

Algorithm Development of Global Land Cover Classification

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■ OBJECTIVE

to develop global land cover classification algorithm for multi-temporal SGLI reflectance products.

■ STUDY DATA

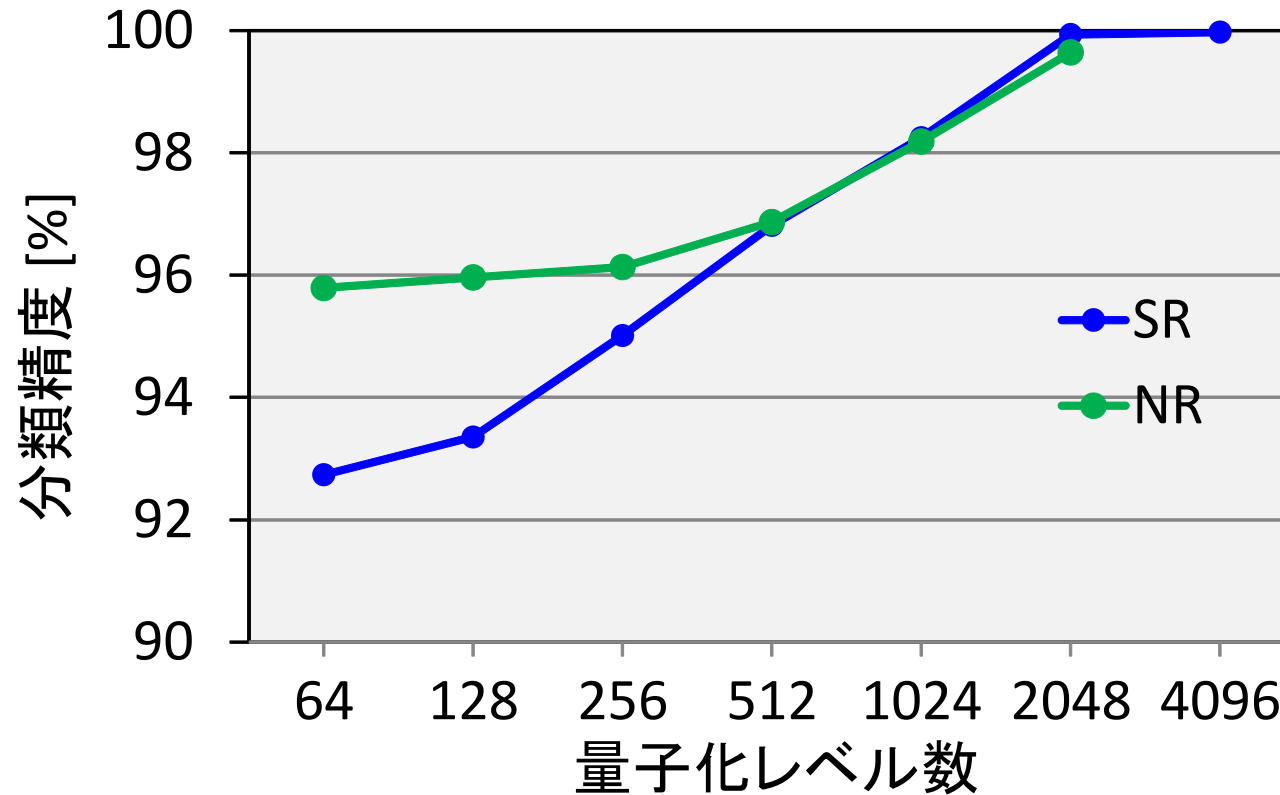
MODIS Reflectance Products

■ PROBLEMS Clarified in The Last Fiscal Year

1. Overlearning
2. Stability of Classification Results
3. Verification Method for Classification Results

1. Overlearning

■ Example



Estimated classification accuracy has no meaning.

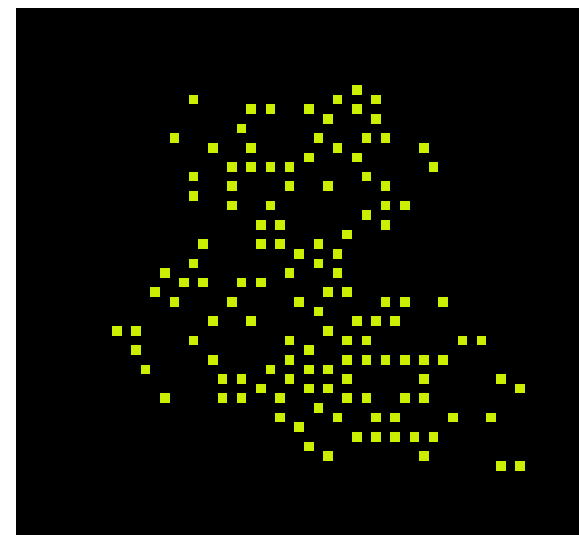
■ Cause

Test Samples = Training Samples

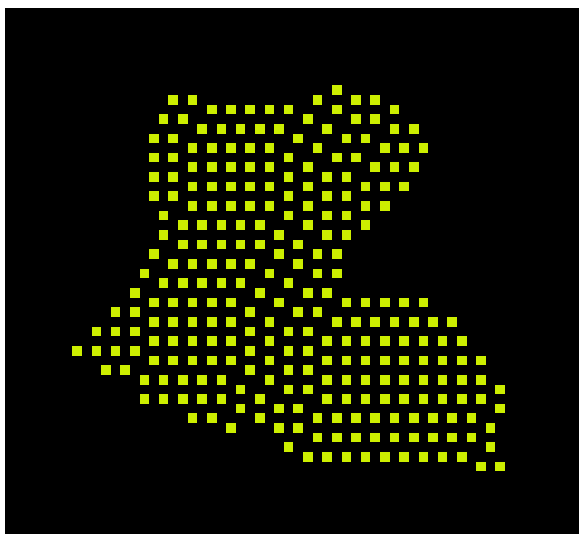
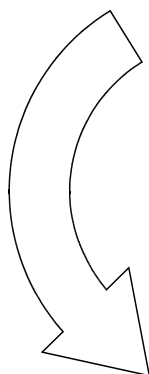
■ Countermeasure



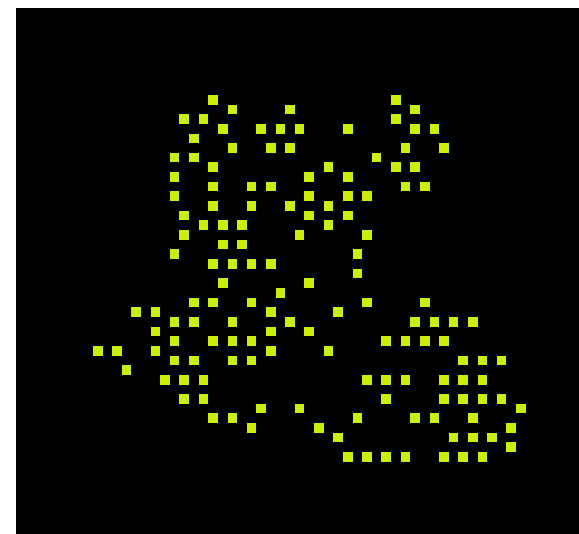
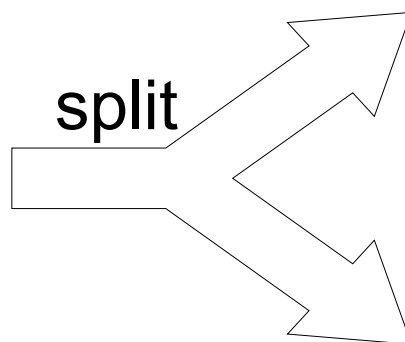
Supervised Area



