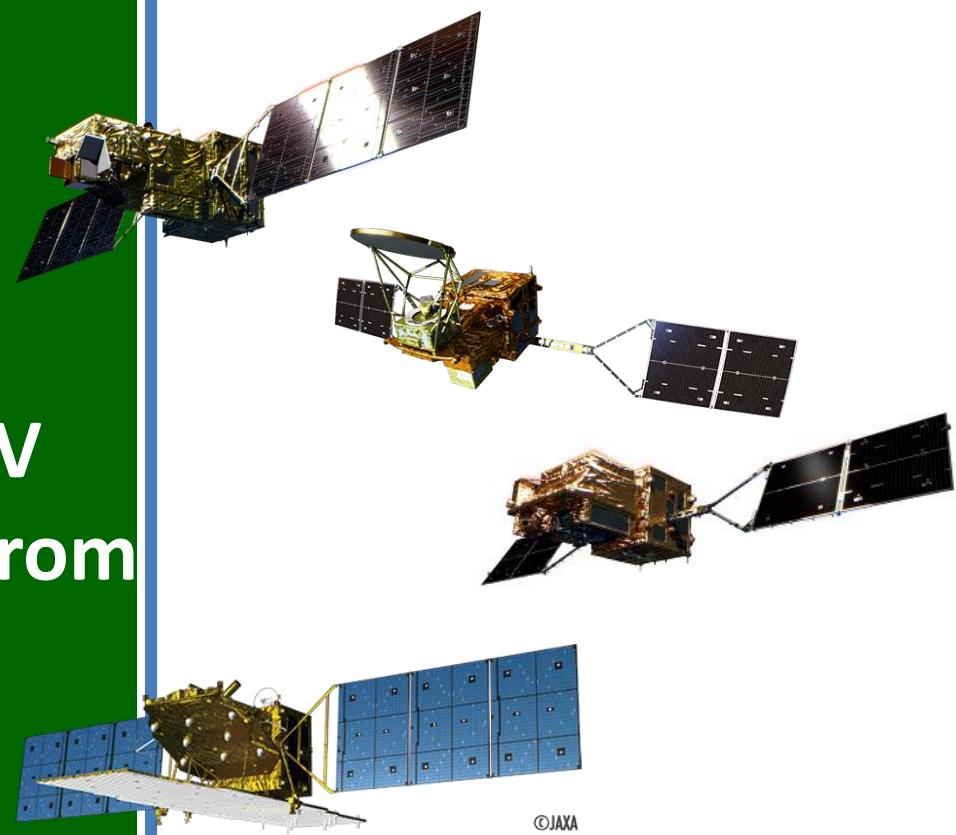




RESPONSIBLE CONSUMPTION AND PRODUCTION



CLIMATE ACTION



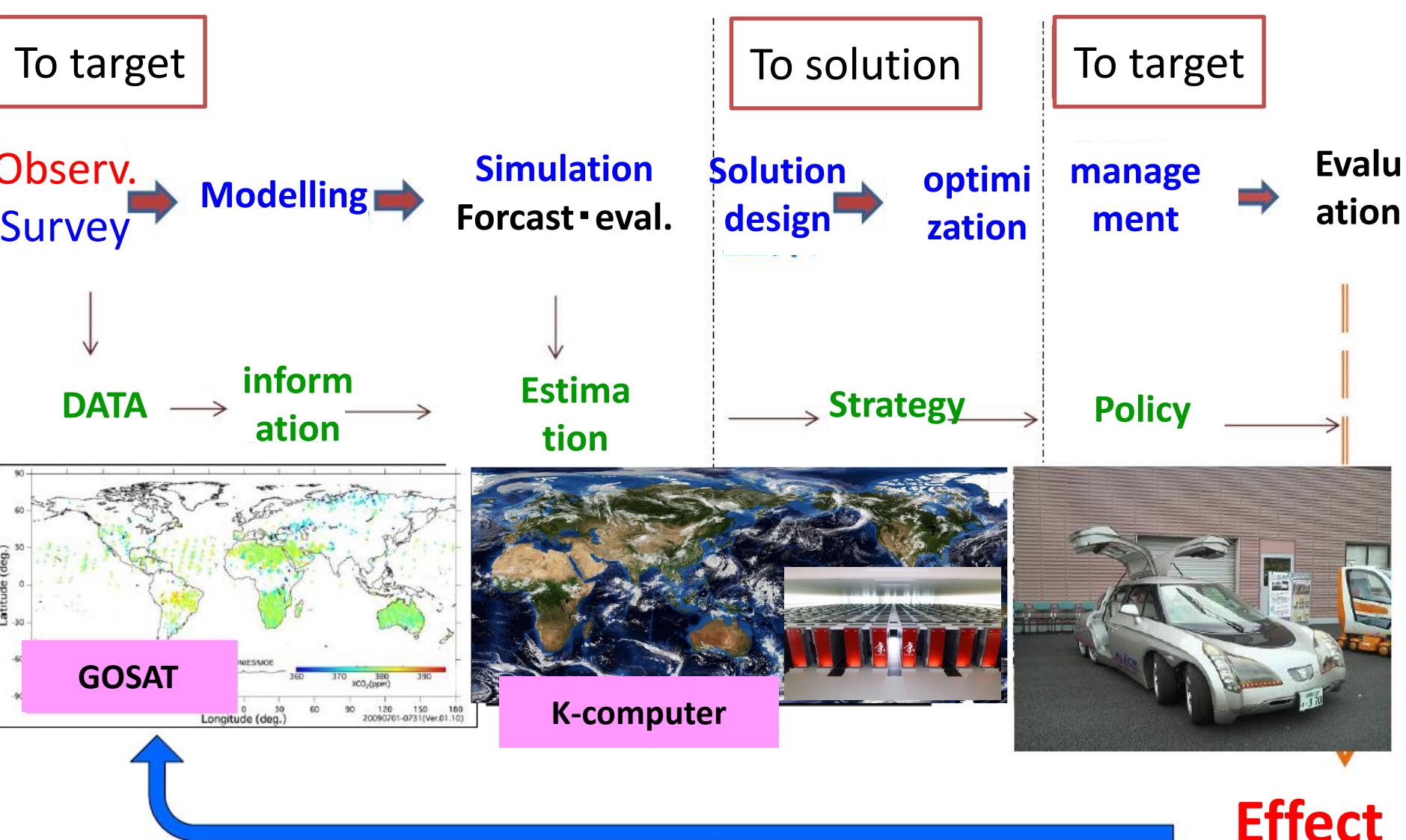
©JAXA

Satellite data based MRV system of GHGs emission from tropical rice paddies

Hironori Arai^{1,3)}, Wataru Takeuchi¹⁾,
Kei Oyoshi²⁾, Lam Dao Nguyen⁴⁾,
Towa Tachibana⁵⁾, Ryuta Uozumi,
Koji Terasaki³⁾, Takemasa Miyoshi³⁾,
Hisashi Yashiro³⁾, Kazuyuki Inubushi⁵⁾



Cycle from Observation to Countermeasure



Observation of the effect

Modified from Yasuoka 2015

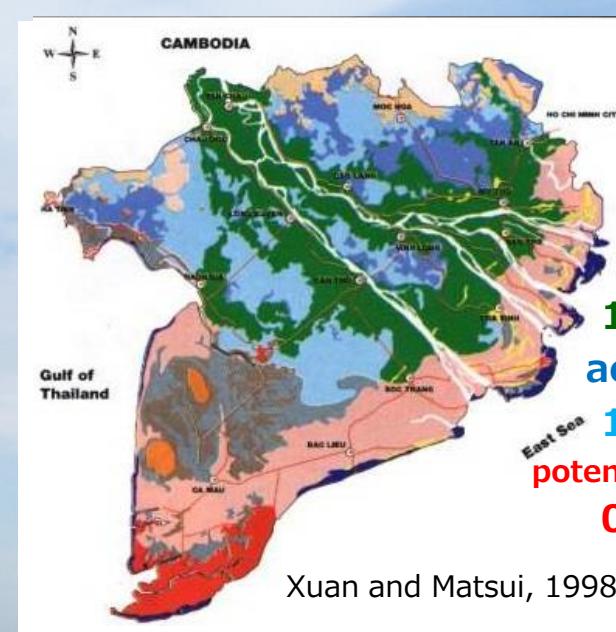
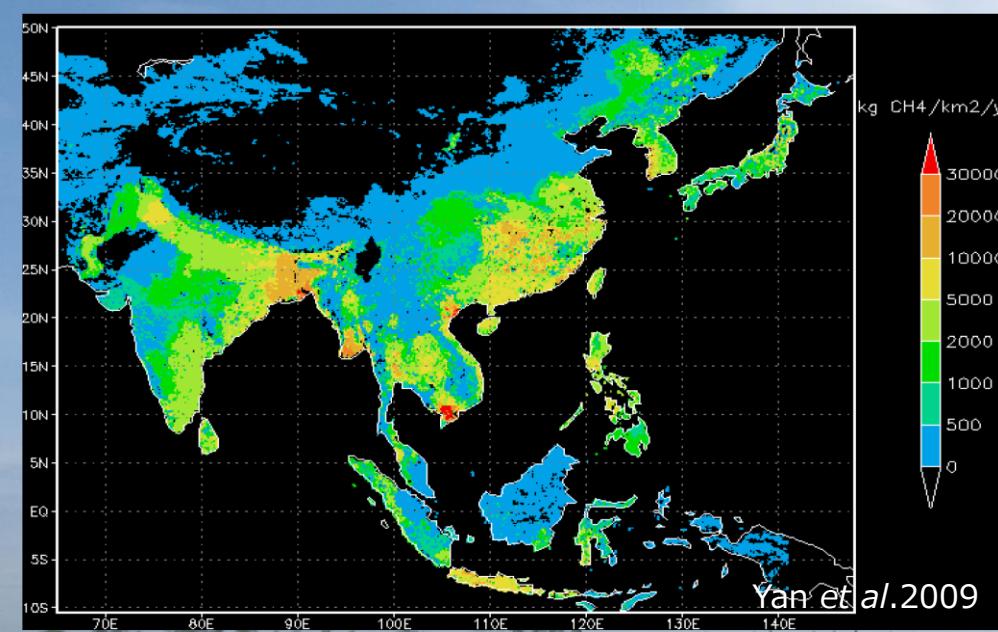
Outline

1. Background & Objective

2. Ground observation of greenhouse gas emission and semi-empirical 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



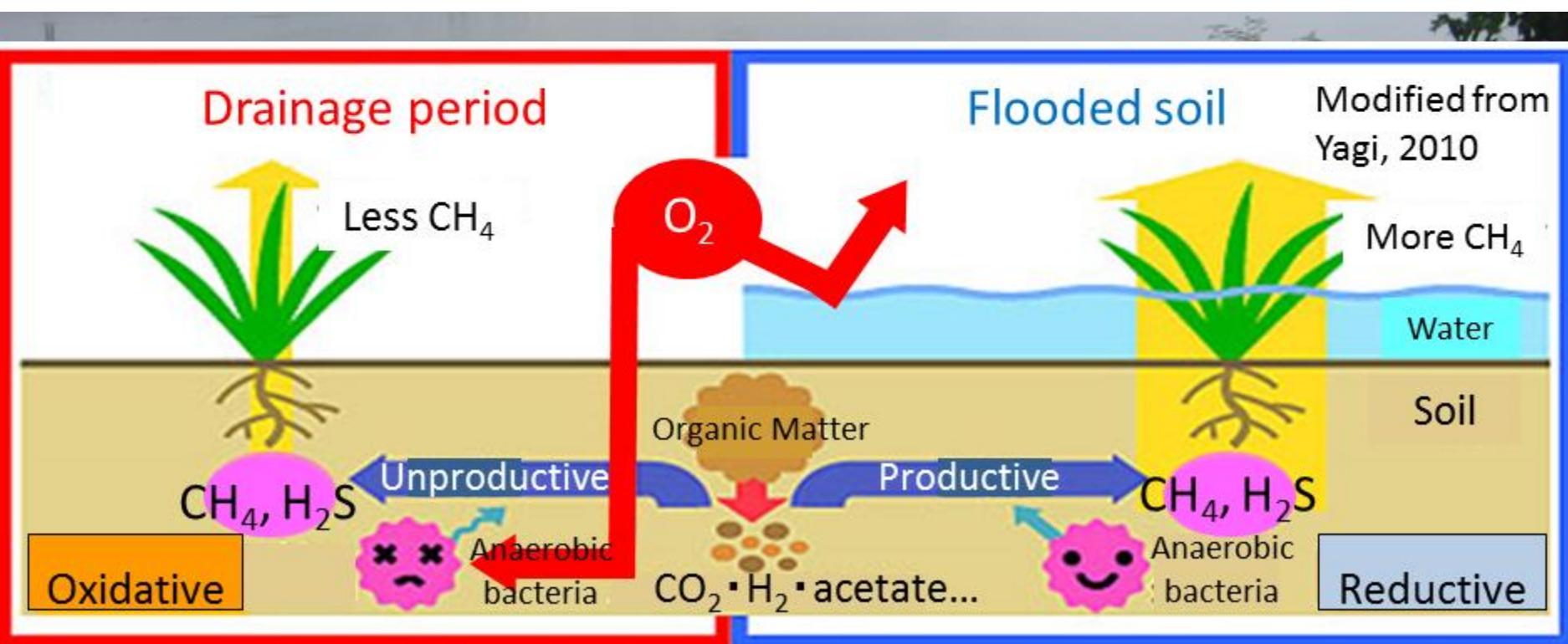
- Continuously flooded nearly through a year +
- High straw production

