

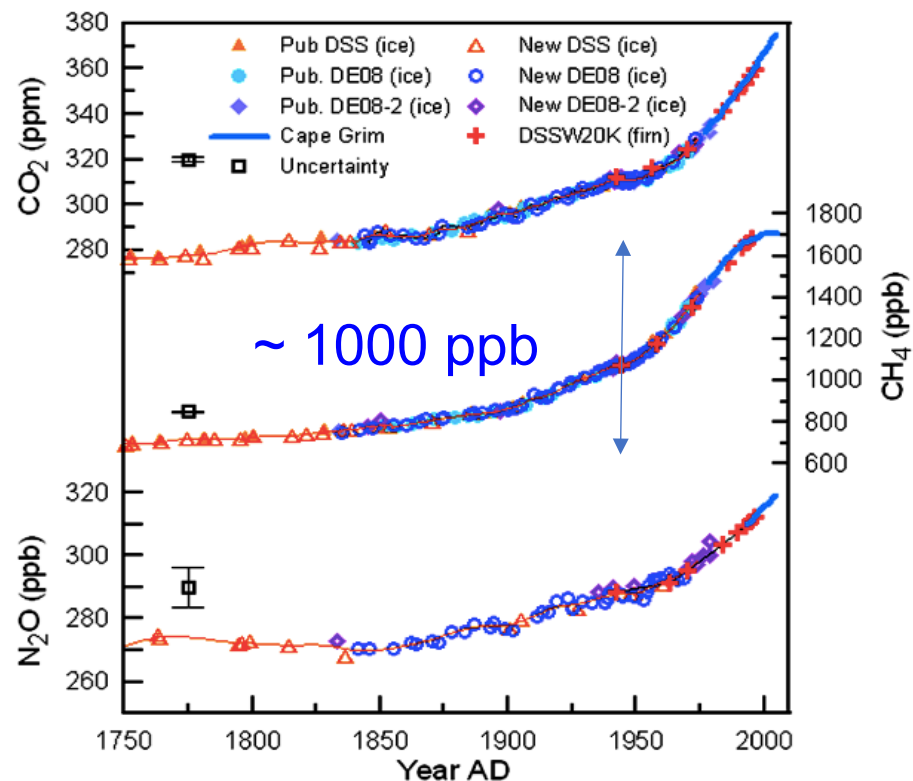
Recent global and Regional Methane Budgets Constrained by Atmospheric Observations

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Greenhouse gas methane (CH₄)

Greenhouse gases in last 300 years



Ice core record

MacFarling Meure et al., GRL, 2006

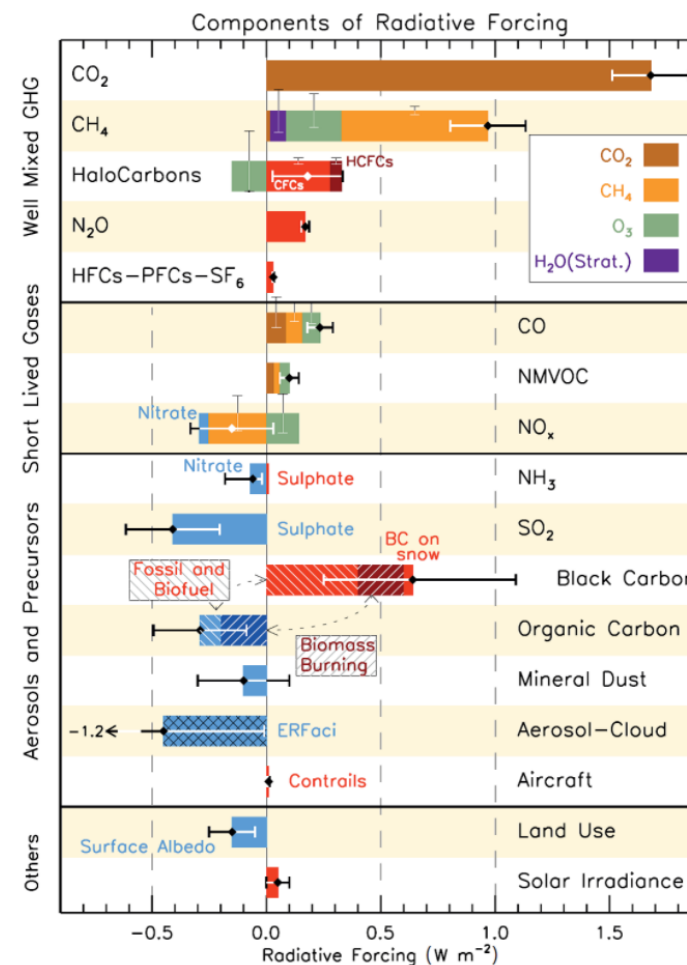
GWP
(~100 years)