Training Area

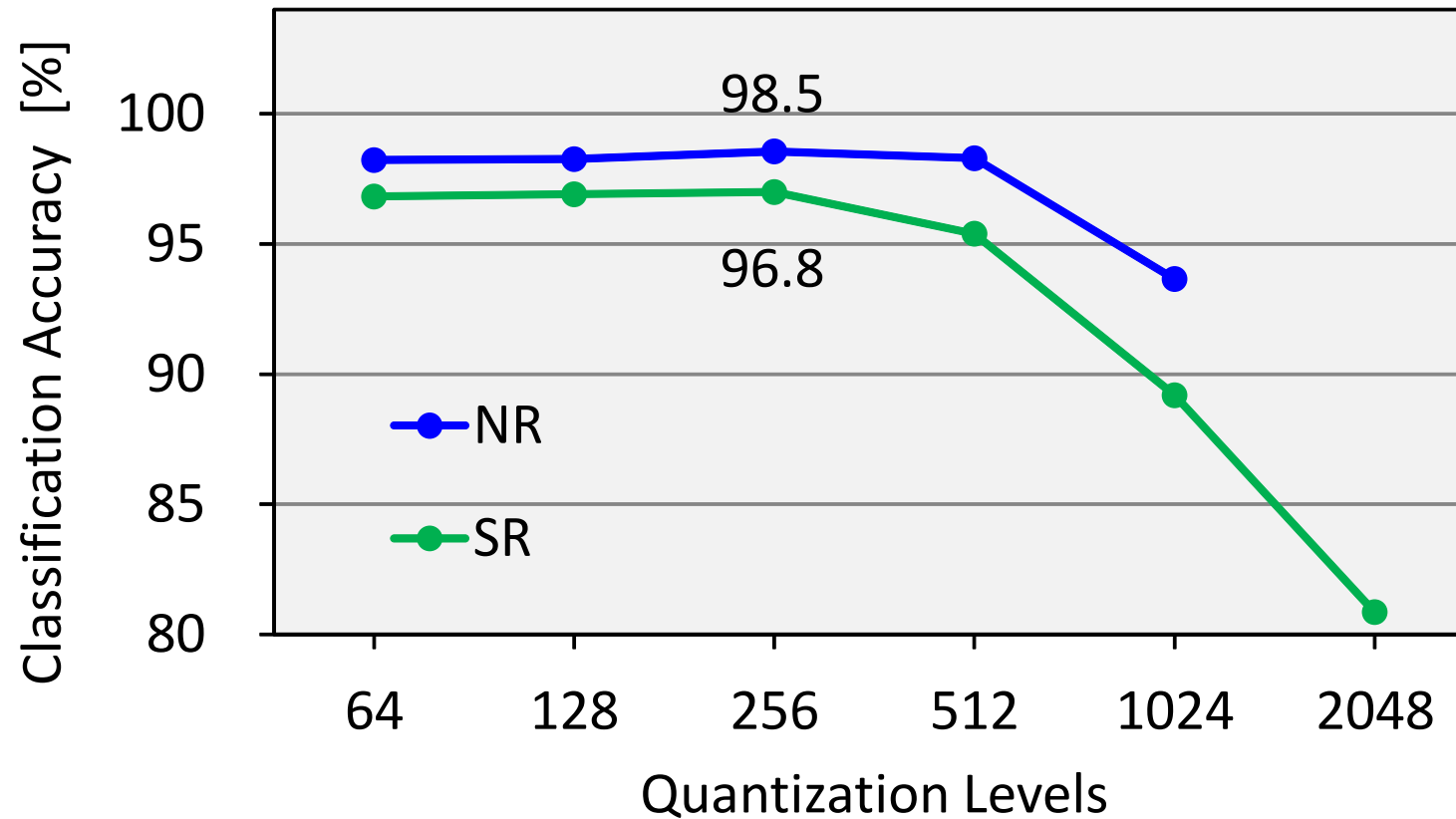


Sampled Supervised Area



Test Area

■ Result



Overlearning does not appear.

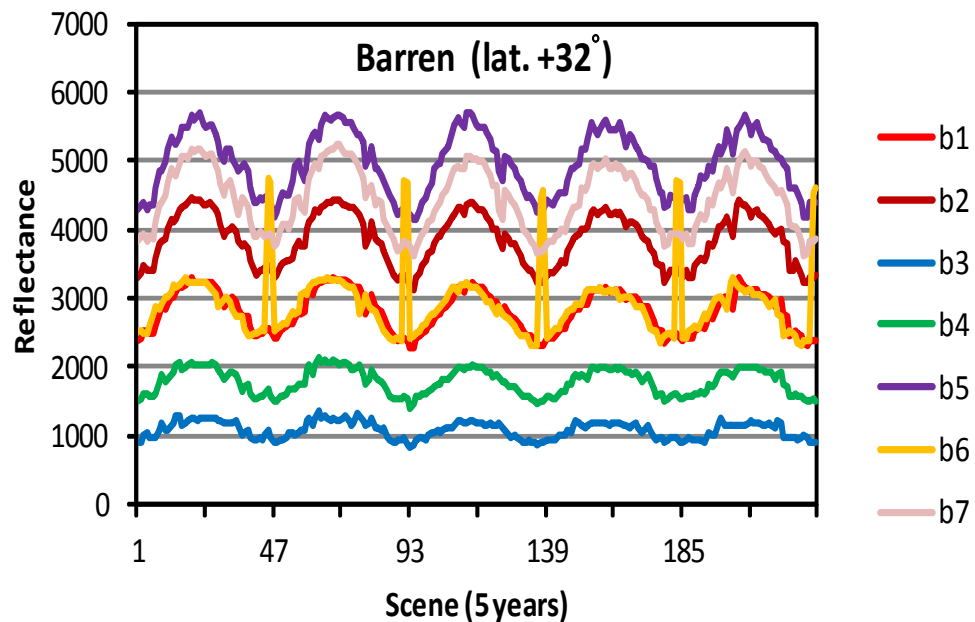
➡ Estimated classification accuracy has meaning.