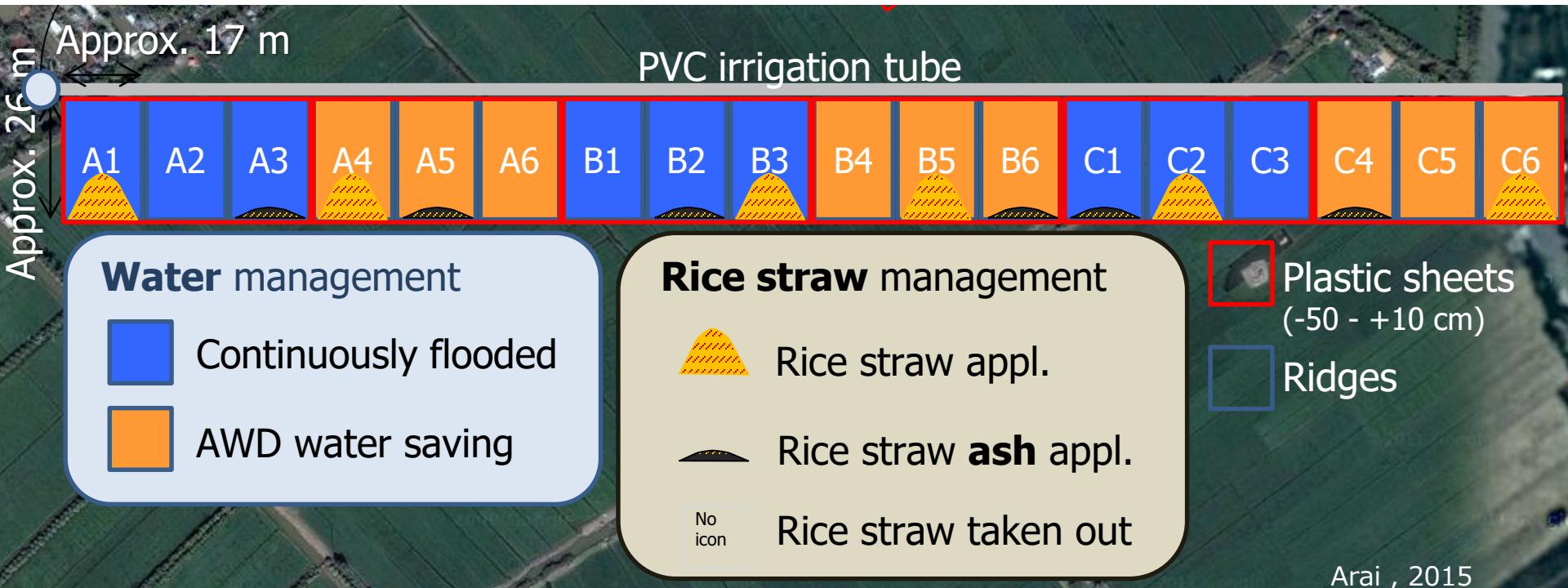
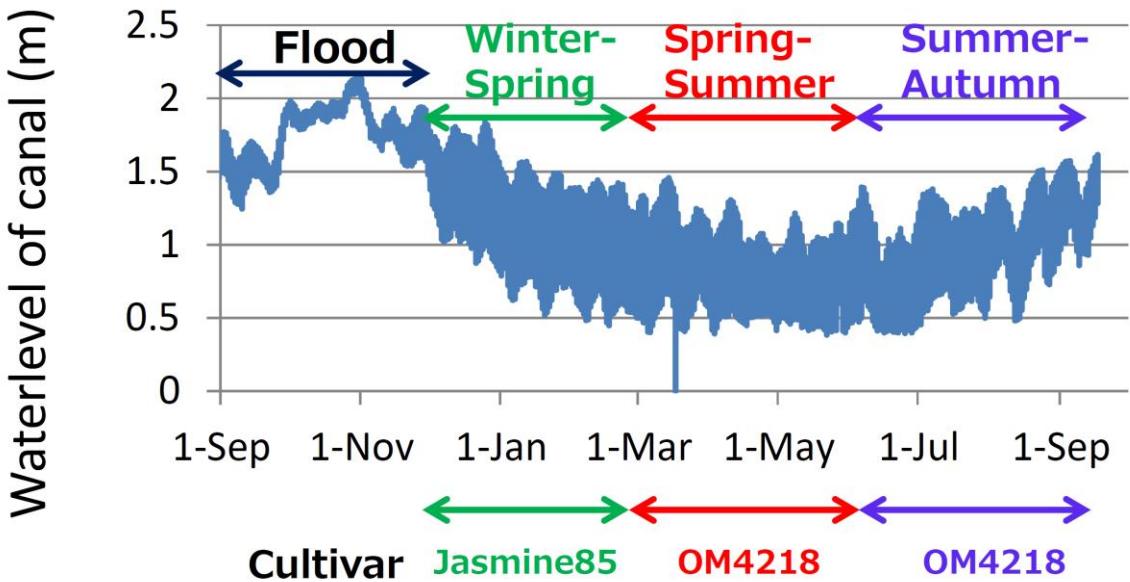


- Anaerobic stress for rice production
- High GHGs emission

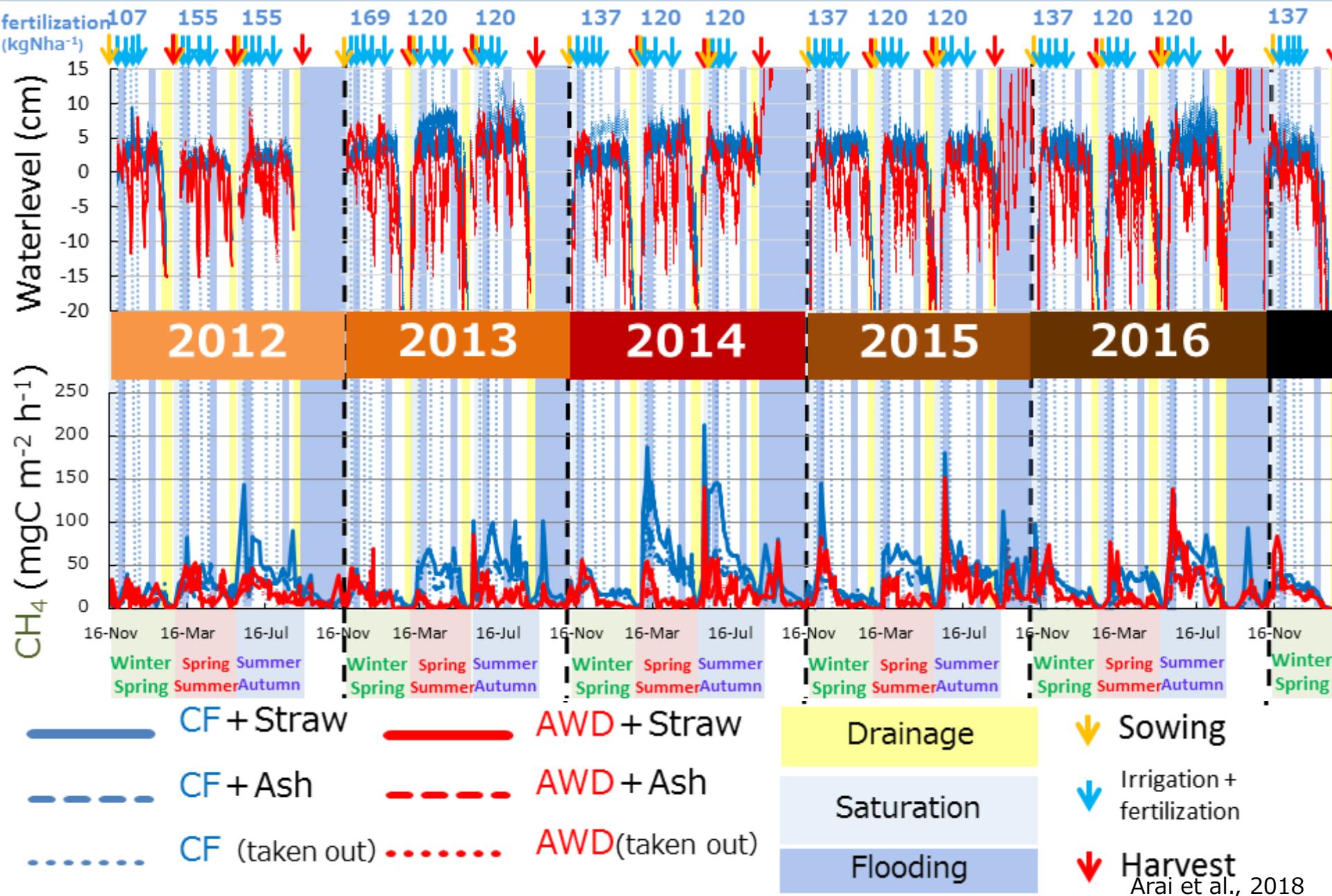
(Alternate Wetting and Drying)

