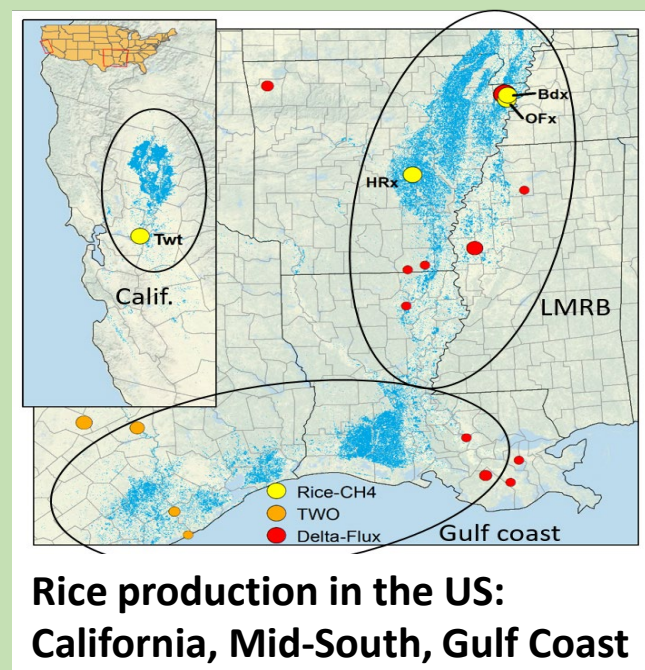


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Researchers: Beatriz Moreno-García, Rita Leavitt, Univ. Arkansas
Graduate Student: Rongzhu Qin, Yawen Deng, Mingtong Zhang, Univ. Ill.
Undergraduate Student: Isabelle Wagenvoort, Colo. College

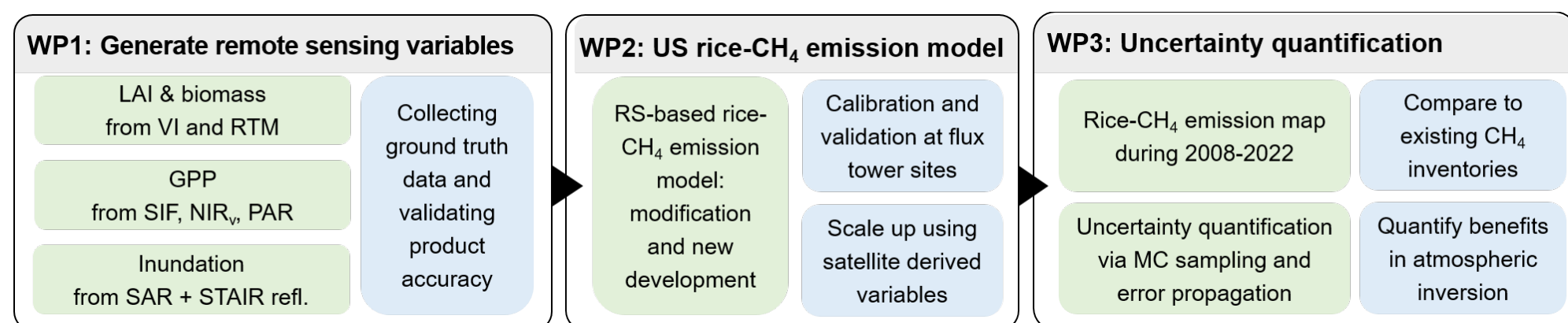
Motivation



Globally, rice is responsible for:
10% of anthropogenic CH₄
25% of all water withdrawals
20% of human caloric intake
The US grows less than 2% of global rice supply, but is 5th largest exporter

- Current inventories of US GHG emissions give huge uncertainty around annual rice-CH₄ emissions (9.2-21.6 MMT CO₂-eq at 95% CI)
- Inventories are annual and not fine frequency
- More information is needed to guide emissions reductions approaches

