



RESPONSIBLE CONSUMPTION AND PRODUCTION



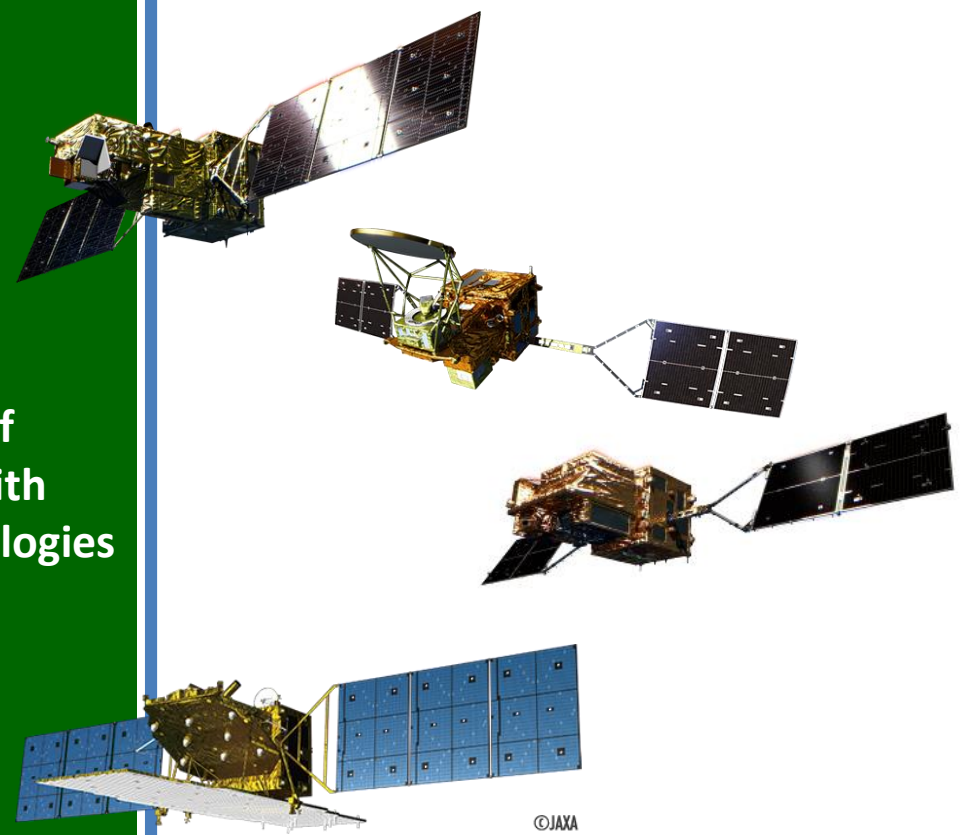
CLIMATE ACTION

Establishing an integrated MRV system of greenhouse gas emission from wetlands with Japanese earth-observation/modelling technologies and a data assimilation technique

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Wataru Takeuchi<sup>1)</sup>,  
Kei Oyoshi<sup>3)</sup>,  
Lam Dao Nguyen<sup>4)</sup>,  
Kazuyuki Inubushi<sup>5)</sup>

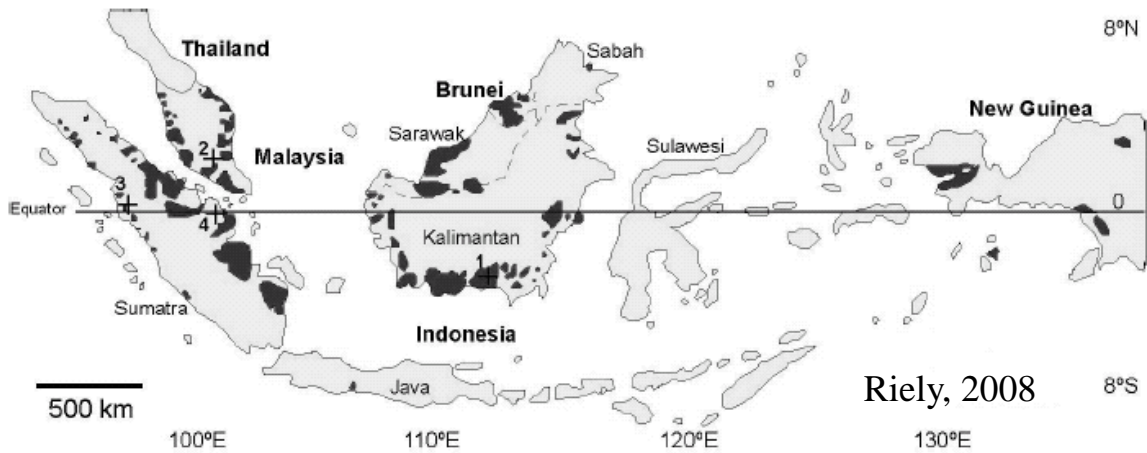
Koji Terasaki<sup>2)</sup>,  
Takemasa Miyoshi<sup>2)</sup>,  
Hisashi Yashiro<sup>2)</sup>



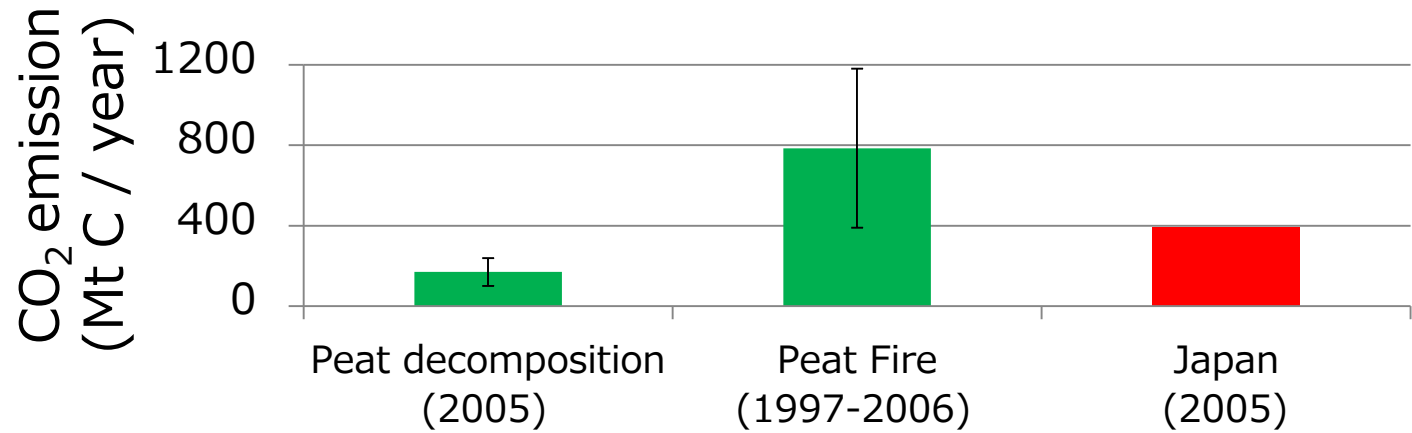
# Outline

- 0. Motivation to DA (Story taking me here today)**
1. Background & Objective
2. Ground observation of greenhouse gas emission and statistical modeling
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  - Cropping calendar & the adjacent fallow length
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  - Top down verification with GOSAT
4. My next work with DA

# Drainage on peatlands in SE asia



CO<sub>2</sub> emission (Mt C/year) from peat in south east Asia and Japanese total emission.



# Target fields

**Natural Forest**



**Drained Forest**



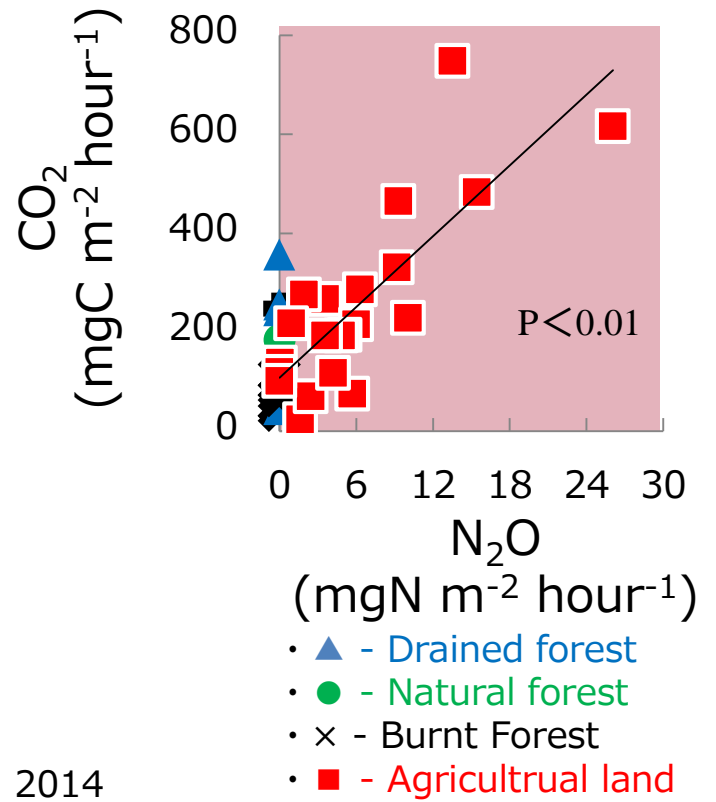
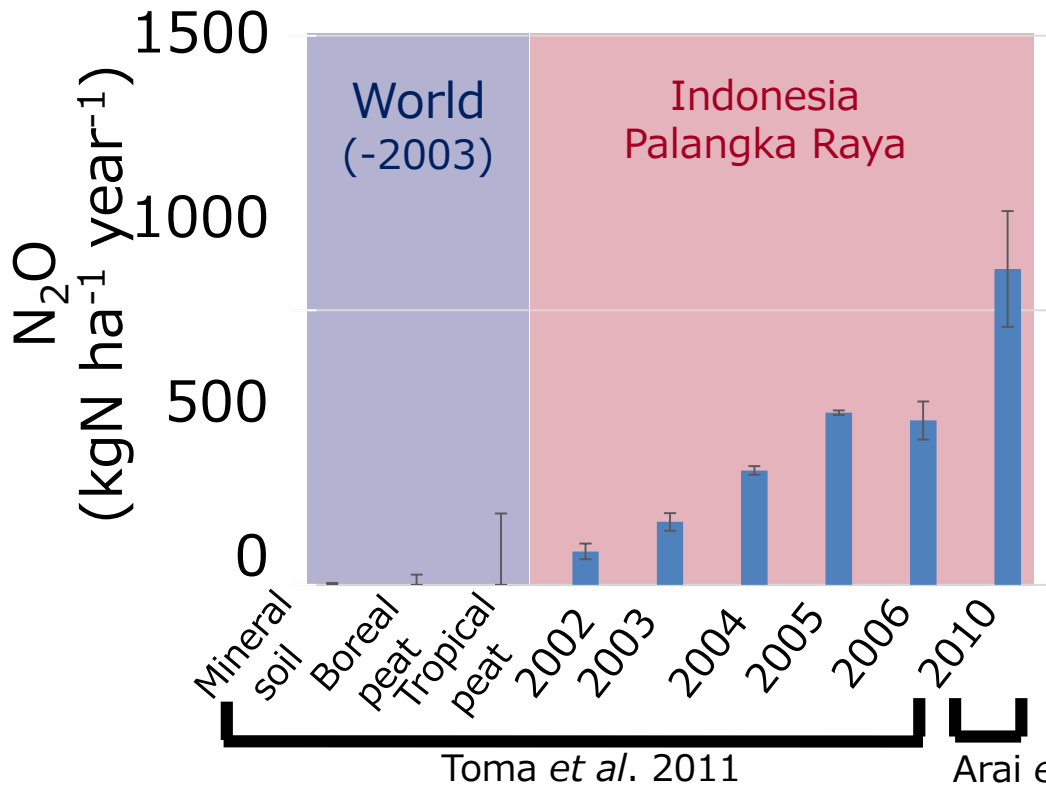
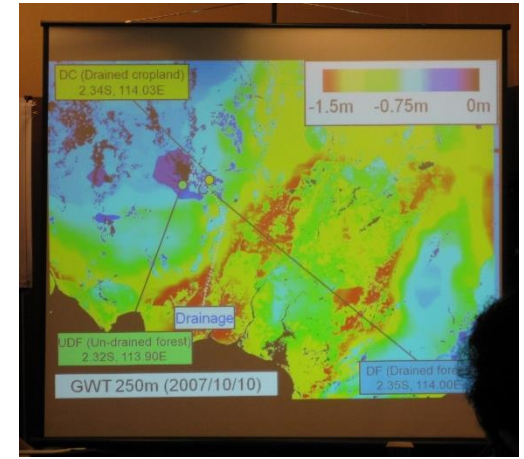
**Agricultural land**



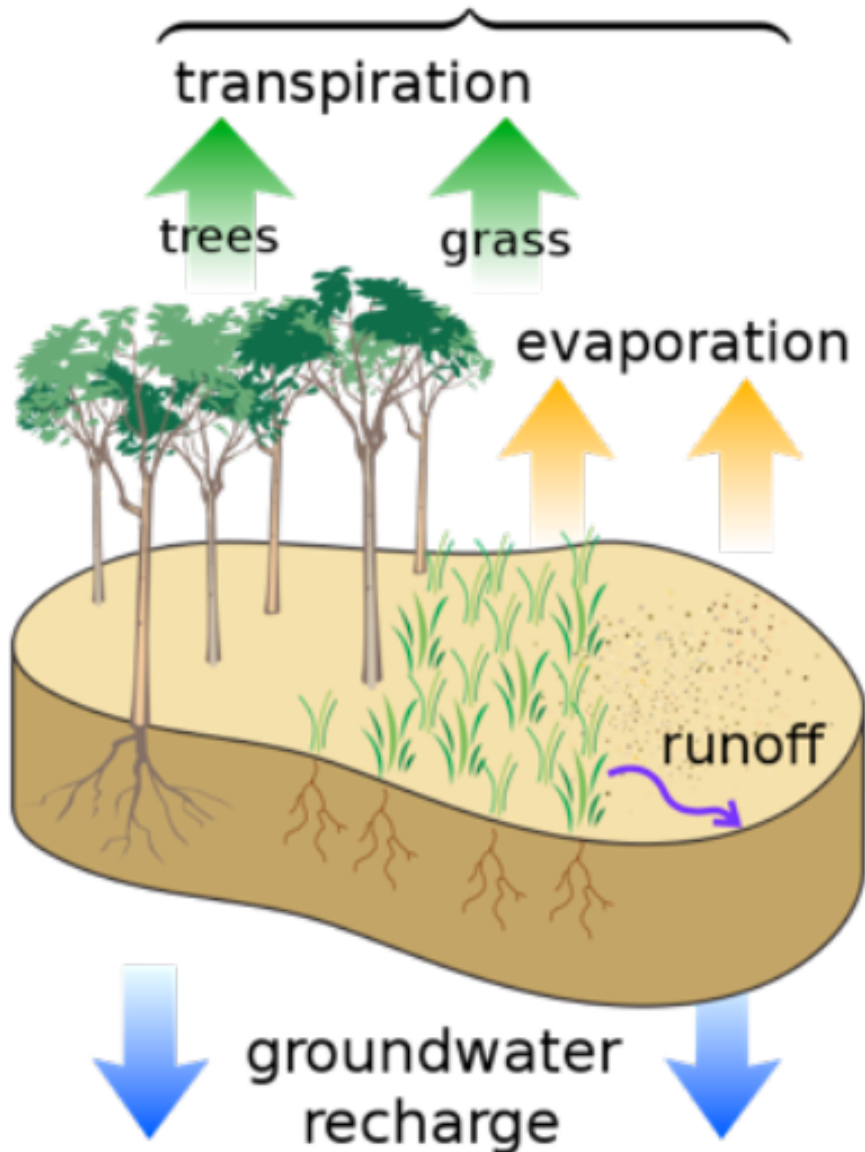
**Burnt Forest**







evapotranspiration =  
transpiration + evaporation



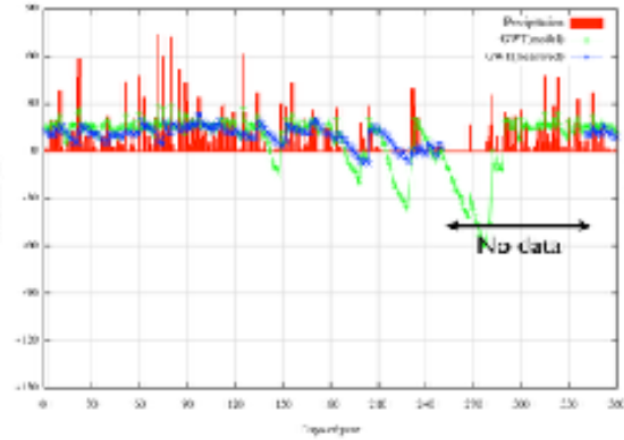
- 🍏 KBDI index are used to compute the balance between evapotranspiration and precipitation. [Keetch et. al, 1965]

$$dQ = \frac{[800 - Q][.968 \exp(.0486T) - 8.30]}{1 + 10.88 \exp(-.0441R)} d\tau \times 10^{-3}$$

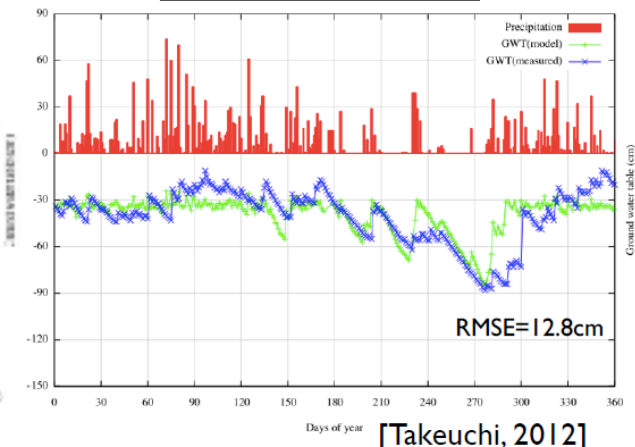
- 🍏 Presently, this index is derived from satellite observation:
  - 🍏 **land surface temperature (LST)** from **MTSAT** received at IIS/U-Tokyo
  - 🍏 **rainfall** from global satellite mapping (**GSMaP**) provided by JAXA EROC.
- 🍏 Ground water table (**GWT**) is modeled as a function of KBDI