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

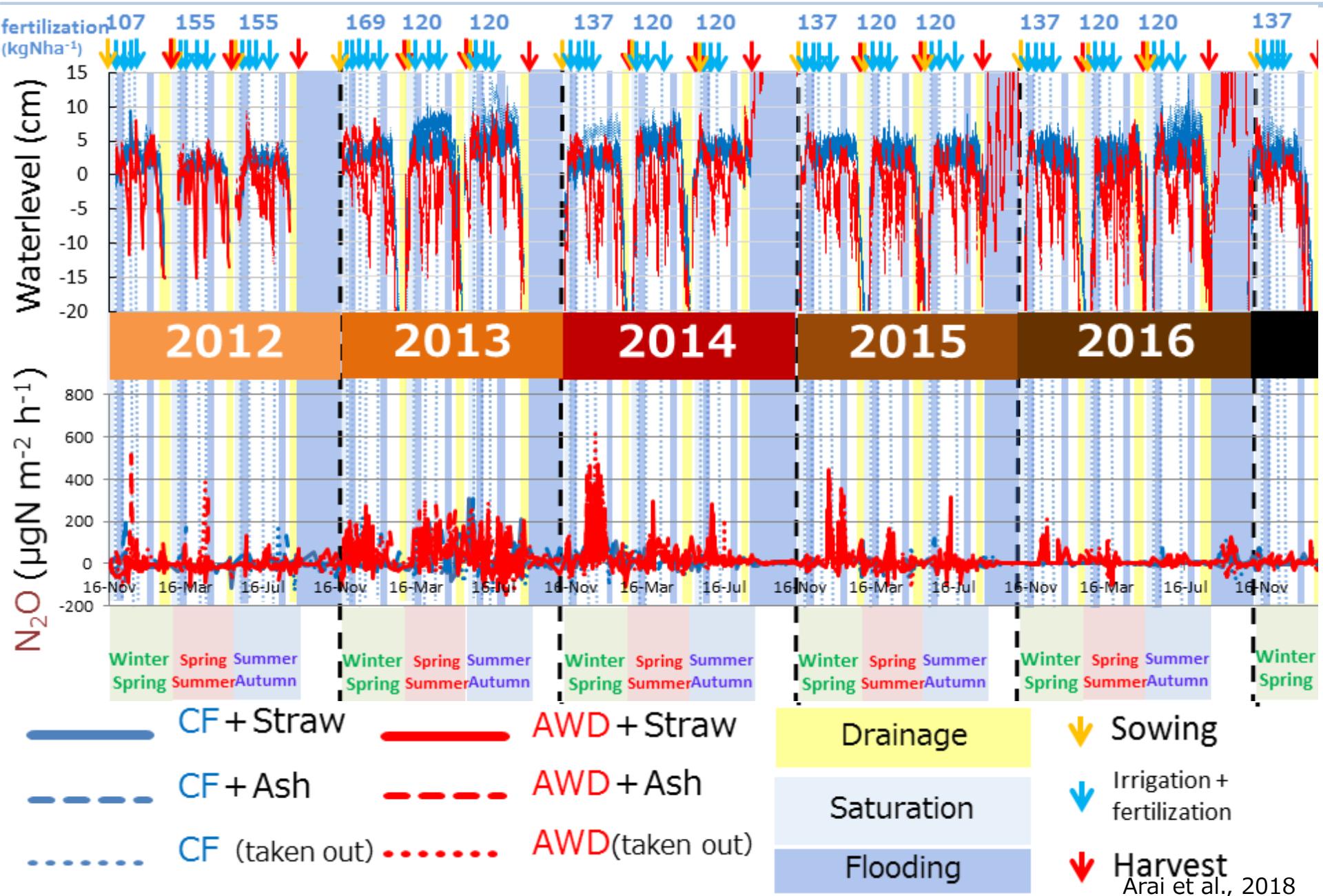




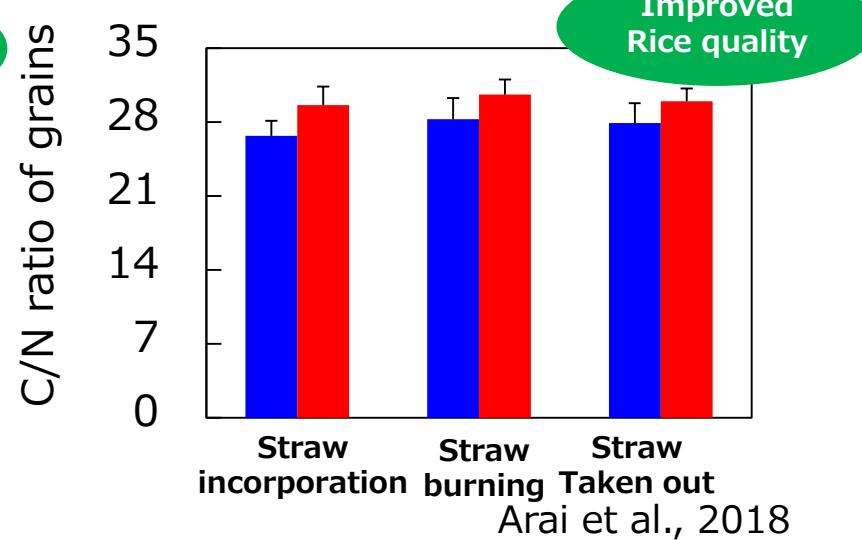
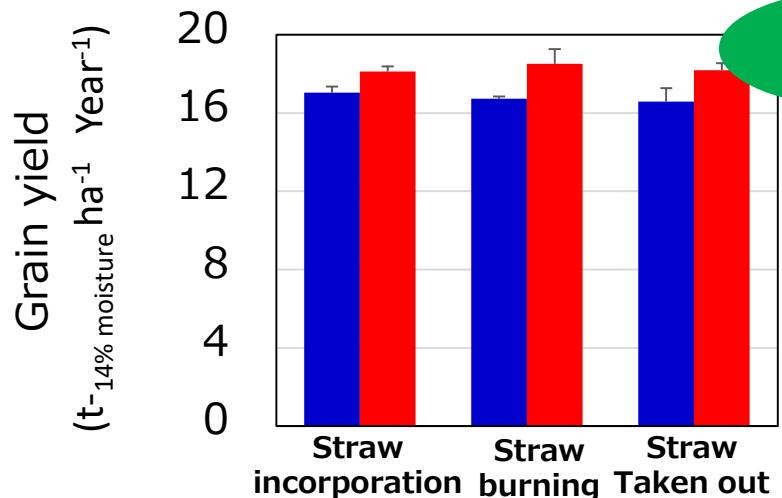
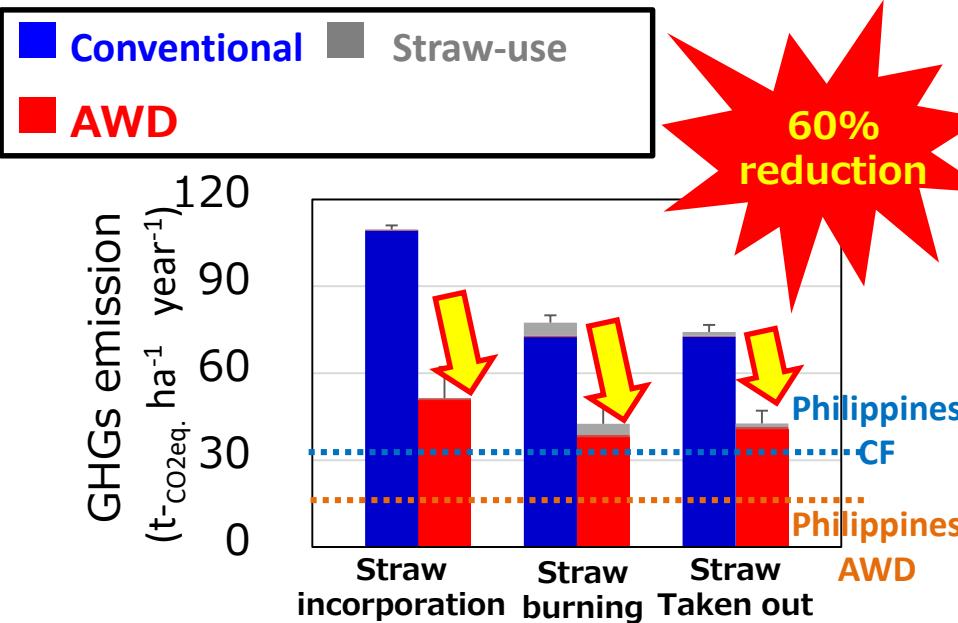
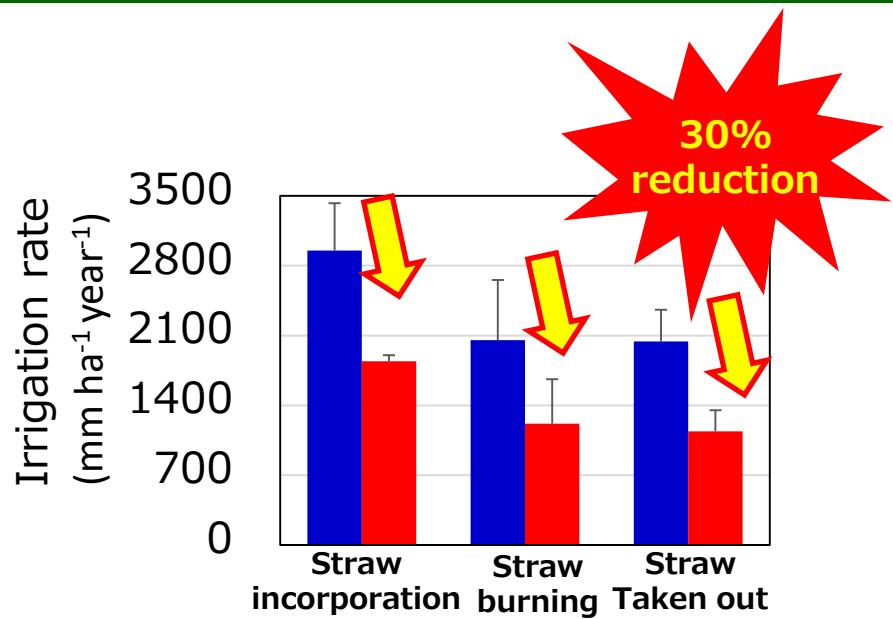
Characteristics of the Mekong delta



Characteristics of the Mekong delta

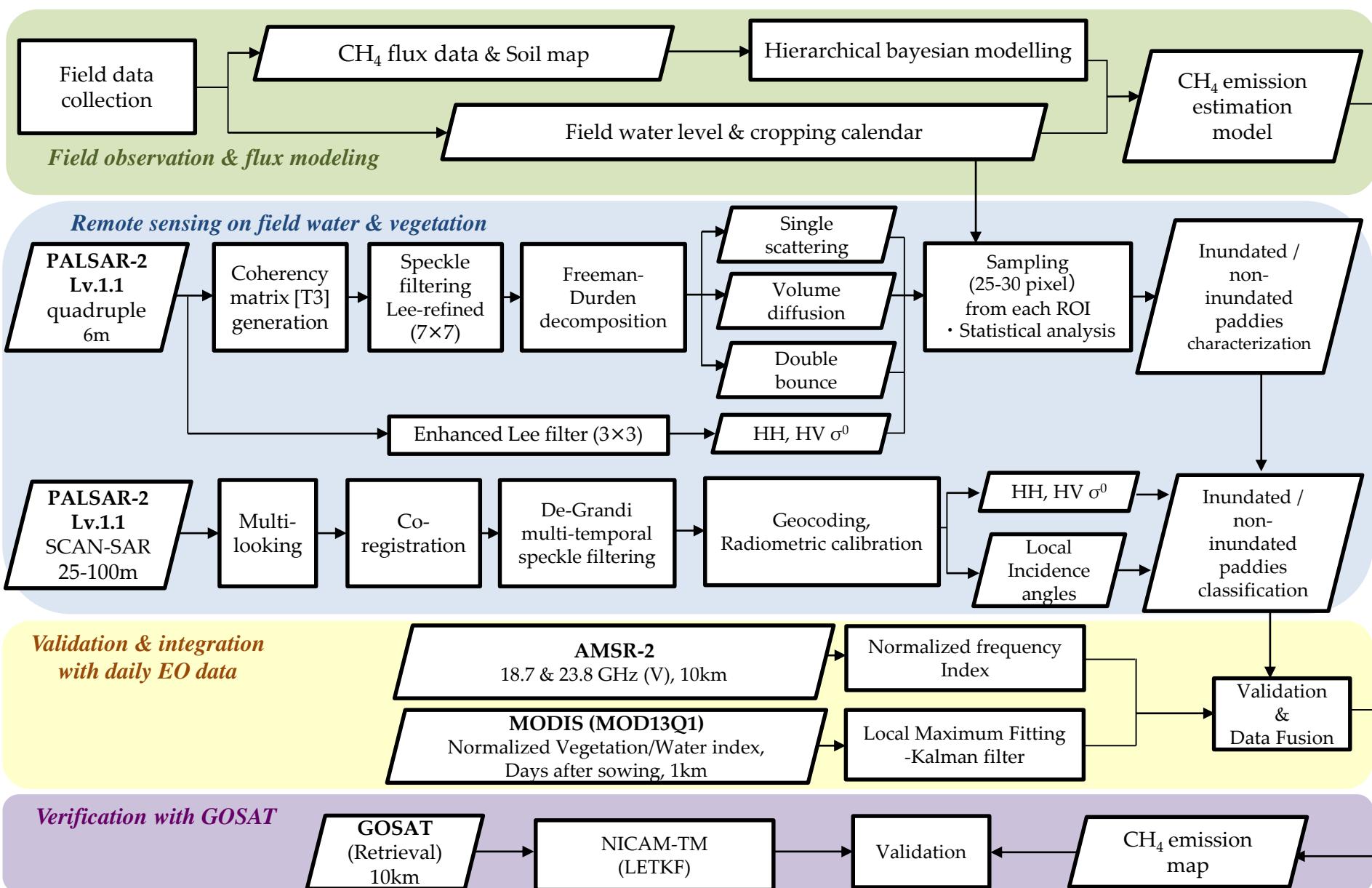


- Reduction of irrigation rate & GHGs (2012-2016)
- Increase of rice grains and its quality



Arai et al., 2018

Flow chart

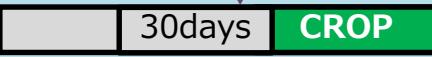
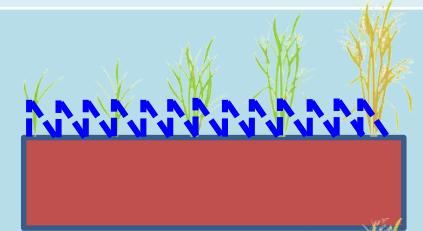
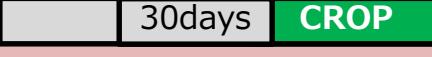
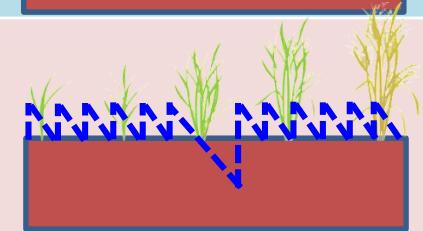
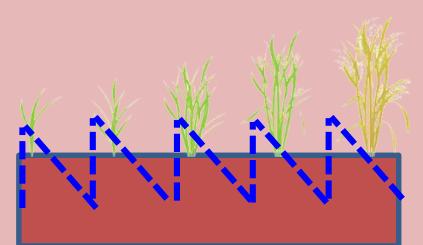


Outline

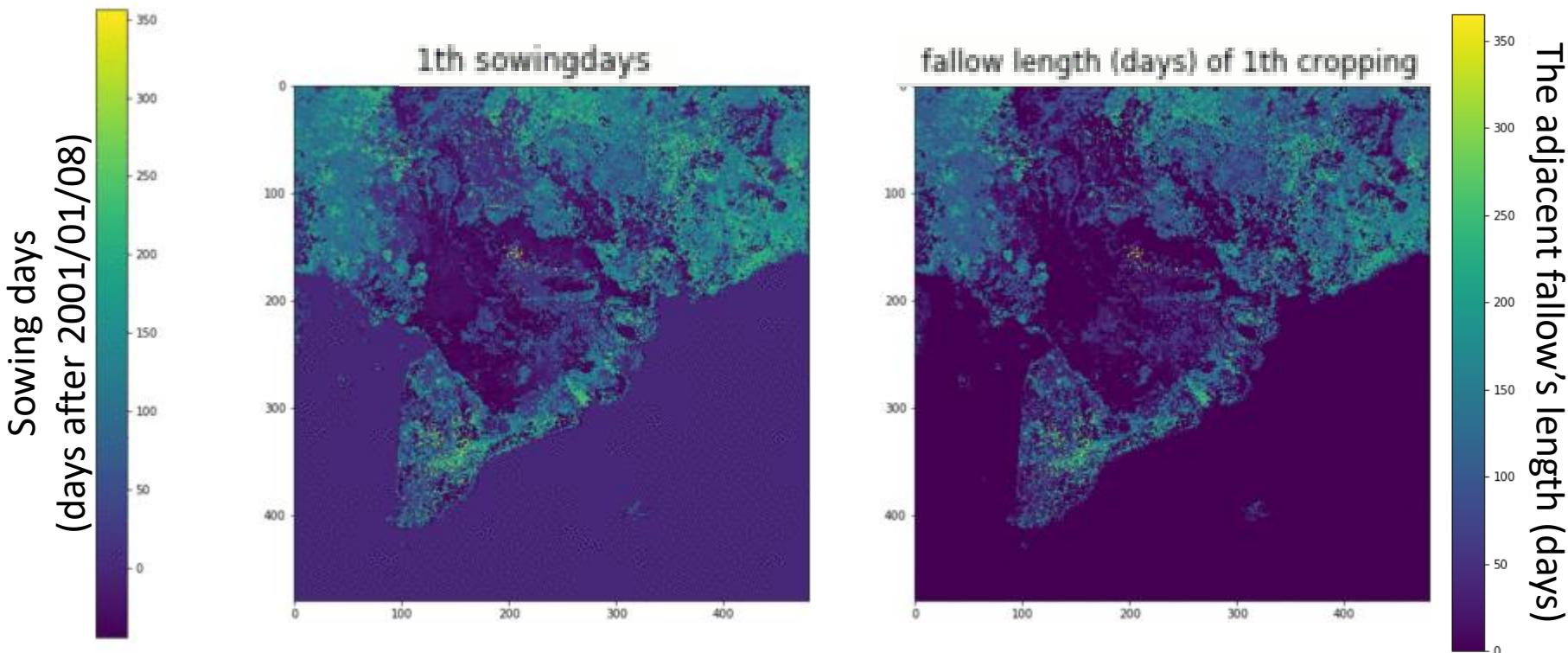
- 1. Background & Objective**
- 2. Ground observation of greenhouse gas emission and semi-empirical modeling**
- 3. Satellite remote sensing of GHG emitters**
 - Cropping calendar & the adjacent fallow length
 - Paddy soil/water covered by rice plants
 - Top down verification with GOSAT

IPCC guideline (Tier1)

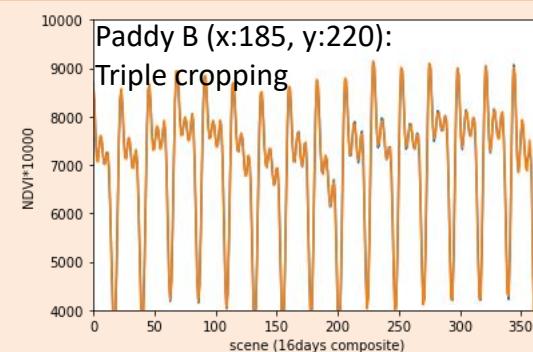
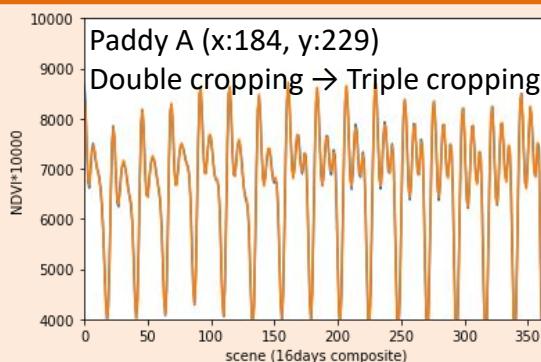
[Emission factor × Scaling factor in IPCC guideline]

Straw incorporation time and amount	Water regime prior to rice cultivation	Water regime during rice cultivation
A.   	<p>①</p>  <p>>30 days flood CROP</p>	
B.  	<p>②</p>  <p><180 days Non-flood CROP</p> <p>③</p>  <p>>180 days Non-flood CROP</p>	 

