

ENERGY & WASTE



Oxygen

CH4 Modeling Natural Wetlands



Figure 2. DNDC-modeled vs. observed CH4 fluxes from a pristine wetland in Marcell, MN (field data from Crill et al. 1988, Dise 1991, Clement et al. 1995, and Shurpali et al. 1993).

Rice studies in southeast China

- Case study 1: rice mapping in Fuyang City, Zhejiang Province (a city in China is an administrative level higher than county, which means that a city usually covers several counties)
- Case study 2: rice modeling and bio-estimating in Haining County, Zhejiang Province (in-process)



• Case study 1: rice mapping

- PALSAR images acquired in 2006

(HH polarization, pixel size - 6.5m)



- Study area (Fuyang City administrative area)

-Covering 1.8×105 hectares;

-a typical subtropical climate. Nearly half a year is rainy and cloudy;

-major LULC:

mountains (~70%), lowland plains (~20%), and water (~10%);

-Rice:

- -- dominant crop in lowland plains (~50% of lowland plain);
- -- single cropping
- -- small fields

-- highly fragmented with urban developments, dryland crops and shrub orchards (tea, mulberry, etc)

- Image classification: Support vector machine (SVM)





		Reference data							Cand
		w	R	0	D	U	Classified	User's	Cond. Kappa
	YV						totals	accuracy(%)	
Classified data	W	161	21	8	5	5	200	80.50	0.7529
	R	9	180	9	0	2	200	90.00	0.8689
	0	34	16	130	8	12	200	65.00	0.5788
	D	4	19	21	137	19	200	68.50	0.6285
	U	3	1	1	2	193	200	96.50	0.9545
	Reference	211	237	169	152	231	1000		
	totals								
	Producer's	76 20	75 05	ግፍ ሰሳ	00 12	07 55			
NAME AND ADDRESS OF	accuracy (%)	70.30	(2.23	10.92	YU. IS	رر.ده	ne water water and the second s		ventormolecular average

Error matrix of the PALSAR-derived class map:

W - Water; R - Rice; O - Orchard; D - Dryland crop; U - Urban.

Overall Classification Accuracy = 80.10%; Overall Kappa Statistics = 0.75.

- Backscatter characteristics



-rice turned to be confused with dryland crop and orchard in each image.

Temporal Changes in L-band Sigma 0



Source: Inoue et al. 2002

- Phenological variation of vegetation backscatter



-rice backscatter increased along the three growing stages. The backscatter of dryland crop and orchard did not change much.



- Study area (Haining County)

- PALSAR FBD images: (HH&HV): -June 28, 2007 (transplanting) -August 13, 2007 (tillering) -September 28, 2007 (heading)

Field measurements (totally 44 fields):
June 29, 2007 (4 fields)
August 15, 2007 (19 fields)
September 30, 2007 (21 fields)

- Measured biophysical parameters:
 - plant height; plant density
 - unit dry weight, unit wet weight (leaf moisture)
 - leaf area index (LAI), leaf orientation angles
 - Water depth

- Rice canopy scattering model

-a 2nd-order radiative transfer function (RTF) model (based on Karam et al. 1995; Wang and Qi, in press)

-Rice was simulated as two-layer (leaves and short stems) continuous vegetation.

-Soil ground was thus simulated as a smooth continuous surface with dielectric constant of water.



(scattering intensity in each component is attenuated by scatterers on top of it.)

- Next steps: biophysical quantification

- Leaf density and plant height are the two major variables that control rice backscatter;

$$\sigma_{\text{mod}el}^0 = f(\text{leaf } _\text{density, plant } _\text{height})$$

- These two variables could be retrieved via model inversion with the criteria:

$$\min(\sigma_{\text{mod}\,el}^0 - \sigma_{\text{PALSAR}}^0)^2$$

- They were finally applied to all rice fields in the image to estimate field-based green biomass and rice yields in the study area.

Java Example



Figure 5 – Rice agriculture dynamics in Java from multiple PALSAR images collected on 6 Dec 2006 and 21 Jan 2007. Green indicates fields planted, and pink indicates fields harvested during the interval between acquisitions. The image is ~5 km across and has a spatial resolution of 6-m.