# Lower ground water table of peatland in Indonesia are prone to fires and large carbon emission sources

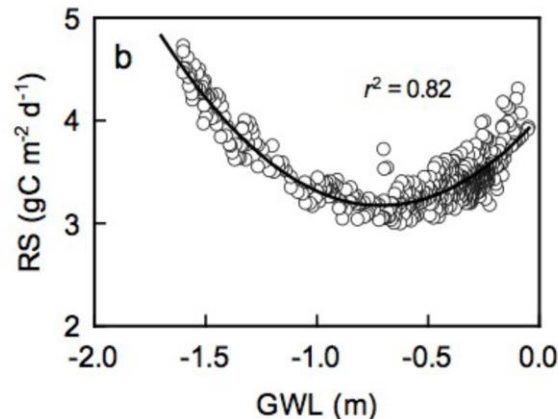
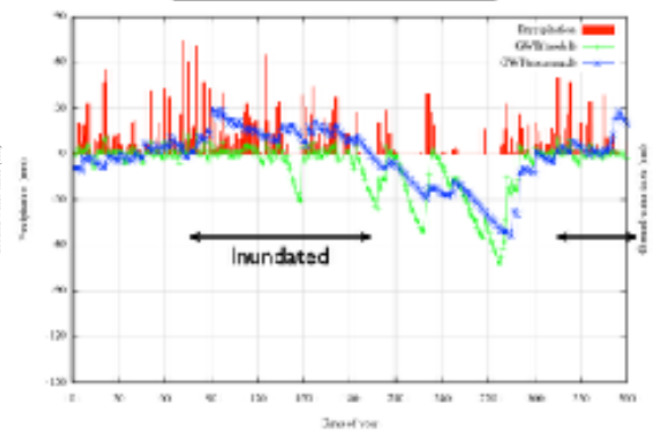
DEF (Drained forest) 2,345, 114,03E



DF (Drained forest) 2,355, 114,00E



UDF (Un-drained forest) 2,325, 113,92E



[Hirano, 2012]

	GWT (m)		SR (gC/m <sup>2</sup> /yr)	
	Satellite-based model (This study)	In-situ (Hirano, 2012)	Satellite-based model (This study)	In-situ (Hirano, 2012)
2007	-0.07	-0.4	1,386	1,238
2008	-0.26	-0.4	1,331	1,236



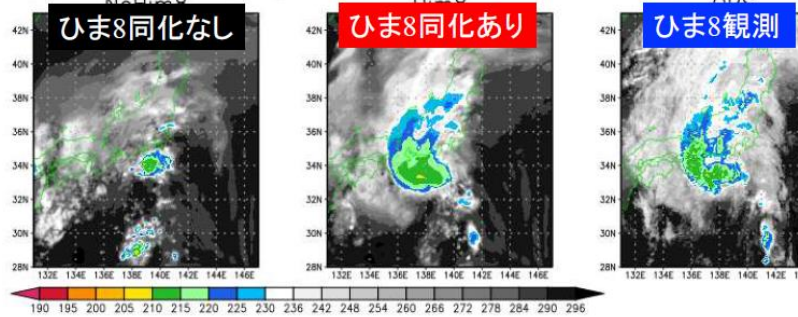
気象衛星ひまわり ～宇宙からの最先端データが～



# 豪雨事例へのひまわり8号同化

色: モデルと実観測それぞれの赤外輝度温度

Simulated/Observed Brightness Temperature B14 (K), at 18:00z08SEP2017



低: 高い雲

高: 低い雲/晴天

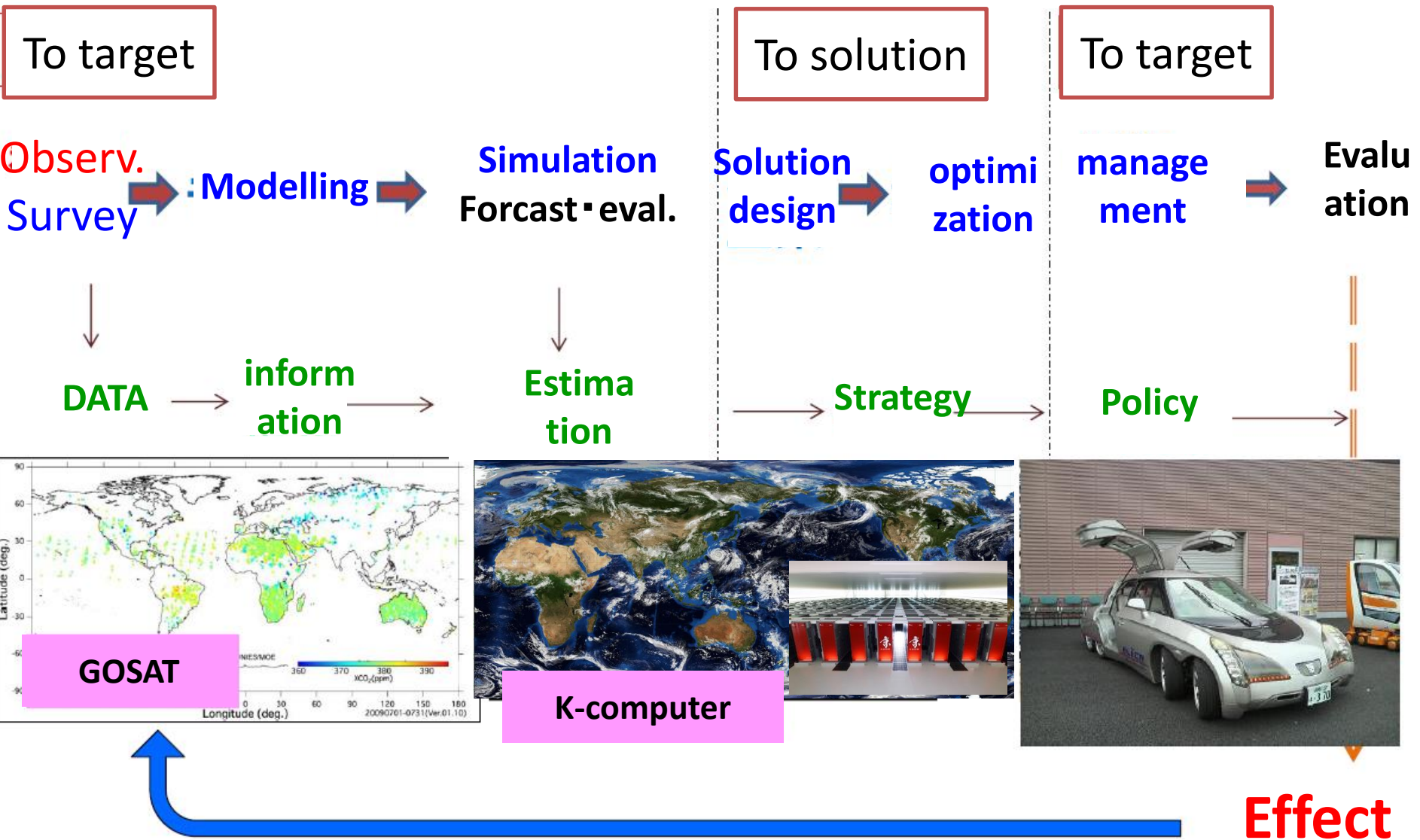
台風本体や関連する雲域が劇的に改善

三好2017





# Cycle from Observation to Countermeasure



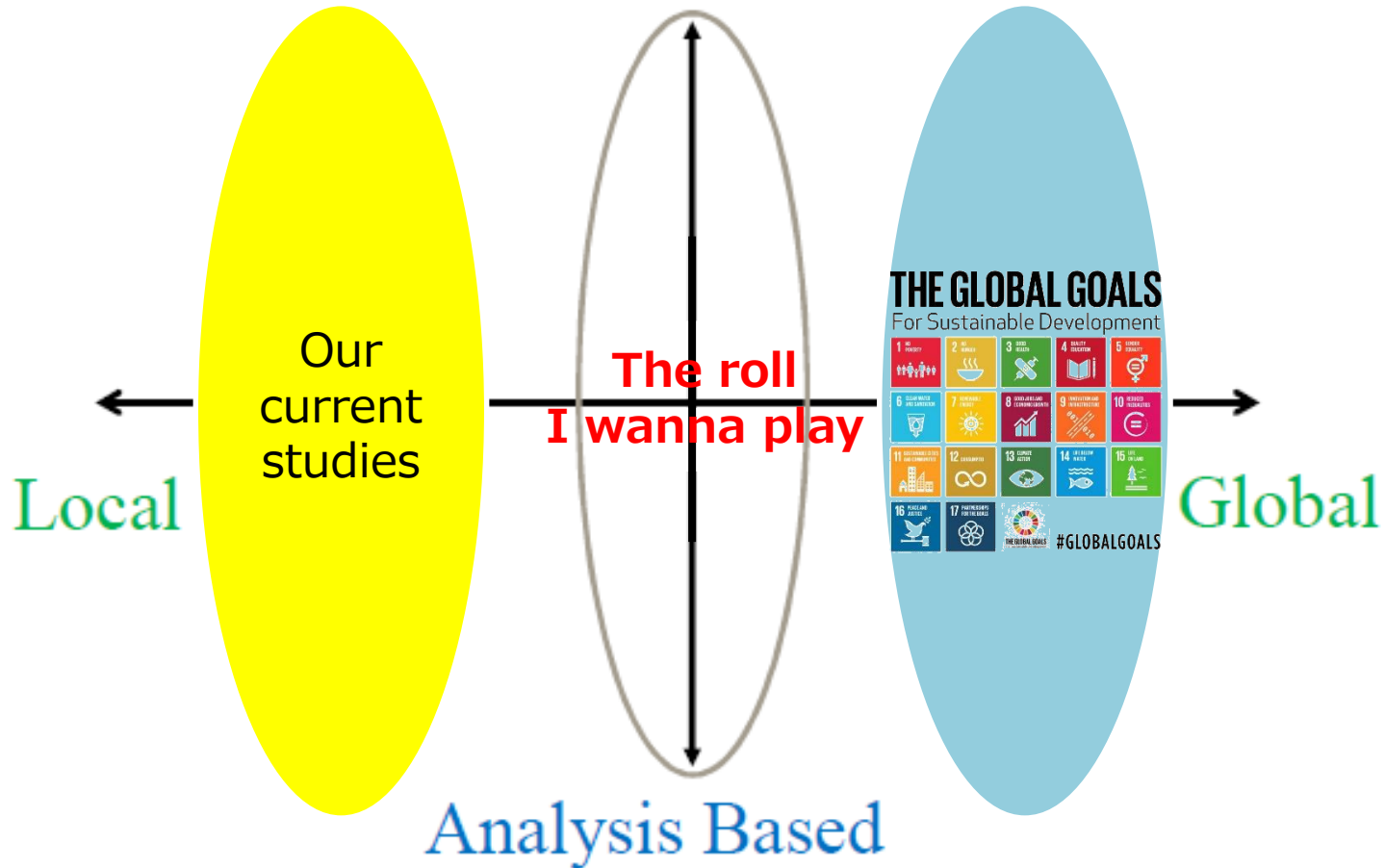
Observation of the effect

Modified from Yasuoka 2015

Customization

Commonized aspect

Solution Based



Customized Research

Commonization

Modified from Yasuoka 2017

# Outline

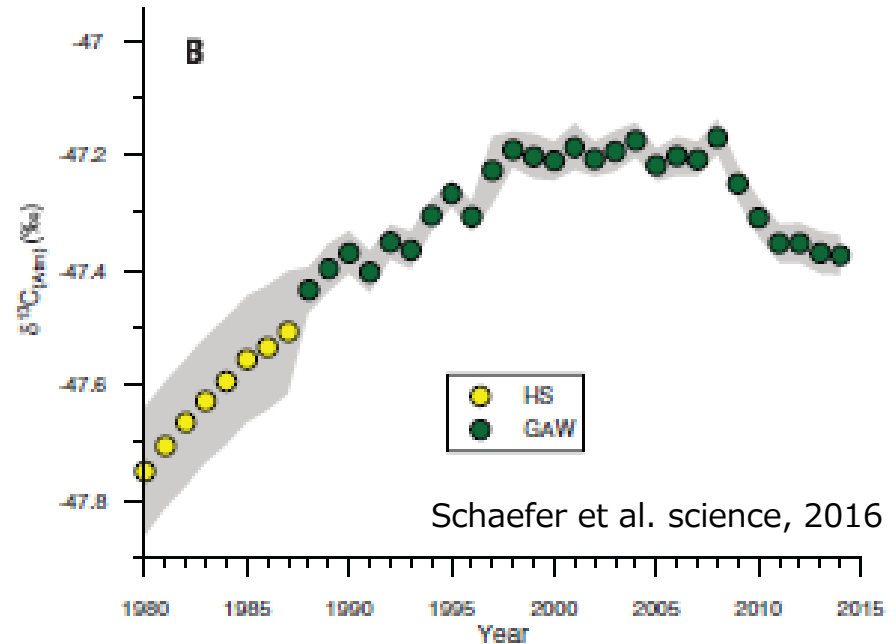
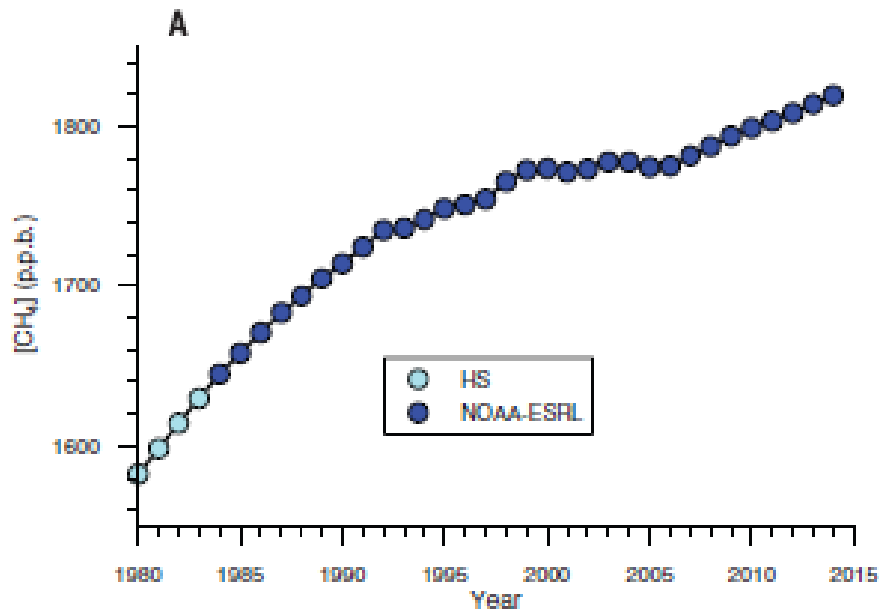
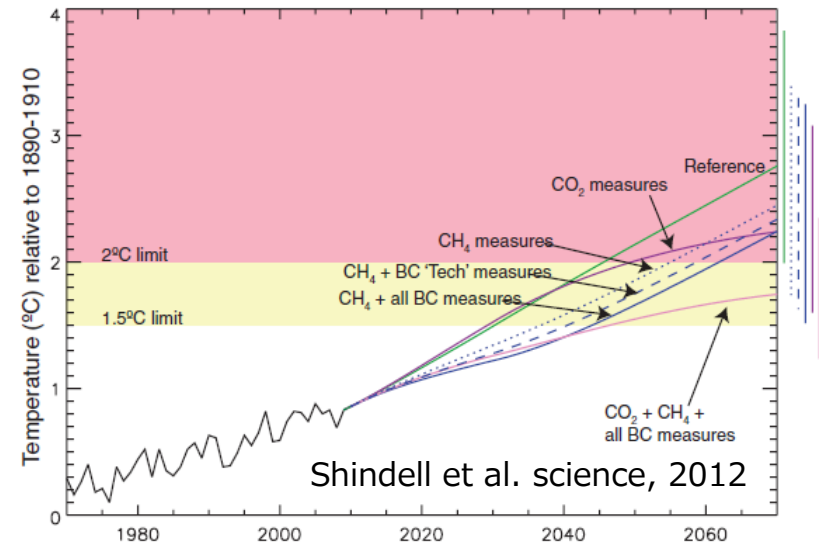
0. Motivation to DA (Story taking me here today)
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# CH<sub>4</sub>

## Global Warming Potential of CH<sub>4</sub> (IPCC) -on a 100-year horizon-

21 → 23 → 25 → 34

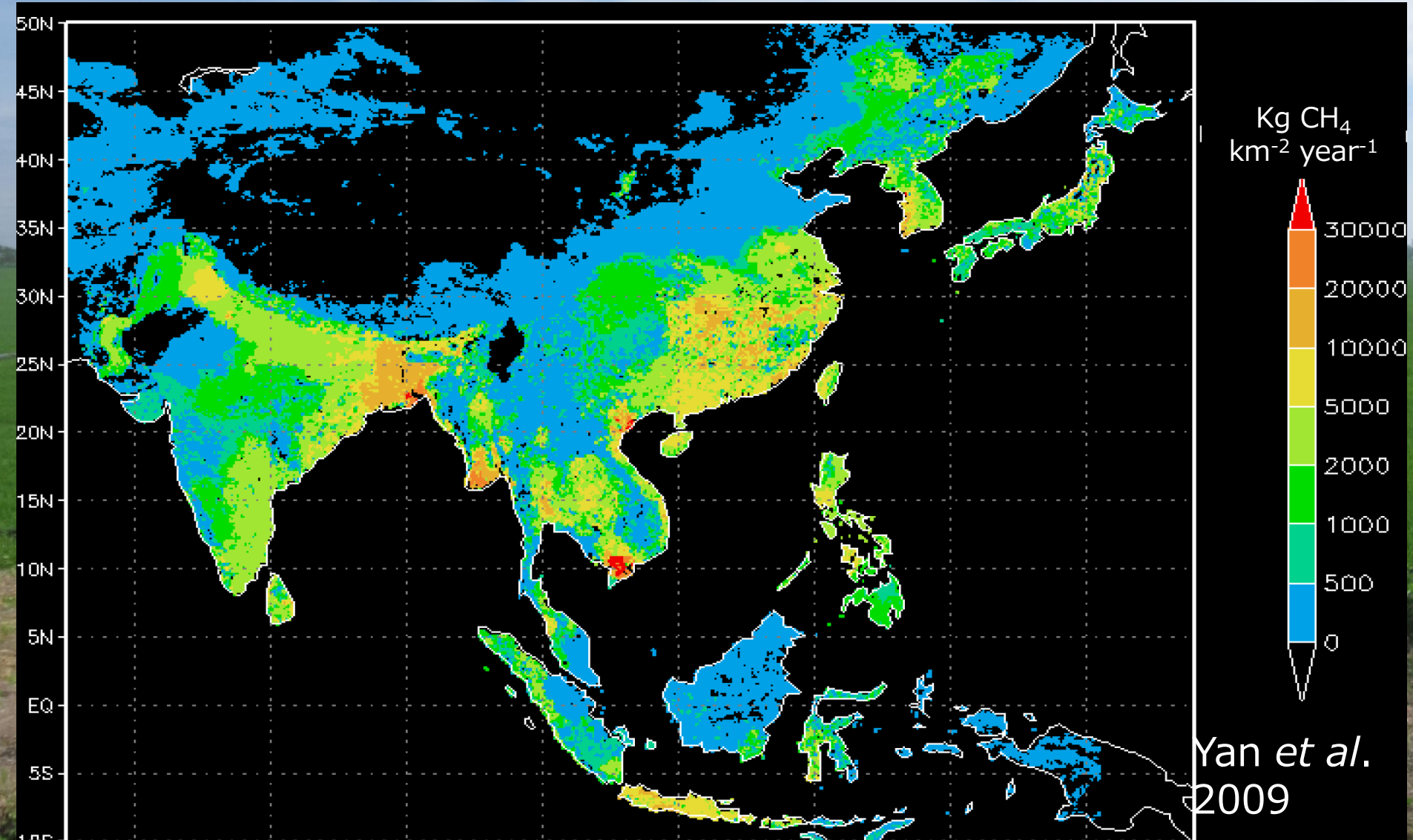
AR2	AR3	AR4	AR5
1996	2001	2007	2013



Schaefer et al. science, 2016



# Characteristics of Agriculture in Monsoon Asia



# Development economic assessment to realize scientific decision making

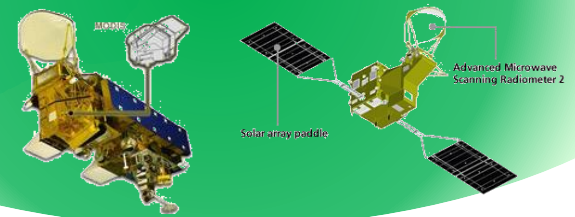
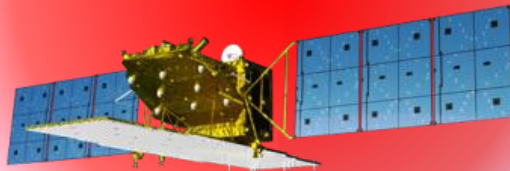
Verification with the GOSAT and atmospheric simulation

