55 | Grassland |
| VALERI  | Laprida | Argentina | –36.99 | –60.55 | Grassland |
| VALERI  | Larose | Canada | 45.38 | –75.22 | MF |
| LPV     | Le Larzac | France | 43.9375 | 3.123056 | Tropical moist forest |
| LPV     | Los Inocentes | Costa Rica | 11.01 | –85.49 | Herbageous |
| LPV     | Maun | Botswana | –19.92 | 23.59 | ENF |
| LPV     | Metolius (old pine) | OR, USA | 44.49 | –121.62 | ENF |
| LPV     | Metolius (young pine) | OR, USA | 44.43 | –121.56 | ENF |
| LPV     | Mongu | Zambia | –15.44 | 23.25 | Shrub |
| LPV     | Nkana | Zambia | 44.57 | –1.04 | ENF |
| LPV     | Okwa River | Botswana | –22.41 | 21.71 | Shrub |
| LPV     | Park Falls | WI, USA | 45.946 | –90.272 | ENF |
| VALERI  | Plan De Dieu | France | 44.2 | 4.95 | Crops |
| LPV     | Puechabon | France | 43.72 | 3.65 | EBF |
| LPV     | Rumilly-sur-Seine | France | 48.44 | 3.77 | Crops |
| VALERI  | Roivaneimi | Finland | 66.46 | 25.35 | ENF |
| LPV     | Ruokolatti | Finland | 61.53 | 28.71 | ENF |
| LPV     | Sevilleta | NM, USA | 34.35 | –106.69 | Shrub |
| LPV     | Siera Chincua | Mexico | 19.82 | –100.28 | ENF |
| VALERI  | Siera Chincua | Mexico | 19.82 | –100.28 | ENF |
| LPV     | Skukuza | South Africa | –25.02 | 31.497 | Shrubland/woodland |
| VALERI  | Sonian | Belgium | 50.77 | 4.41 | ENF |
| VALERI  | Sud-Ouest | France | 43.51 | 1.24 | Crops |
| LPV     | Tapajos | Brazil | –2.857 | –54.959 | EBF |
| LPV     | Ticino | Italy | 45.201 | 9.058 | Poplar plantation |
| LPV     | Tshane | Botswana | –24.16 | 21.89 | Herbaceous |
| LPV     | Turco | Bolivia | –18.24 | –68.18 | Shrub |
| VALERI  | Wankama | Niger | 13.65 | 2.64 | Grassland |
| LPV     | Watson Lake | YK, Canada | 60.09 | –129.38 | MF |
| LPV     | Whitecourt | AB, Canada | 54.03 | –115.78 | MF |
| VALERI  | Zhang Bei | China | 41.28 | 114.69 | Grassland |
was assigned to each pixel in global POLDER (Pisek et al., 2010a) very little meaningful information. Additionally, a single CI value good quality remote sensing data (Pisek et al., 2013). The intercom-
parison retrievals were thus simply centered over each validation site similarly to what was done in the previous Section 3.1.

3.2. Comparing clumping products with in situ measurements

The CI products were validated over an additional global validation dataset with measured vertical or temporal profiles of foliage clumping from Pisek et al. (2013) that was further expanded with additional sites to represent all the main biomes. Detailed site descriptions are provided in Table 3. The methodology to obtain in situ CI estimates has been previously described in detail in Pisek et al. (2013). The field data should be optimally integrated with high resolution imagery to allow a real product validation (Morissette et al., 2006). Unfortunately, with only one exception of a limited extent high resolution map of clumping index (<1 km²) by Simic et al. (2010), no such maps are currently available, allowing only a limited evaluation. Due to the current absence of high-resolution CI maps, the remote sensing retrievals were simply centered over each validation site similarly to what was done in the previous Section 3.1.

