Development of Global land cover classification algorithms and validation methods

全球土地被覆分類アルゴリズムと 分類精度検証方法の開発

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Tasks of GCOM-C1 (FY2013)

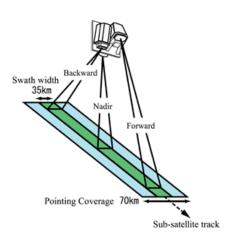
- Development of the classification method for broadleaf forest and needleleaf forest using multi-direction satellite data
 - Study the differences between needleleaf and broadleaf trees using multi-direction data of ALOS/PRISM
- Development of validation methods for global land cover products
 - Produce validation data sets with reliable accuracy using SACLA data sets
 - Produce validation data sets for low resolution data using middle high resolution satellite data sets (ALOS/AVNIR-2).

The characteristic differences between needleleaf and broadleaf trees using multi-direction data of ALOS/PRISM

- The aim of this study is developing an classification method for broadleaf forest and needleleaf forest using multi-direction satellite data
 - Distinguish differences in shape of a canopy of forests using multidirection data
 - Study about characteristic differences among needleleaf and broadleaf trees using multi-directional data of ALOS/PRISM.

PRISM	Characte	rietice
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Number of Bands	1 (Panchromatic)		
Wavelength	0.52 to 0.77 micrometers		
Number of Optics	3 (Nadir; Forward; Backward)		
Base-to-Height ratio	1.0 (between Forward and Backward view)		
Spatial Resolution	2.5m (at Nadir)		
Swath Width	70km (Nadir only) / 35km (Triplet mode)		
S/N	>70		
MTF	>0.2		
Number of Detectors	28000 / band (Swath Width 70km)		
Number of Detectors	14000 / band (Swath Width 35km)		
Pointing Angle	-1.5 to +1.5 degrees		
Folitting Angle	(Triplet Mode, Cross-track direction)		
Bit Length	8 bits		



Copywrite:(JAXA PRISM) http://www.eorc.jaxa.jp/ALOS/en/about/prism.htm

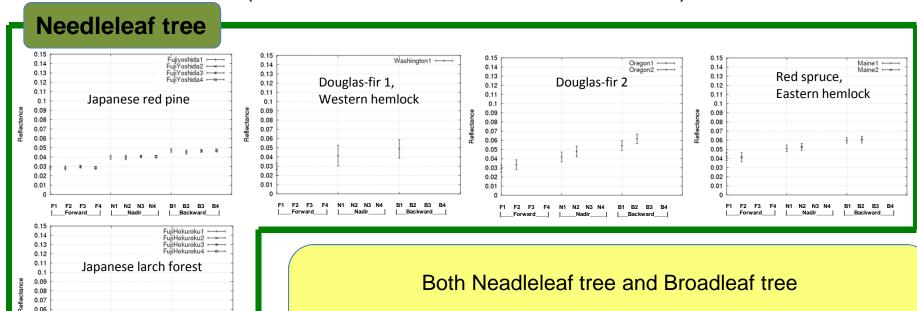
Characteristic of sample

- Selecting sample points from Fluxnet(http://fluxnet.ornl.gov/)
- Land of sample point is relatively flat surface
- Measured under same illumination condition (Principal plane)
- Use scenes of only summer season (High vegetation condition)

Vegetation	Location	Species	SunZenith(PRISM)	Sun azimuth(PRISM)
Evergreen Needleleaf tree	Fujiyoshida	Japanese red pine	69.18	134.47
	USA/Washington	Douglas-fir , Western hemlock	64.3	145.64
	USA/Oregon	Douglas-fir	56.8	151.16
	USA/Maine	Red spruce, Eastern hemlock	64.22	143.51
Deciduous Needleleaf tree	Fuji Hokuroku	Japanese larch forest	69.18	134.47
Evergreen Broadleaf tree	Brazil	Macaranduba,Jatoba, brazilnut,Taxi	54.98	43.99
	USA/Wisconsin	Sugar maple	56.33	152.09
Broadleaf tree	USA/Ohio	Northern red oak,White oak	51.64	155.14
	USA/Tennessee	Oak/Hickory	64.56	136.69

Average and standard deviation of PRISM reflectance (Forward, Nadir, backward) for each sample area (1 blocks size : 20pixel x 20pixel)

Average of PRISM forward nadir backward (Needleleaf and Broadleaf tree)