Project design

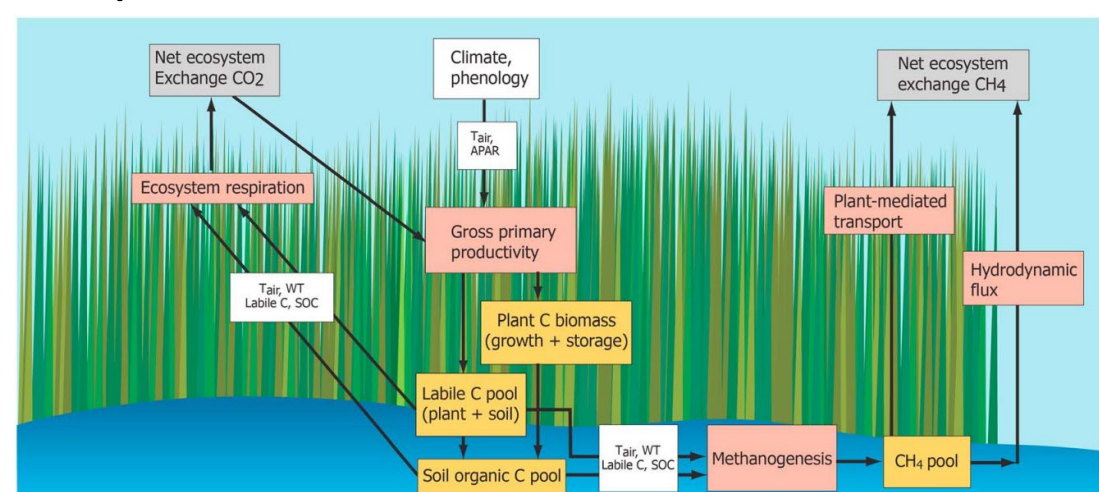


The project objective is to:

- generate a consistent national, daily, 500m rice-CH₄ emission dataset (2008-2022),
- based on a daily satellite-based inundation, GPP, leaf area index (LAI) and biomass dataset,
- process-based modeling, with uncertainty estimates, and
- benchmarked to eddy covariance EC observations of CH₄ flux.

Science area 1). Daily GPP modeling

The **PEPRMT model** estimates CH₄ Flux via GPP and Reco/NEE models with labile and soil organic C pools after ecosystem respiration, from LAI, PAR, Ts observations



Okawa et al. 2017, IGR-Biogeosciences

GPP is modeled as: $GPP = LUE \cdot APAR \cdot f(T_k)$
LUE: maximum light use efficiency constrained by $f(T_k)$
APAR: absorbed photosynth. active radiation $(1 - \exp(-k \cdot LAI)) \cdot PAR$
Where k is the light extinguishing term

$$f(T_k) = 1 - \frac{H_{0r} \cdot \exp\left(\frac{H_{0r}(T_r - T_{opt})}{T_r \cdot H_{0r} + T_{opt}}\right)}{H_{0r} - H_{0c} \left(1 - \exp\left(\frac{H_{0c}(T_r - T_{opt})}{T_r \cdot H_{0c} + T_{opt}}\right)\right)}$$

H_{0r}, H_{0c}, k, T_{opt} tuned using MCMC

Science area 2). SIF measurement system development and adaptation for rice

- System assembly and extended testing over grass, including field calibration
 - SIF/hyperspectral instruments performing well under ideal conditions
 - Finishing last modifications for field deployment & improvements for radiometric accuracy (see Ph.D. student **Will Richardson's** poster here)
- In the field, we now have A/C line power at two eddy covariance sites

Science area 3). Mapping inundation dynamics

Approach:

1. Use **ground truth data** to label daily flood status (flooded/not flooded)
2. Compare Reflectance/Index-based thresholding to phenology-based detection, eventually ML approach
3. Analyze **backscatter patterns** and vegetation indices, from Harmonized Landsat and Sentinel-2 (HLS)