Cropping calendar evaluation with MODIS-NDVI (LMF-KF) for GCOM-C



Samples of paddies



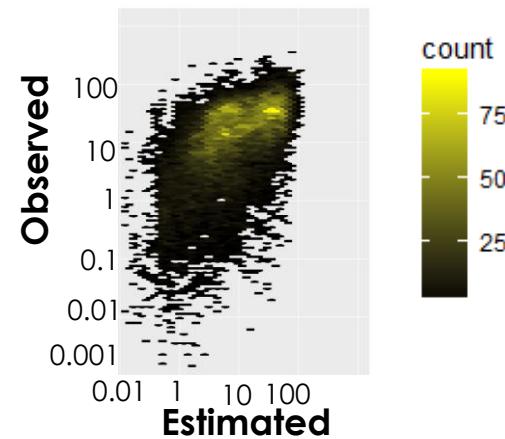
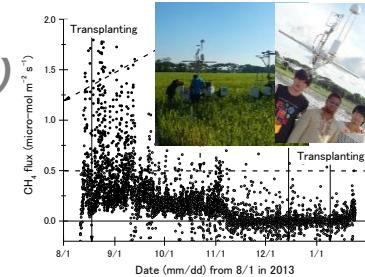
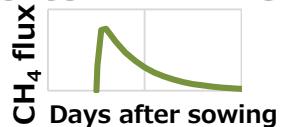
Semi-empirical daily CH_4 flux ($\text{mg C m}^{-2} \text{ hr}^{-1}$) Model

CH_4 emission on a specific date

$$= y * \text{carbon_management} / \text{non-inundated_fallow} / \text{inundated_fallow} * \text{water_management} * \alpha * \beta$$

carbon_management (Michaelis-Menten KINETICS)

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



non-inundated_fallow (OXYDATION CAPACITY)

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



inundated_fallow

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

water_management

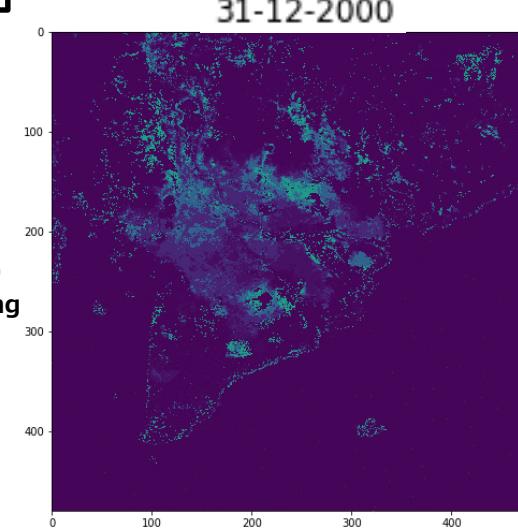
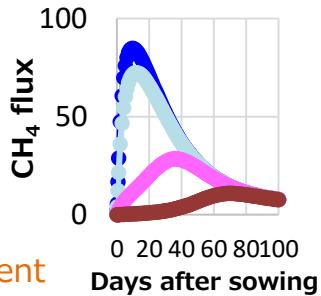
$$= \exp(\eta * \text{inundated days during the last 10days})$$

DAS \leftarrow days after sowing

α \leftarrow straw incorporation coefficient

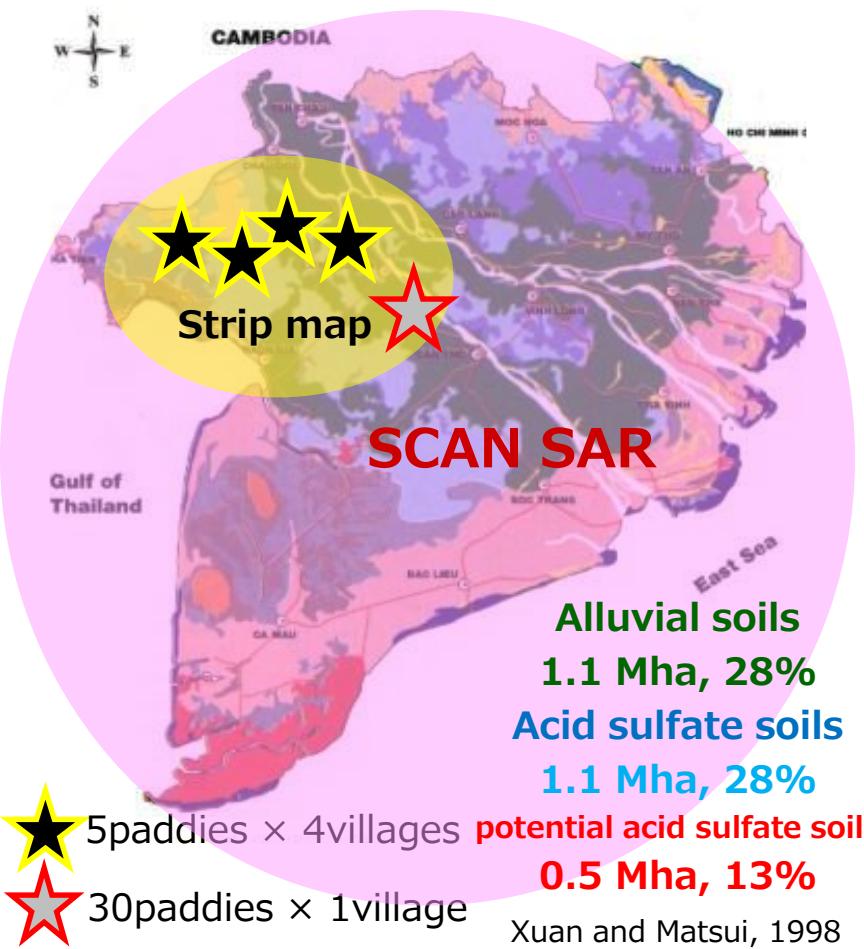
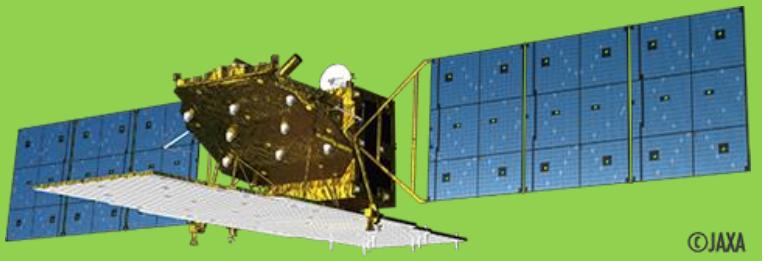
β \leftarrow acid sulfate \cdot coastal sandy soil coefficient

$\gamma, \eta, \delta, \varepsilon, \omega, \zeta, l, \kappa \leftarrow$ constant (> 0)



ALOS-2/PALSAR-2

- Lband-Synthetic Aperture Radar -



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

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

Multilooking

Co-registration

De Grandi
multi-temporal
filtering

Geocoding
&
Radiometric
calibration

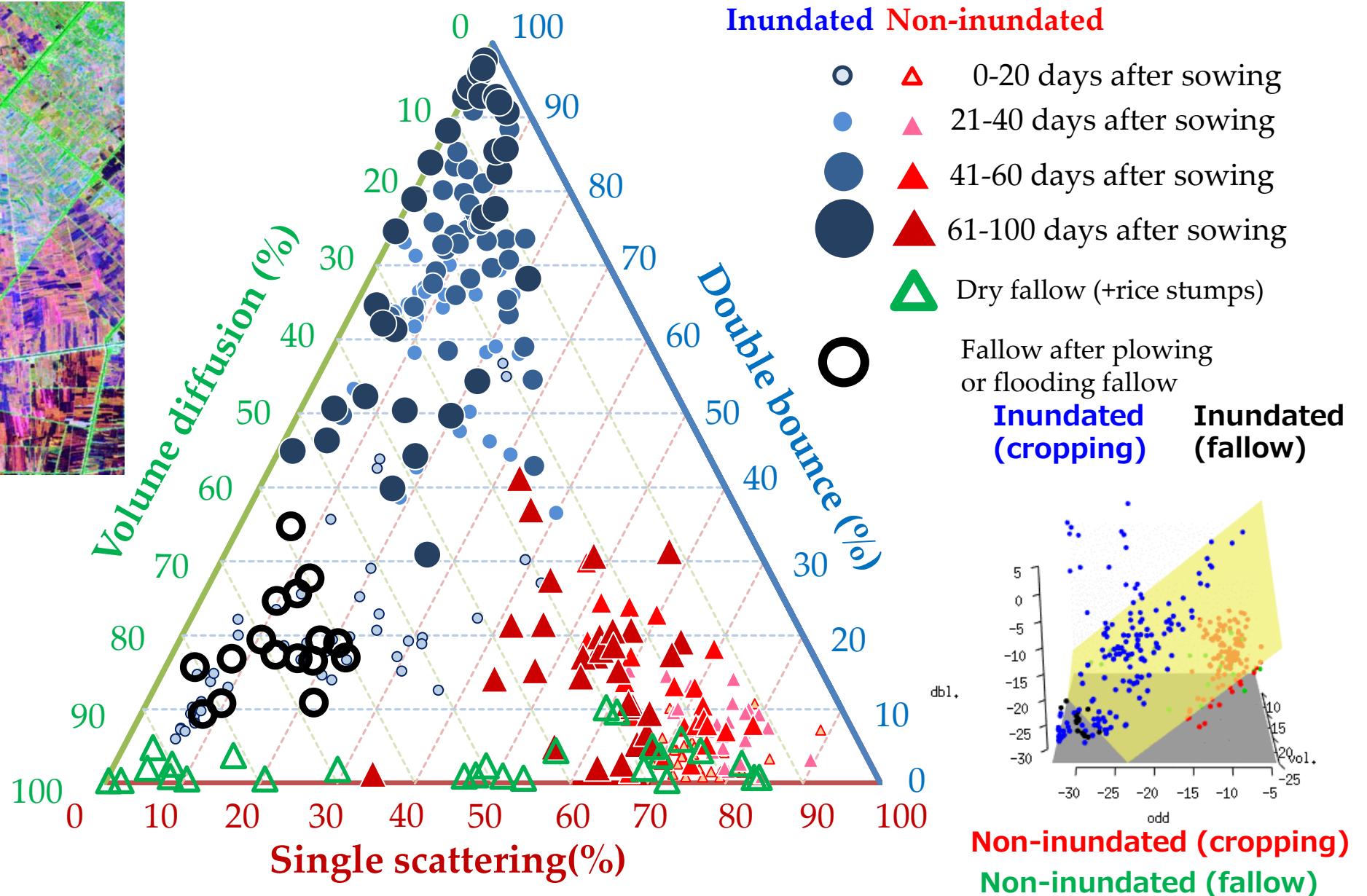
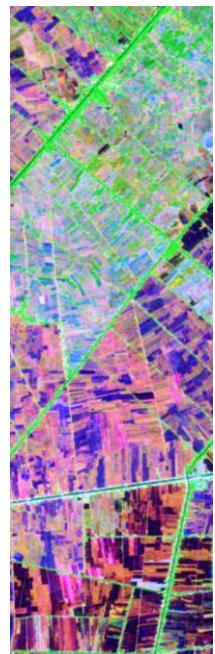
HH **HV** **Incidence angle**

Rice paddy masking
&
Statistical analysis

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

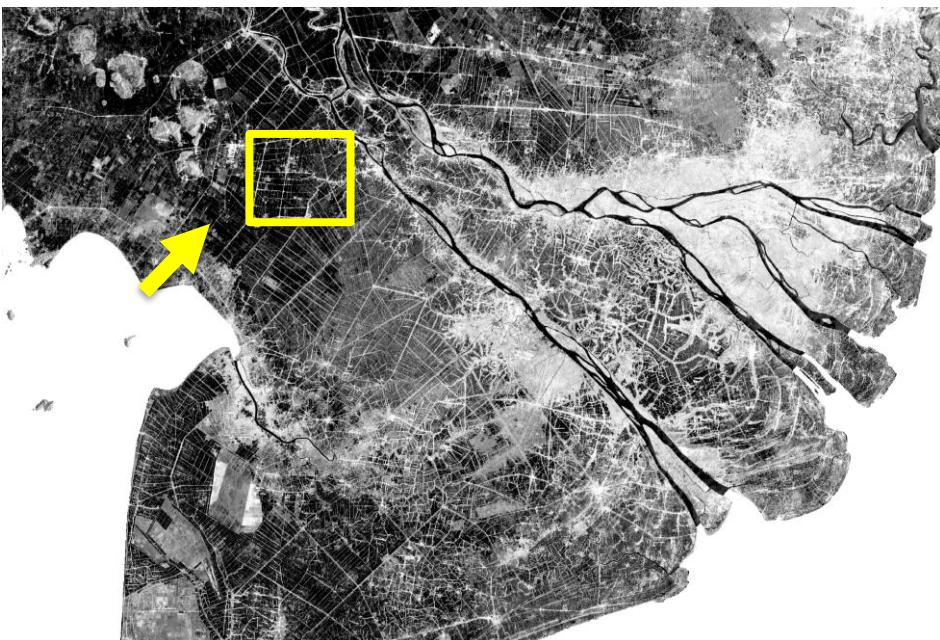
Modified from Avtar et al. 2012

-Freeman-Durden decomposition-

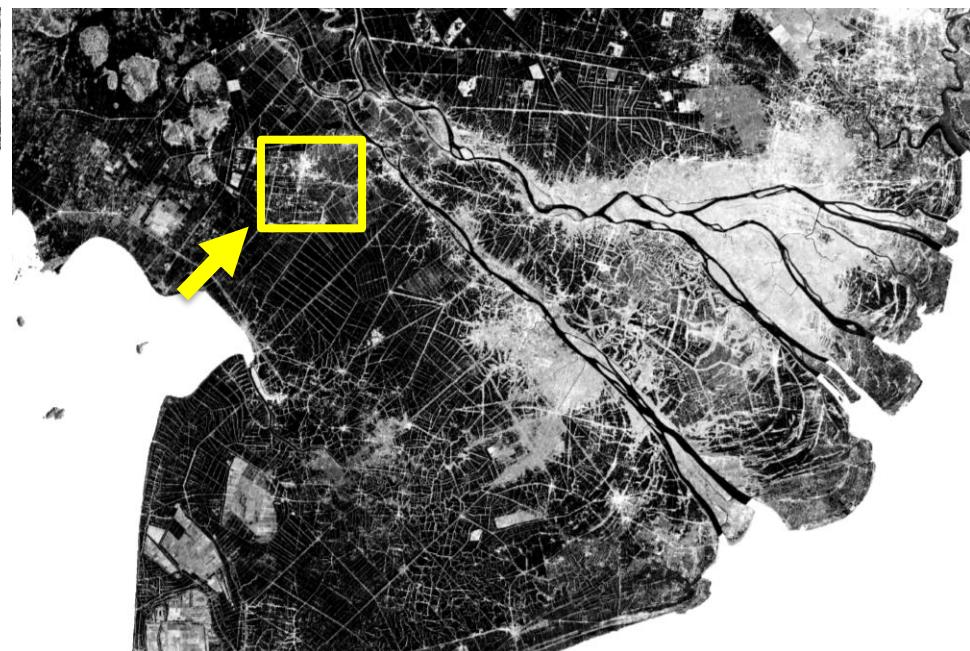


SCANSAR (intensity - HH σ^0)