2. Stability of Classification Results

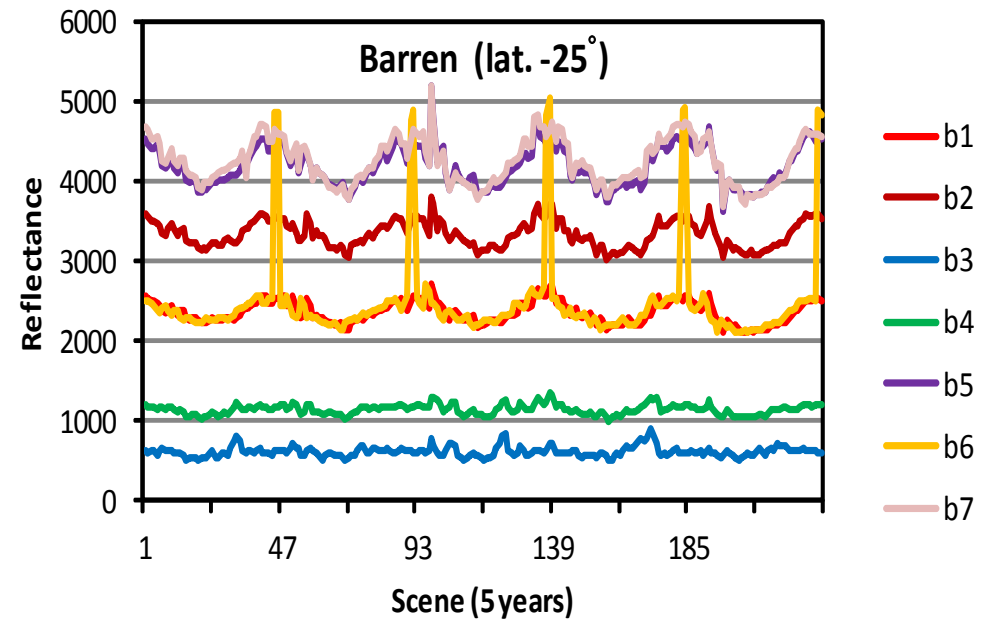
2.1 Spatial Stability of Reflectance in MODIS products

■ Example

[clear case]

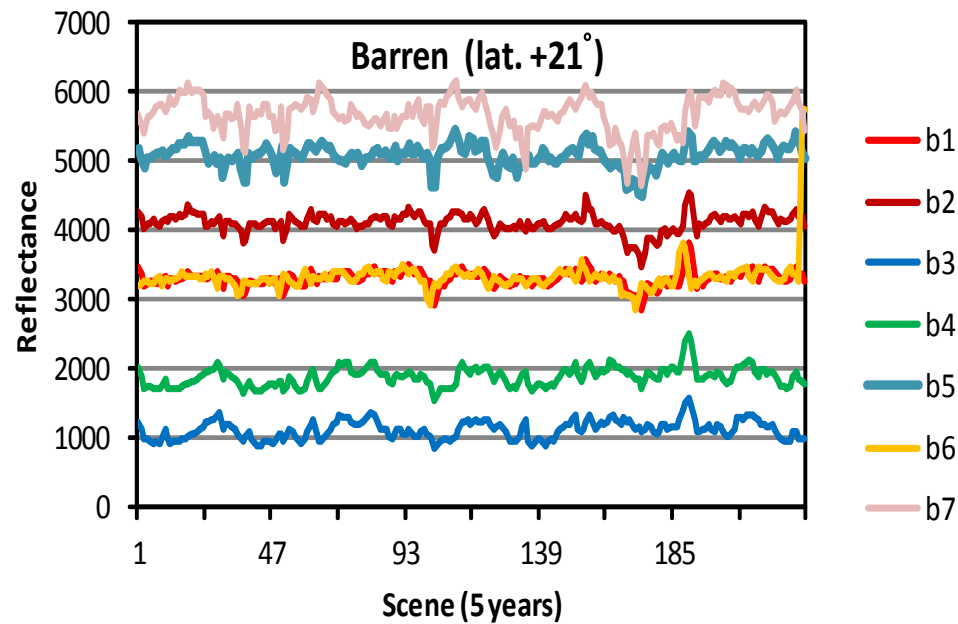


northern hemisphere

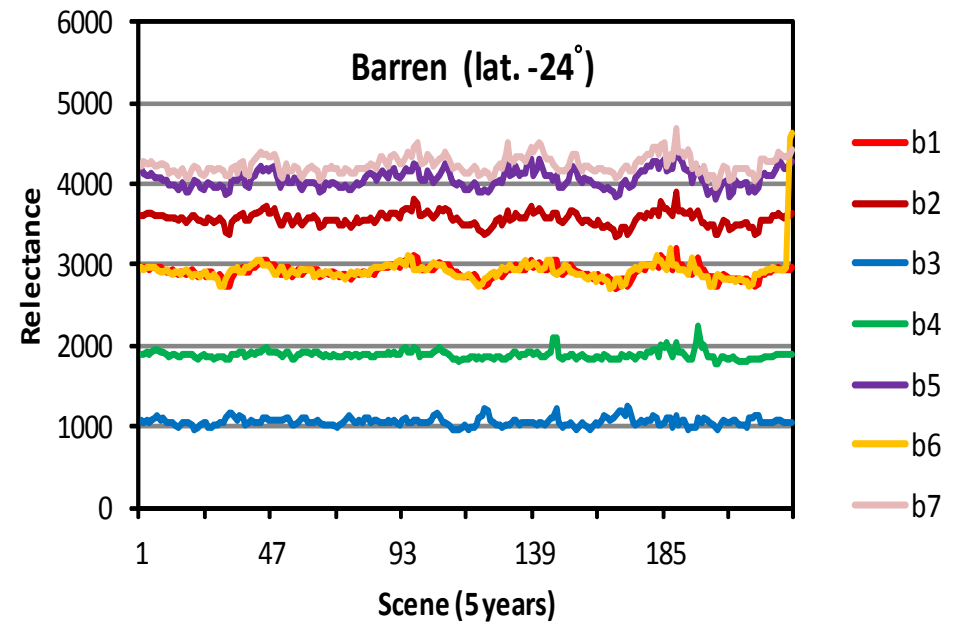


southern hemisphere

[unclear case]



northern hemisphere



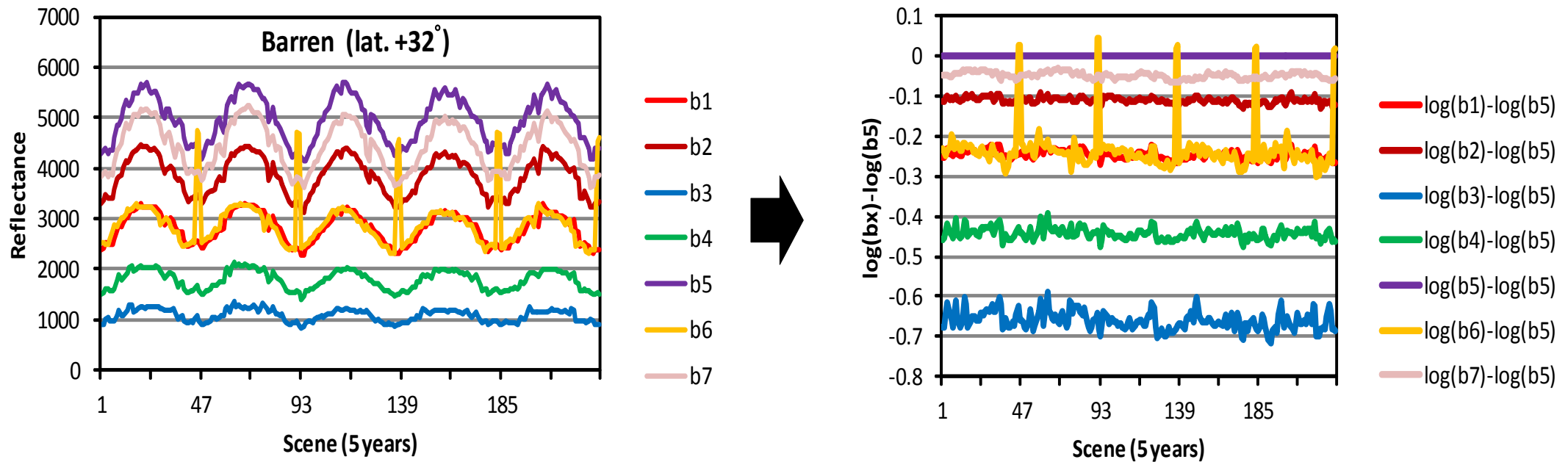
southern hemisphere

■ Cause

Solar Irradiance, BRDF, Topography,

■ Countermeasure
Solar Irradiance Correction:
 $\log(\text{reflectance}) - \log(\text{reflectance of band5})$