Unveiling the potential of CH<sub>4</sub> reduction and the baseline

Future prediction of CH<sub>4</sub> emission in global scale

Monitoring present status of water management

Monitoring/Reporting long-term changes of rice cropping frequency, fallow season management and inundation status



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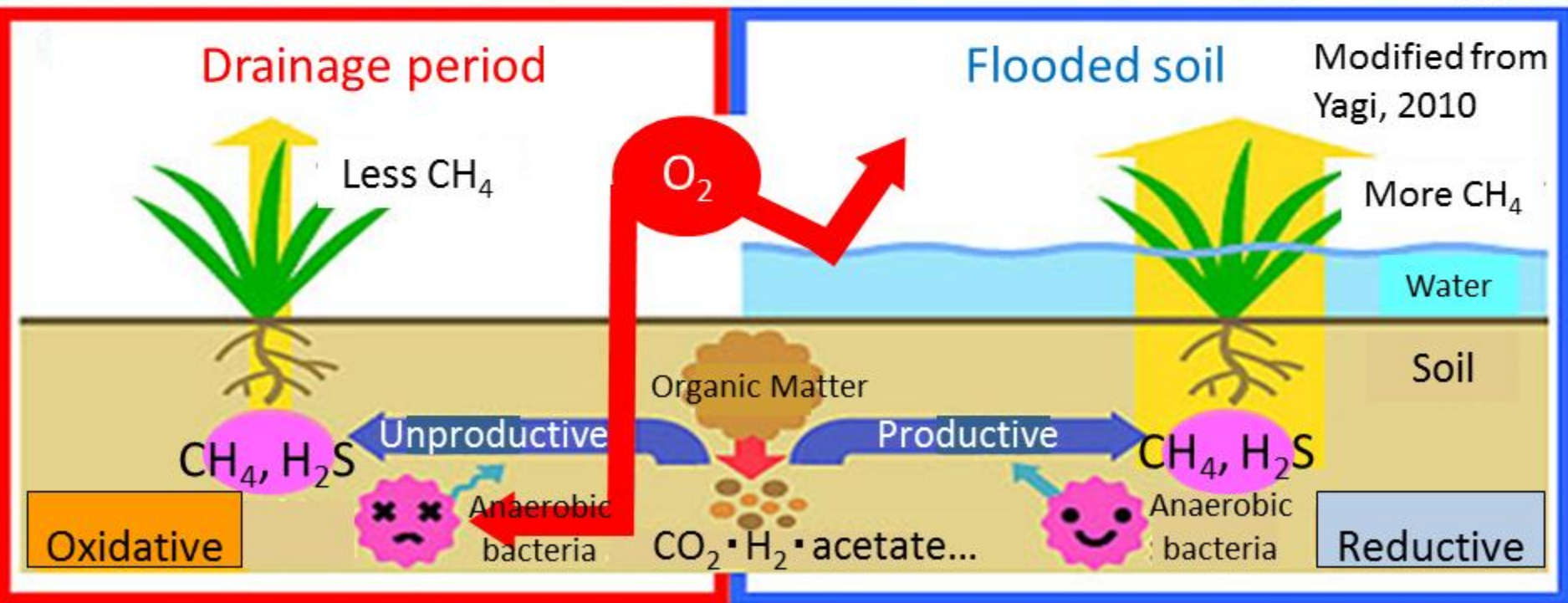
- Continuously flooded nearly through a year
- +
- High straw production



- Anaerobic stress for rice production
- High GHGs emission

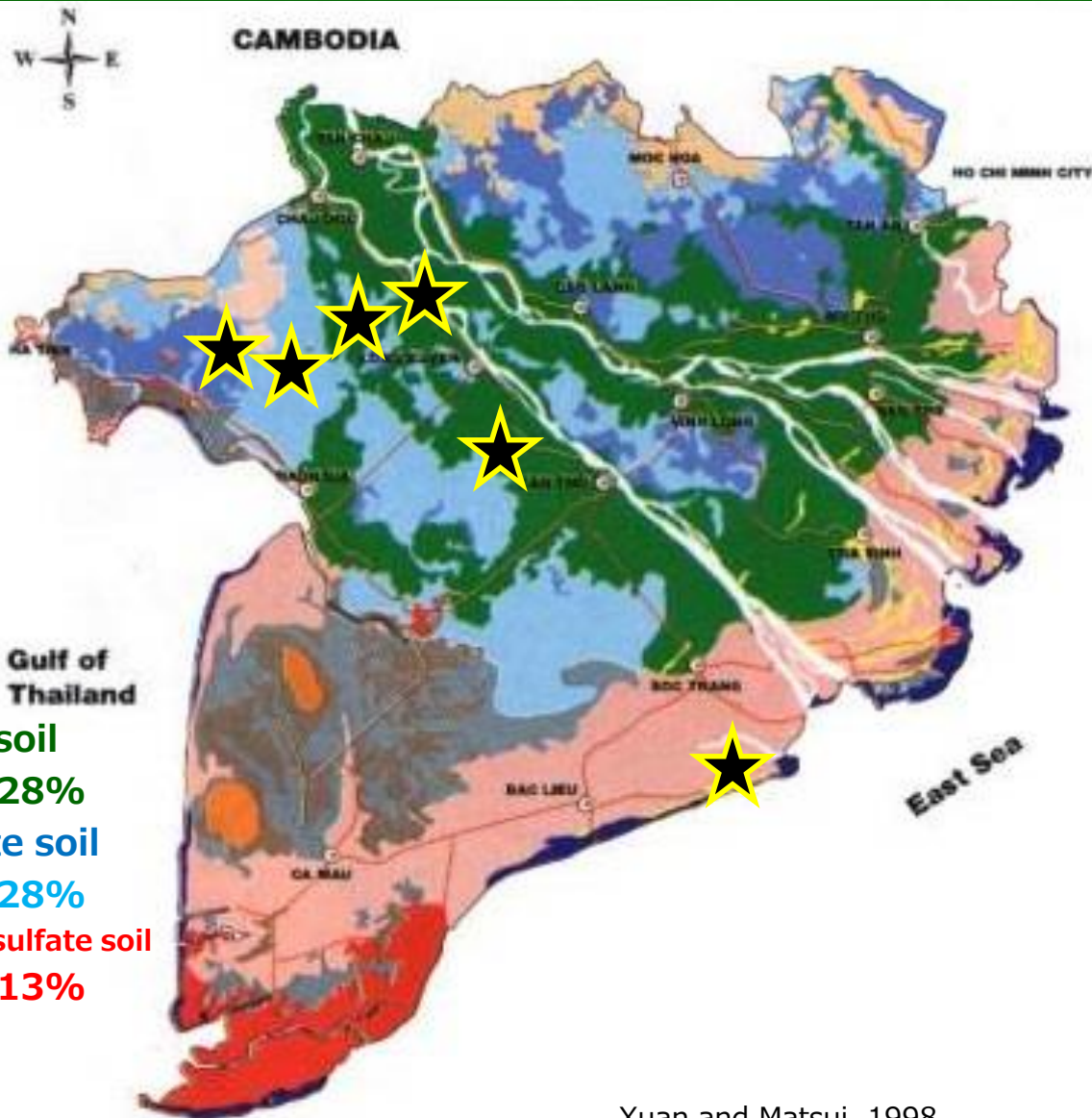
(Alternate **W**etting and **D**rying)

- Irrigation-water saving
- Anaerobic-stress mitigation
- GHGs mitigation





# Obtained annual CH<sub>4</sub> emission data so far



EF baselines in  
philippines  
(GoP 2014, Basak 2016)

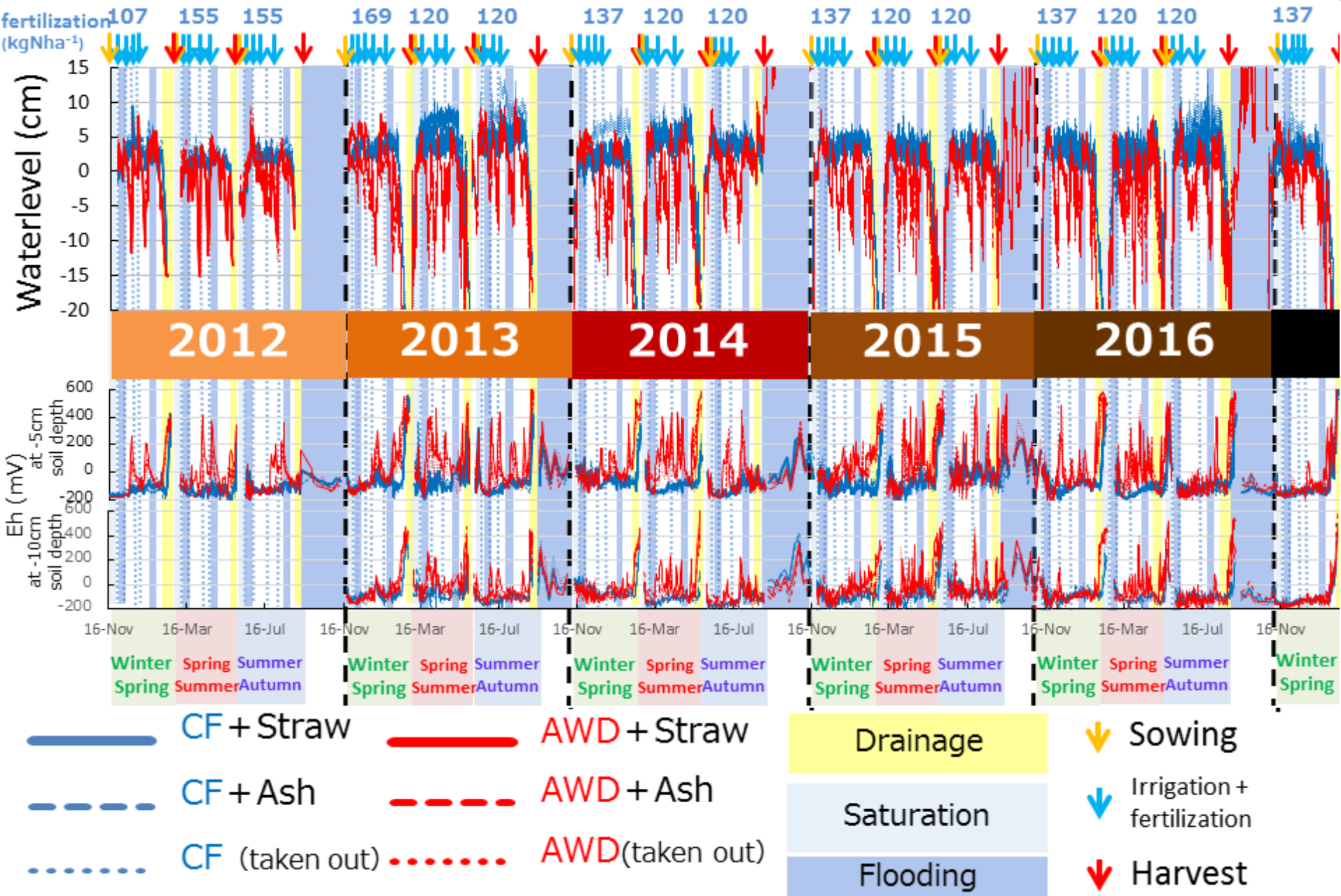
- Single crop (rainy season)
- Single crop (dry season)
- double crop

(n=3)

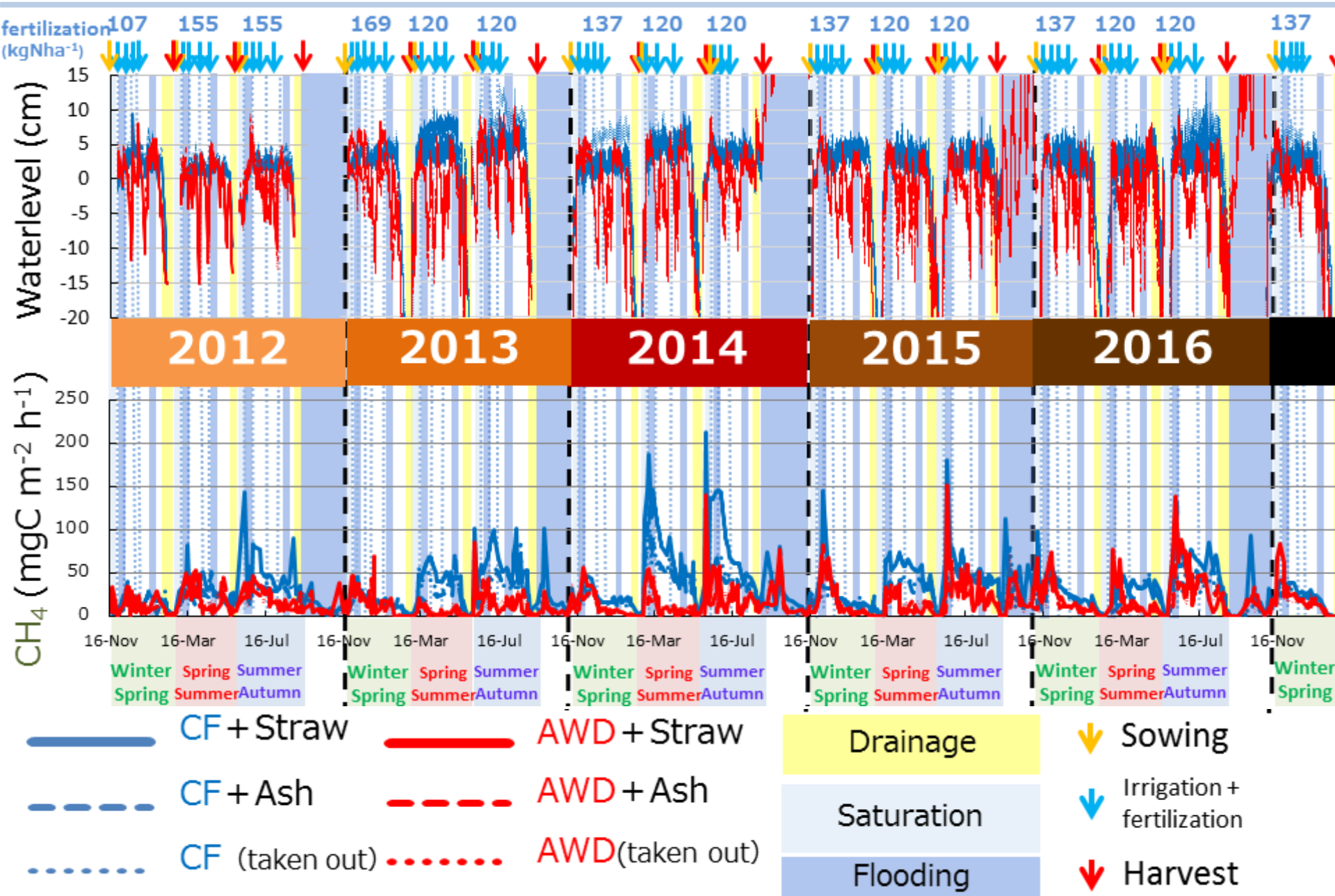
**alluvial soil**  
**1.1 Mha, 28%**  
**acid sulfate soil**  
**1.1 Mha, 28%**  
**potentially acid sulfate soil**  
**0.5 Mha, 13%**