Dry season (2015 Apr. 10)



Flooding season (2015 Oct. 23)

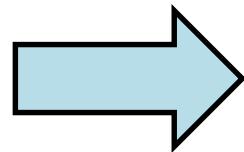


Double bounce detection by SCANSAR (intensity - HH σ^0)

Dry season (2015 Apr. 10)



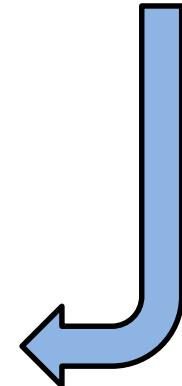
Rainy season (2015 Jul. 03)



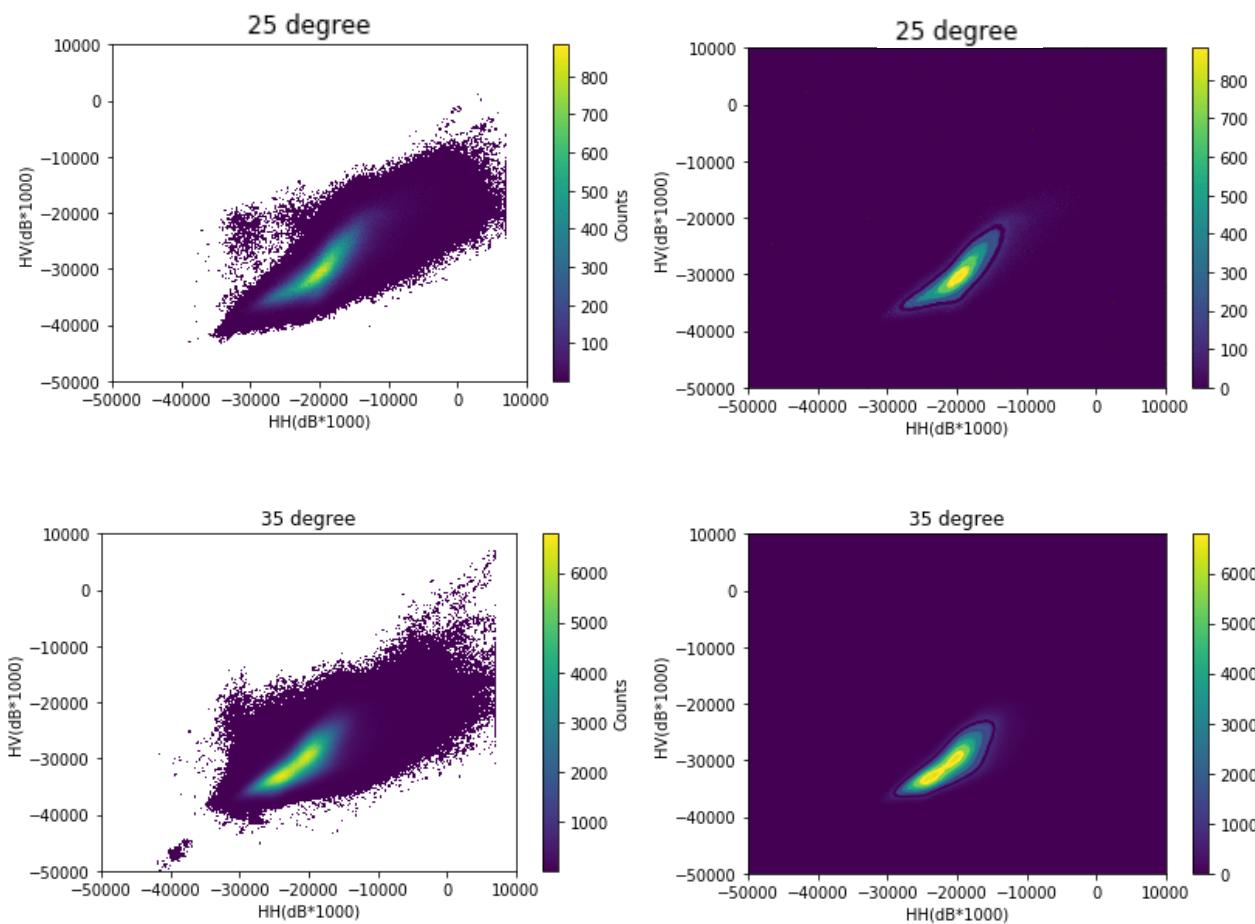
Flooding season (2015 Oct. 30) -LANDSAT-8-



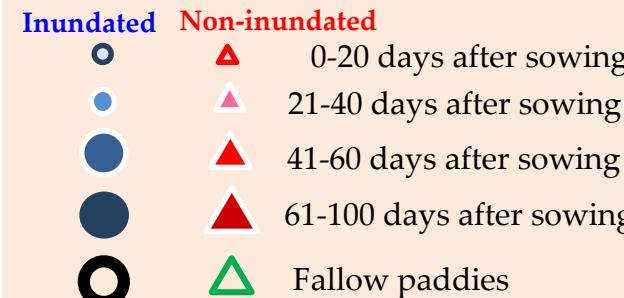
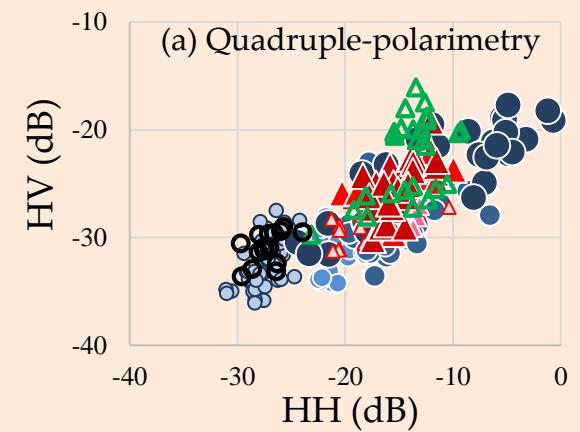
Flooding season (2015 Oct. 23)



SCAN-SAR (25m)

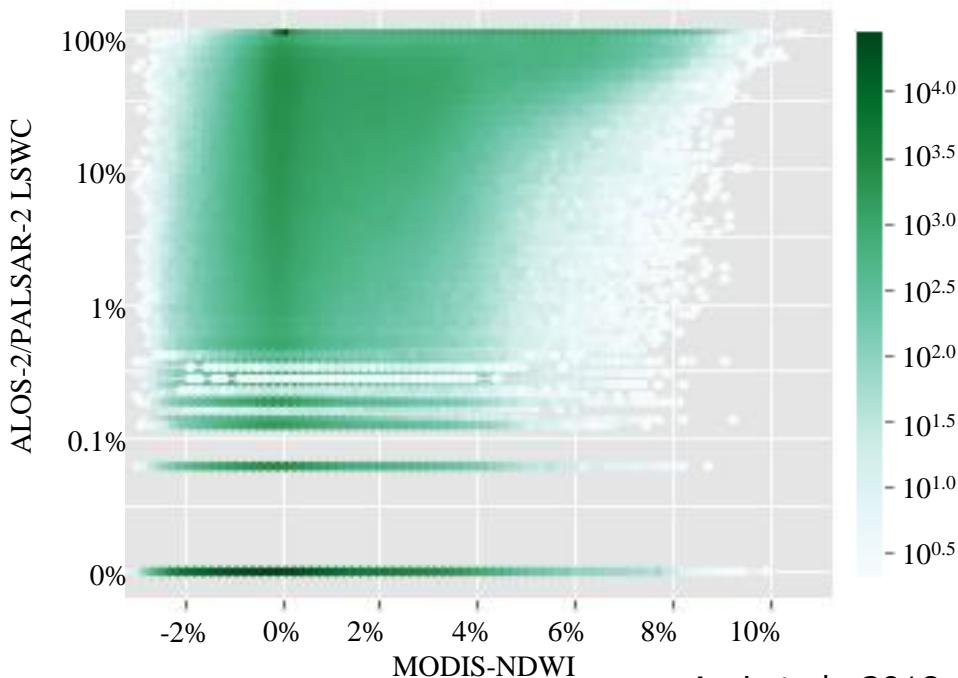
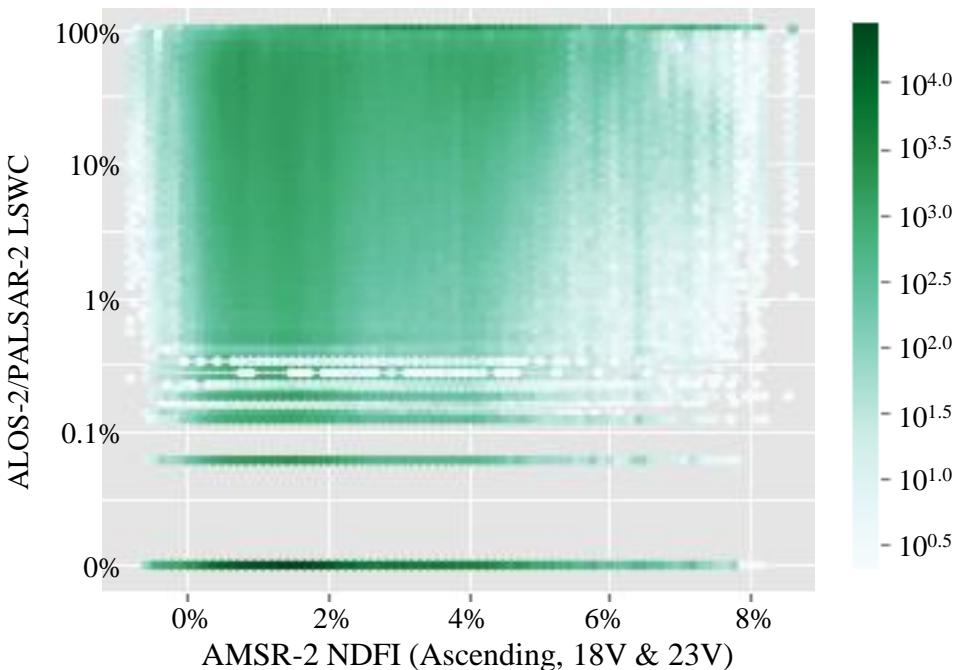
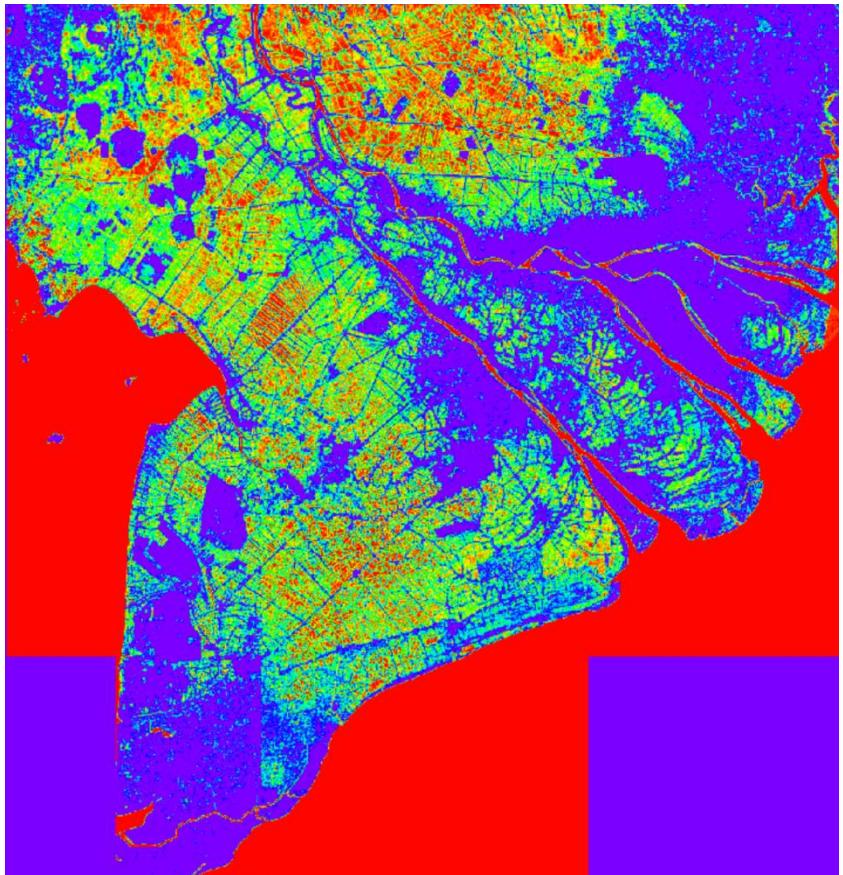


Full-polarimetry (3m)