■ Result



example of northern hemisphere

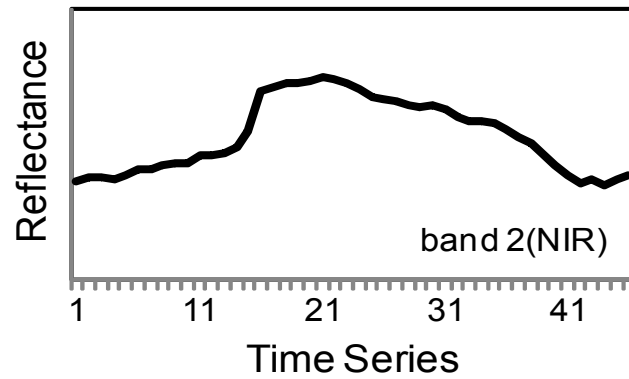
➡ Dimentions of feature space decreases from 7 to 6.

2.2 Stability of Time-domain Co-occurrence Matrix

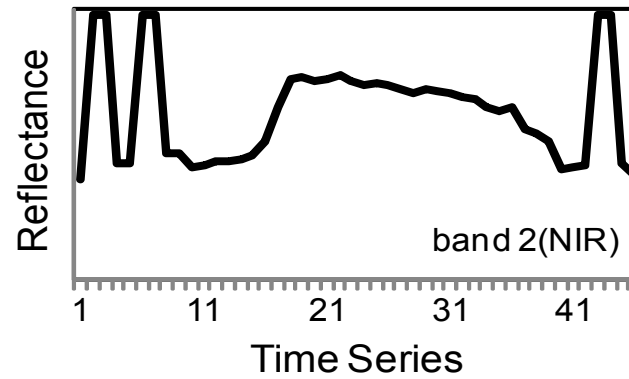
■ Example : cloud/snoe noize



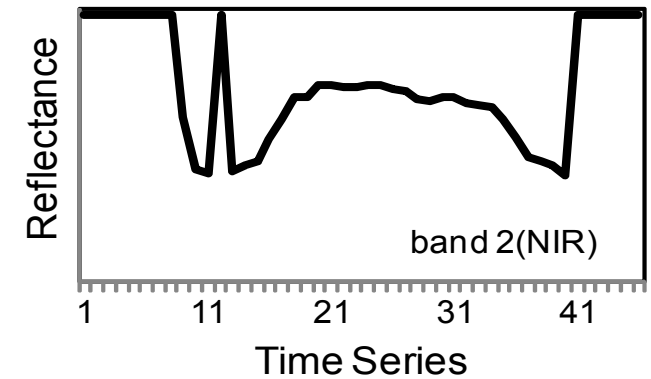
year long clear



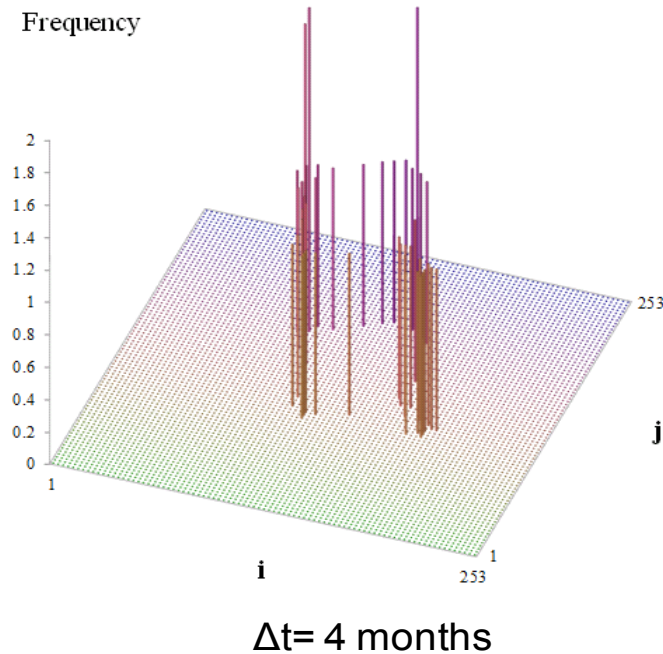
some times cloudy



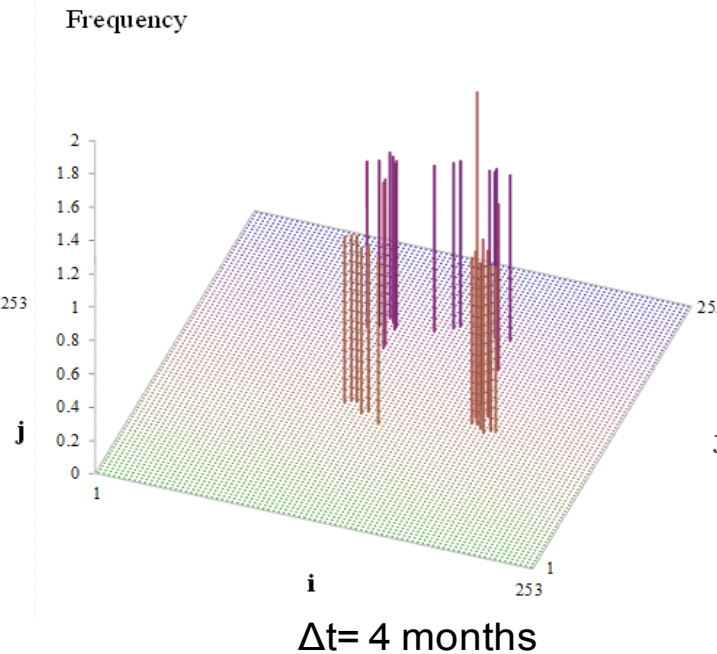
long term cloudy



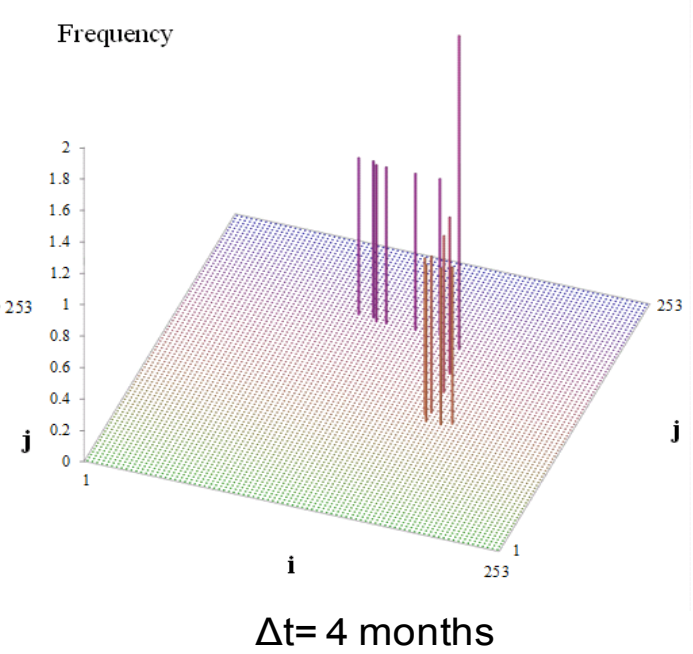
Frequency