Xuan and Matsui, 1998

# Characteristics of the Mekong delta

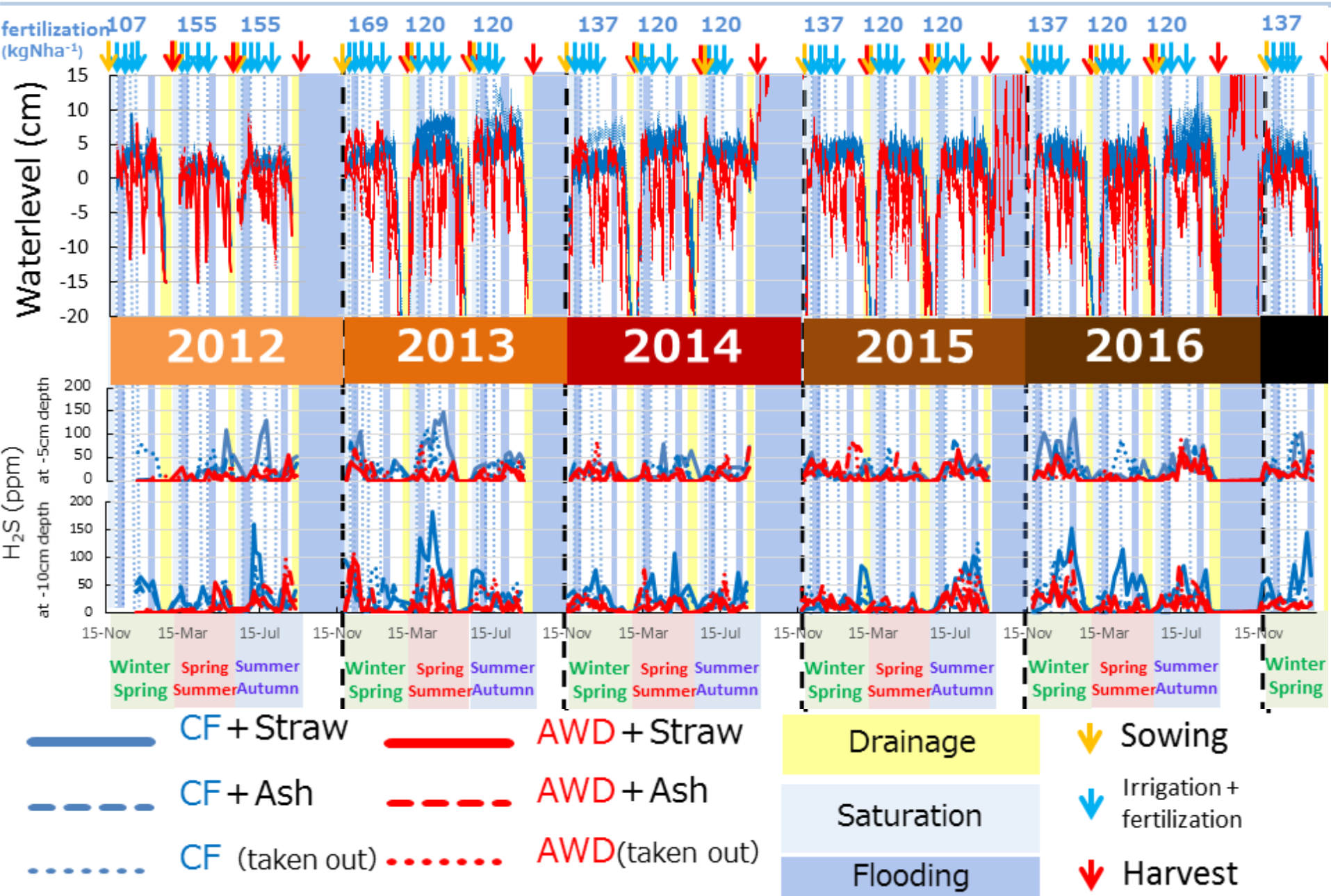


# Characteristics of the Mekong delta



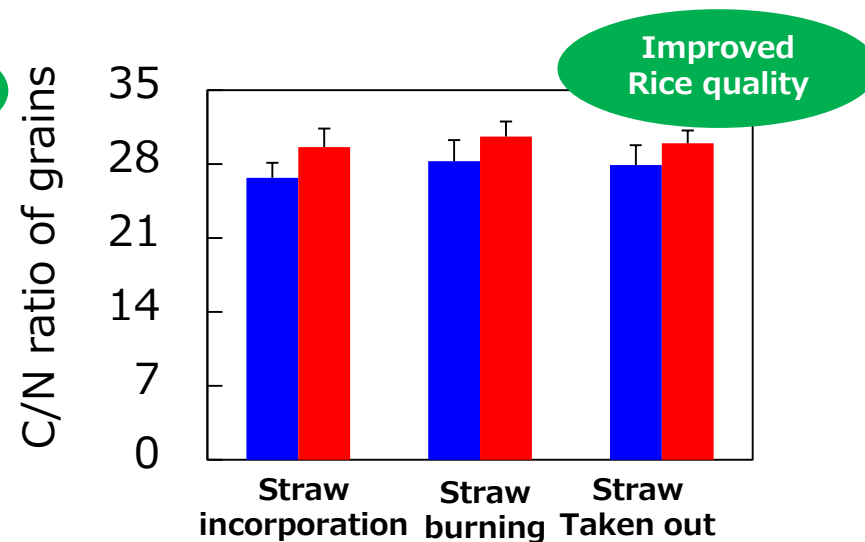
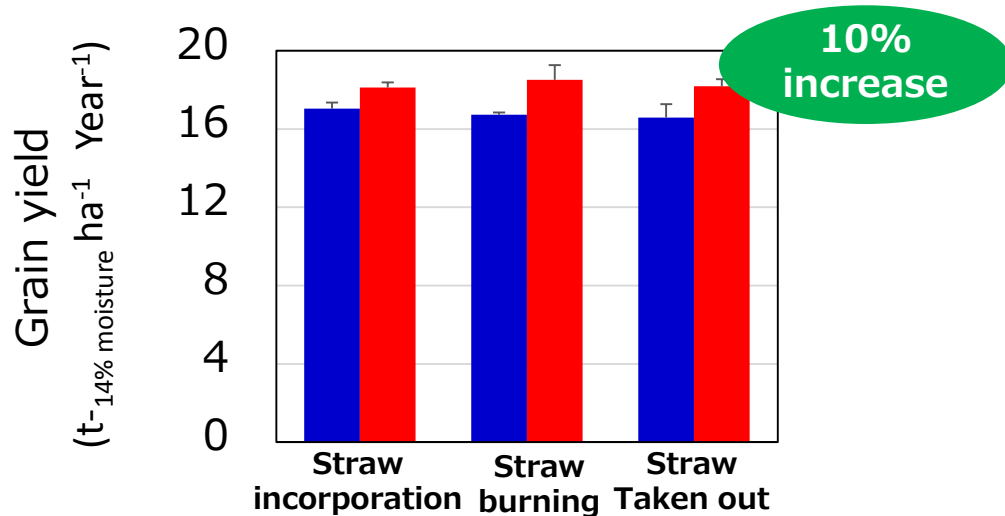
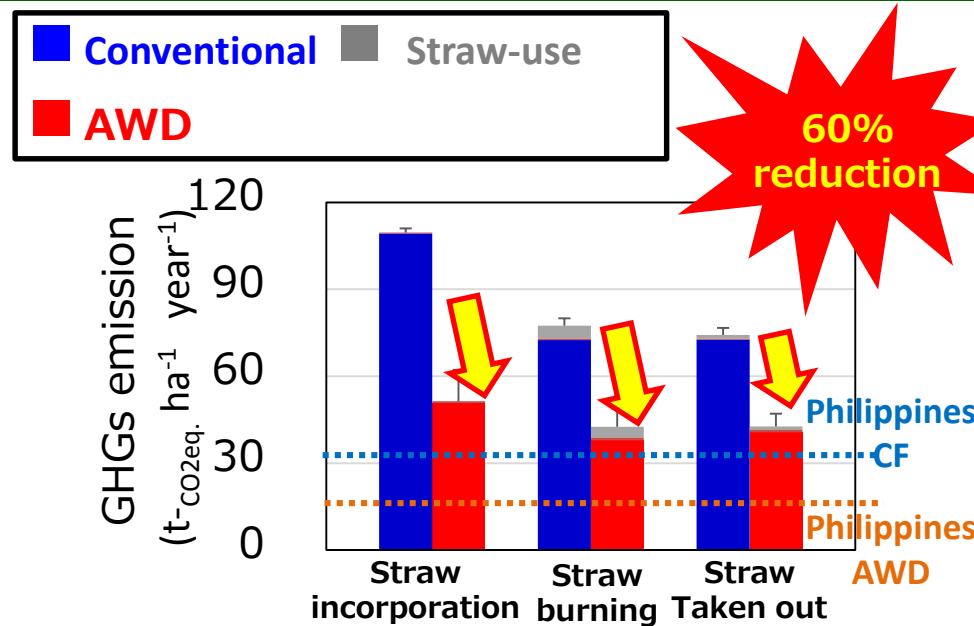
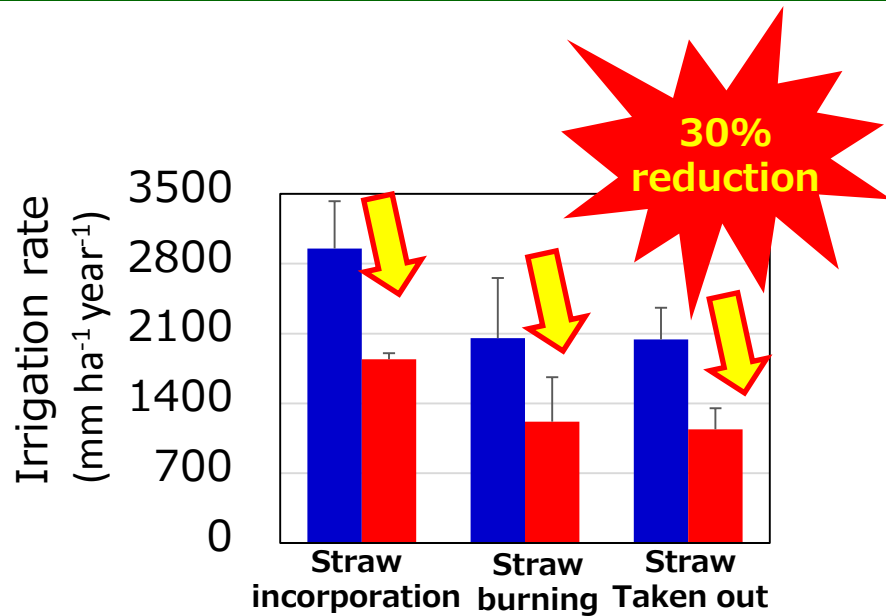


# Characteristics of the Mekong delta

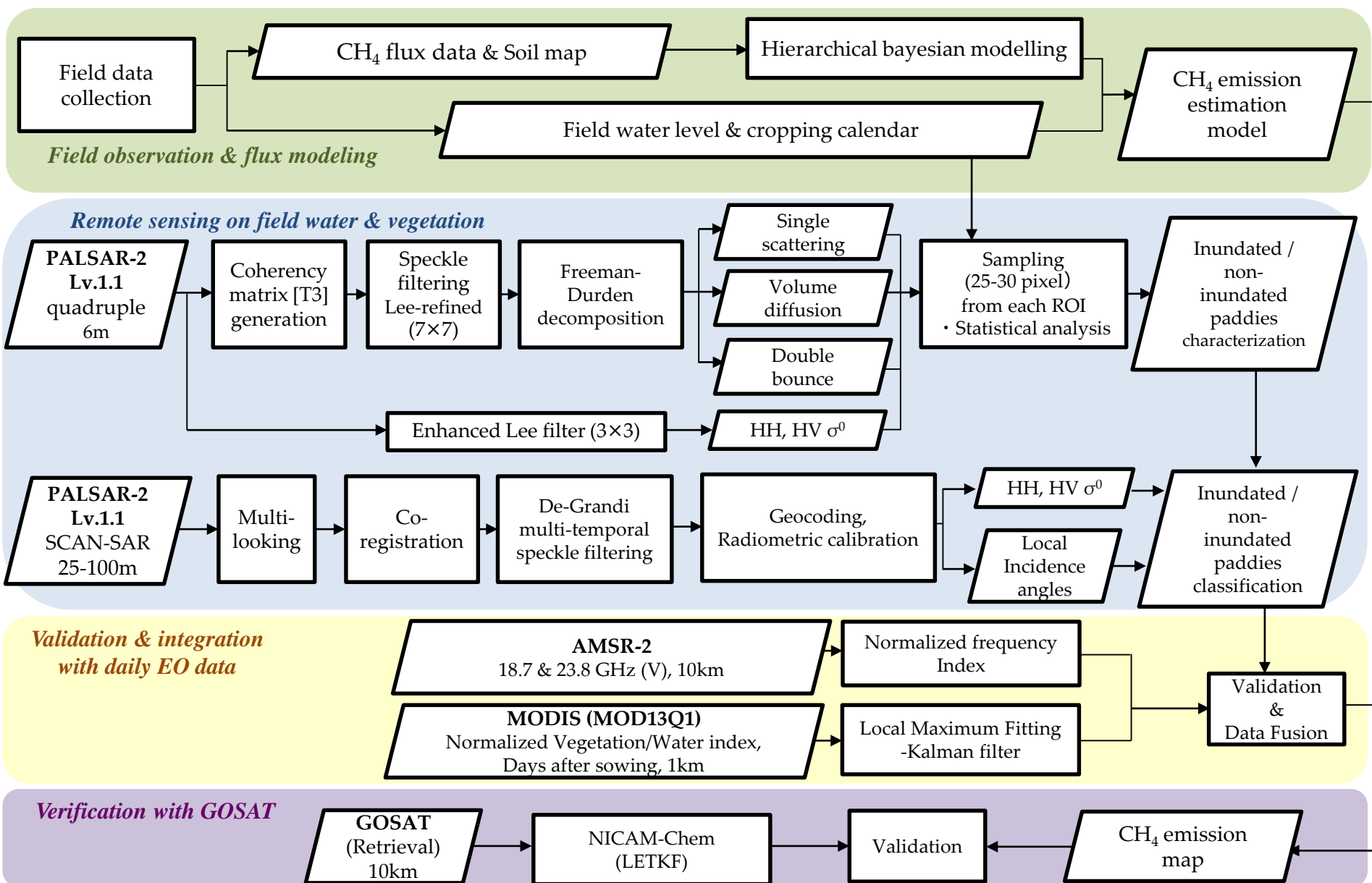




- Reduction of irrigation rate & GHGs (2012-2016)
- Increase of rice grains and its quality

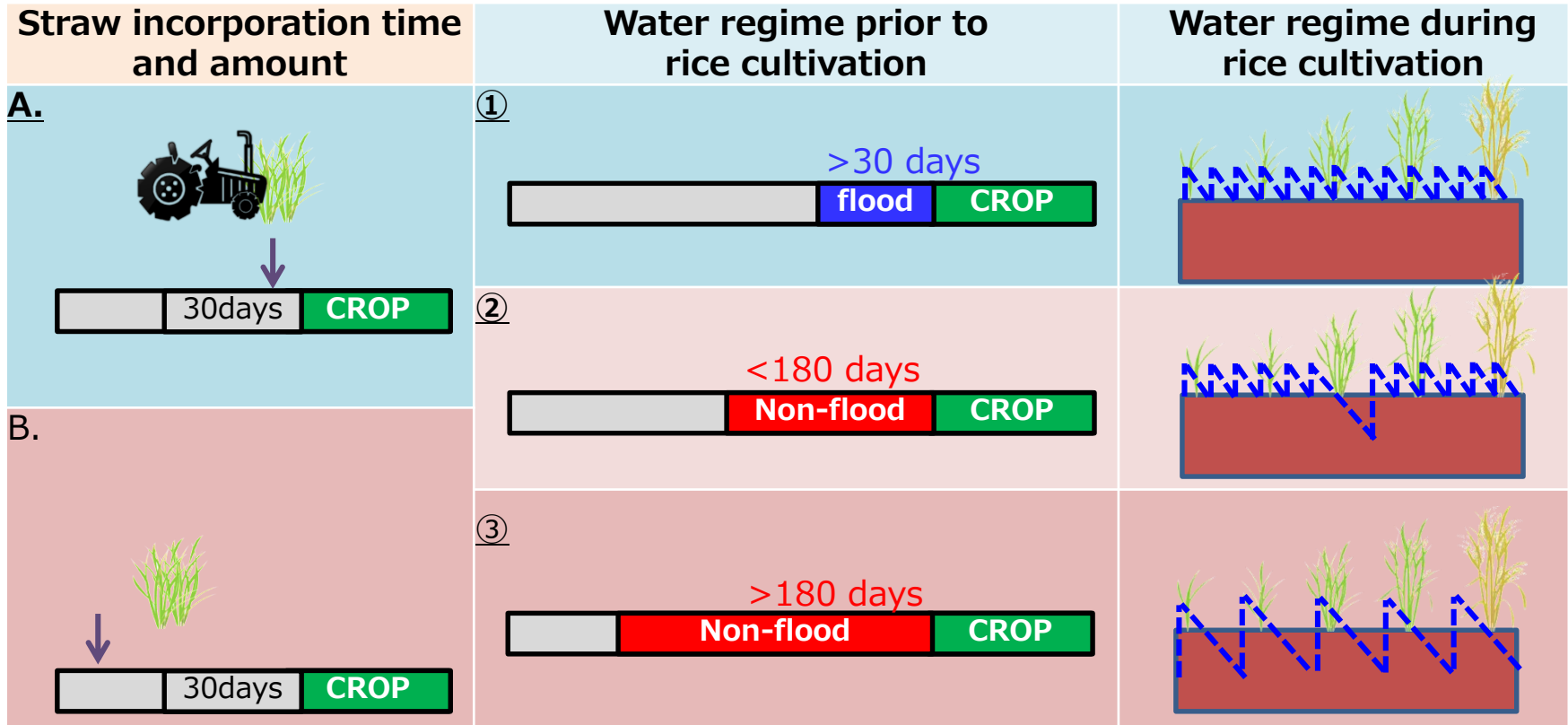


# Flow chart

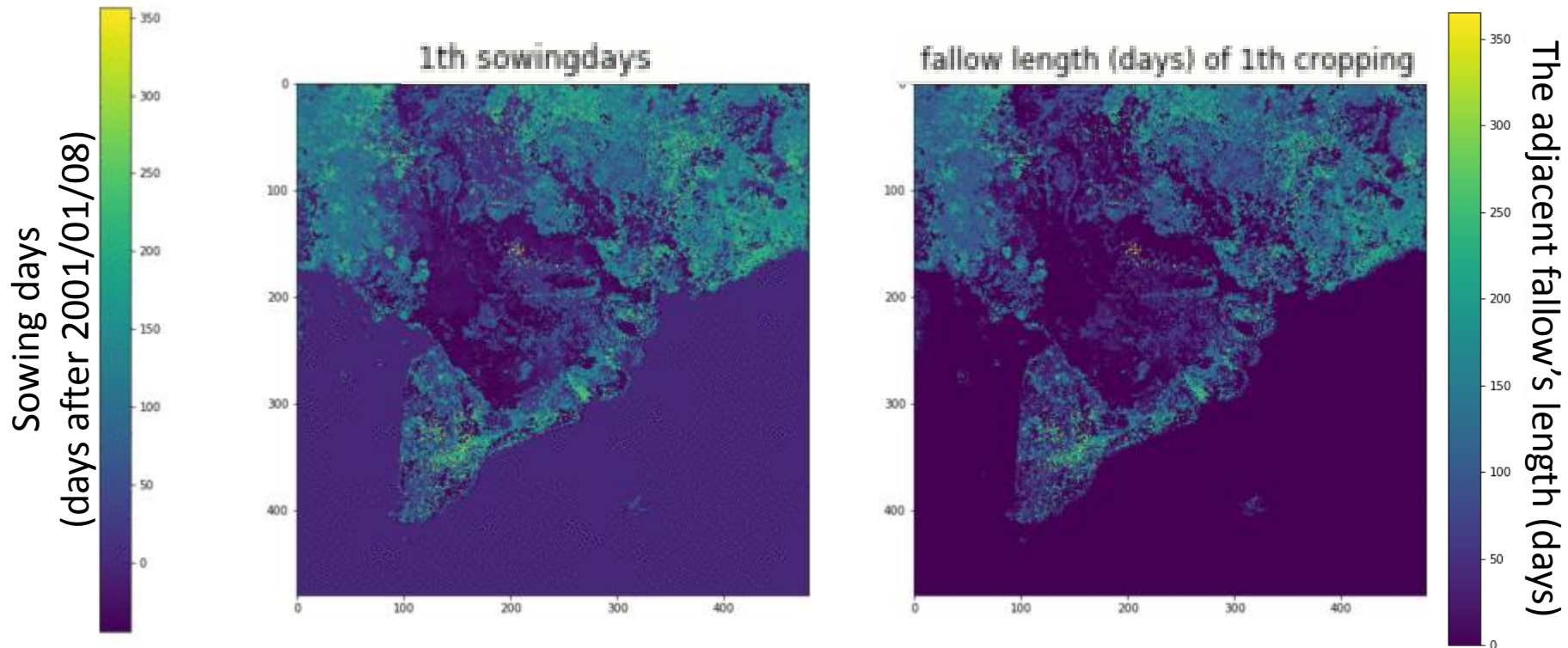


# IPCC guideline (Tier1)

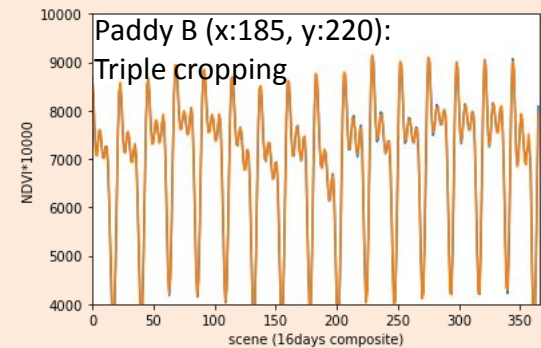
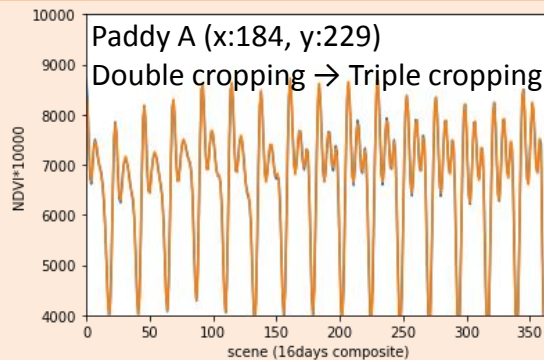
[Emission factor × Scaling factor in IPCC guideline]