3.2.1. Vertical profiles of clumping index

In-situ measurements of CI at different heights using towers were available for twelve of the field sites. We measured two northern boreal evergreen needleleaf stands in Hyytiälä (61.85° N, 24.29° E) and Sodankylä (67.36° N, 26.64° E), Finland. Scots pine (Pinus sylvestris L.) was dominant in both stands. The forest floor vegetation in Hyytiälä was dominated by lingonberry, blueberry, lichens and mosses (Ilvesniemi et al., 2009) and by fork moss with lichens at Sodankylä (Manninen et al., 2012). The third boreal ever-

green needleleaf stand was near Sudbury, Canada (47.16° N, 81.75° W). The overstory vegetation was formed by short black spruce (Picea mariana) trees (~5.6 m); the understory vegetation consisted mainly of feather moss (Hylocomium splendens) with contributions from labrador tea (Ledum groenlandicum) and leather leaf (Chamaedaphne calyculata) (Pisek et al., 2010b). All three boreal forest sites lacked a tall understory vegetation.

The understory was also virtually missing at the three ever-

green broadleaf forest sites. The Mediterranean oak (Quercus ilex) stand in Castelporziano, Italy (41.71° N, 12.38° E) contained only a few Pistacia lentiscus bushes in the understory layer. The Wombat forest research site (~37.42° S, 144.09° E) is located in the Wombat State Forest, Victoria, SE Australia. The site is a secondary re-
growth Eucalyptus forest that was last harvested in 1980. Dominant tree species are Messmate Stringybark (Eucalyptus obliqua), Narrow Leaf Peppermint (Eucalyptus radiata) and Candlebark (Eucalyptus rubida) with an average canopy height of 25 m. The understory consists mainly of patchy grasses. The second dry sclerophyll site at Whroo (~36.67° S, 145.03° E) in Victoria, Australia is box iron-
bark woodland with lower tree height and canopy cover. The veget-
ation was dominated by two main Eucalypt species: Gray Box (Eucalyptus microcarpa) and Yellow Gum (Eucalyptus leucoryzon). The mean tree height at Whroo was 15.3 ± 0.2 m.

Warra Long Term Ecological Research (LTER) site (~43.09° S, 146.66° E; Neyland et al., 2000) is located in SW Tasmania, Australia. It represents a tall E. obliqua wet forest with rainforest understory and a dense man-fern (Dicksonia antarctica) ground-layer. The for-
est stands around the Warra site had mature heights in excess of 55 m: the tallest E. obliqua within the LTER reaches a height of 90 m.

Two native cloud forest stands were located in Thurston Lava Tube (19.41° N, 155.23° W) and Laupahoehoe (19.93° N, 155.29° W), Hawaii, USA. The Thurston Lava Tube site consists

3.1. Pair-wise comparison over LPV/VALERI sites

First we retrieved CI values from the corresponding products over a set of 63 globally distributed Land Product Validation (LPV) and Validation of Land European Remote sensing Instruments (VALERI) sites (Table 2) that represent a recommended pool of sites for the systematic intercomparison of land biophysical products (Baret et al., 2006; Garrigues et al., 2008; Nightingale et al., 2011). For analysis requiring direct product-to-product compari-
ison, the common approach is to resample them to 3 × 3 pixel size to reduce effects from co-registration inaccuracies and Point Spread Function differences (see e.g. Camacho et al., 2013; D’Odorico et al., 2014). However, here intercompared products contain clumping retrievals at very different spatial scales, ranging from 275 m (MISR) to ~6 km (POLDER). Further resampling to 3 × 3 pixel size would produce values at a scale (~18 km) with very little meaningful information. Additional, a single CI value was assigned to each pixel in global POLDER (Pisek et al., 2010a) and MODIS maps (He et al., 2012), while MISR results have been presented so far in the form of temporal trajectories over selected areas, where temporal resolution varied based on the availability of good quality remote sensing data (Pisek et al., 2013). The intercom-
parison retrievals were thus simply centered over each validation site. MISR retrievals were made using the closest in time available good quality MISR BRF data to the date of ground measurements of biophysical parameters at each site. Given the fact that all three products used the identical NDHD–CI algorithm by Chen et al. (2005) and MODIS and POLDER maps are already in the same (inverted sinusoidal) projection, the differences between the indi-
vidual products should highlight mainly the changes in vegetation heterogeneity with spatial scale. The consistency is further evalu-
ated by intercomparing the bulk distribution of the available global product values per biome type.