1

80

300

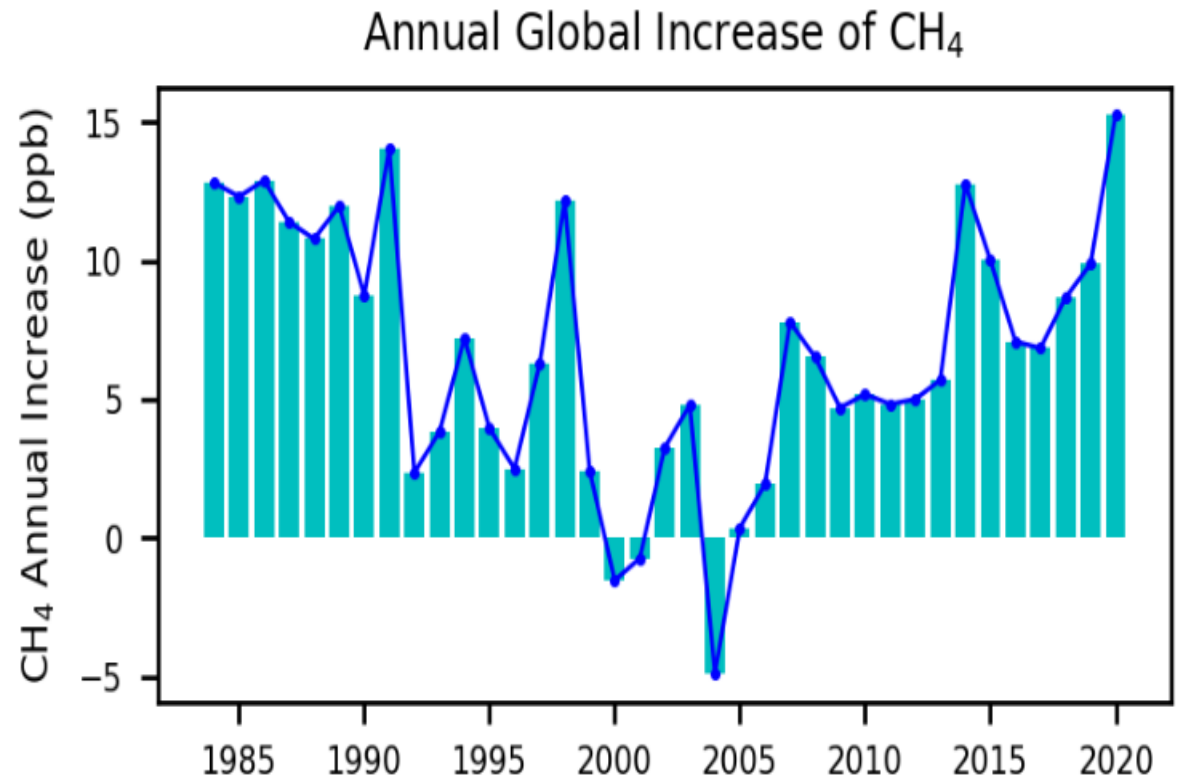
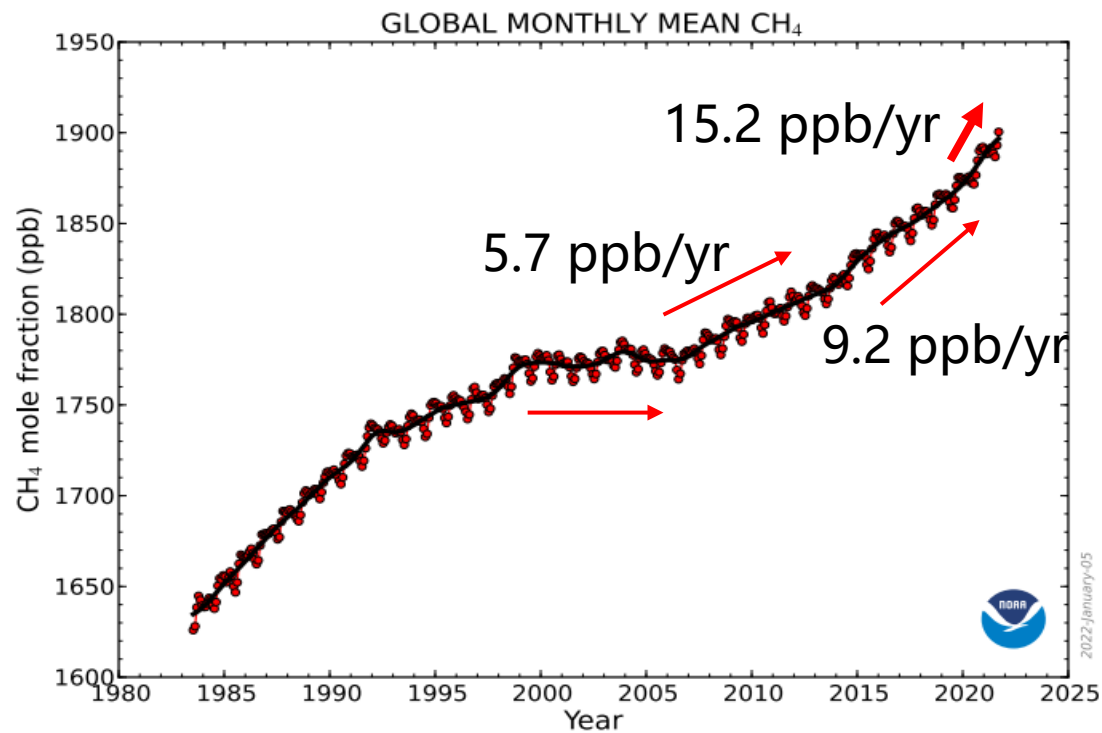
Radiative forcing