# Cropping calendar evaluation with MODIS — NDVI (LMF-KF)



Samples of paddies





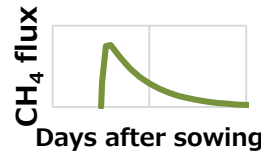
# Semi-empirical daily CH<sub>4</sub> flux (mg C m<sup>-2</sup> day<sup>-1</sup>) Model

## CH<sub>4</sub> emission on a specific date

$$= \gamma * \text{carbon\_management} / \text{non-inundated\_fallow} / \text{inundated\_fallow} * \text{water\_management} * \alpha * \beta$$

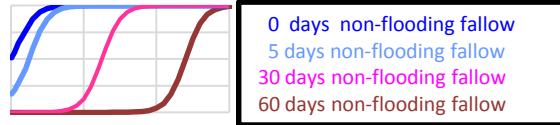
### carbon\_management

$$= [\exp(-DAS * \delta) - \exp(-DAS * (\delta + \omega)) + \kappa]$$



### non-inundated\_fallow

$$= [1 + \exp(-1 * \zeta * (DAS - l * \text{days of nonflooding days of the former fallow}))]$$



### inundated\_fallow

$$= \exp(\epsilon * \text{days of flooding days of the former fallow})$$

### water\_management

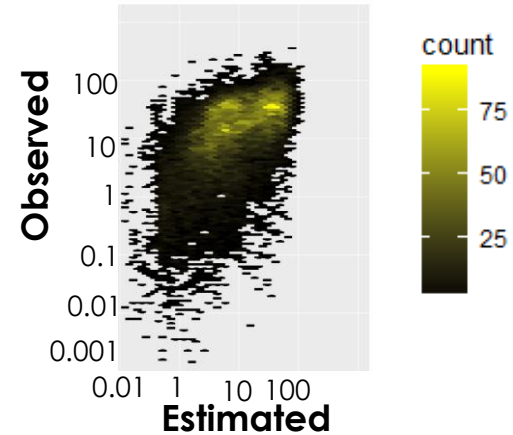
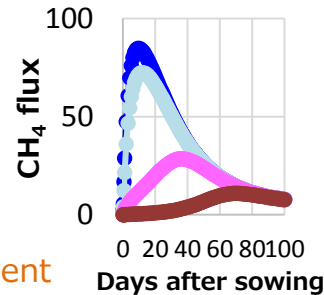
$$= \exp(\eta * \text{inundated days during the last 10 days})$$

DAS ← days after sowing

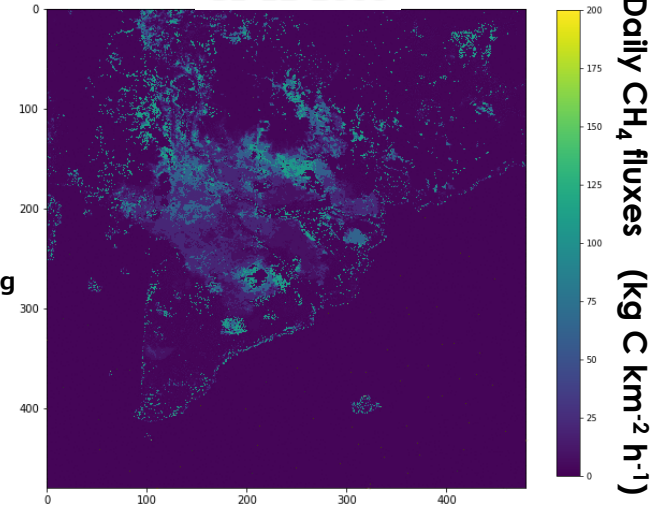
$\alpha$  ← straw incorporation coefficient

$\beta$  ← acid sulfate · coastal sandy soil coefficient

$\gamma, \eta, \delta, \epsilon, \omega, \zeta, l, \kappa$  ← constant (>0)



31-12-2000



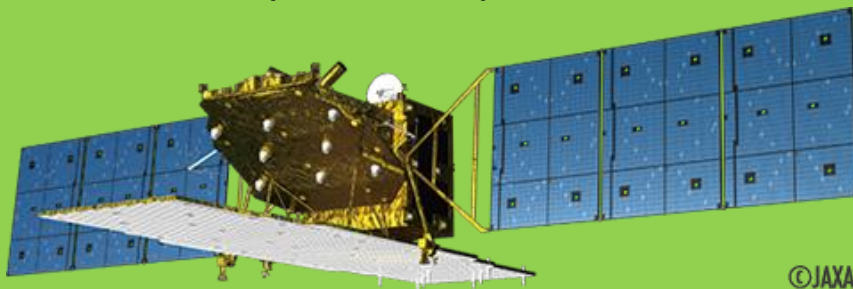
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# Satellite remote sensing of soils

## ALOS-2/PALSAR-2

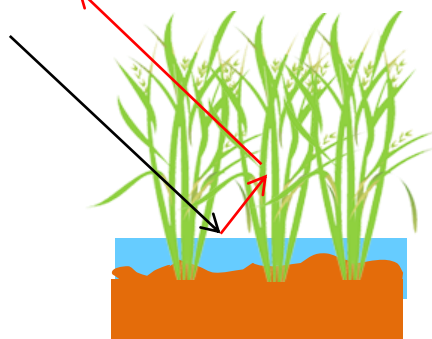
- Lband-Synthetic Aperture Radar -



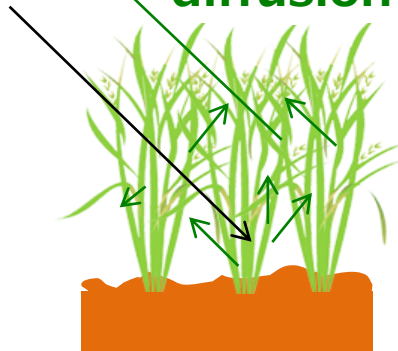
### Scattering model decomposition



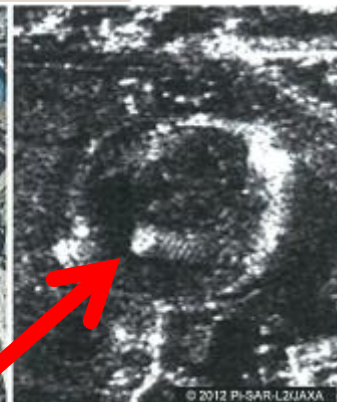
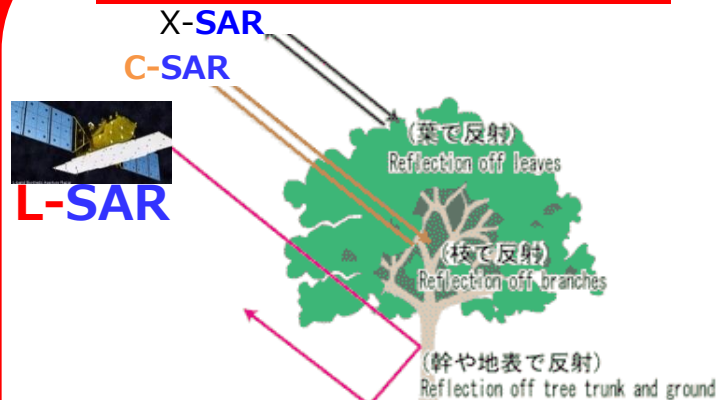
**Double bounce**



**Volume diffusion**



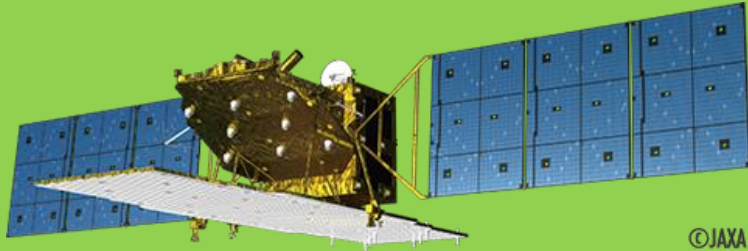
### High transparency



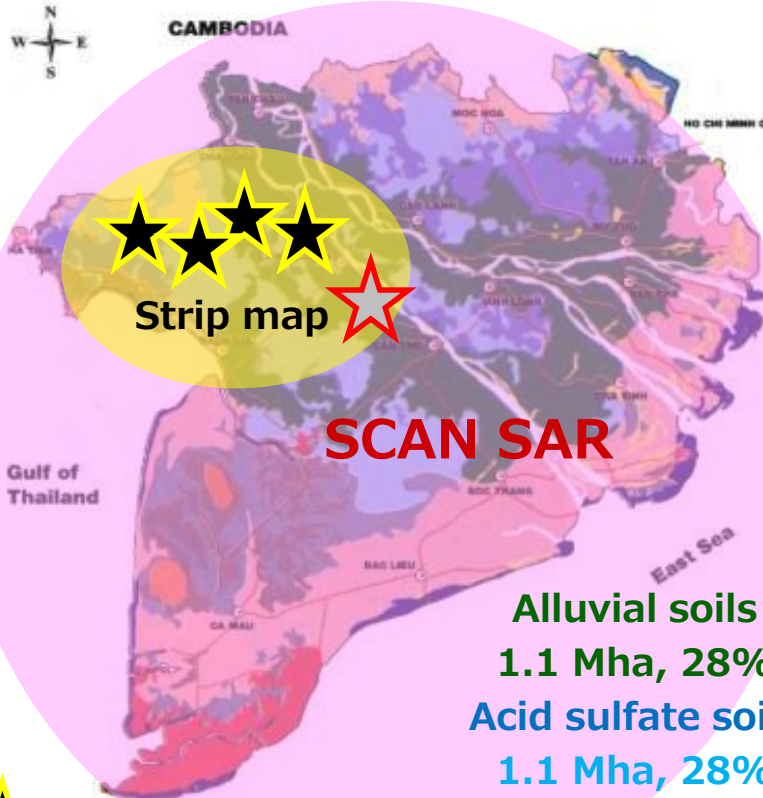
MITSUBISHI CO, LTD.

# ALOS-2/PALSAR-2

- Lband-Synthetic Aperture Radar -



©JAXA



**Alluvial soils**  
1.1 Mha, 28%

**Acid sulfate soils**  
1.1 Mha, 28%

**potential acid sulfate soils**  
0.5 Mha, 13%

★ 5paddies × 4villages  
★ 30paddies × 1village

Xuan and Matsui, 1998

**PALSAR-2 Lv.1.1**  
(quad. CEOS)  
23 scenes

Coherency matrix [T3]  
generation

Speckle filtering  
LEE refined  
(7×7)

**Polarimetric decomposition**

Freeman -Durden      Cloud -Pottier

Sampling (25-30pixel)  
from each ROI  
&  
Statistical analysis

**Classification of *inundated paddies* and *non-inundated paddies*  
which is covered by rice plants**

**PALSAR-2 Lv.1.1**  
(SCANSAR CEOS)  
105 scenes

Multilooking

Co-registration

De Grandi  
multi-temporal  
filtering

Geocoding  
&  
Radiometric  
calibration

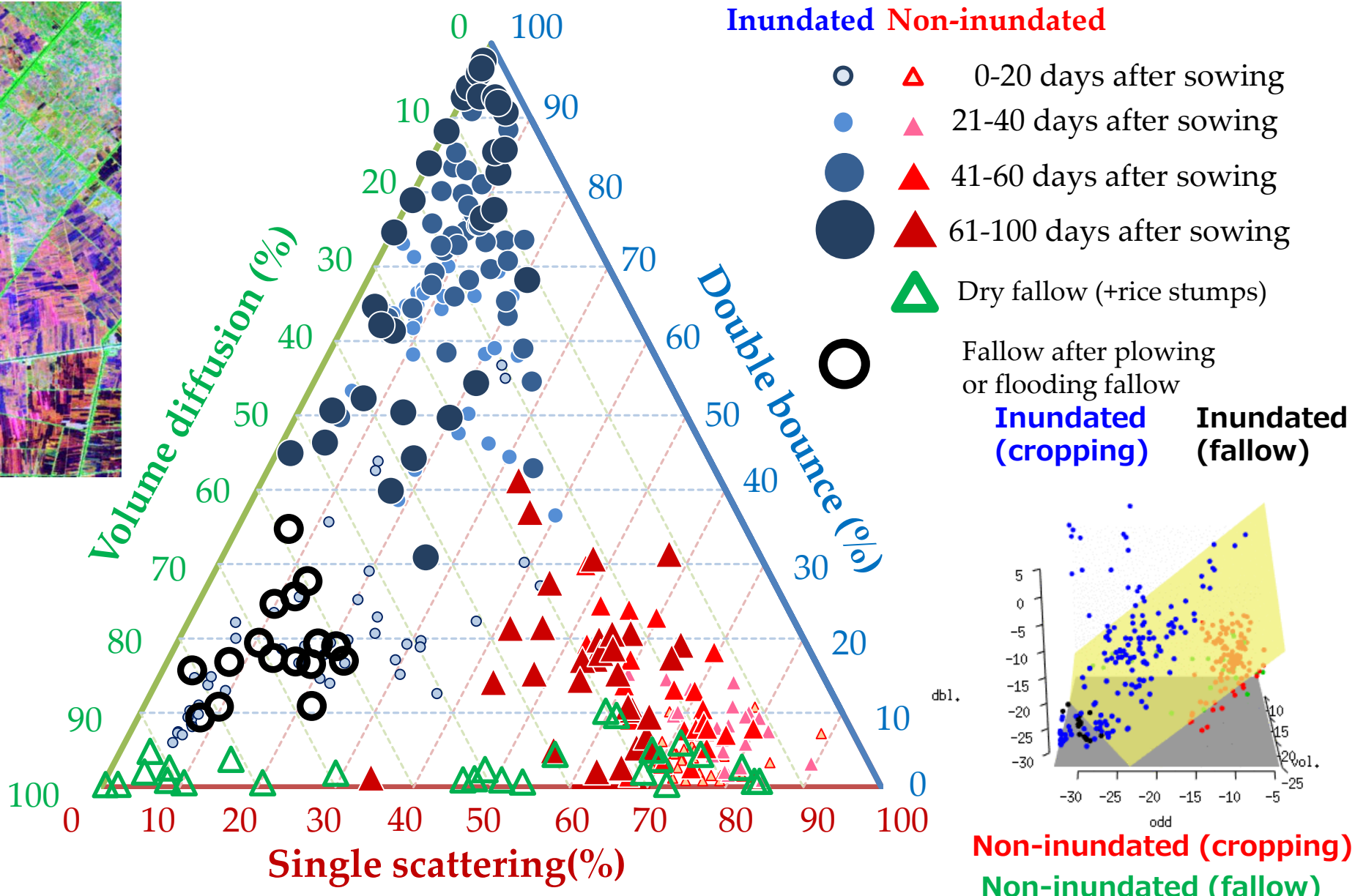
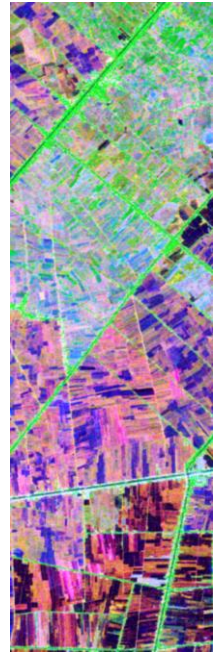
HH    HV    Incidence angle