Northern hemisphere: Forward < Nadir < Backward Southern hemisphere: Backward < Nadir < Forward

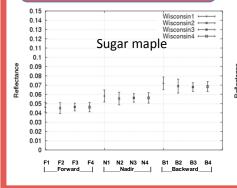


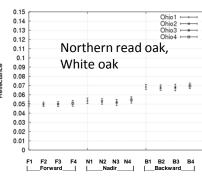
F1 F2 F3 F4 N1 N2 N3 N4

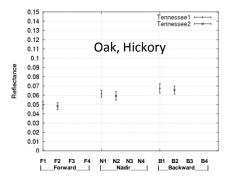
0.05

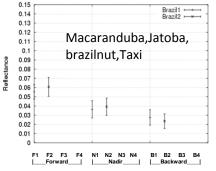
0.03

0.01









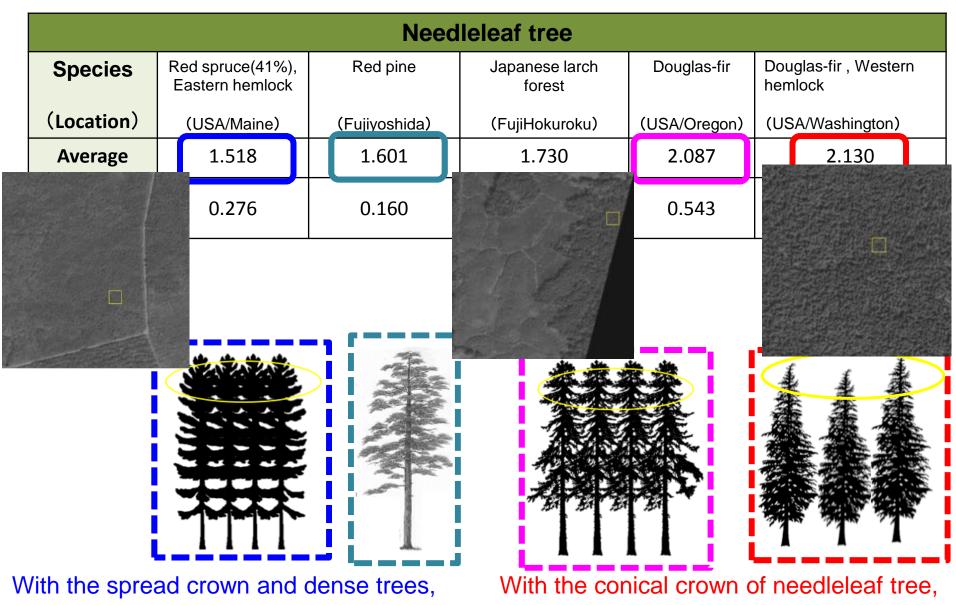
Ratio of Forward to Backward

Needleleaf tree						
Species	Red spruce, Eastern hemlock	Red pine	Japanese larch forest	Douglas-fir	Douglas-fir , Western hemlock	'n
(Location)	(USA/Maine)	(Fujiyoshida)	(FujiHokuroku)	(USA/Oregon)	(USA/Washington)	
Average	1.518	1.601	1.730	2.087	2.130	
Standard deviation	0.276	0.160	0.342	0.543	0.811	

Broadleaf tree				
Species	Oak/Hickory	Northern red oak, White oak	Sugar maple	Macaranduba, Jatoba, brazilnut, Taxi
(Location)	(USA/Tennessee)	(USA/Ohio)	(USA/Wisconsin)	(Brazil)
Average	1.314	1.390	1.543	2.651
Standard deviation	0.151	0.151	0.284	8.341