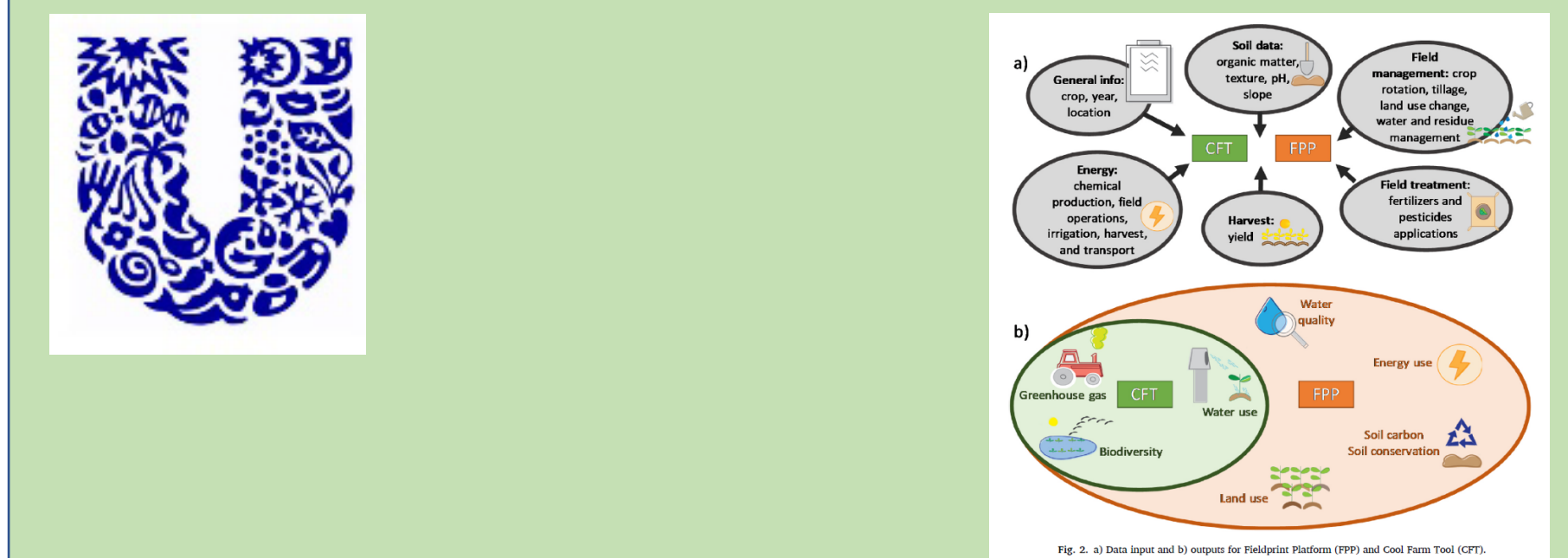
Stakeholder engagement

NRCS: Adam Chambers



Unilever: Stefani Millie Grant

Work by Beatriz Moreno García, University of Arkansas



Moreno-García et al. 2021, JCP

Next: Environmental Defense Fund: Joe Rudek & team



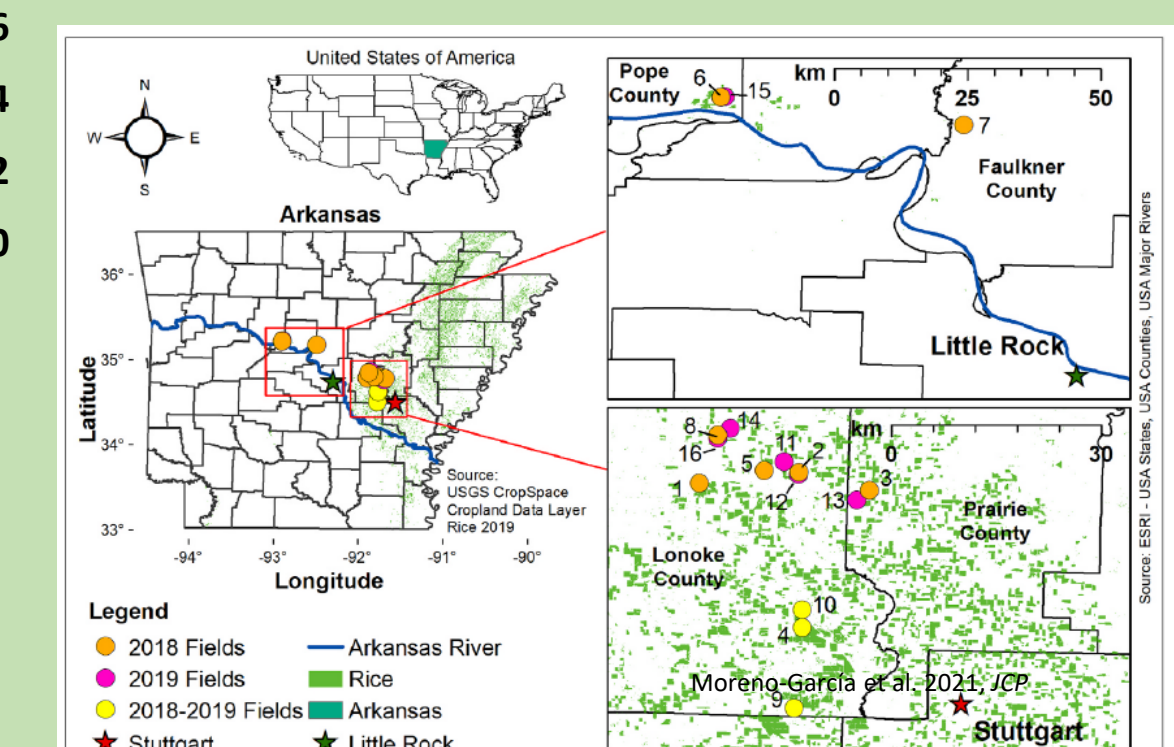
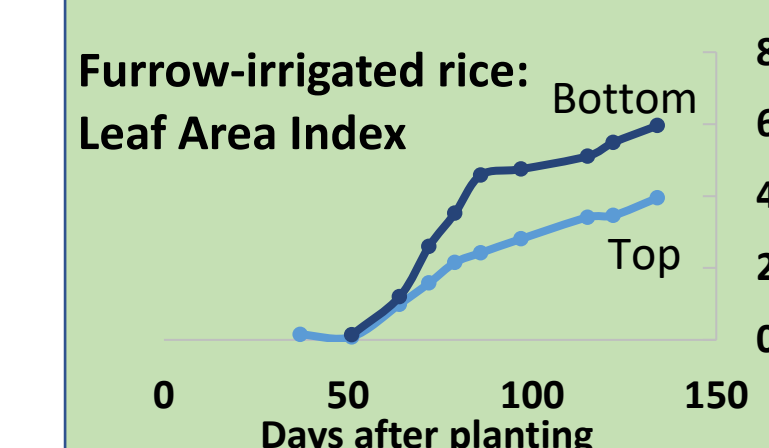
Ongoing synthesis work

Optical sensing workshop + follow-up



Prasad Bandaru's Crop CMS

- Contributing data from 20+ field sites in Arkansas, multiple years
- LAI, soil moisture, inundation, crop staging, etc.
- Data from 'newer' production practices like furrow-irrigated rice



Science area 4). CH₄ flux prediction

Here I highlight new work from our group, demonstrating the effectiveness of simple flood inundation duration in predicting CH₄ fluxes