Rice paddy masking  
&  
Statistical analysis

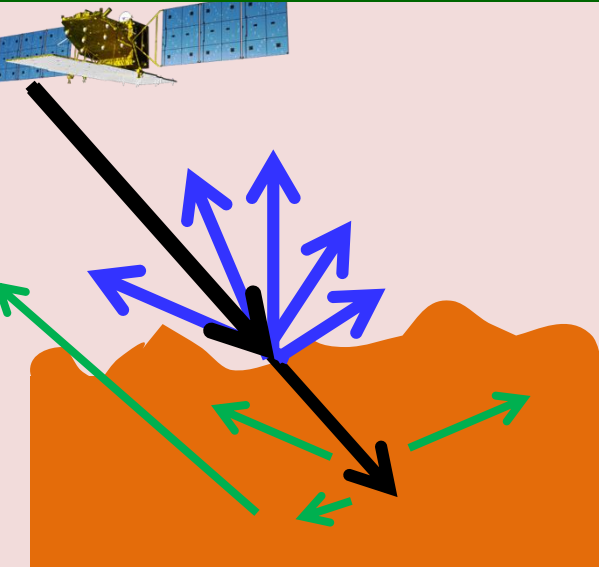
Modified from Avtar *et al.* 2012



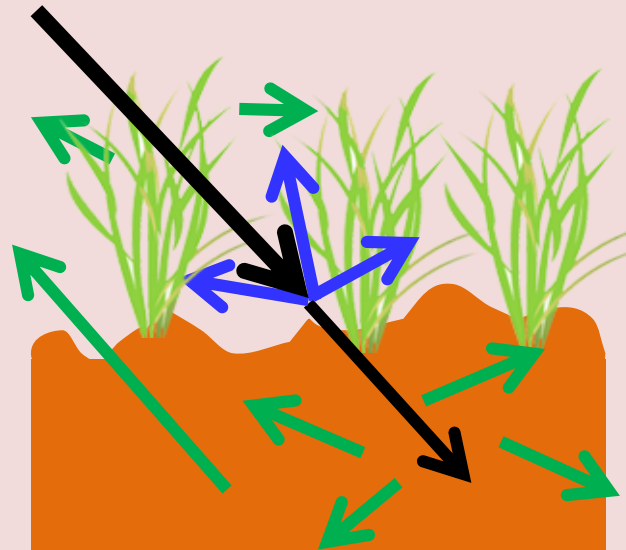
# -Freeman-Durden decomposition-



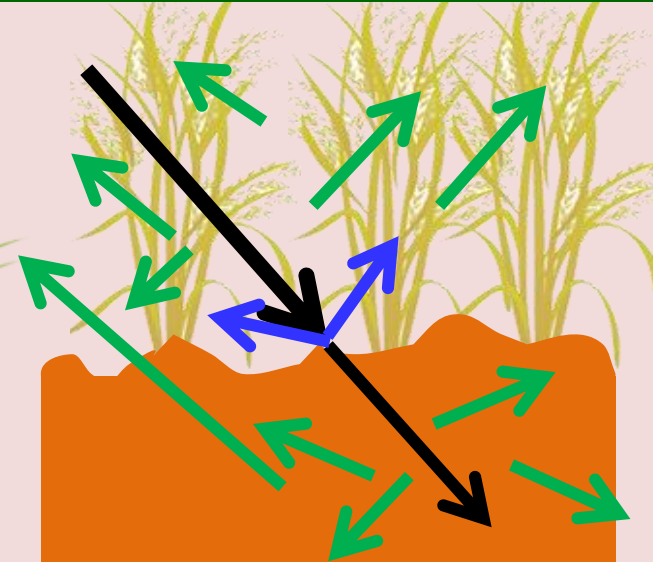
# Dominant scattering type



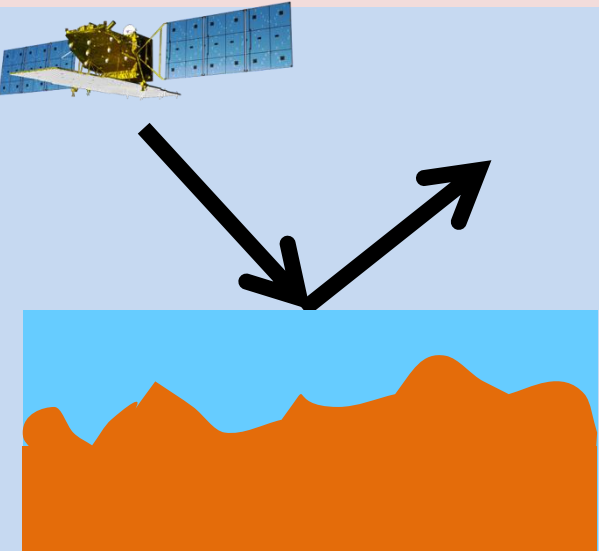
Single scattering



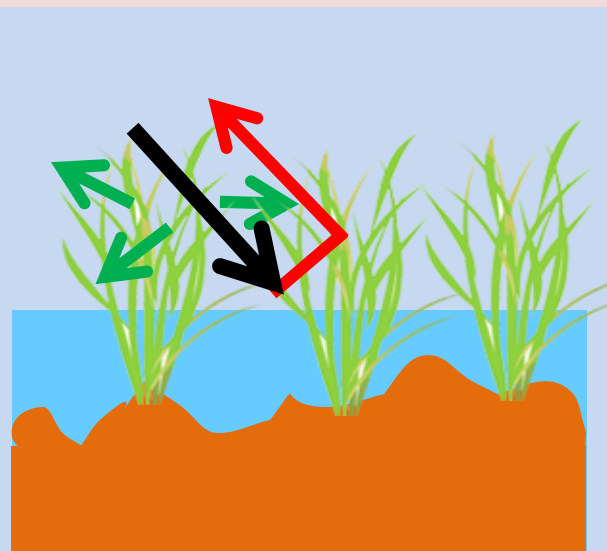
Single (+ Volume)



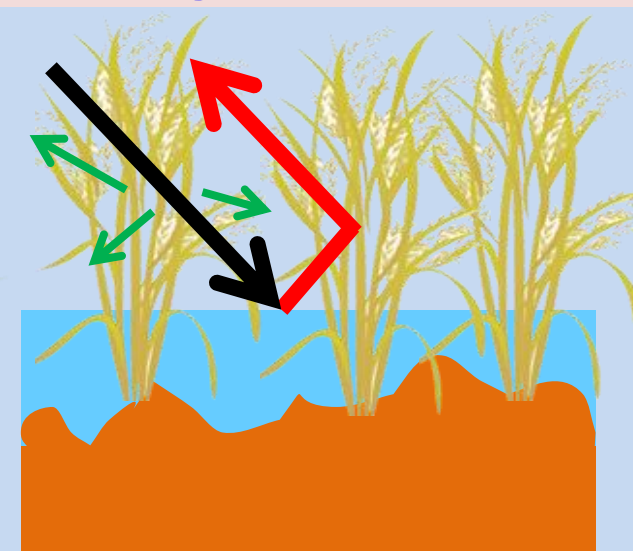
Single + Volume



Specular reflection



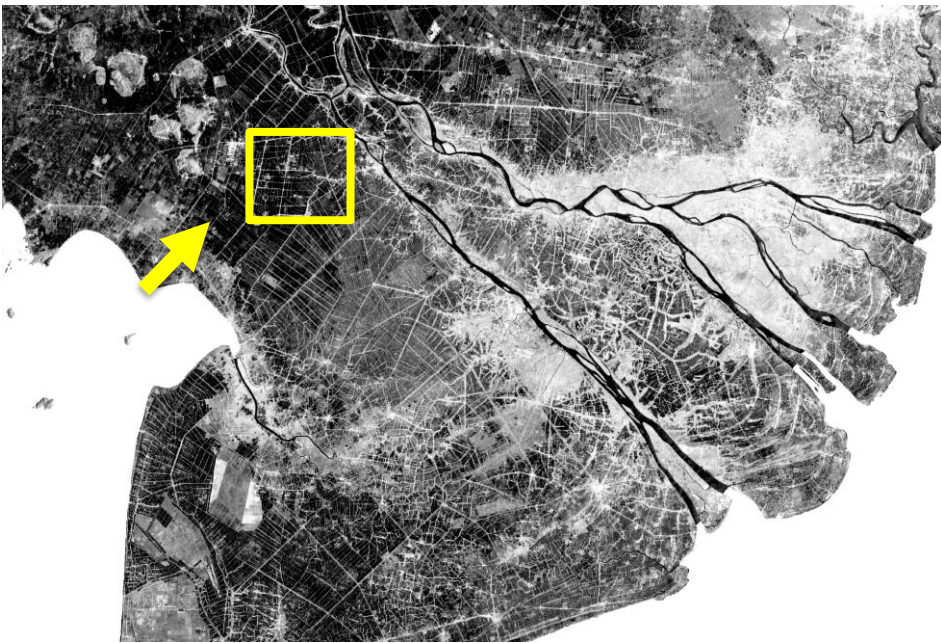
Volume + Double



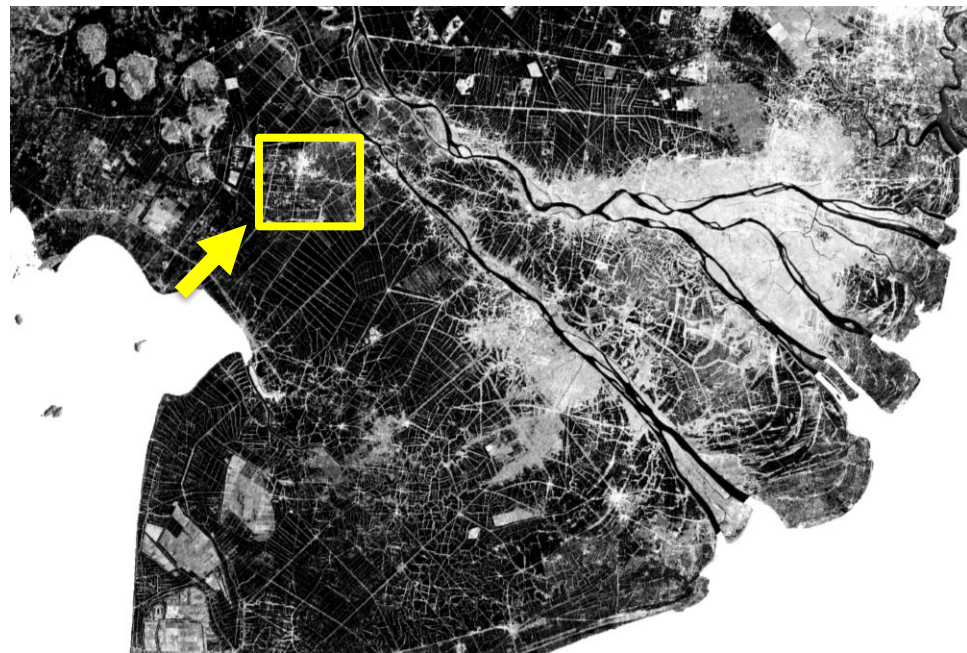
Double (+ Volume)

# SCANSAR (intensity - $HH\sigma^0$ )

**Dry season (2015 Apr. 10)**



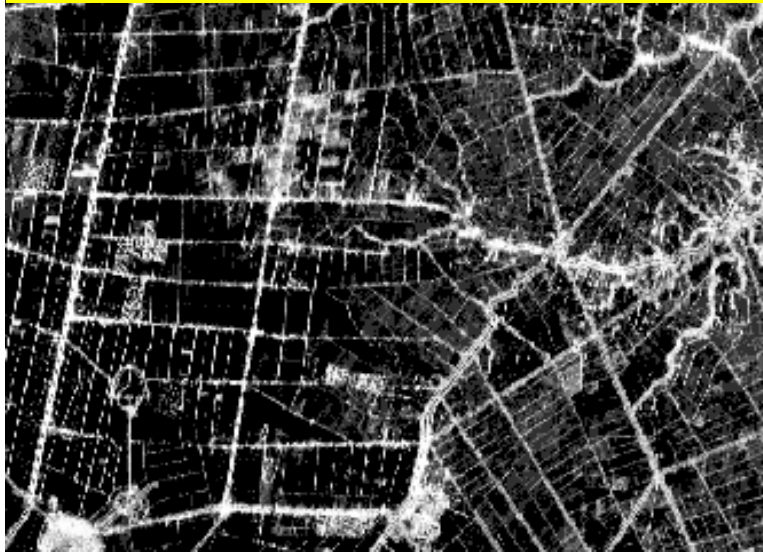
**Flooding season (2015 Oct. 23)**



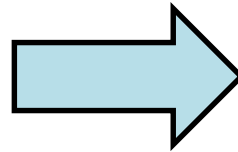


# Double bounce detection by SCANSAR (intensity - $HH\sigma^0$ )

**Dry season** (2015 Apr. 10)



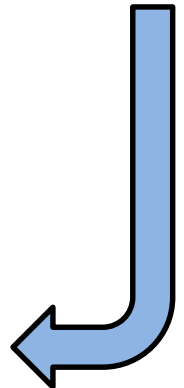
**Rainy season** (2015 Jul. 03)



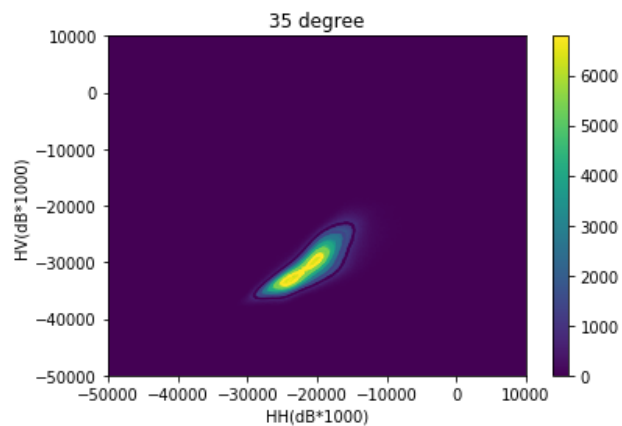
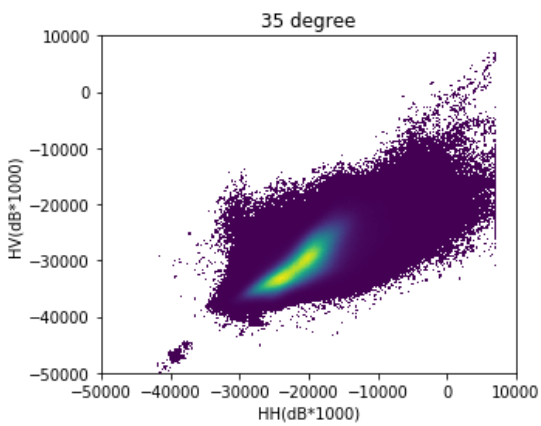
**Flooding season** (2015 Oct. 30) -LANDSAT-8-



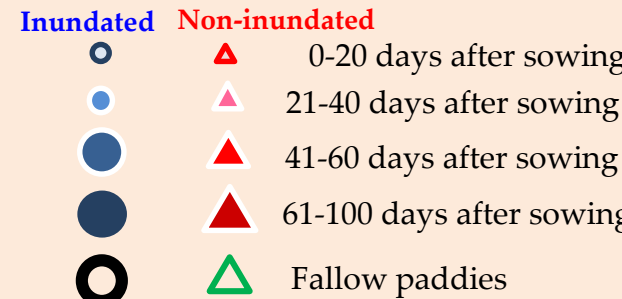
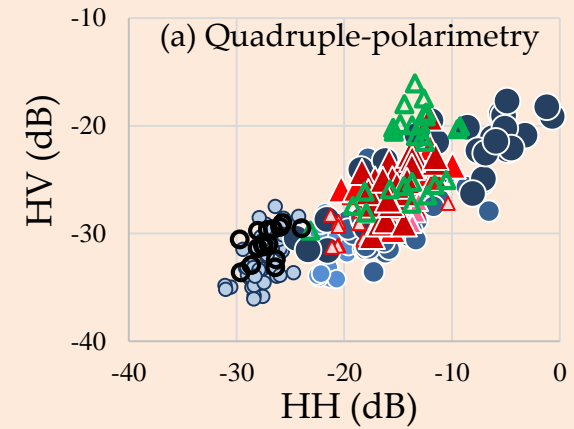
**Flooding season** (2015 Oct. 23)

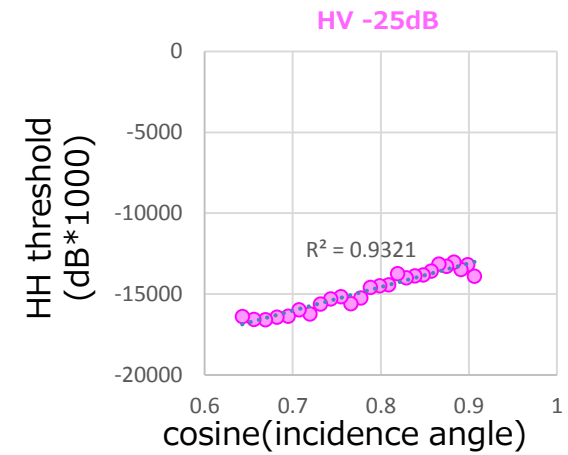
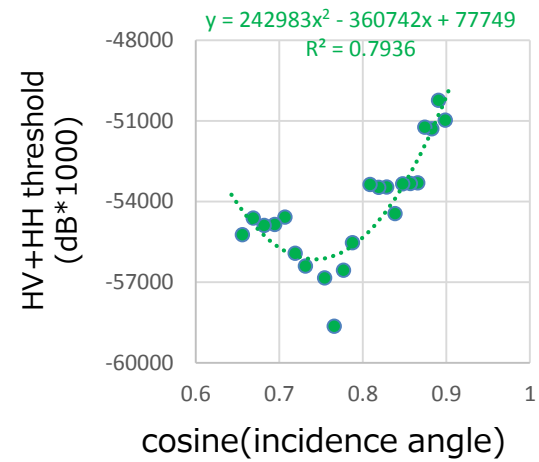
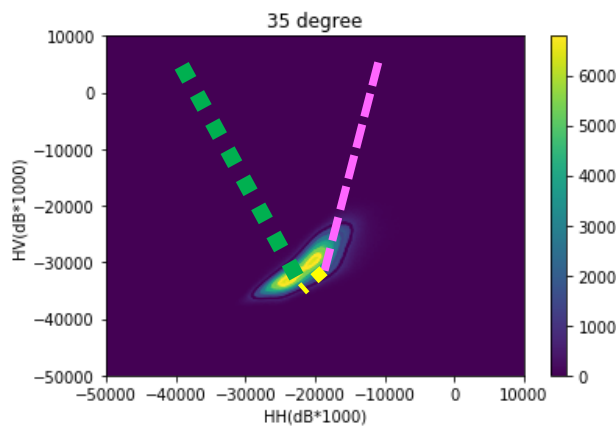
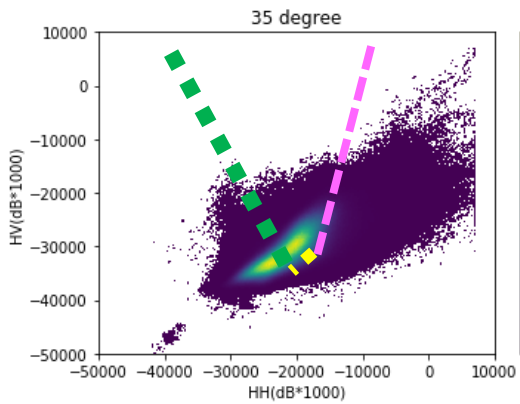