![Fig. 1. Global maps of foliage clumping from POLDER (A) and MODIS (B).](image-url)
Table 3
Characteristics of the validation sites with vertical or seasonal profiles of clumping.

| Site Description | Location | Lat   | Lon   | Forest type | Overstory | Mean tree height (m) | Understory | Reference          | In-situ data collection |
|------------------|----------|-------|-------|-------------|-----------|----------------------|------------|---------------------|------------------------|
| Hyytiala         | Finland  | 61.85 | 24.29 | NEF         | SP        | 16                   | Lingonberry, blueberry and mosses | Ilvesniemi et al. (2009) | 2001/11               |
| Sodankyla        | Finland  | 67.36 | 26.64 | NEF         | SP        | 12                   | Lichen, fork moss | Rautainen et al. (2007) | 2007/9                 |
| Sudbury          | Canada   | 47.16 | –18.75| NEF         | BS        | 5.6                  | Feather moss, labrador and leather tea | Pisek et al. (2010b) | 2007/6                 |
| Laupahoehe       | HI, USA  | 19.93 | –155.3| EBF         | M, K      | 19                   | Cibotium spp. | Kellner and Assner (2009) | 2013/1                 |
| Thurston Lava    | HI, USA  | 19.41 | –155.2| EBF         | M         | 14.5 ± 1.4           | Cibotium glaucum (3.8 ± 2.7 m) | Giambelluca et al. (2009) | 2010/9                 |
| Tube Castelporziano | Italy    | 41.71 | 12.38 | EBF         | MEO       | 16                   | Pistacia lentiscus | Fares et al. (2014) | 2014/6                 |
| Wombat           | Australia| –37.42| 144.09| EBF         | EO, ERA,  | 25                   | Patchy grass | Haverd et al. (2013) | 2013/7                 |
| Whroo Warra      | Australia| –36.67| 145.03| EBF         | EM, EL,   | 55                   | Patchy grass | New site | 2013/7                 |
|                  |          |       |       |             | EO        |                      | Nothofagus cunninghamii, Atherosperma moschatum, Eucryphia lucida, Phylici cladus asplenifolius | Neyland et al. (2000) | 2013/8                 |
| Jarvselja        | Estonia  | 58.27 | 27.27 | MF          | SB, BA, NS| 17                   | Suppressed tree layer (mean height of 6.4 ± 0.6 m) | Noe et al. (2011) | 2011/7                 |
| Morgan-Monroe    | IN, USA  | 39.32 | –86.41| BDF         | SM, TP, S | 27                   | Max. understory height 10 m | Oliphant et al. (2006) | 2005/6                 |
| State Forest     | Hesse    | France| 48.67 | 7.06   | BDF       | 22                   | Sparse grass (Nov-Apr) | Longdoz et al. (2008) | 2014/8                 |
|                  |          |       |       |             | EB        |                      |                        | Sprintsin et al. (2011) | 2005                   |
|                  |          |       |       |             |          |                      |                        | Ryu et al. (2012) | 2009/7–2010/3           |
|                  |          |       |       |             |          |                      |                        | Kuusk et al. (2013) | 2011/4–2010/10          |
| Honghe           | China    | 47°39.11′ | 133°31.31′ | CRO |             | –                    | Ledum palustre, Eriophorum vaginatum, continuous Sphagnum spp. moss layer | Fang et al. (2014) | 2012/6–2010/10          |