Frequency



Frequency

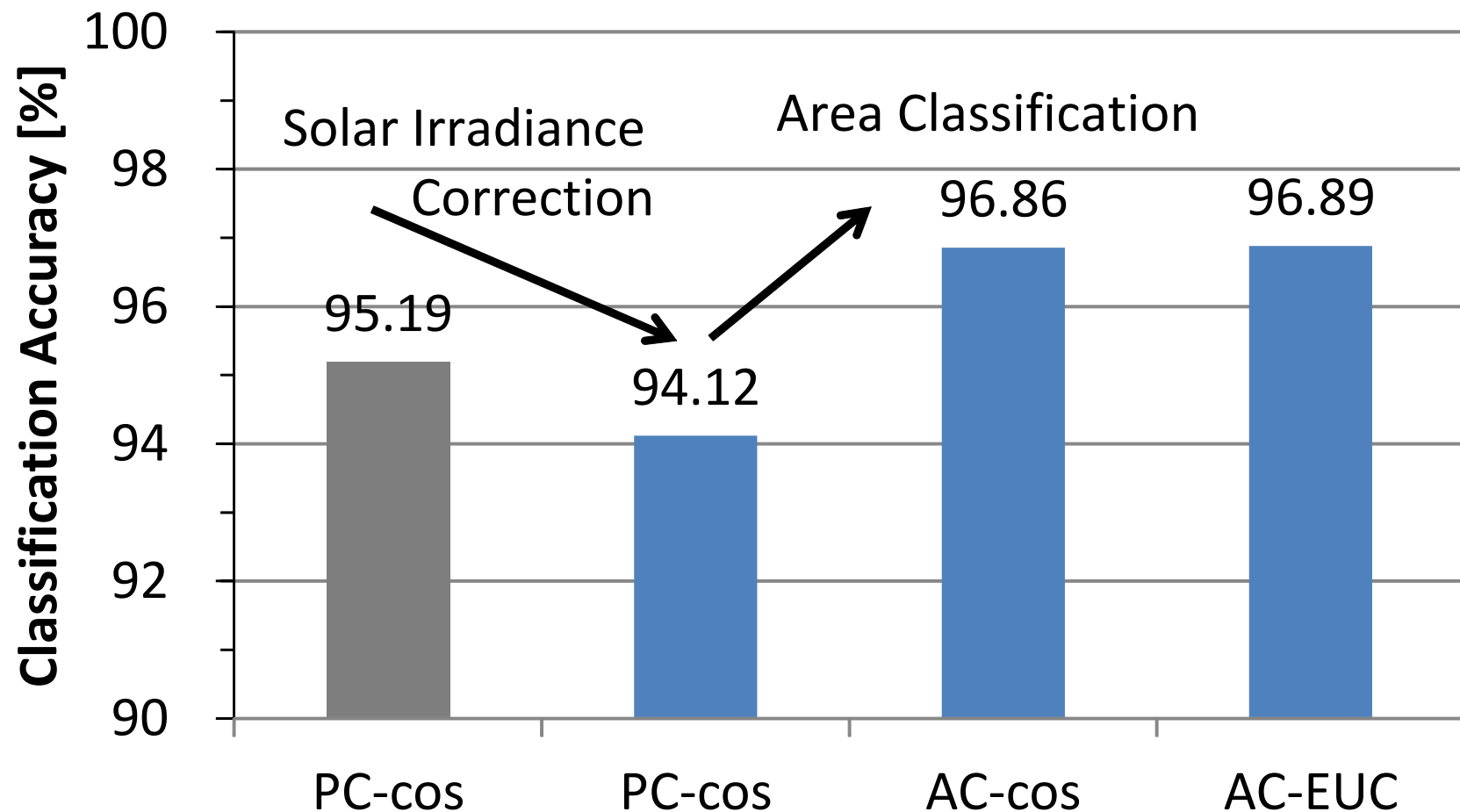


➡ time-domain co-occurrence matrix : sparse

■ Countermeasure

Area Classification : using 8 neighbour pixels with weight

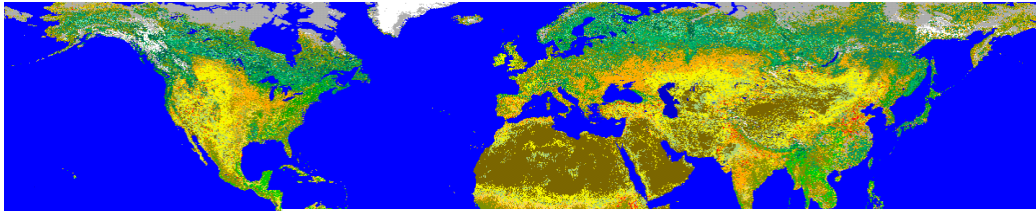
■ Result for Classification Accuracy



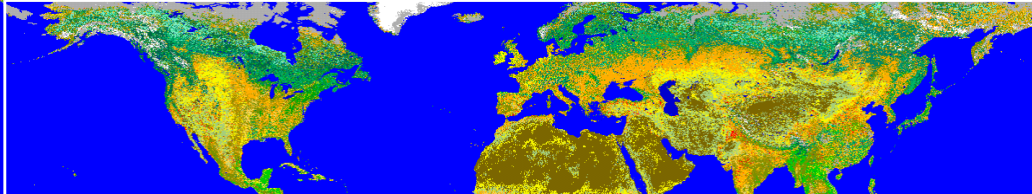
2.3 Stability for Multiple Year

■ Example

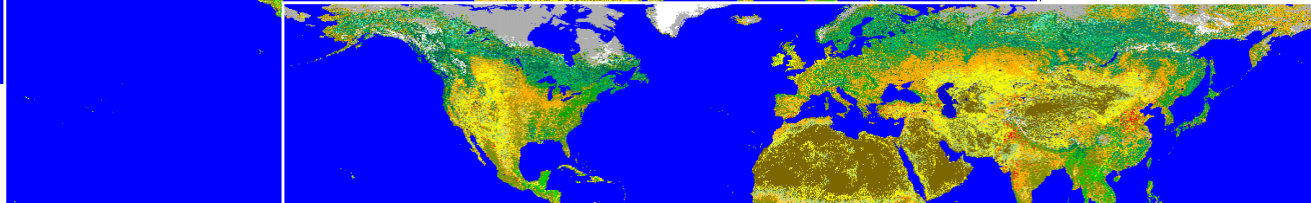
2007



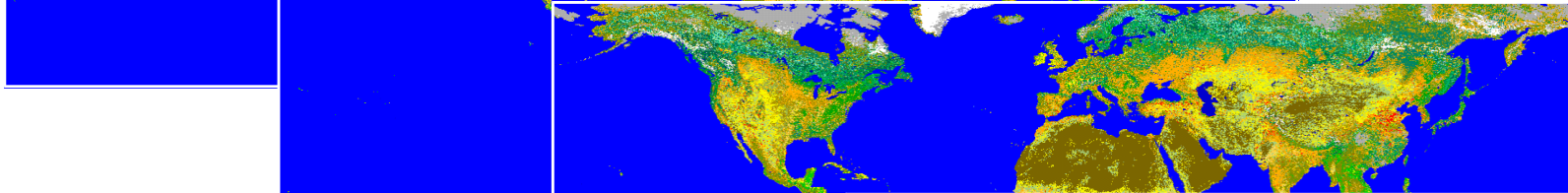
2008



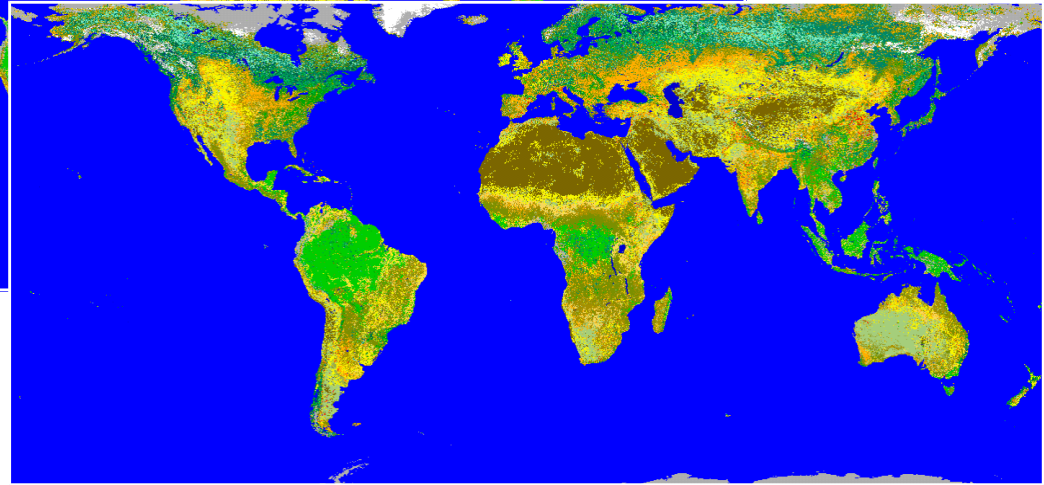
2009