## Full-polarimetry (3m)

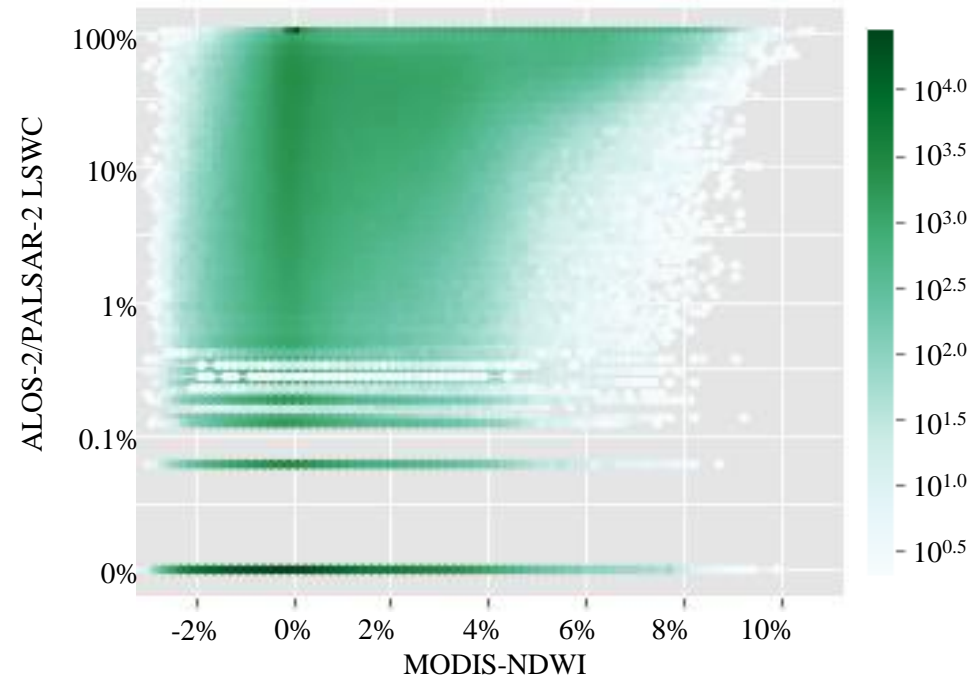
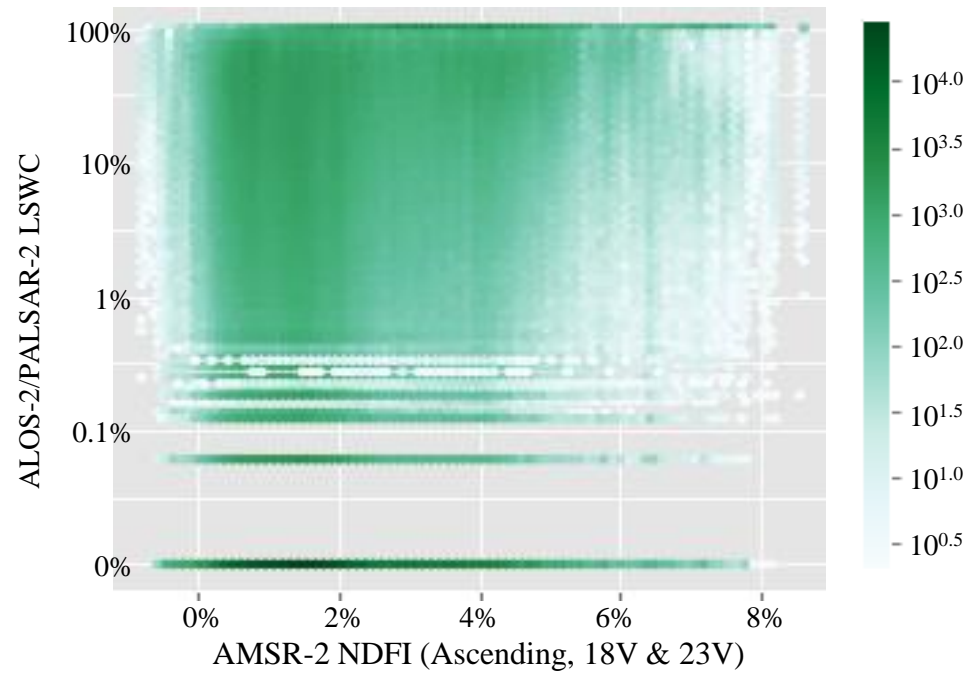
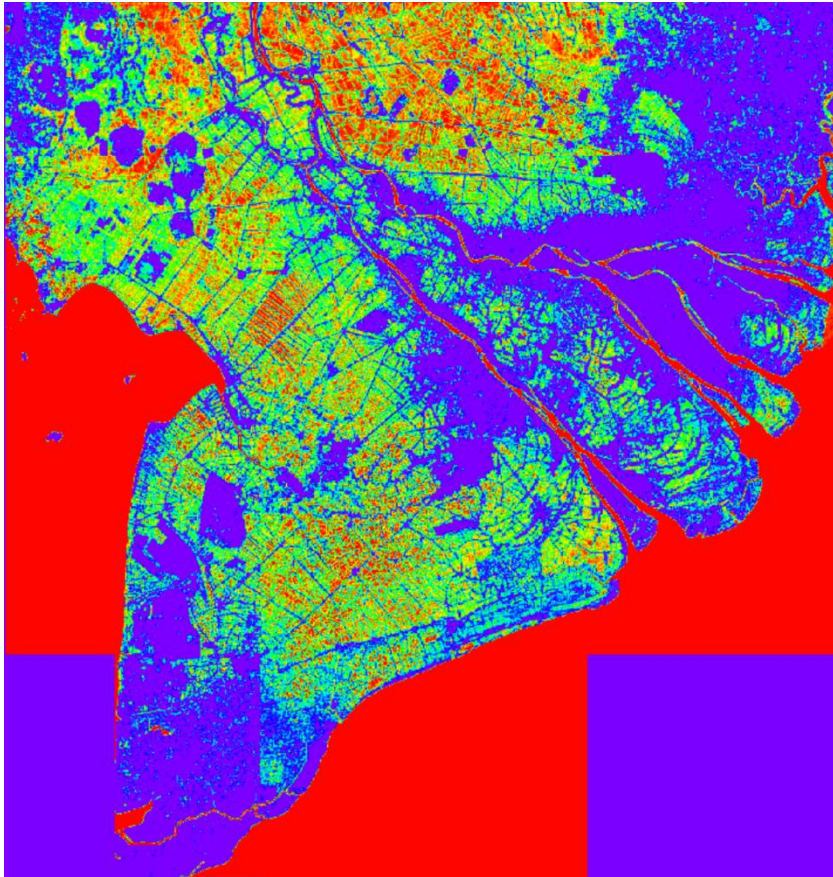




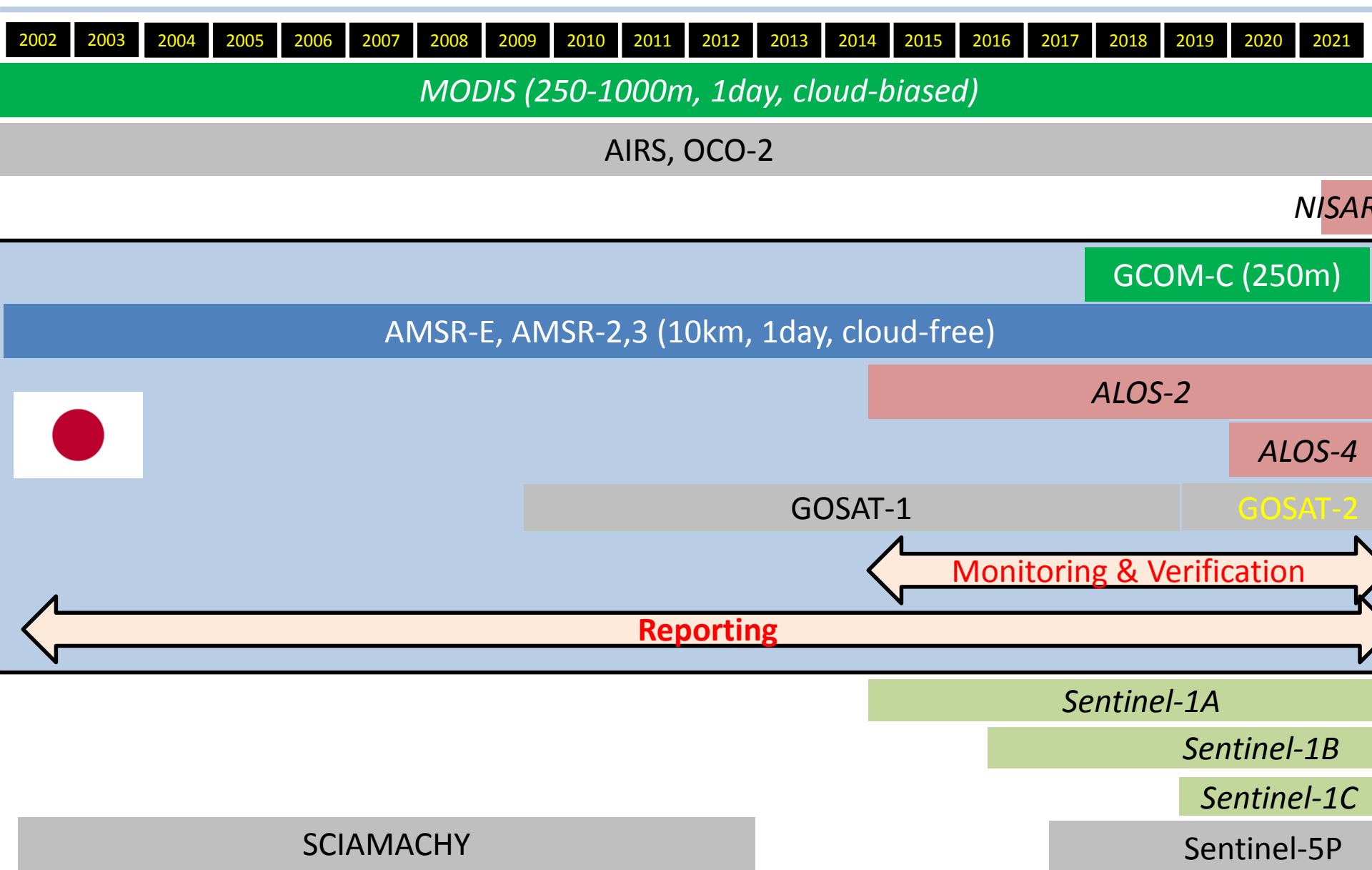
HH threshold (dB) =  $0.550 \cdot HV + 12.9 \cdot \text{cosine}(IA) - 11.2$

# Floodability analysis

(Cumulative LSWC/  
observation scenes)



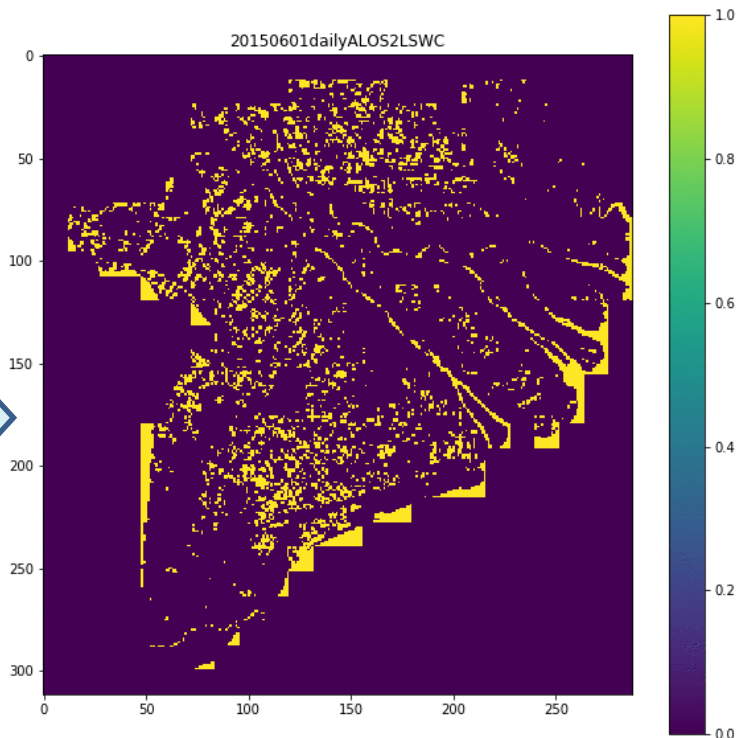
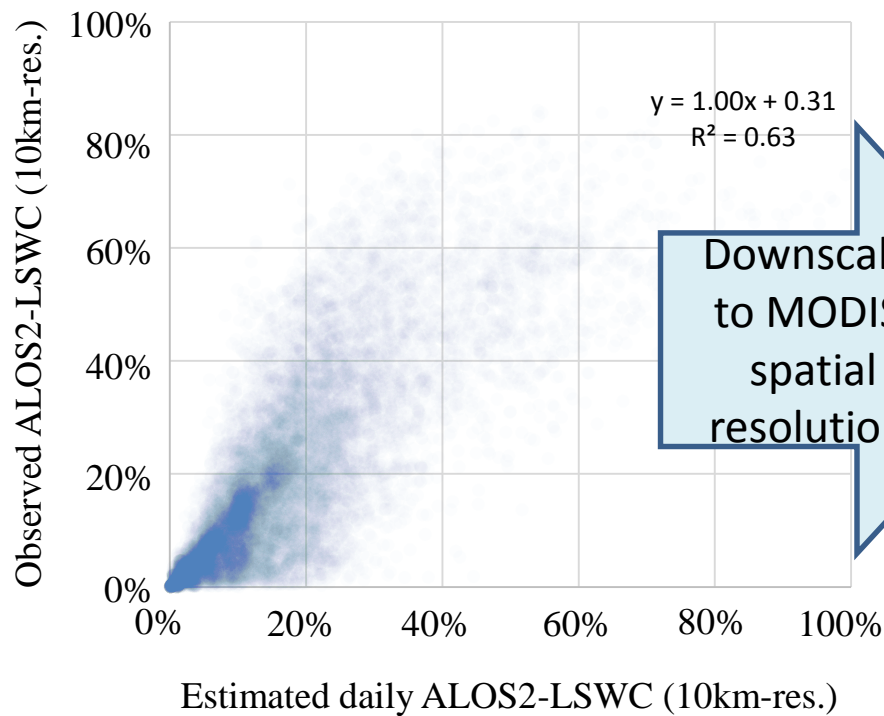
# MRV and available data





# Daily ALOS2-LandSurfaceWaterCoverage estimation

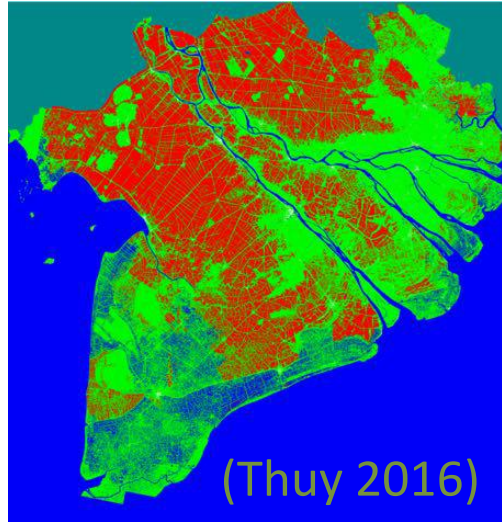
$$= (\text{ALOS2floodability} * \omega + \zeta) * \exp(\text{AMSRNDFI} * \delta - \text{MODISLSVC} * \delta)$$



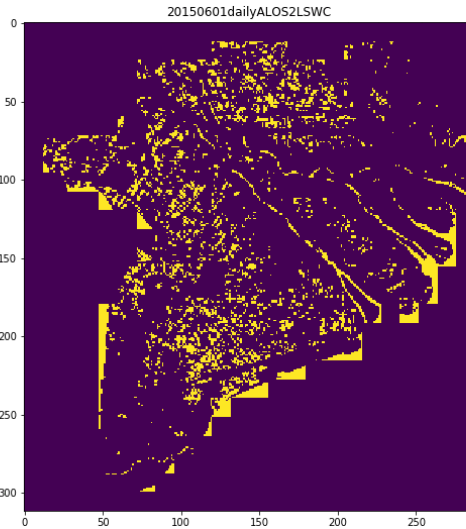
# MONITORING with ALOS2 (since 2014)



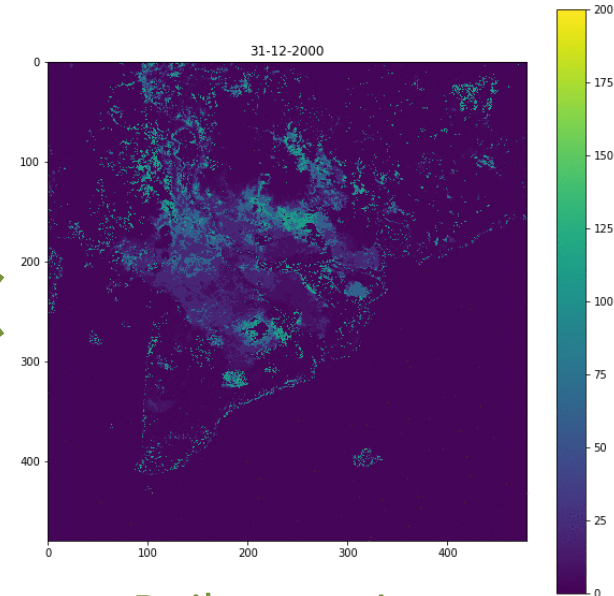
# REPORTING with AMSR, MODIS, GCOM-C/W (since 2002,daily)



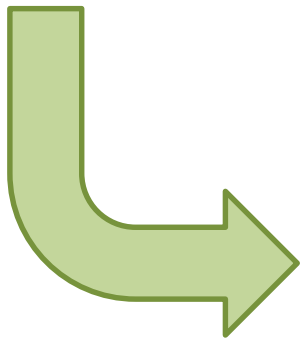
Paddy distribution



Daily inundation status



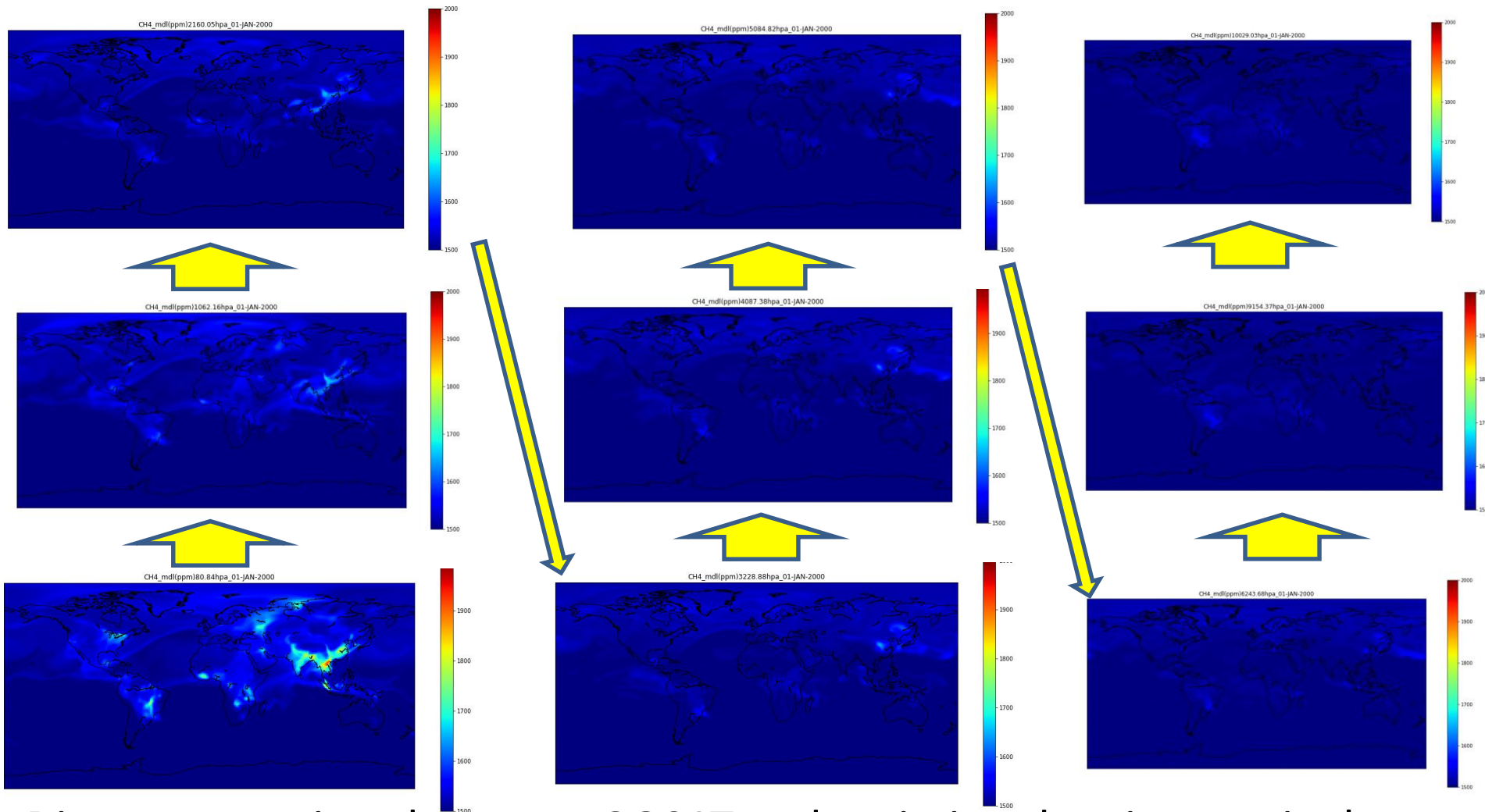
Daily cropping calendar



Estimate daily CH<sub>4</sub> emission REPORT (250m res., 2002-)

Need to be **VERIFIED !**

# NICAM-TM # with different altitudes (2000/Jan.)

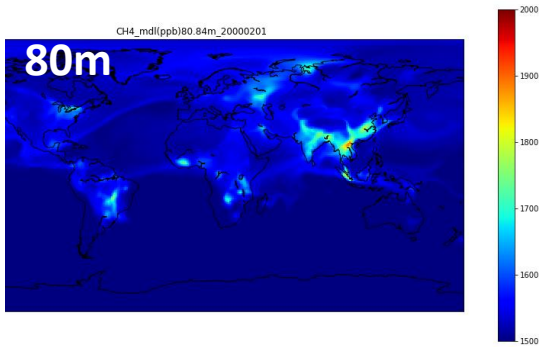


Direct comparison between GOSAT and emission data is meaningless...