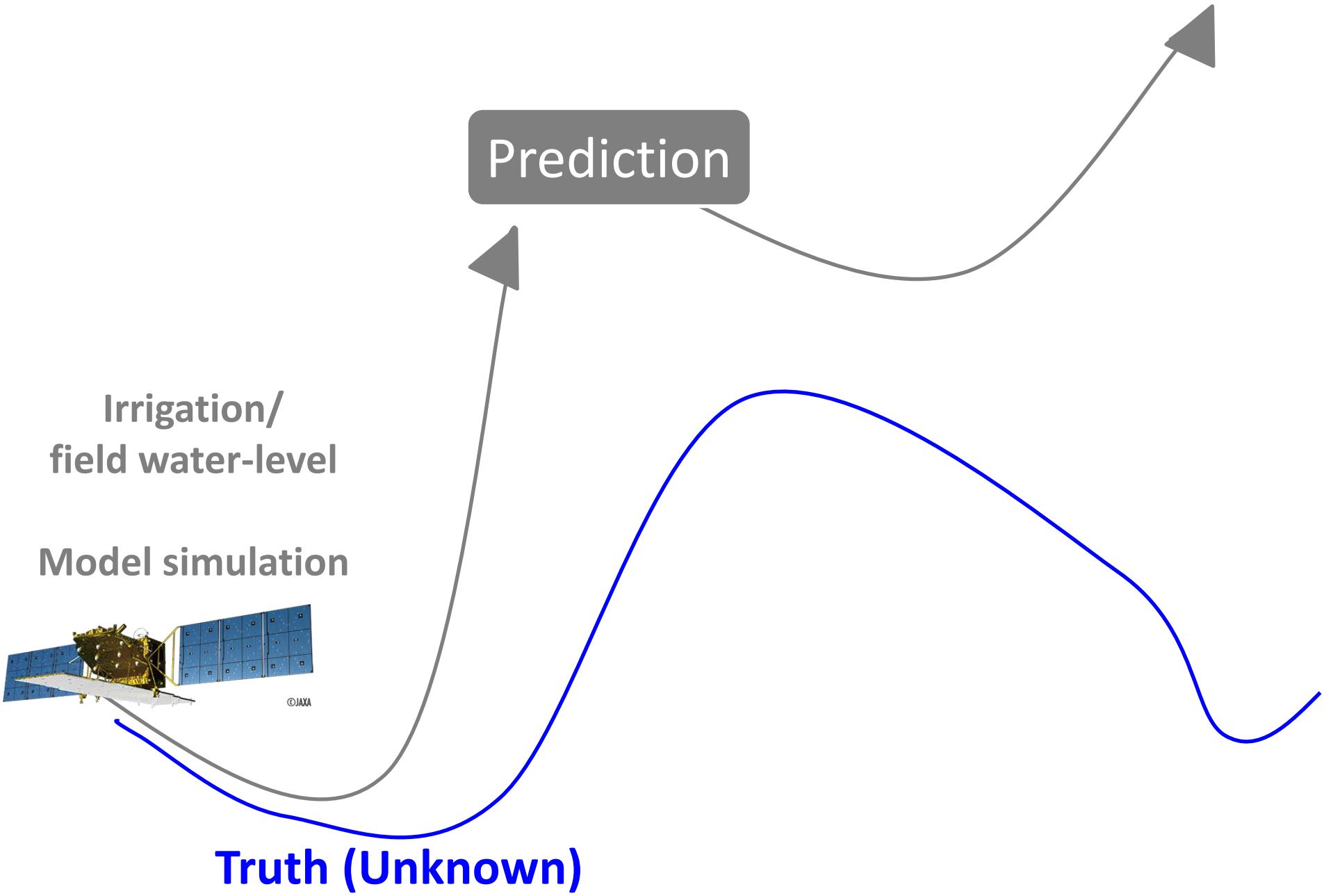


Floodability analysis

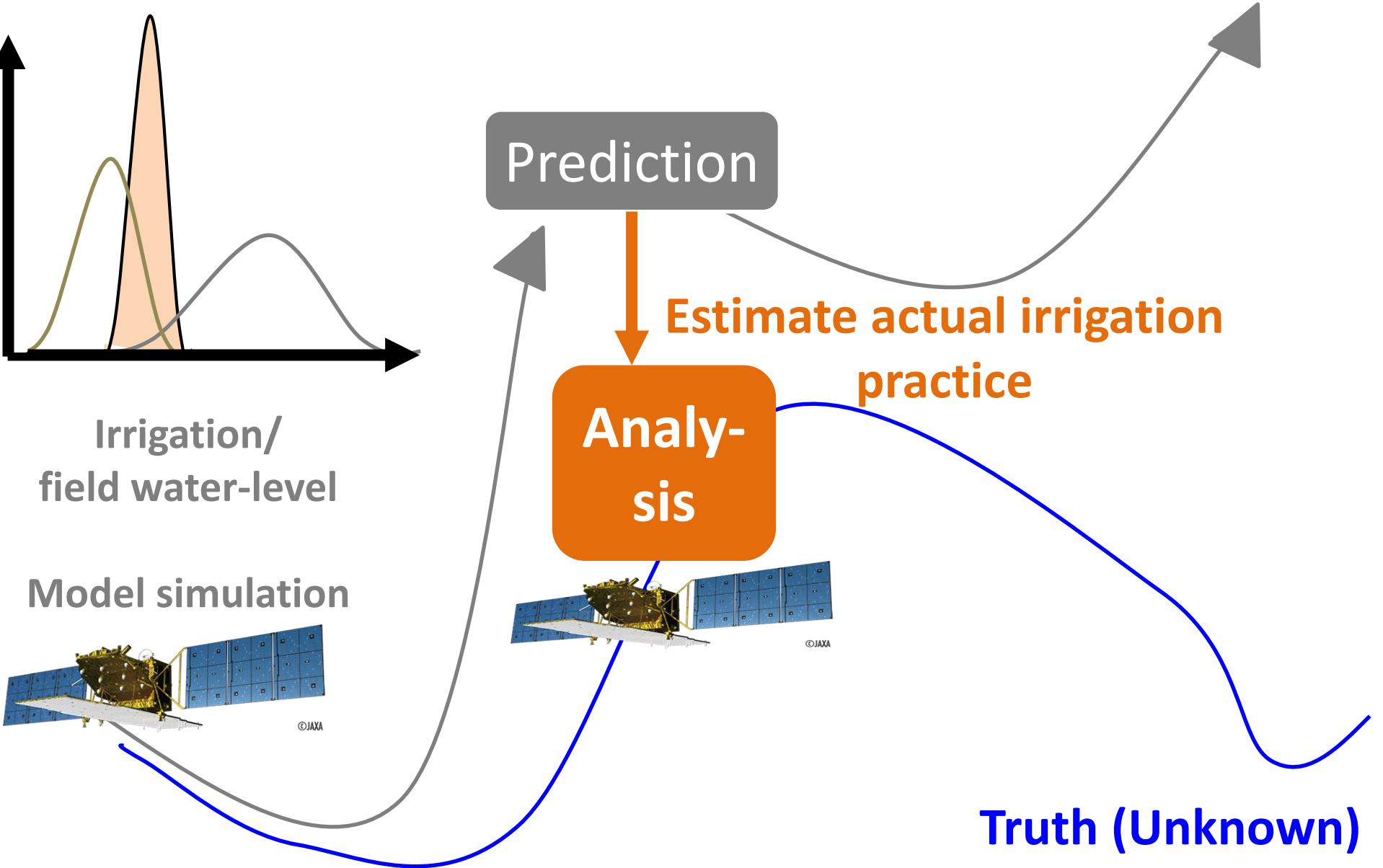
(Cumulative LSWC/
observation scenes)