IPCC AR5

Global methane concentration

What is driving the increase of methane concentration and its variability?



Methane emissions



Multiple sources

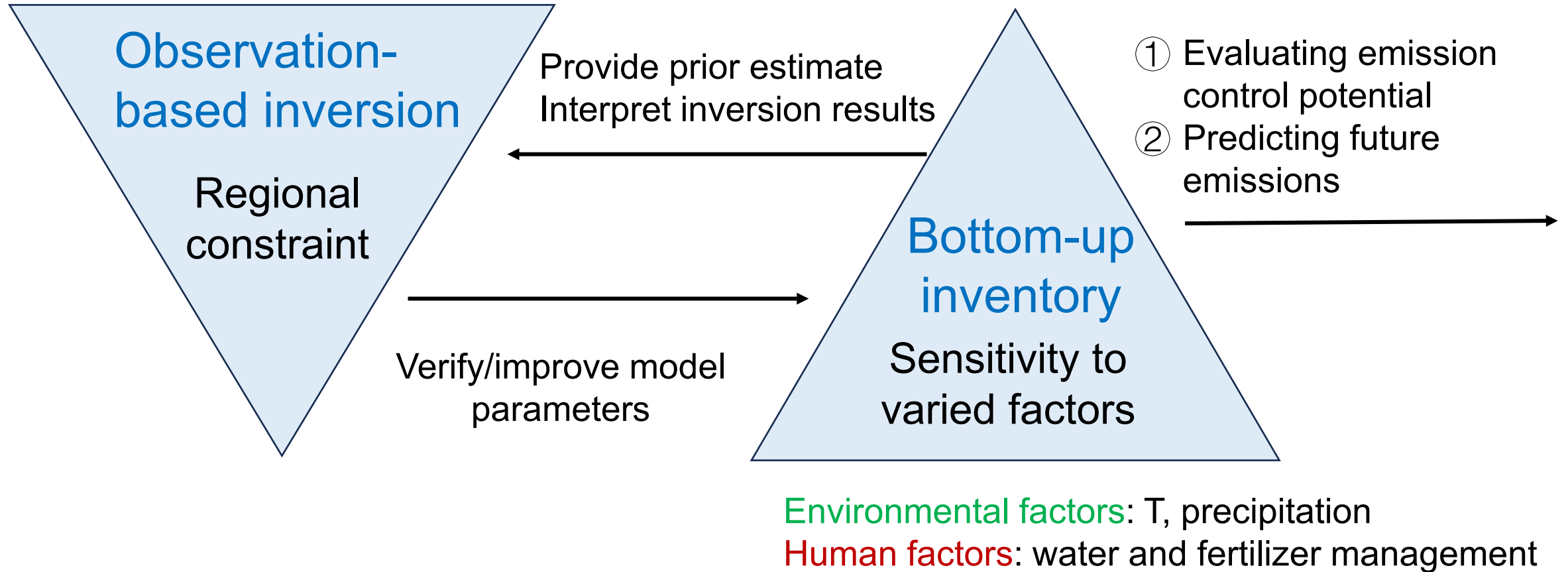
Energy transition
Food security
Waste treatment