Needleleaf tree > Broadleaf tree

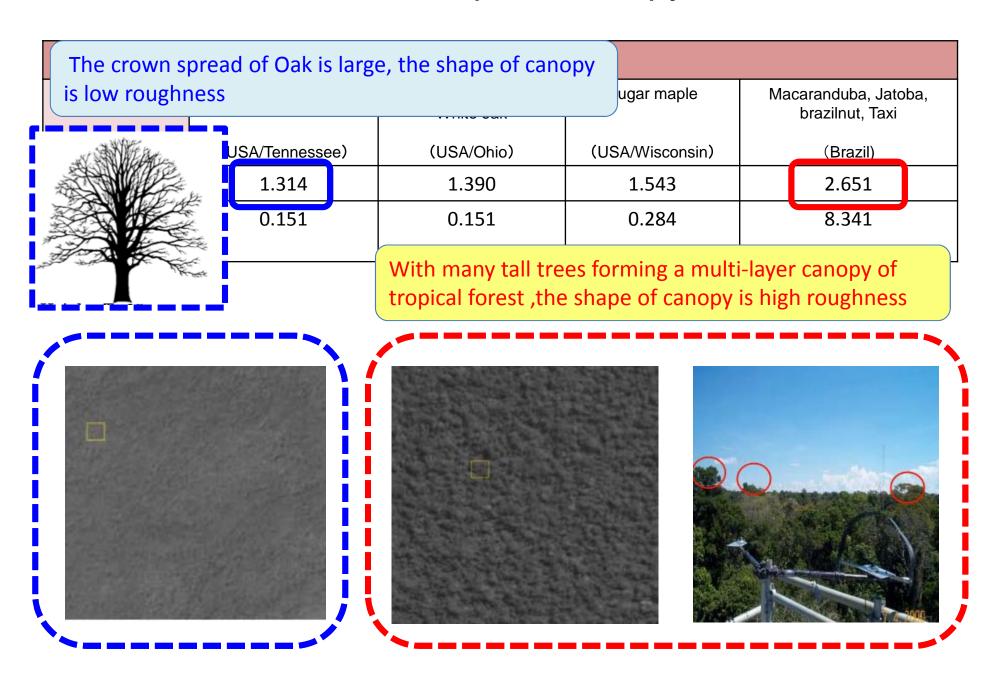
Differences in shape of canopy of trees



the shape of canopy is low roughness

the shape of canopy is high roughness

Differences in shape of canopy of trees



The classification applicability of forests using multi-direction data

Distinguish differences in shape of a canopy using multi-direction data



The shape of canopy is related to species, ages, forming layer of forest canopy and etc.



Need to collect training data from various forest types, and develop the method of classification between needleleaf forest and broadleaf forest using multi-direction satellite data

Validation data for global land cover products - data with reliable accuracy using SACLA -

Produce validation data sets with reliable accuracy using SACLA data sets

SACLA: several persons determined the class item of the site by watching photos on Degree Confluence Project (DCP) web site (provided by Dr. Sasai)

Merit

Collect validation data around the world

Demerit

Accuracy is not high since there are less information to determine a class.

<Information of the web >

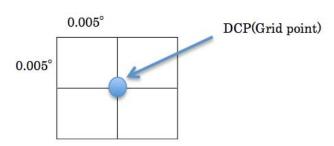
- Visit season
- Some photos (East,West,North,South)
- Description of the way of visiting to the point

Degree Confluence Project (DCP) http://confluence.org/index.php



Produce validation data sets with reliable accuracy using SACLA data sets

- Confirm four pixels, corresponding to a size of SGLI, centered on the DCP(grid point) using the following three check points by several persons.
 - Check Photos and descriptions on Degree Confluence Project (DCP) web site
 - ✓ See the surrounding area using Google Earth
 - Check the seasonal change characteristic of the site using Vegetation Index and spectral reflectances



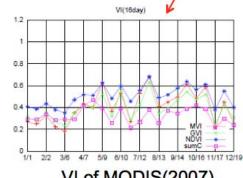
Problems

- Difficult to distinguish between forests with accuracy
- Difficult to distinguish between openshrub and grassland covered with low vegetation
- ✓ Difficult to determine the class item of a mixel land cover

In IGBP classes, cropland/natural vegetation mosaic is only mixel land cover type.

Google Earth





VI of MODIS(2007)

Validation data for global land cover products - expanding point data of field survey -

- ➤ In a location having no view from standing place such as forest, validation data collected by field survey may be low accuracy with a size according to low resolution satellite data
- Produce validation data sets for low resolution data with middle high resolution satellite data sets such as ALOS/AVNIR-2, Landsat8
 - ✓ Use middle high resolution satellite data sets (ALOS/AVNIR-2)
 - Analyze the land cover characteristic of well-known site using some index values (Vegetation Index, UPDM coefficients)
 - Calculate the average and the standard deviation of area (250mx250m) centered on the field point.
 - Determine the condition of the land cover by the data distribution
 - Distinguish pixels satisfying the condition from area(1kmx1km) centered on the field point
 - ✓ Determine the location for the validation data using the occurrence ratio(70%) of the class with a size according to low resolution (SGLI)

Determine the location for the validation data using occurren

Data: ALOS/AVNIR-2 produced by JAXA/EORC D

Wakakusayama(Nara): Grassland

Distinguish using MVIUPD of November

In the circle, there are many evergreen trees (high values) and shops(low values).