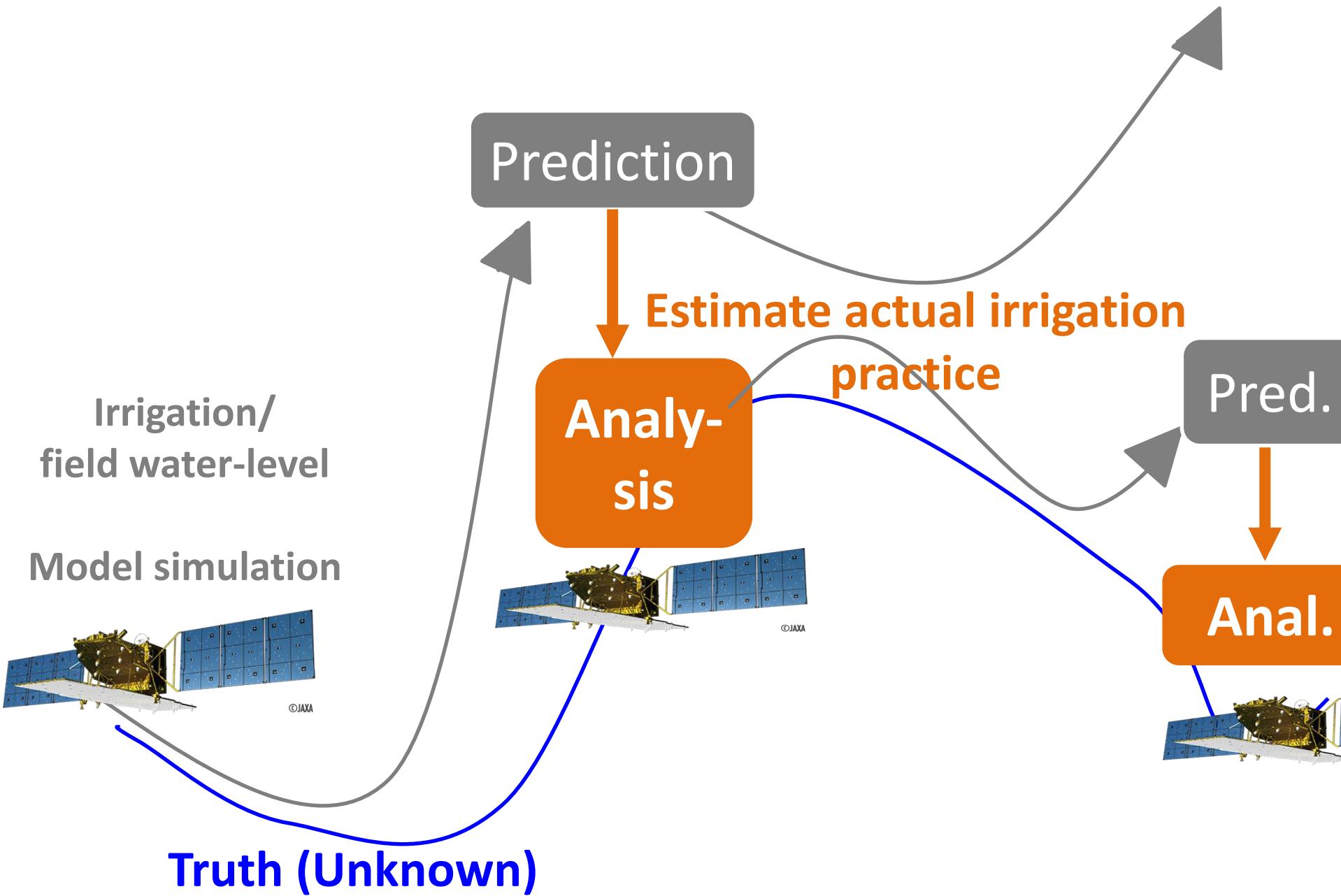
Our data integration scheme



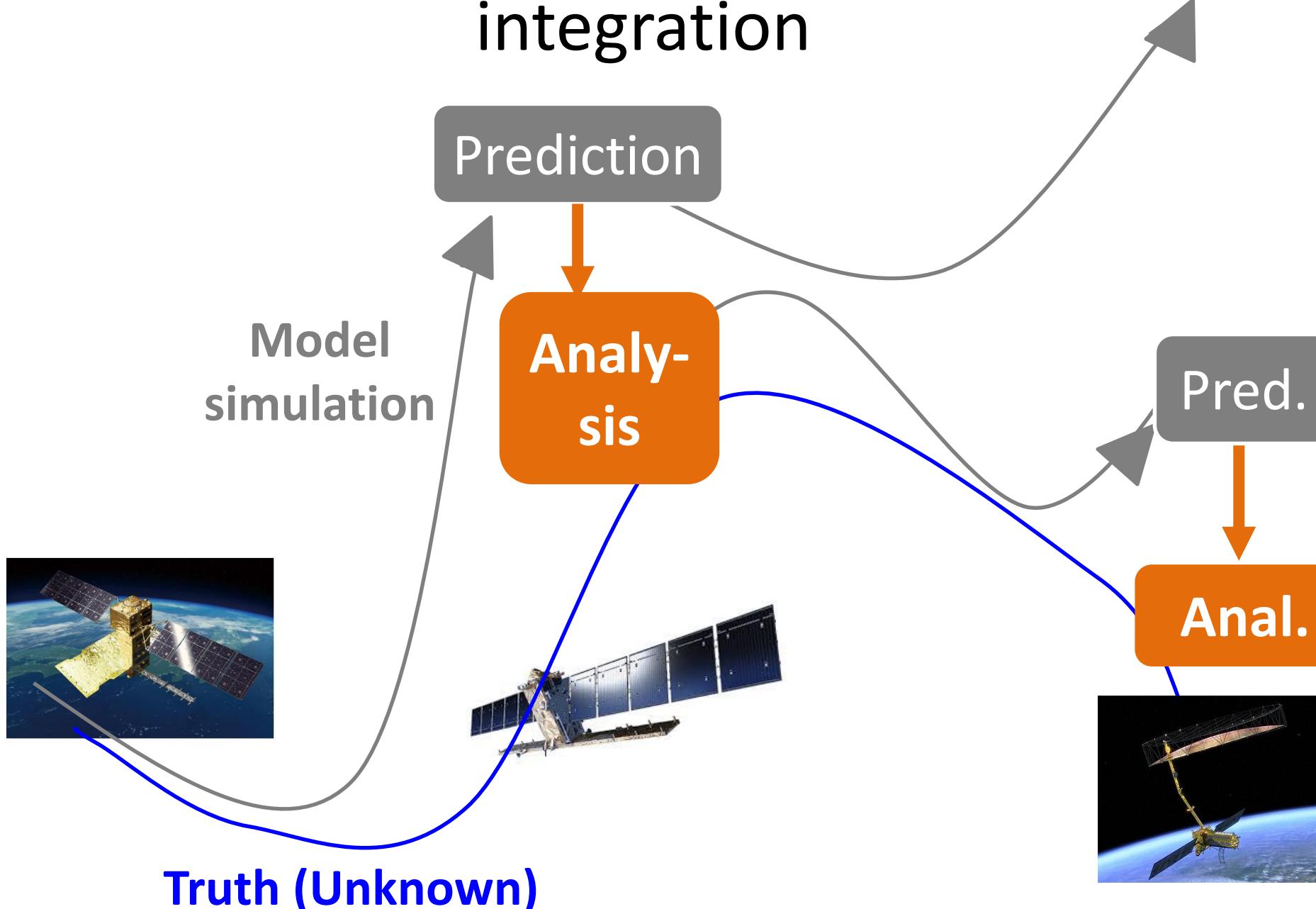
Our data integration scheme



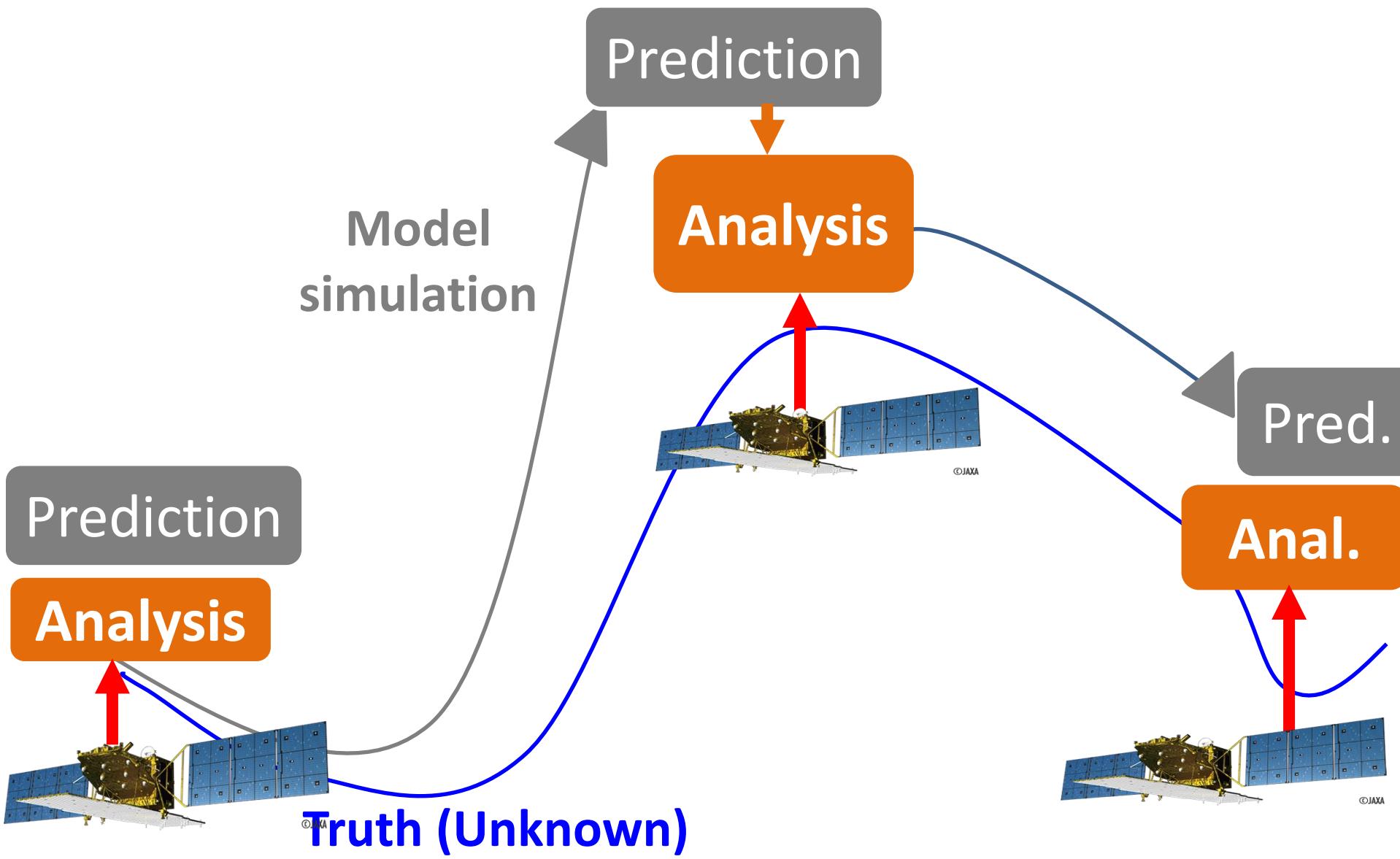
Our data integration scheme



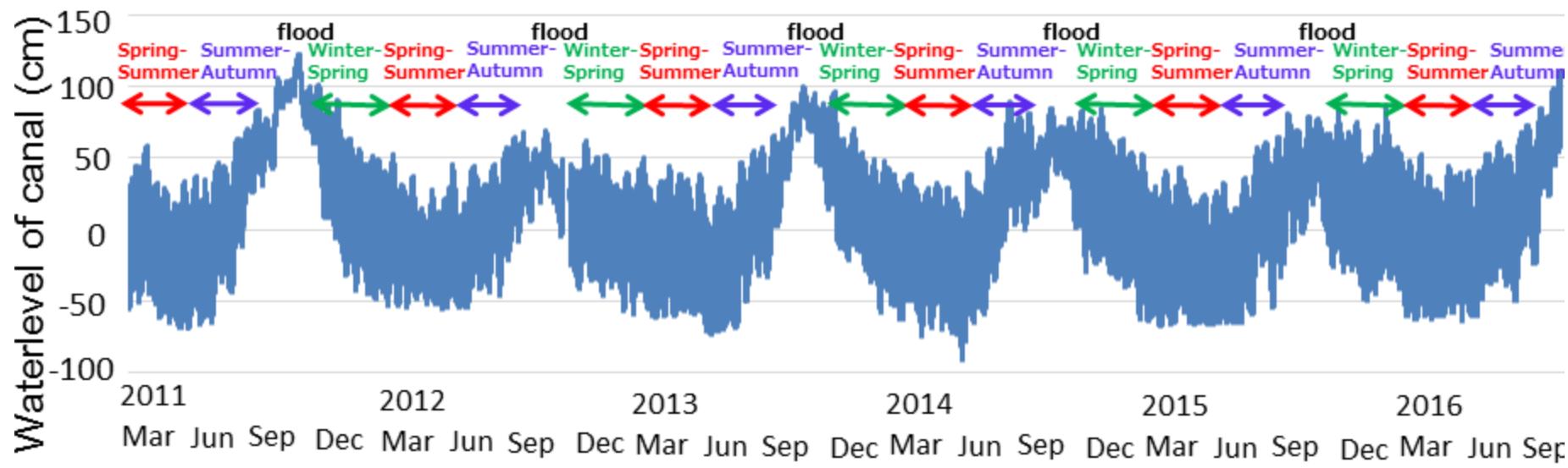
Different frequency SARs / optical sensors integration



Speckle noise filtering considering geo-process

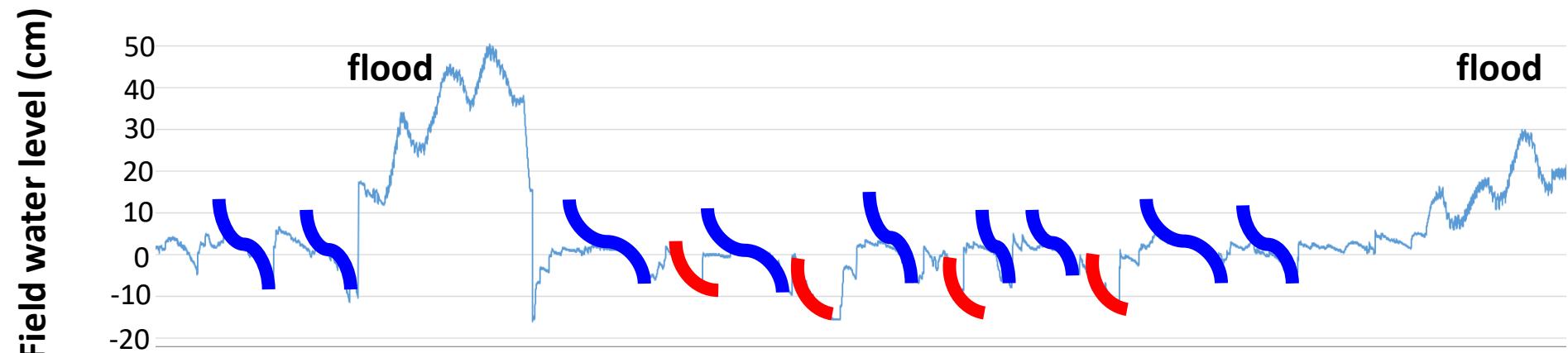
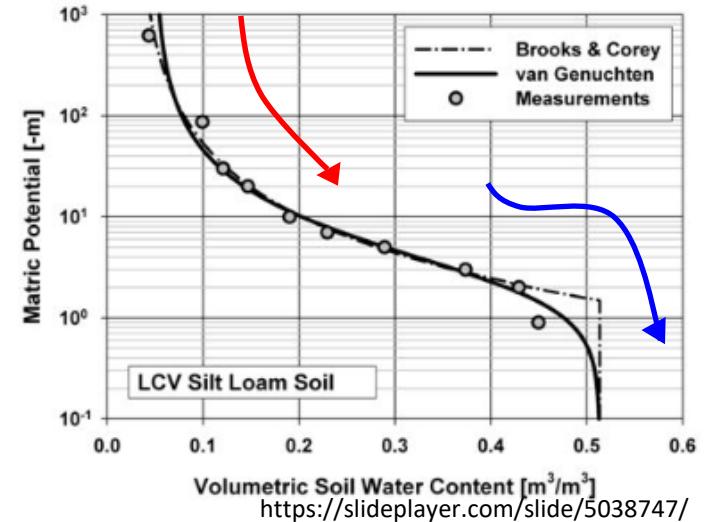


Simulation scheme considering actual states of the delta



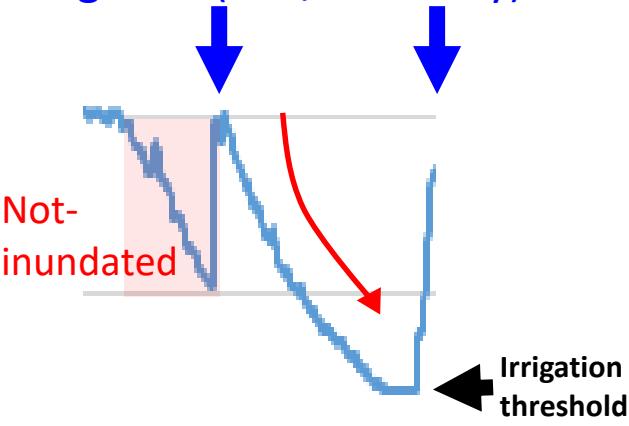
Simulation scheme

Hysteresis of soil matric potential

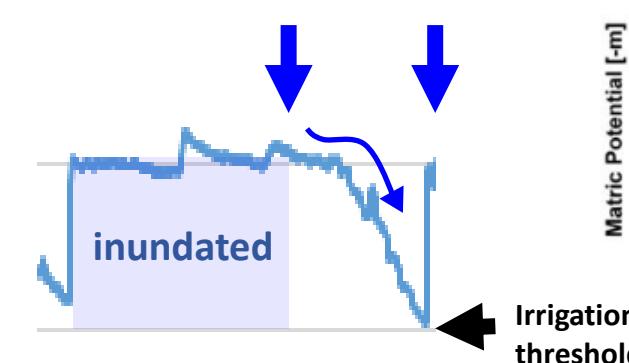


Model structure

Irrigation (init./boundary)



Irrigation (init./boundary)



Implicit RK4 integration model

WL = field water level

Matric-potential at irrigation index (Di) = $\Sigma(\text{soil inundation rate before the irrigation, days after sowing, clay content}) \cdot \alpha_i$

t = days after irrigation

Gravitational-potential at irrigation index (G) = field water level after irrigation * β

$$\frac{dWL}{dt} = \gamma * \exp\left(\delta * \{1 - \log[\exp(Di * (t - G)) + 2 + \exp(-Di * (t - G))]]\} * Di * (t - G)\right)$$

$$- \frac{\delta * [\exp(Di * (t - G)) - \exp(-Di * (t - G))] * Di * (t - G)}{\exp(Di * (t - G)) + 2 + \exp(-Di * (t - G))}$$

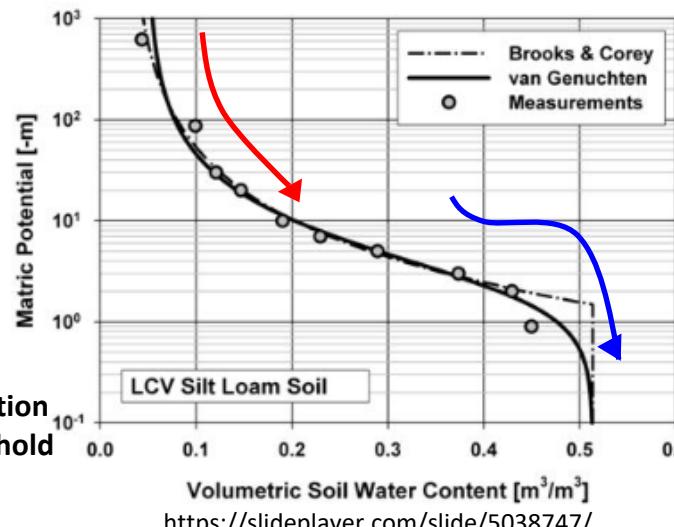
$$+ Di * \{1 - \log[\exp(Di * (t - G)) + 2 + \exp(-Di * (t - G))] + \text{rain-fall}\}$$