Bottom-up estimate

Super emitters
Highly heterogeneous
Complex dependence on environment

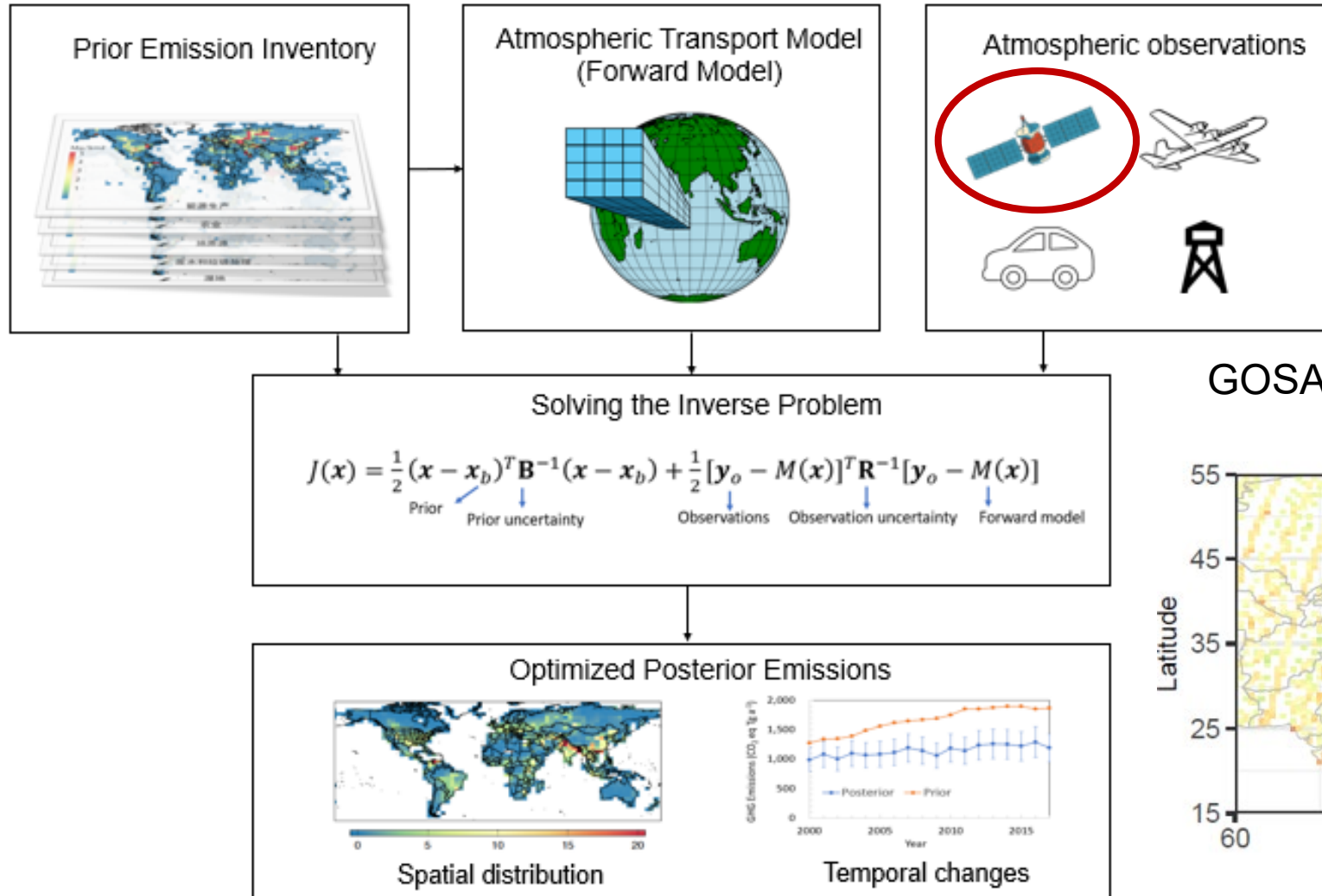
Top-down approach using atmospheric observations