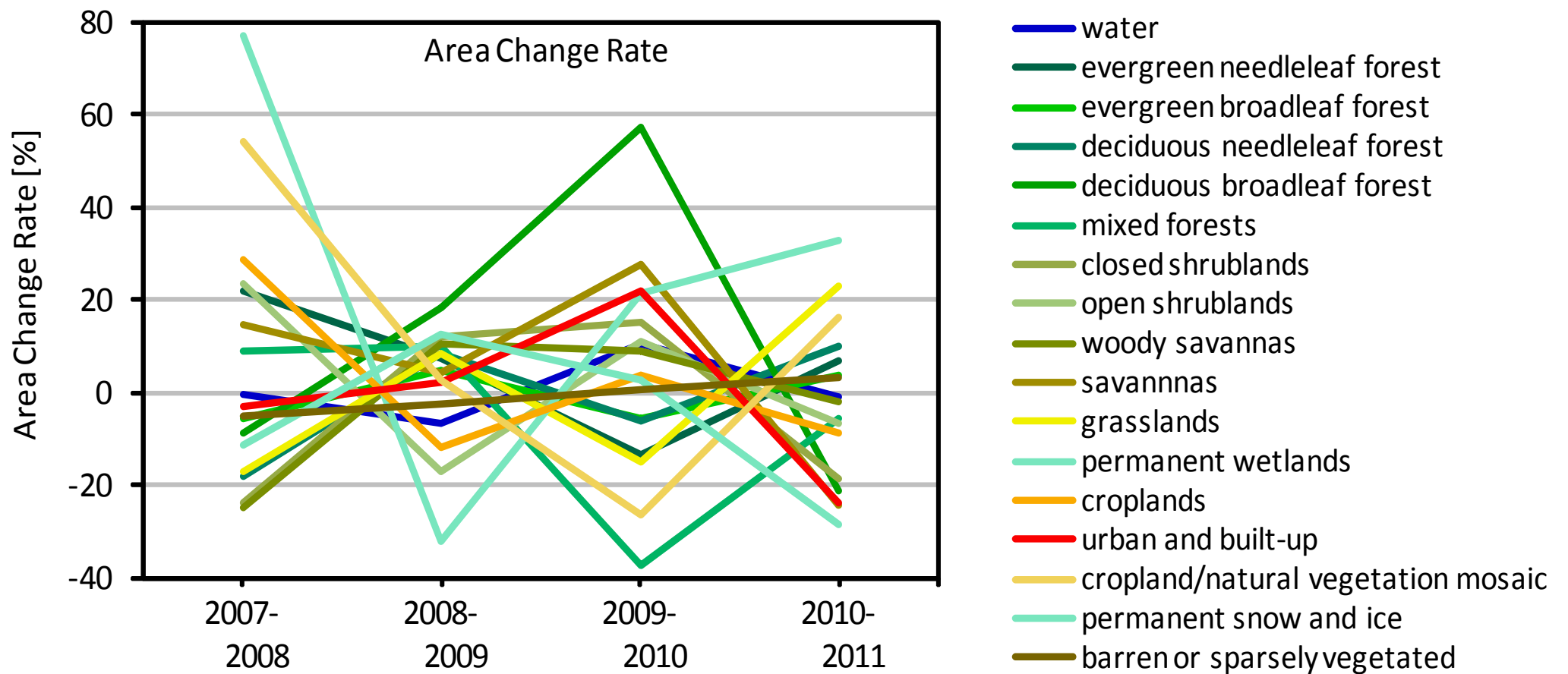


2010



2011



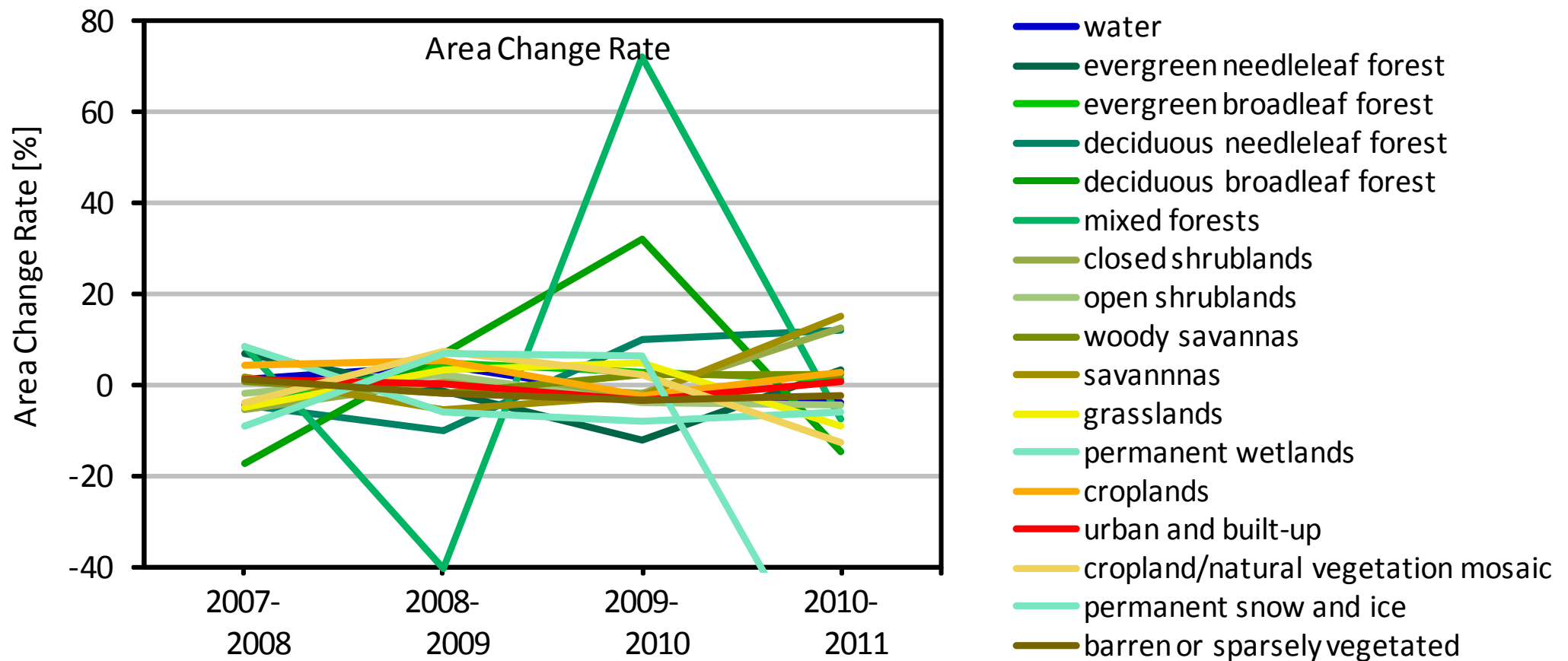


Area Change Rate (the last FY)

Countermeasure

unify/integrate some classification results (cos, EUC)

Result

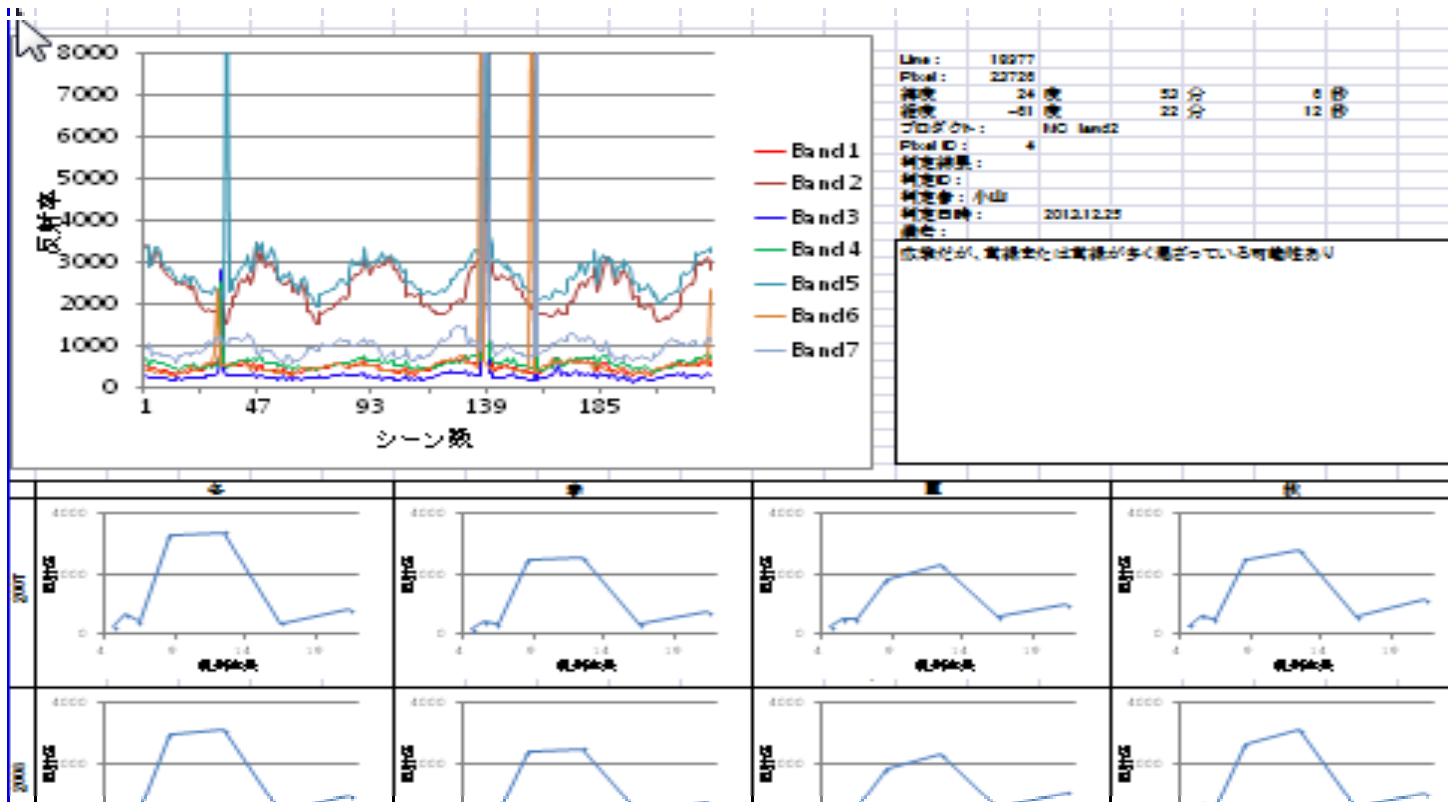


3. Verification Method for Classification Results

■ Strategy

- (1) random sampling of verification points for each category in the classification result
- (2) human interpretation for those verification points

■ method



Sheet of
Spectral
Characteristic

Sheet of the point Image

												
2007	冬			春			夏			秋		
	R,G,B:143	R,G,B:214	R,G,B:5,6,7	R,G,B:143	R,G,B:214	R,G,B:5,6,7	R,G,B:143	R,G,B:214	R,G,B:5,6,7	R,G,B:143	R,G,B:214	R,G,B:5,6,7
2008	冬			春			夏			秋		
	R,G,B:143	R,G,B:214	R,G,B:5,6,7	R,G,B:143	R,G,B:214	R,G,B:5,6,7	R,G,B:143	R,G,B:214	R,G,B:5,6,7	R,G,B:143	R,G,B:214	R,G,B:5,6,7

■ Trial Result and Workload Assessment

10 points/category : 2 ~ 4 man-week

100 points/category : 5 ~ 10 man-month
(It seems not to be impossible work.)