In the column “Forest type” NEF – needleleaf evergreen forest, EBF – evergreen broadleaf forest, BDF – broadleaf deciduous forest, MF – mixed forest, S – savanna, CRO – cropland. In the column “Overstory” SP – Scots pine, M – Metrosideros polymorpha, K – Koa, MEO – Mediterranea oak, EO – Eucalyptus obliqua, Era – Eucalyptus radiata (narrow leaf peppermint), ERAu – Eucalyptus rubida (Candlebark), EM – Eucalyptus microcarpa (Gray Box), EL – Eucalyptus leucoxylon (Yellow Gum), SM – sugar maple, TP – tulip poplar, S – sassafras, WO – white oak, BO – black oak, SB – silver birch, BA – black alder, NS – Norway spruce, EB – European beech, AP – Aleppo pine, BIO – blue oak, GP – gray pine.

primarily of a single canopy species, ohia lehua (Metrosideros polymorpha), with a dense understory layer of hapu‘u ferns (Cibotium spp.) [Giambelluca et al., 2009]. Laupahoehe had similarly comprised overstory with an additional dominant species, Koa (Acacia koa) (Kellner and Assner, 2009). The successional deciduous broadleaf stand in Morgan–Monroe State Forest (39.32° N, 86.41° W) in Indiana, USA, was comprised predominantly of sugar maple (Acer saccharum), tulip poplar (Liriodendron tulipifera), sassafras (Sassafras albidum), white oak (Quercus alba), and black oak (Quercus nigra). The canopy vertical structure was fairly consistent around the tower with peaks in L_t occurring at the crown level at approximately 20–30 m and at the undergrowth level at approximately 0–10 m (Oliphant et al., 2004).

The deciduous broadleaf type was also represented by an experimental plot located in the state forest of Hesse (48.67° N, 7.06° E) in north east of France. The stand was comprised mainly (90%) of European beech (Fagus sylvatica) with a mean tree height ~22 m. Due to canopy closure, understory vegetation is very sparse. Granier et al. (2000) provide a more detailed description of the site.

A scaffolding tower in Jarvselja, Estonia (58.31° N, 27.30° E) was located in a hemiboreal-mixed stand with co-dominant species of silver birch (Betula pendula Roth.), black alder (Alnus glutinosa L.) and Norway spruce (Picea abies (L.) Karst.). A suppressed tree layer (mean height of 6.4 ± 0.6 m) was present around the tower and surrounding forest (Noe et al., 2011). 3.2.2. Seasonal variation of clumping index

Seasonal trajectories of CI were available for four additional sites. Yatir forest, Israel (31.35° N, 35.03° E), is a monoculture plantation which is dominated by Aleppo pine (Pinus halepensis Mill.). Sparse understory vegetation develops only during the rainy season (November–March) and disappears shortly thereafter (Grünzweig et al., 2003).

An oak-savanna ecosystem in California, USA (Tonzi; 38.43° N, 120.96° W), was dominated by blue oak trees (Quercus douglasii) with occasional (<10%) gray pines (Pinus sabinina) (Baldocchi et al., 2004).

The third site with a seasonal trajectory of CI was a Scots pine (P. sylvestris L.) stand in Järvselja, Estonia (58.31° N, 27.30° E). The site was very homogeneous with respect to its horizontal structure and gap fraction (Pisek et al., 2011). Forest understory vegetation was composed of sparse labrador tea and cotton grass, and a continuous Sphagnum moss layer. The site is included in the Radiation transfer Model Intercomparison (RAMI, http://ramibenchmark.jrc.ec.europa.eu/HTML/Home.php) exercise (Kuusk et al., 2010).