Research framework: bottom-up modeling + observation



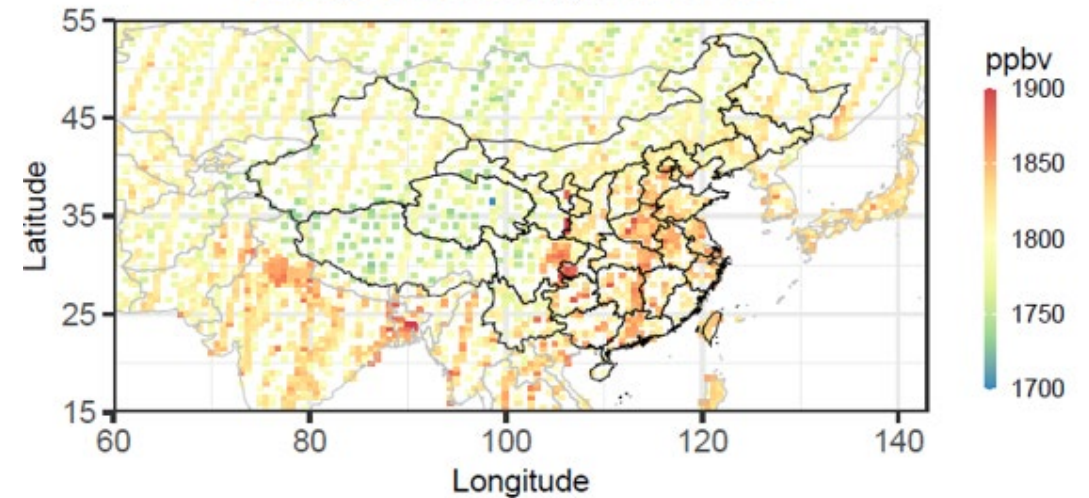
Estimating emissions from atmospheric observations