Concept of the Proposed Classification Method

(1) Classification Feature

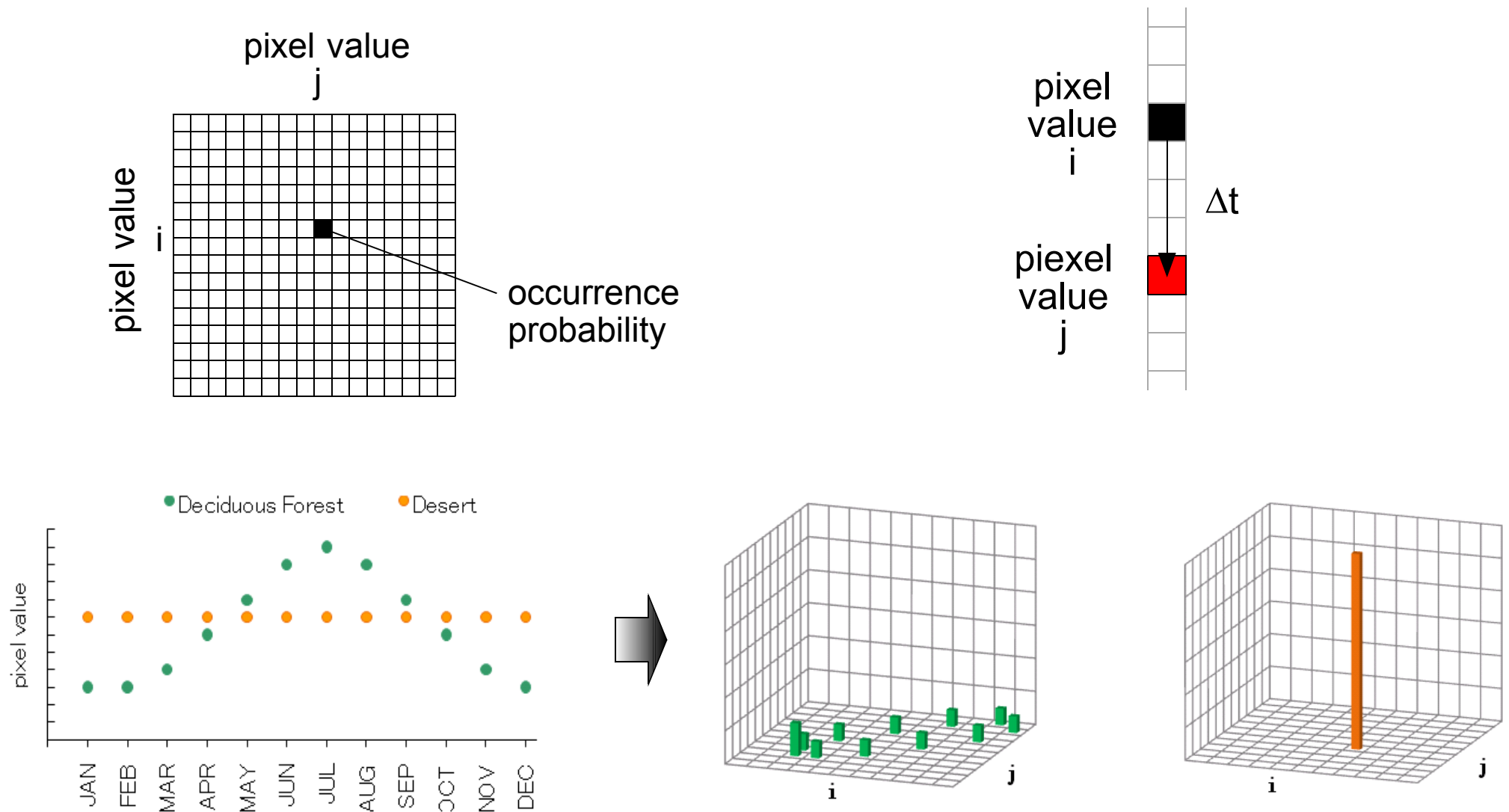
⇒ time domain co-occurrence matrix

(2) Classifier

⇒ non-parametric minimum distance classifier

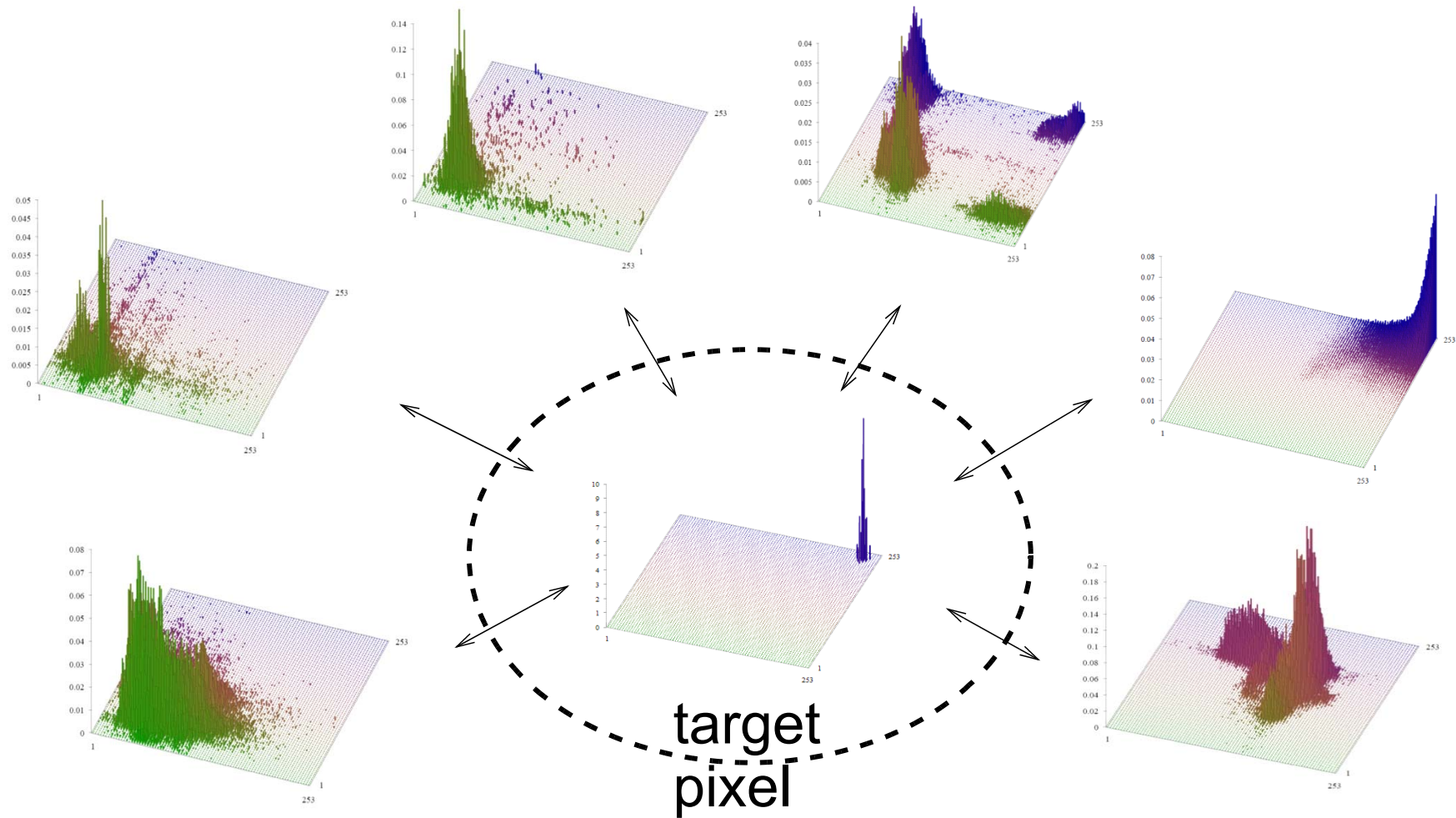
■ Proposed Time Domain Co-occurrence Matrix

Each elements (i, j) of the time-domain co-occurrence matrix is defined as probability that two pixels with a specified time-separation Δt in the same spatial position have pixel value i and j .



■ Proposed Non-Parametric Minimum Distance Classifier

pattern matching for time-domain co-occurrence matrix



■ Training Area