Irrigation function

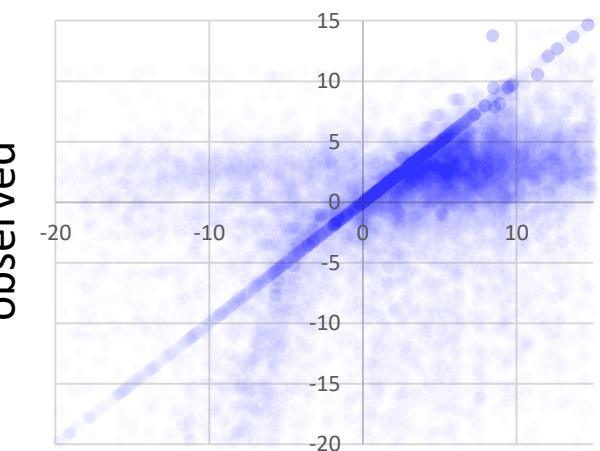
if $WL < \text{threshold}$

$WL = x \text{-cm above the soil}$

Parameter update by the analysis with EO data

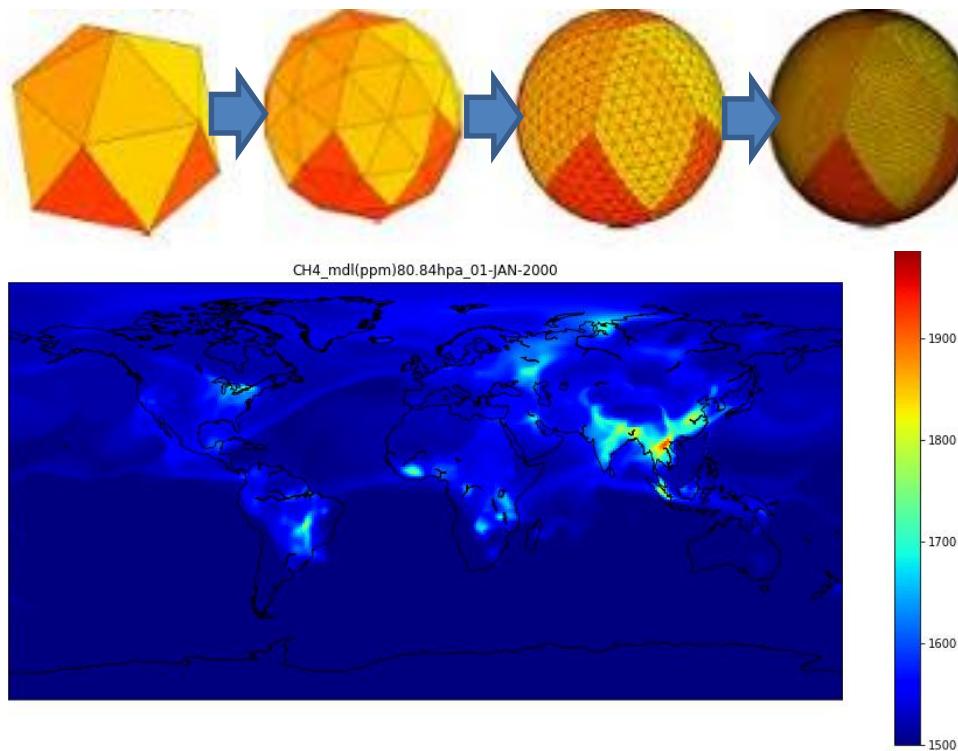


observed

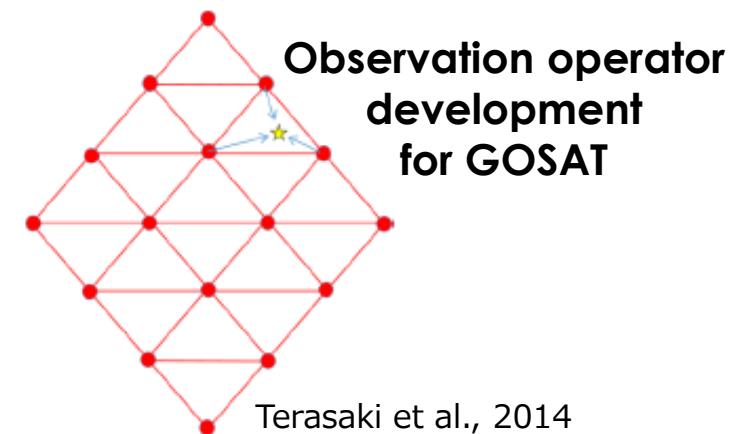
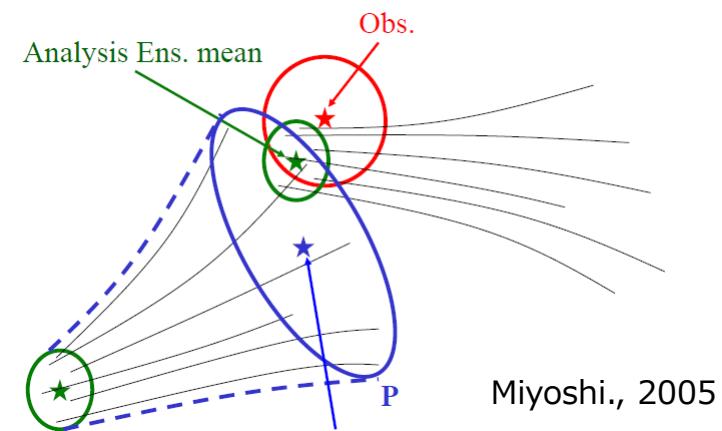


NICAM-TM(Chem)-LETKF with AMSU, PREPBUFR and GOSAT

Nonhydrostatic ICosahedral Atmospheric Model-TM(Chem)



Local Ensemble Transform Kalman Filter



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

→GOSAT data assimilation with NICAM-TM!

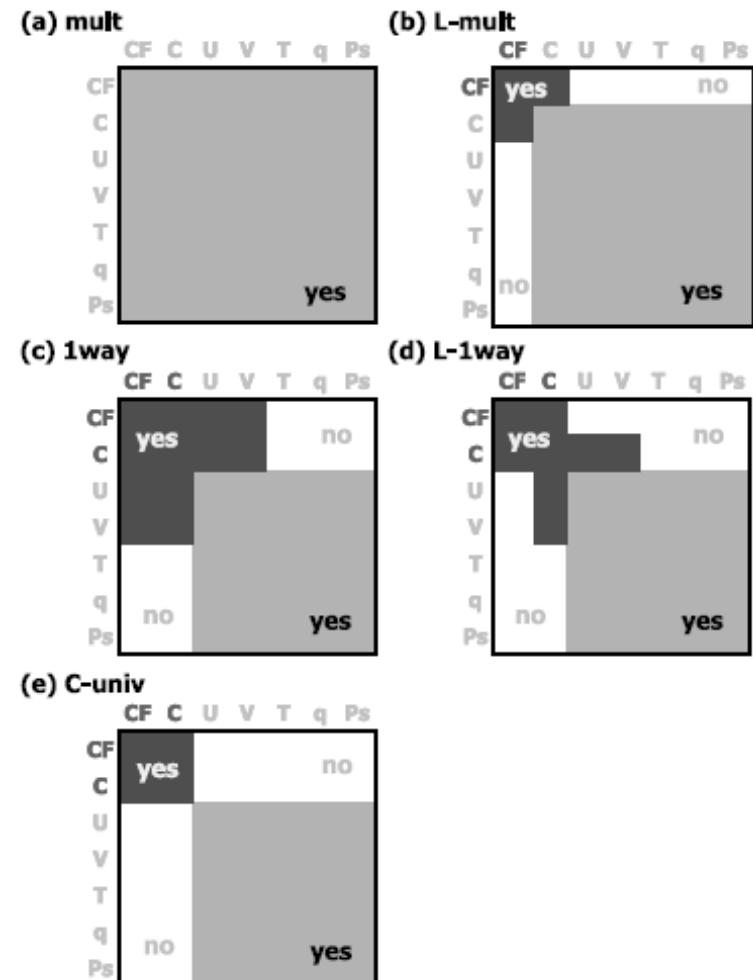
“Variable localization” in an ensemble Kalman filter: Application to the carbon cycle data assimilation

Ji-Sun Kang,¹ Eugenia Kalnay,¹ Junjie Liu,² Inez Fung,² Takemasa Miyoshi,¹
and Kayo Ide¹

Flux estimation from
atmospheric concentration
by omitting multi-collinearity

- No direct emission or
apriori info. is required!

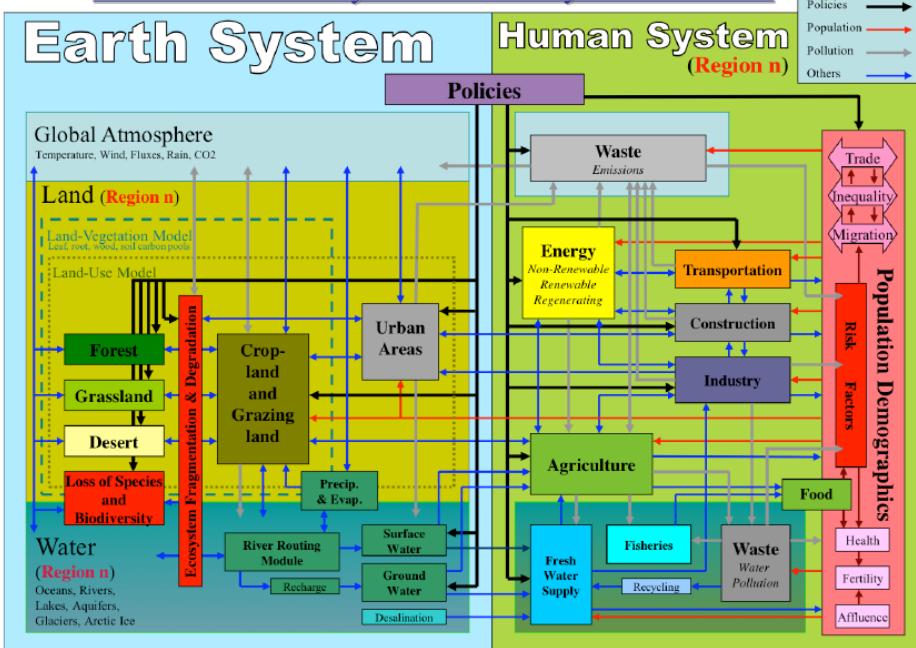
Transparent MRV!!



Back ground covariance matrices

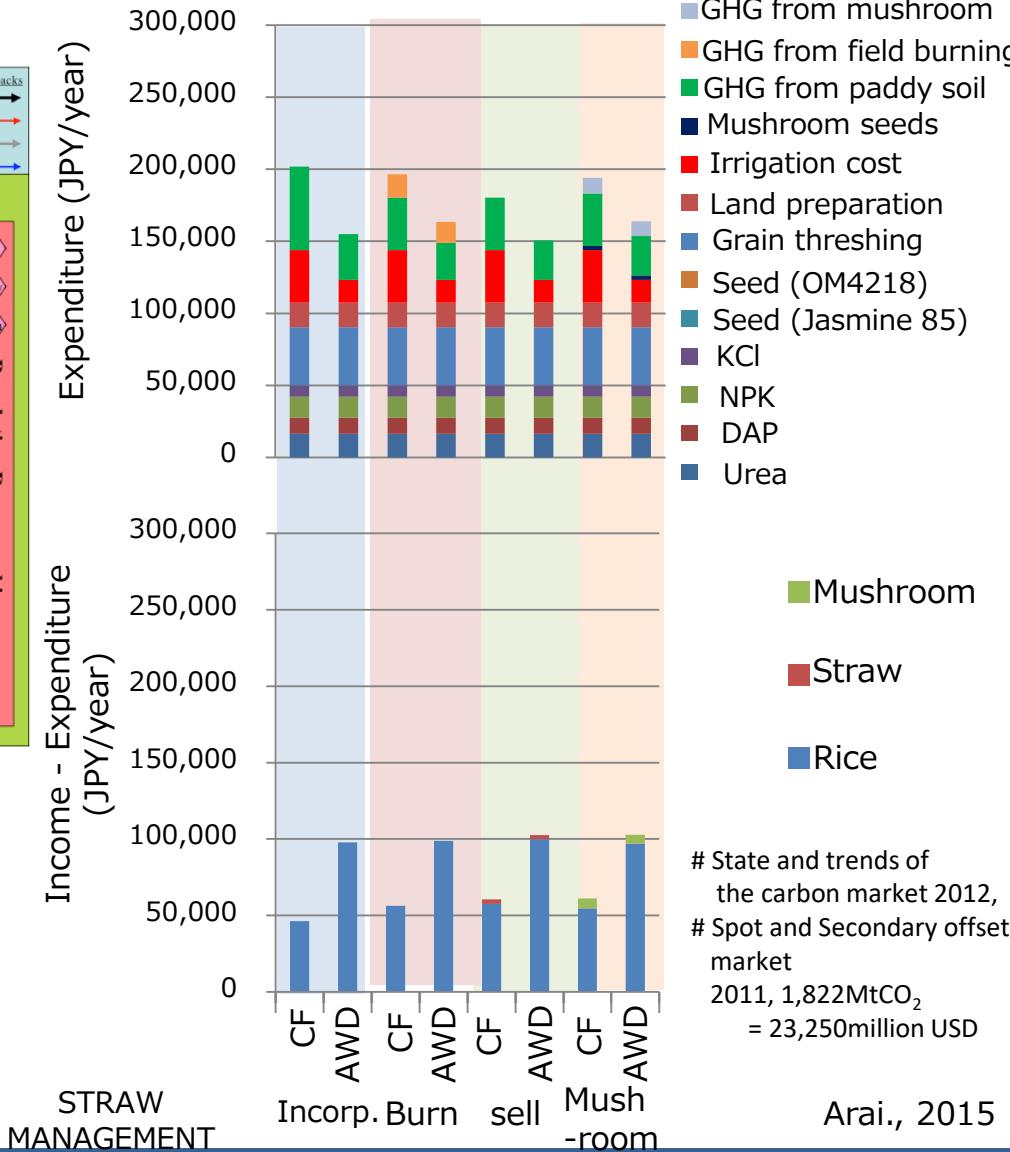
Economic assessment of GHG mitigation under various uncertainties

Schematic of Earth System - Human System Feedbacks

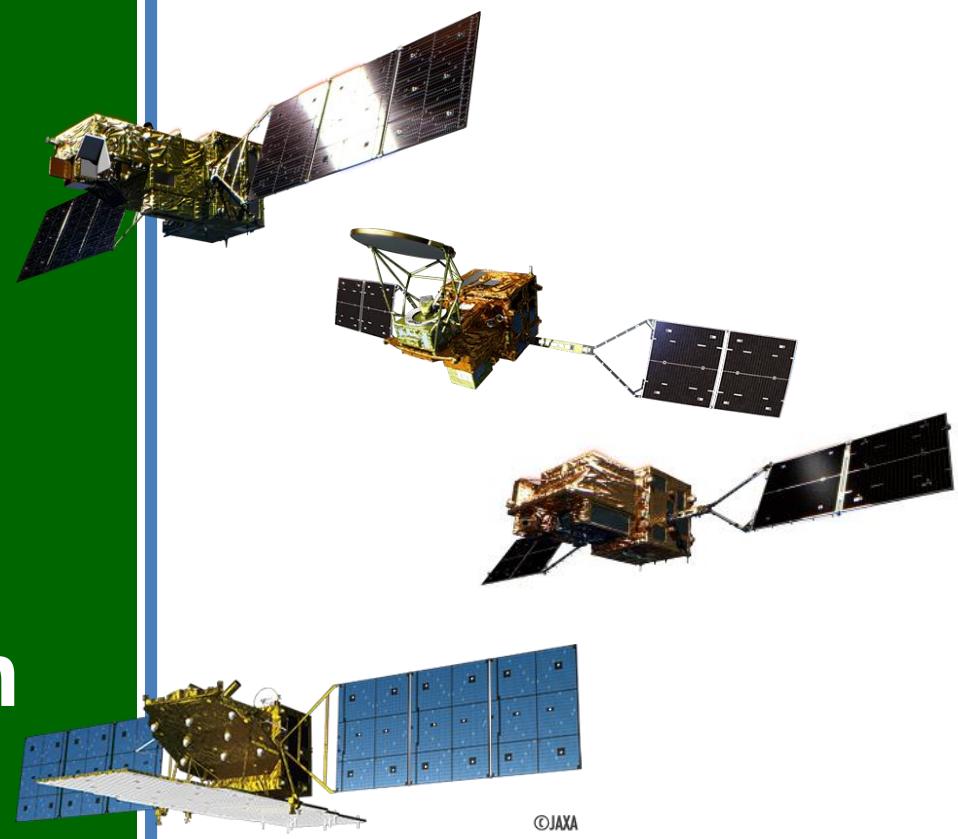


Kalnay et al. 2017

Transparent MRV system on baselines/mitigation-effects with satellite data is the key !



Thank you for your attention



©JAXA