Inverse Modeling of GHG Atmospheric Observations



GOSAT retrieval from University of Leicester

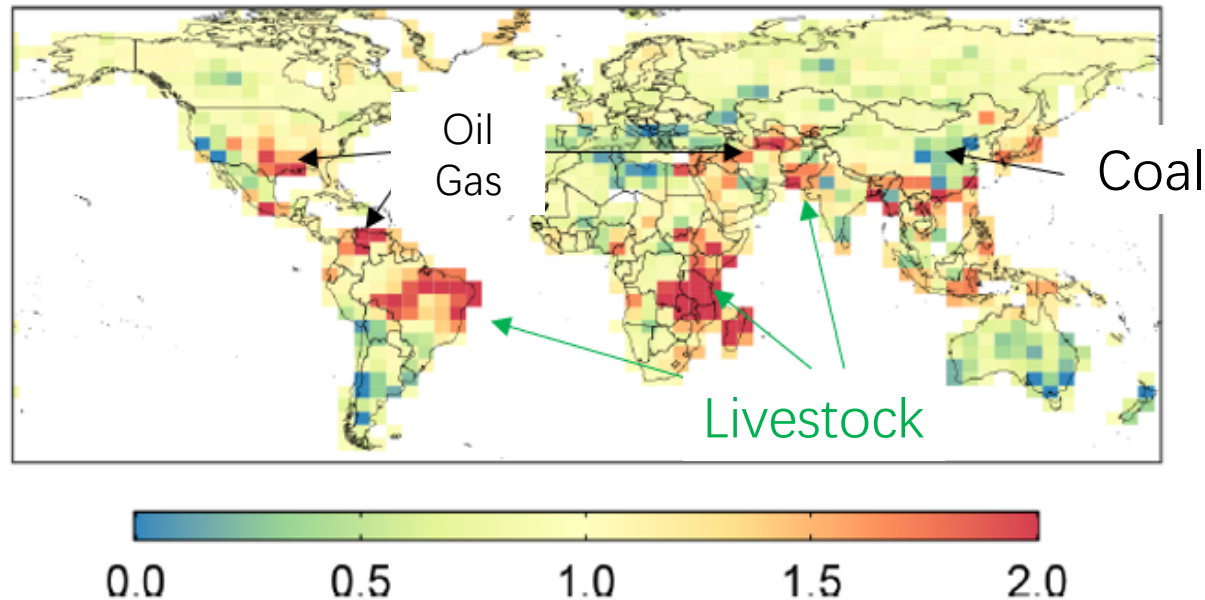
Average GOSAT XCH₄ for 2010–2017



Global methane emissions 2010-2018

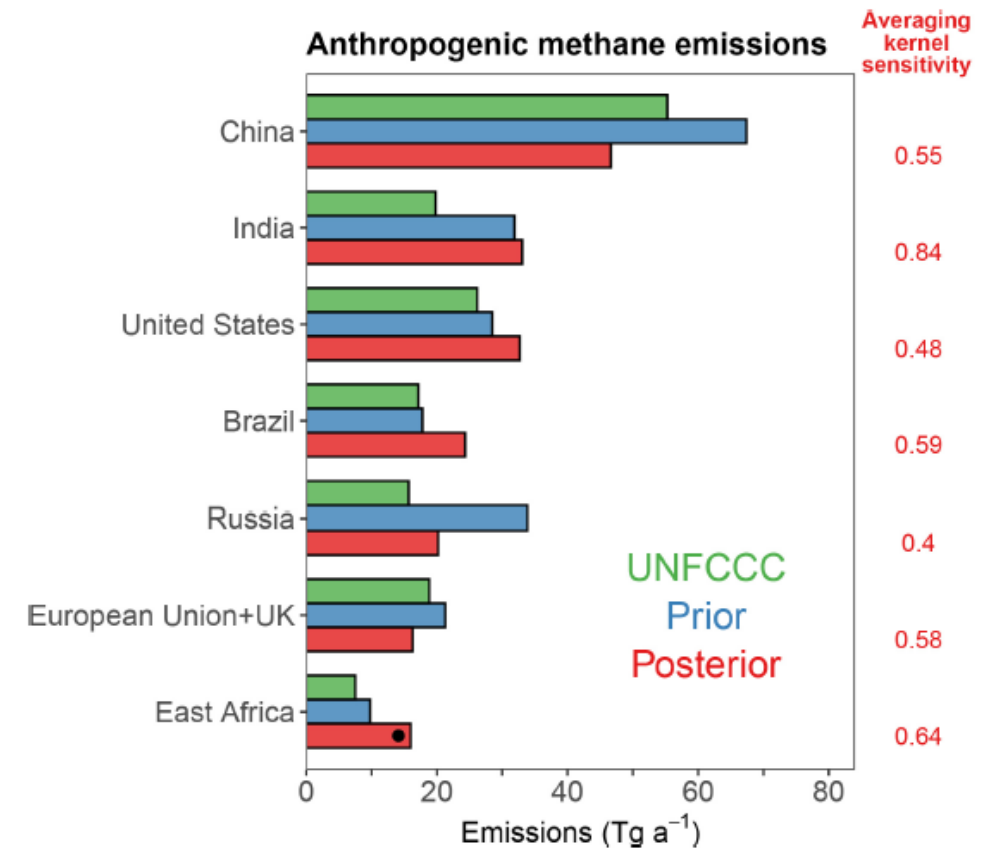
Emissions inferred from GOSAT satellite observations

Posterior / prior scaling factor



Zhang et al., Atmos. Chem. Phys. 2021

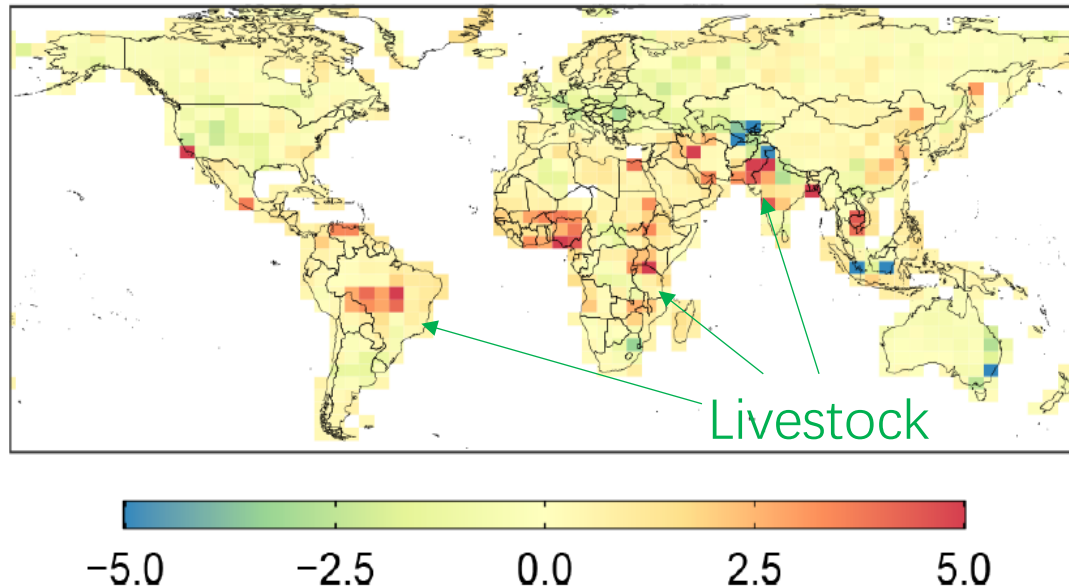
Verifying national inventory



Which sector dominates the increase in emissions?