The last validation site representing croplands was located at the Honghe Farm (47.65° N, 133.52° E) in the Heilongjiang province, NE China. The area was dominated with large homogeneous paddy rice fields (>5 km² homogeneity). The rice-cropping practices were uniform, growing a single rice variety (Japanica) once a year during the summer season (May to September).
4. Results and discussion

4.1. Pair-wise product intercomparison over the LPV/VALERI sites

The pair-wise comparisons between CI values from different products centered over the 63 sites from LPV/VALERI networks are shown in Fig. 2. CI values from MODIS and MISR showed the best overall agreement (Fig. 2A). This is not surprising, since the two products use the same wavelength domain (visible red) for the CI retrieval and they are also the closest in resolution scale (500 m vs. 275 m). The relative distribution of CI values between vegetation types coincided between the two products as well. Needleleaf forests are the most clumped vegetation type, followed by mixed and deciduous forests, shrubs and crops, and with grasslands appearing to be the least clumped (closest to the random distribution). The MODIS CI product indicated a much wider CI range over intercomparison sites with needleleaf forests (0.47–0.72) compared to MISR (0.52–0.59). Depending on the land cover, different coefficients were applied to estimate CI from Eq. (1) (Chen et al., 2005). All MISR retrievals coincided with the LPV/VALERI needleleaf designations over the respective sites. The MODIS CI product used the GLC2000 land cover map (Bartholomé and Belward, 2005), which can differ from the actual vegetation types present at individual LPV/VALERI sites. He et al. (2012) previously noted that CI can be seriously biased by using a wrong land cover type. On the contrary, in our study the MISR retrievals varied markedly over grasslands (0.52–1.0), while MODIS retrievals were confined to much narrower range (0.72–0.86). There is no specific algorithm for CI retrieval over non-forested areas; coefficients for broadleaf trees are applied in the respective products over these areas, instead. The modeled results by Chen et al. (2005) suggested that areas with less than 25% vegetation coverage or fragmented land cover should be treated with caution. Our results confirm this as well, especially retrievals over non-forested areas should be taken with pre-caution. Interestingly, MODIS and MISR retrievals agree quite well over fields with crops (Fig. 2A), which confirmed similarly homogeneous vegetation coverage at both scales of 275 m and 500 m over the majority of the intercomparison sites.

The distribution of CI values for needleleaf forests was more similar along the 1:1 line in case of the MODIS–POLDER pair-wise comparison (Fig. 2B). This is not surprising, since both products use the same GLC2000 land cover map, and most of the intercomparison sites with forests were located within larger areas with homogeneous vegetation. Fig. 2B also confirmed that using different bands for the CI retrieval (red from MODIS vs. NIR from POLDER) introduced no systematic bias. The POLDER CI product included two clear outliers with very low CI values (high clumping) (Fig. 2B and C): BOREAS BERMS (0.34) and Watson Lake (0.44). Given the location of the respective sites, such values indicate the effects of topography on CI retrievals from POLDER data had not been entirely removed by Pisek et al. (2010a). The MISR–POLDER comparison offered the least agreement (Fig. 2C). The gradually decreasing agreement between products from Fig. 2A to C confirmed the importance of using CI value appropriately matched to the scale of the application in question (Ryu et al., 2010).

The mean CI and its one standard deviation for each GLC2000 land cover type from the available MODIS and POLDER global CI maps are shown in Table 4. There was not much difference in CI value distributions from the two maps in Fig. 1 with exception of
needleleaf forests, where MODIS CI values appeared to be lower by ~0.1 than POLDER retrievals. This agreed with the observations from Fig. 2B, suggesting that despite the limited number of inter-comparison sites Fig. 2 offered a good initial overview of the differences between the CI products.

4.2. Variation of clumping index across canopy depths

It should be acknowledged that satellite measurements respond primarily to the structural effects in upper levels of canopies. Chen et al. (2005) developed the CI algorithms using $L_t$ input and the resulting gap fraction simulated by the geometrical optical model 4-Scale (Chen and Leblanc, 1997). The CI algorithm should enable the correct retrieval of the stand average clumping of the leaves from remote sensing data. On the other hand ground level CI measurements may be biased by presence of an understory at forest sites, since the two layers can be differently spatially aggregated. There was no pronounced tree/shrub understory layer at the first three validation sites, representing boreal needleleaf forests (Fig. 3A–C). In situ CI estimates from different heights in the canopy were then very similar, which also confirms previous modeling results by Nilson et al. (2011). The three needleleaf sites with vertical profiles represent forests with different age, height, and species. There was a close agreement of MISR CI retrievals with in situ CI values in all three cases. The MISR CI values also agreed quite well with available in situ measurements over VALERI inter-comparison sites (Table 2). This suggests that MISR was indeed capable of producing quality CI estimates over this vegetation type.