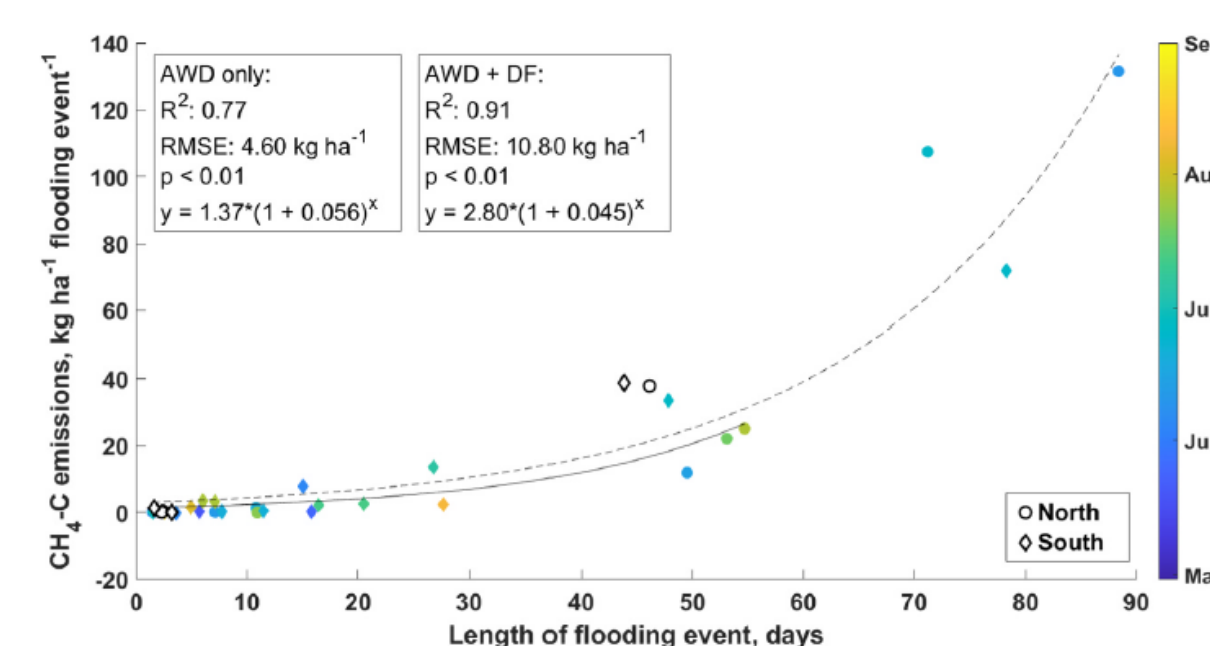
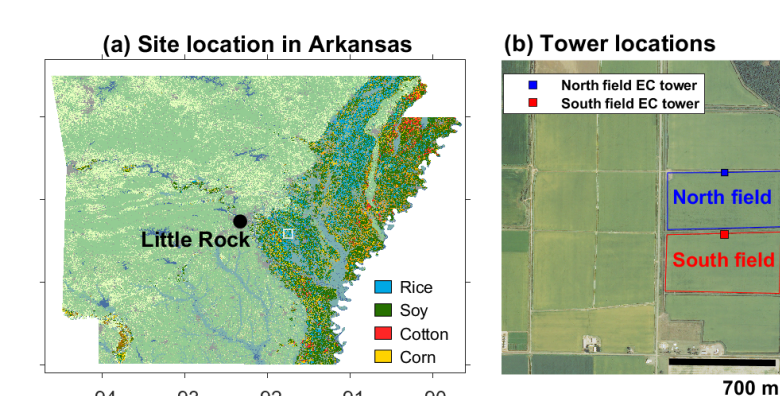


Fig. 7. Cumulative CH₄ emissions per each flooding event for all field-seasons managed. The dashed line indicates the relationship for all main field-seasons (26 independent flooding events) while the solid line indicates the relationship for all main field-seasons managed with AWD (23 independent flooding events). The black outlined points indicate flooding events during the ratoon season and were not included in model creation. Each main season flooding event is color-coded by the month at the midpoint of the event. Note that the cumulative emissions include the 2 days following each event to catch any spike in emissions at the beginning of the drying event that is attributable to the release of trapped CH₄ generated during the flooding event.

Leavitt et al. 2023, Agric. Ecosys. Env.