Emissions inferred from GOSAT satellite observations

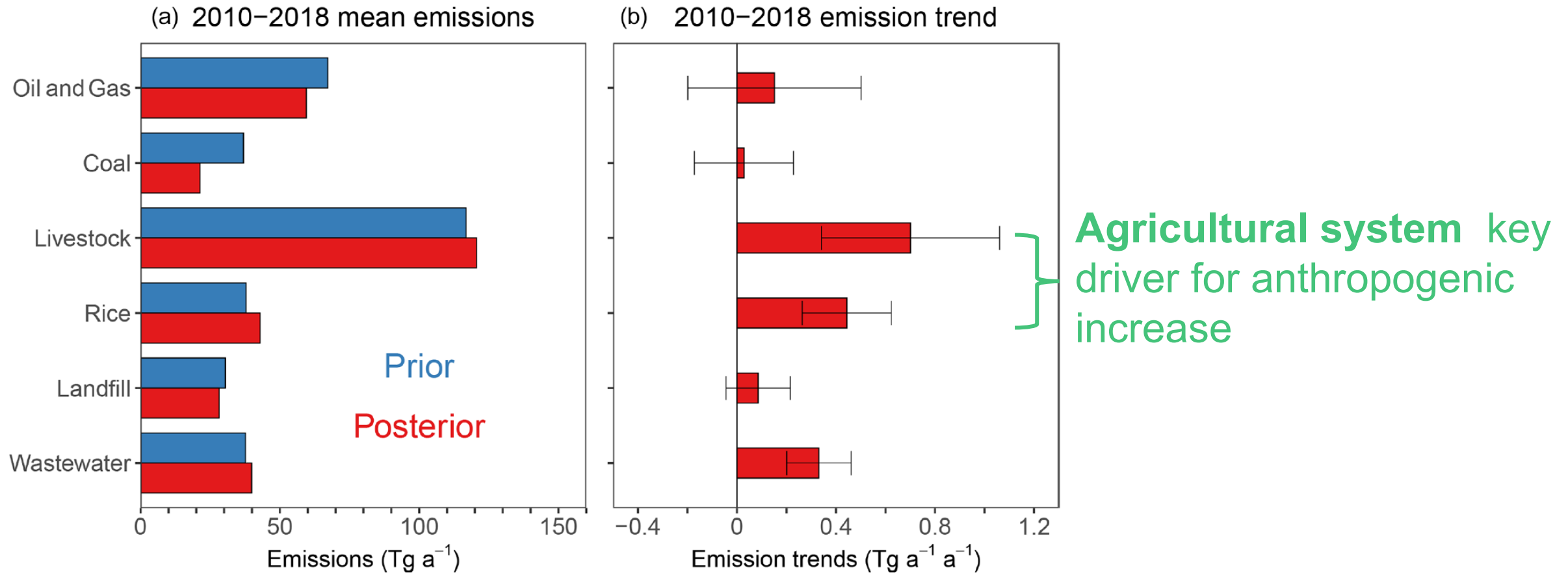
Growth rate of anthrop. emissions (%)



Countries with the fastest growing livestock population (UNFAO)

| Country | Trend (million head per year) |
|-----------|----------------------------------|
| Pakistan | 1.4 |
| Ethiopia | 1.2 |
| Tanzania | 1.1 |
| Brazil | 0.9 |
| Argentina | 0.7 |