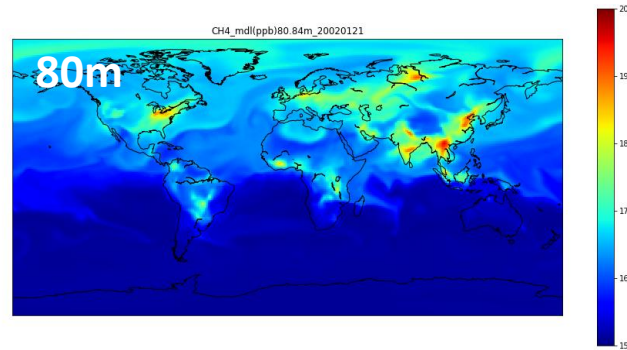
→Need transport model! But,,,,,

# Check spin-up status

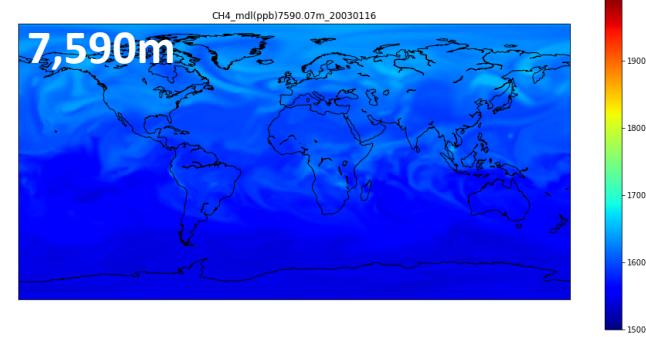
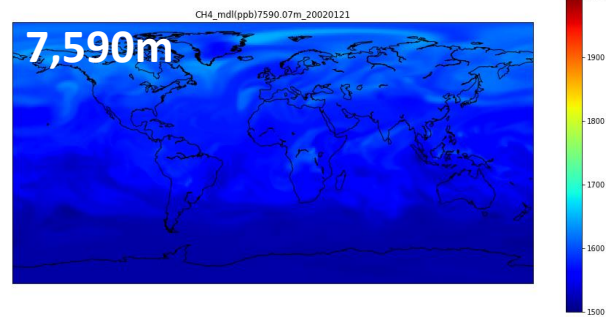
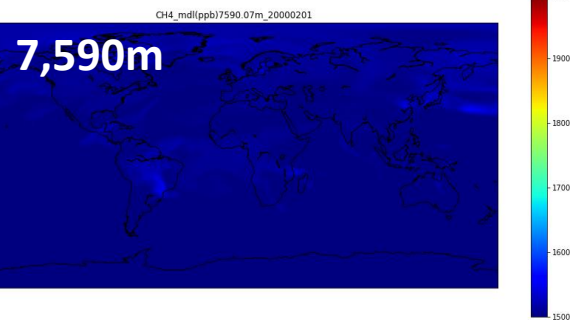
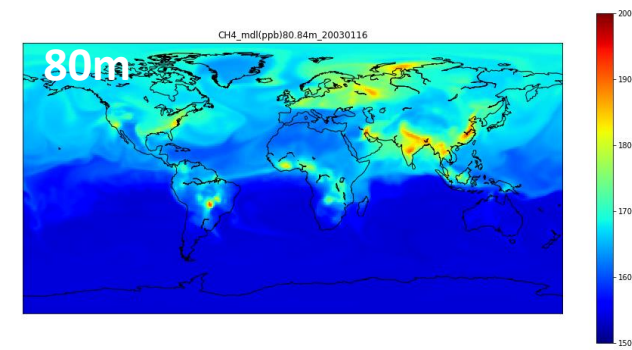
2000/2/1 (1 month after)



2002/1/21 (2 years after)



2003/1/16 (3 years after)



Long years are needed for spin up  
+ strong dependency on initial condition,,,  
→DA is essential!





Geosci. Model Dev., 10, 2201-2219, 2017  
<https://doi.org/10.5194/gmd-10-2201-2017>  
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Volume 10, issue 6



Article

**Assets**

Peer review

Metrics

Related articles

Development and technical paper

15 Jun 2017

## **A 4D-Var inversion system based on the icosahedral grid model (NICAM-TM 4D-Var v1.0) – Part 2: Optimization scheme and identical twin experiment of atmospheric CO<sub>2</sub> inversion**

Yosuke Niwa et al.

### **Model code and software**

#### **NICAM-TM 4D-Var**

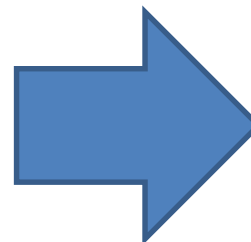
1. H. Tomita and M. Satoh; 2. M. Sato, T. Matsuno, H. Tomita, H. Miura, T. Nasuno and S. Iga; 3. M. Satoh, H. Tomita, H. Yashiro, H. Miura, C. Kodama, T. Seiki, A. T. Noda, Y. Yamada, D. Goto, M. Sawada, T. Miyoshi, Y. Niwa, M. Hara, T. Ohno, S. Iga, T. Arakawa, T. Inoue, and H. Kubokawa

<http://nicam.jp/hiki/?Research+Collaborations>

**GOSAT + NICAM-TM**



**GOSAT-2 + NICAM-TM-4DVAR**



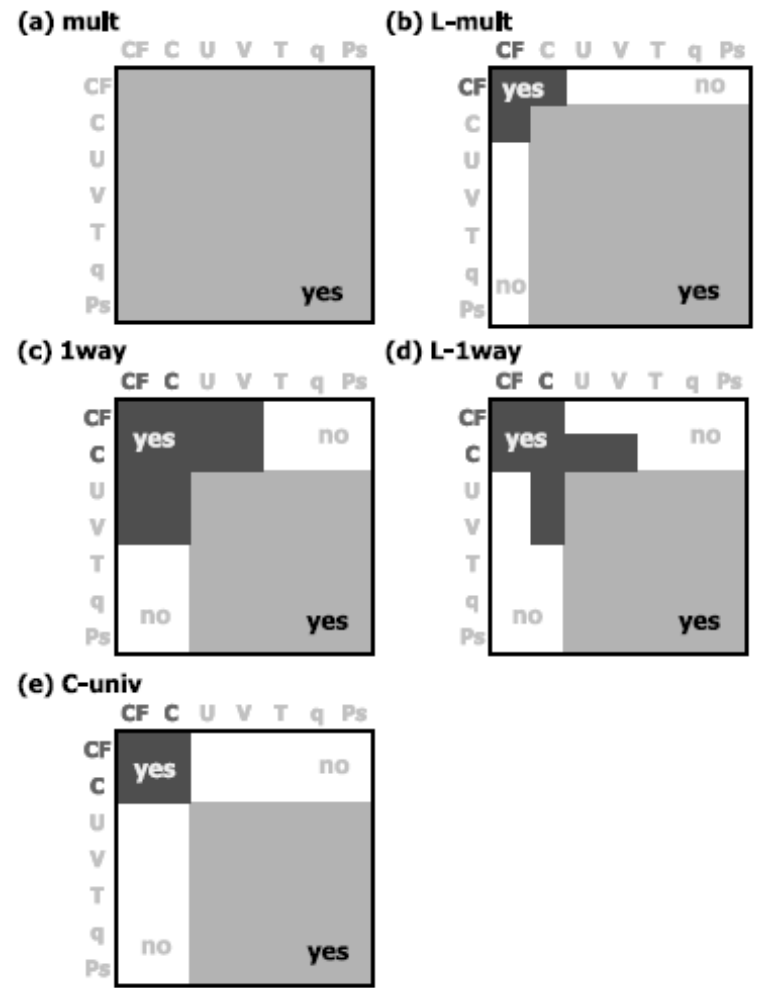
# “Variable localization” in an ensemble Kalman filter: Application to the carbon cycle data assimilation

Ji-Sun Kang,<sup>1</sup> Eugenia Kalnay,<sup>1</sup> Junjie Liu,<sup>2</sup> Inez Fung,<sup>2</sup> Takemasa Miyoshi,<sup>1</sup>  
and Kayo Ide<sup>1</sup>

Flux estimation from  
atmospheric concentration  
by omitting multi-collinearity

- No direct emission or apriori info. is required!

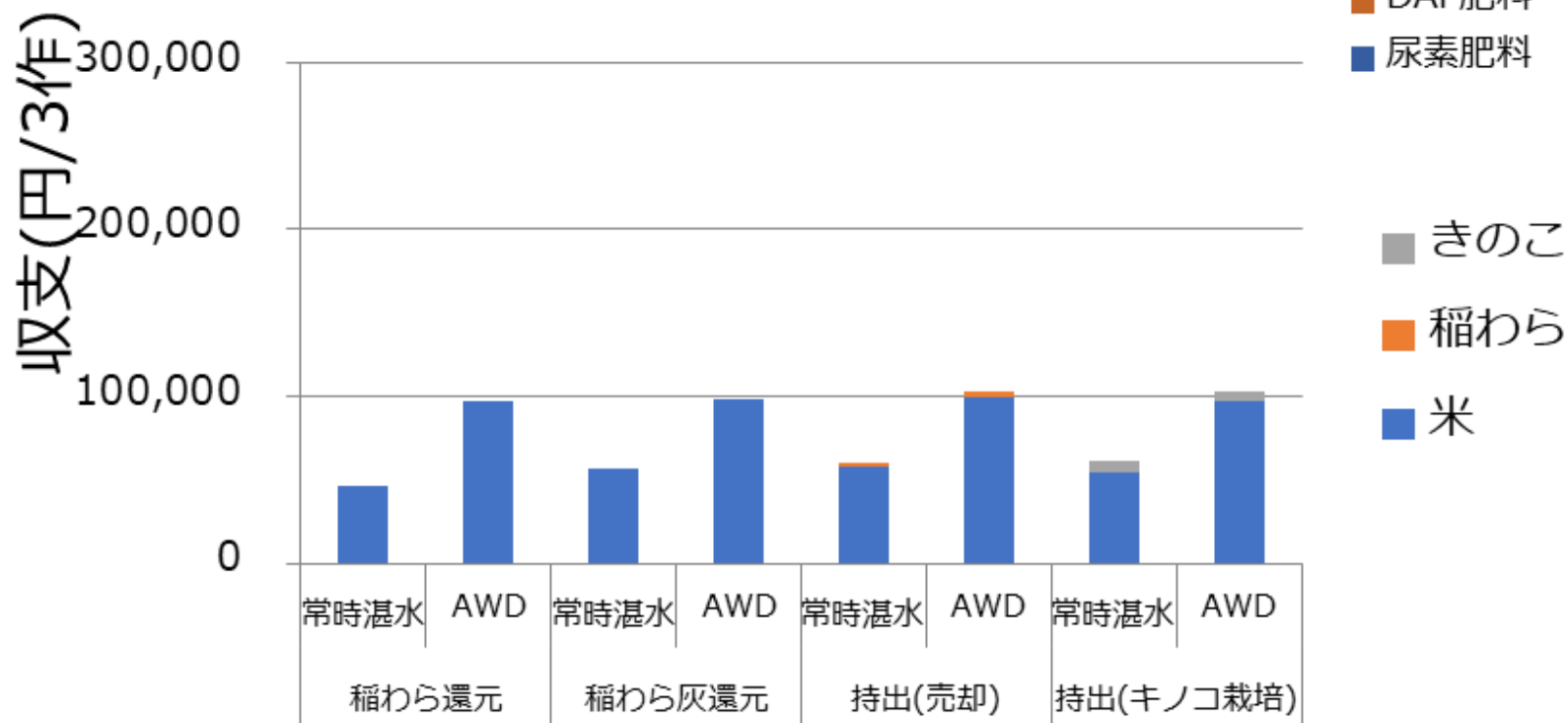
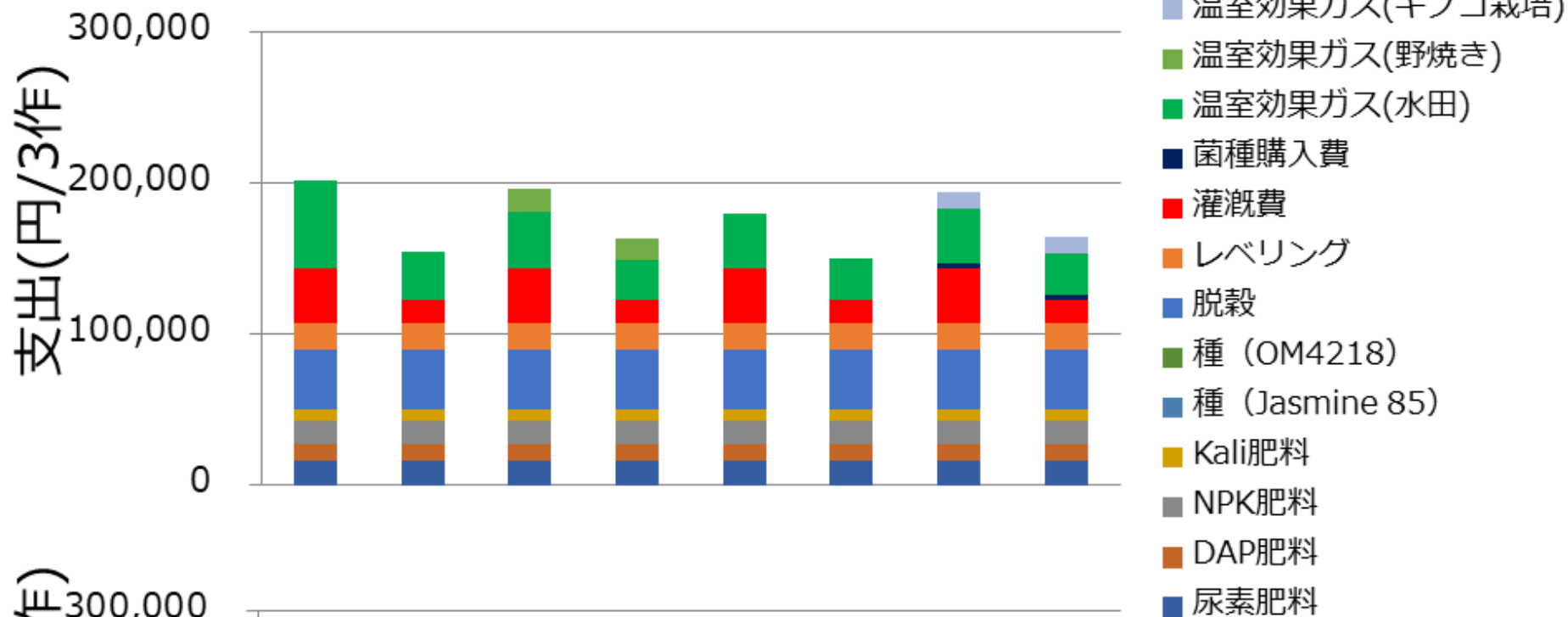
Transparent MRV  
with NICAM-LETKF!



Back ground covariance matrices

# Outline

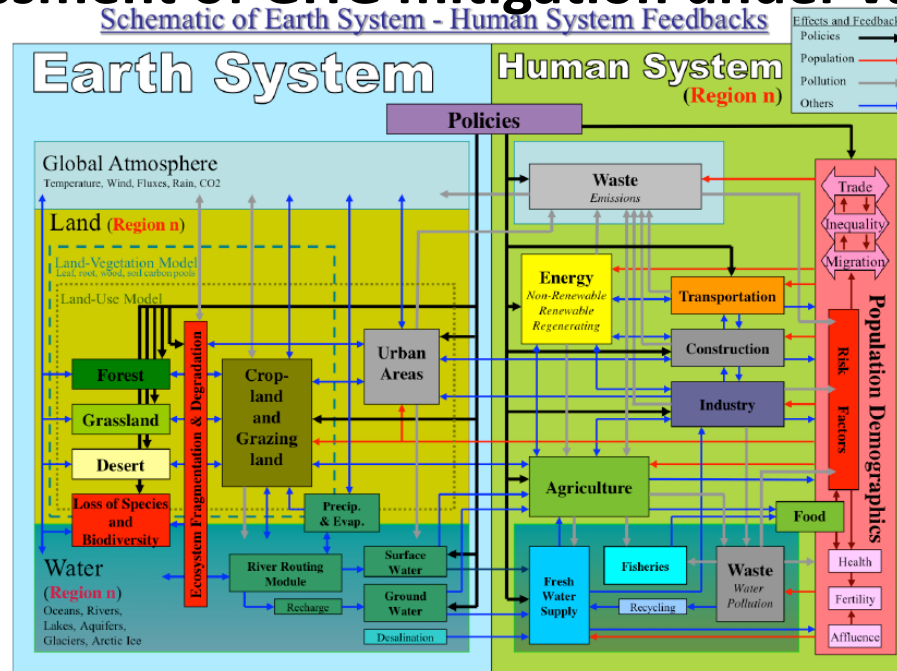
0. Motivation to DA (Story taking me here today)
1. Background & Objective
2. Ground observation of greenhouse gas emission and statistical modeling
3. Satellite remote sensing of GHG emitters
  - Cropping calendar & the adjacent fallow length
  - Paddy soil/water covered by rice plants
  - Top down verification with GOSAT
- 4. My next work with DA**





# My next work with DA

## Economic assessment of GHG mitigation under various uncertainties



Kalnay *et al.* 2017

And if possible...

Soil moisture/Drought assessment/GHG emission estimation with AHI-LST and its DA with atmospheric observation data

PM2.5 emission status estimation with AHI & NICAM-LETKF

Thank you  
for your attention