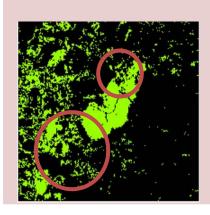
Average is obtained by canceling

each others.

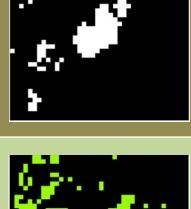
True color image

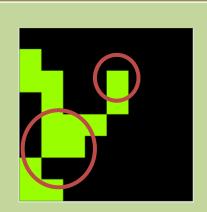


10m/pix



Occurrence Ratio: 80%





Average

The validation data using occurrence ratio

Paddy land condition:

Seika-cho in Kyoto Pref.

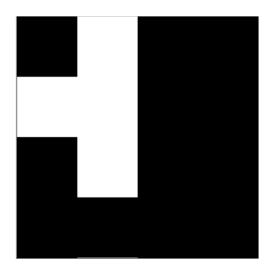
Google Maps

White pixels are paddy area



resolution size: 10m

Distinguish paddy area with a size according to a low resolution pixel (250m) in which occurrence ratio is 70%



resolution size: 250m

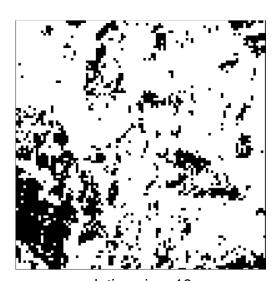
The validation data using occurrence ratio

Evergreen Needleleaf forest condition: 0.8998 < MVIUPD of November < 1.0086

Totsukawamua in Nara Pref.

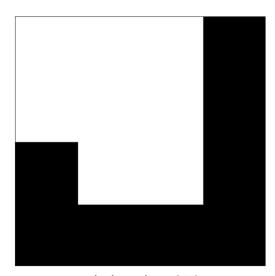
Google Maps

White pixels are evergreen needleleaf trees area



resolution size: 10m

Distinguish evergreen needleleaf trees area with a size according to a low resolution pixel (250m) in which occurrence ratio is 70%



resolution size: 250m

Evergre

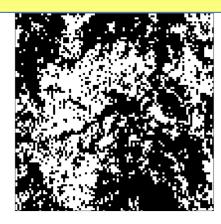
KasugaMt

To determine the location of validation data automatically, distinguish that by shifting of 250m from just location according with SGLI. No pixel distinguished as evergreen broadleaf trees area with a size according to a low resolution pixel (250m) in which occurrence ratio is 70% by shifting of 250m

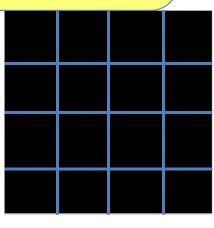
tio







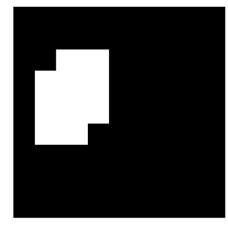
resolution size: 10m



resolution size: 250m

Not to overlook evergreen broadleaf trees area distinguished as data with reliable accuracy, the shifting size is reduced to 100m or 50m.

White Area is evergreen broadleaf trees distinguished with a size according to a low resolution pixel (250m) in which occurrence ratio is 70% by shifting of 100m



resolution size: 250m shifting size: 100m

Summary

- We can distinguish differences in shape of a canopy using multi-direction data, but the shape of canopy is related to species, ages, forming layer of forest canopy and etc.
 We need to collect training data from various forest types
- Produce validation data sets with reliable accuracy using SACLA data sets
- Produce validation data sets for low resolution data using middle high resolution satellite data sets (ALOS/AVNIR-2).

Future works

- Improve the global land cover algorithm
- Continue to produce the validation data sets using SACLA (including Flux site information)
- Continue to produce the validation data using middle high resolution satellite data sets (ALOS/AVNIR-2, LandSat 8)