Fig. 3. Vertical profiles of foliage clumping from in-situ measurements with ±1 standard deviation bars. Clumping index values from MISR data obtained around the same time are marked with vertical dashed red line; MODIS – blue dotted line; POLDER – gray thick dashed line. The mean height of undergrowth layer is marked by a green horizontal dashed line; green areas mark the ±1 standard deviation area. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)
MODIS CI values were also comparable with in situ measurements over two needleleaf sites with exception of Hyytiälä (Fig. 3A). However, Pisek and Oliphant (2013) previously noted that this disagreement was due to wrongly assigned broadleaf forest GLC2000 land cover type over this site. If the MODIS pixel with the Hyytiälä site was classified correctly as a needleleaf forest, the retrieved MODIS CI value (0.53) agreed well with the CI measured at the ground (0.52). The differences between assigned land cover types in the GLC2000 map and identified in immediate area around towers in Hyytiälä and Sodankylä (Table 1) also explained the very different POLDER CI values. When the land cover is very homogeneous and matching dominant vegetation type was assigned (Sudbury; Fig. 3C), POLDER CI value was very close to the in situ CI measurements as well.

Similarly good agreement between ground level in-situ, MISR, and MODIS CI retrievals was observed at eucalypt sites in Australia (Fig. 3D–F). The Whroo tower (Fig. 3E) was located rather close to the edge of the forest – the MISR CI value at 275 m still picks up the signal from the surrounding forest area, while MODIS and POLDER signals at coarser resolutions are already clearly influenced by the surrounding non-forested (less clumped) area. In the POLDER case such influence was evident over the Wombat site as well (Fig. 3D). Castelporziano represented another validation site without pronounced understory vegetation. The state forest reserve around the tower was large enough to occupy the full footprint of POLDER sensor, and POLDER CI value then offered the best match with the in situ measurements (Fig. 3G).

There was a pronounced understory layer present at the remaining sites with evergreen vegetation and measured vertical profiles of clumping (Fig. 3H–I). The remotely sensed CI values can then differ from CI measured at the ground. In the native cloud forests in Hawai‘i (Fig. 3H–I), large gaps between tree crowns at upper levels of the canopy may not be measured near the ground due to occlusion by lower vegetation fern branches. Instead, the MISR CI values in particular were close to in situ measurements obtained above the nearly uniform understory fern layer at ~6 m (Fig. 3H–I). The vertical profile at Thruston Lava Tube site differed from Laupahoehoe because there was an additional dominant species A. koa in the overstory at Thruston Lava Tube. There was a frequent cloud cover over both Hawai‘i sites which severely limits the opportunities for acquiring good quality remote sensing data. Due to the missing data, the POLDER CI value (0.64) over the two sites was originally filled and corresponded to mean CI value calculated from the successful retrievals over the same biome (Pisek et al., 2010a). Such values need to be treated with caution, as they obviously cannot correctly reflect possible local variations in vegetation structure (Fig. 3H–I).

A suppressed tree layer was also present at the two broadleaf stands in Indiana (Fig. 3J) and Estonia (Fig. 3K). The best agreement between satellite and field CI values both at MMSF and in Järvesela was again achieved for observations taken above the understory layer (Fig. 3J–K). Results over Hesse (Fig. 3L) documented that CI retrievals from remote sensing data can match the ground in situ measurements in deciduous broadleaf forests if the land cover vegetation type is correctly assigned, the forest area is sufficiently large, and there is no pronounced shrub/tree understory layer.

4.3. Seasonal variation of clumping index

The land surface modeling community has assumed that clumping is constant over seasons (Baldocchi et al., 2002; Houborg et al., 2009; Sampson et al., 2006) and thus its temporal variation has been ignored. Fig. 4 shows the clumping may change with season even for evergreen needleleaf forests due to needle phenology (Sprötsin et al., 2011). At Yatir and Tonzi the POLDER CI values matched quite well the seasonal clumping minima (Fig. 4A and B). Clumping was underestimated in both maps from POLDER and MODIS over Järvesela RAMI stand (Fig. 4C). Similarly to the Hyytiälä case (Fig. 4A), this disagreement was caused by wrongly assigned forest GLC2000 land cover type over this site (He, pers. comm.). MODIS and POLDER CI values closely matched the in situ measurements in Järvesela RAMI stand after assigning the correct land cover type (needleleaf). The POLDER CI value did not agree with seasonal clumping minima from in situ measurements at Honghe site (Fig. 4D). This was due to insufficient vegetation coverage for correct clumping information retrieval from the satellite data at the beginning and the end of season. MISR CI seasonal minima also leveled off around the same value as POLDER (0.84; Fig. 4D), confirming the effect of insufficient vegetation coverage during the non-growing season at non-evergreen sites. Fig. 4 illustrates the potential of MISR to track successfully the seasonal trajectories of clumping. The retrieved trajectories were also rather stable. No smoothing algorithm was applied within the MISR CI processing chain (Pisek et al., 2013). The stability is not surprising since the primary mission of the MISR instrument was to study the Earth atmosphere and, in particular, to characterize atmospheric aerosols and clouds (Diner et al., 1998; Verstraete et al., 2012). Significant efforts have been invested to address these issues in great detail (Diner et al., 2005) and provide quality Level 2 products that are utilized to convert MISR TOA BRF into surface BRF values at 275 m resolution. At the same time, the MISR product quality requirements are rigorous (Bothwell et al., 2002), and no good quality Level 2 data may be available for extended periods of time (Fig. 4C and D). Fig. 4C and D demonstrates that gaps in seasonal trajectory of CI can be effectively filled using MISR observations from other years over sites with stable land cover, when the atmospheric conditions were more favorable.

5. Conclusion

In this study we compared the most recent CI products from the space-borne multi-angular POLDER, MODIS, and MISR sensors for
the first time. This exercise corresponded to Stage 1 validation as defined by CEOS (Nightingale et al., 2011; Weiss et al., 2014). Our main results highlight the following:

1. Satellite measurements responded to the structural effects near the top, while ground measurements may be biased by the lower vegetation understory layers.
2. POLDER CI map (Pisek et al., 2010a) may be used to predict the upper boundary of seasonal clumping. The coarse spatial resolution of the POLDER map at ~6 km presents the main challenge for the product validation with in situ measurements.
3. CI values in MODIS global clumping map by He et al. (2012) corresponded to the median CI from the seasonal trajectories of clumping. The MODIS CI map by He et al. (2012) with its spatial resolution at 500 m compared to ~6 km from POLDER might be also more suitable given the spatial resolution of current land surface models (e.g. Houborg et al., 2009).
4. If more detailed information is required, MISR retrievals can track correctly seasonal developments of clumping as well.
5. Correct land cover information (deciduous vs. needleleaf) is crucial for retrieving accurate CI value. Furthermore, spaceborne sensors cannot provide correct CI estimates over areas with insufficient vegetation coverage (<25%).

The field data should be optimally integrated with high resolution imagery to allow a more thorough product validation and intercomparision (Morsiette et al., 2006). The current lack of such high resolution CI maps presents the main challenge for the next, more in-depth validation stages as defined by CEOS. Given the previously documented importance of foliage clumping on correct estimation of global terrestrial gross primary productivity (e.g. Chen et al., 2012; Ryu et al., 2012), production of higher resolution maps of foliage clumping, such as using UAVs equipped with BRF sensors (Kuusk et al., 2014), is strongly encouraged.

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