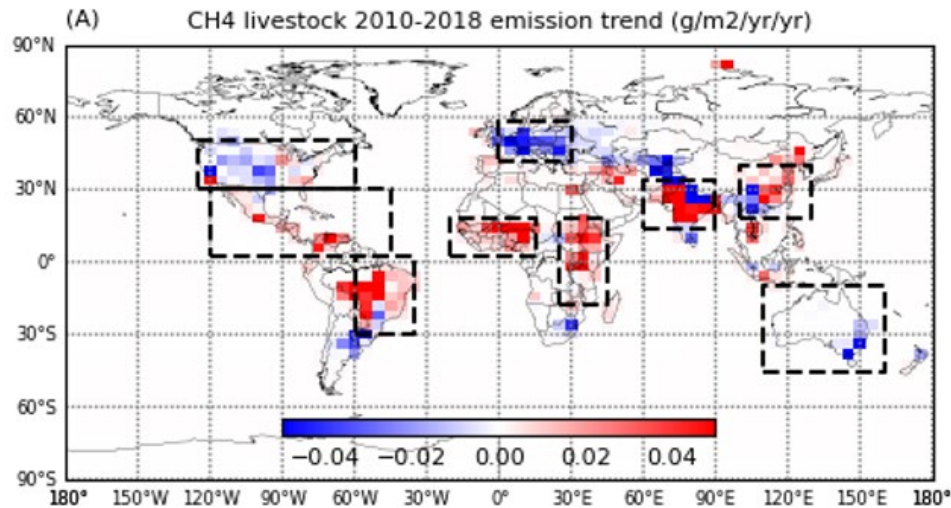
Sector attribution of emission trend



Satellite analysis of multiple species

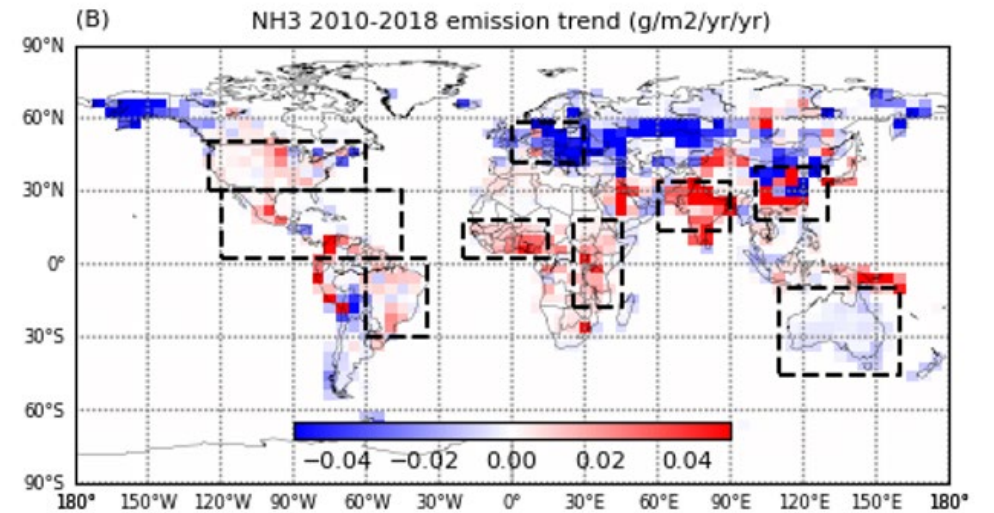
Livestock sector emits both methane and NH₃

GOSAT livestock methane trend



Zhang et al., *ACP*, 2021

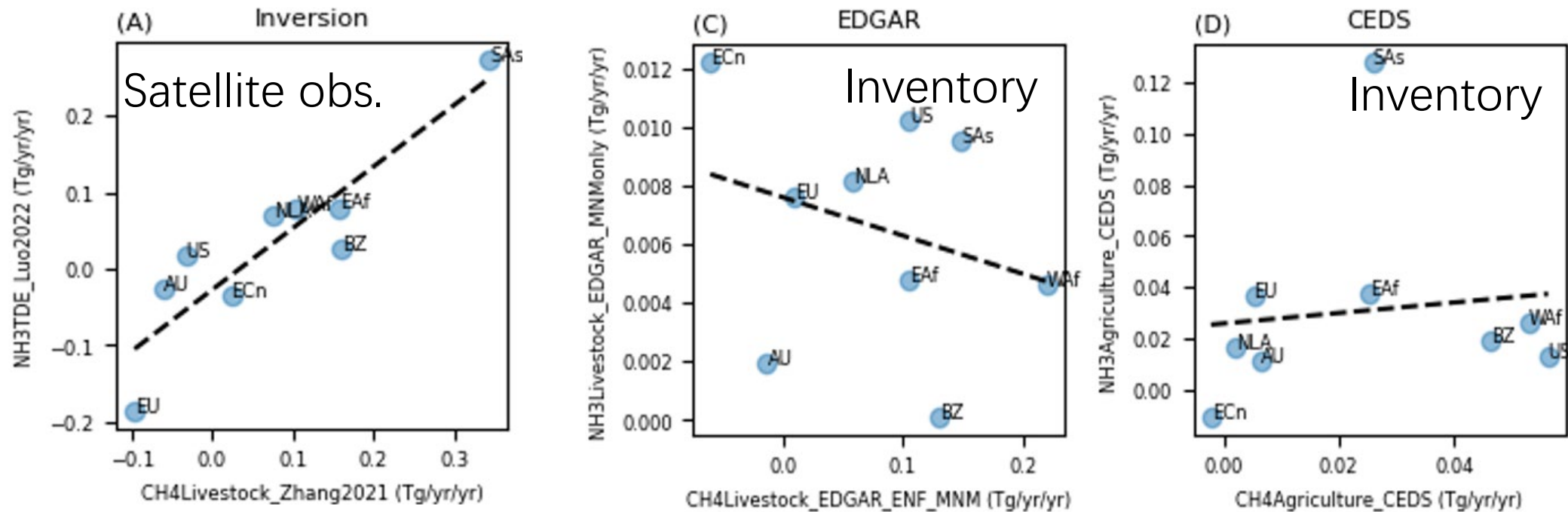
IASI NH₃ emission trend



Luo et al., *ACP*, 2022

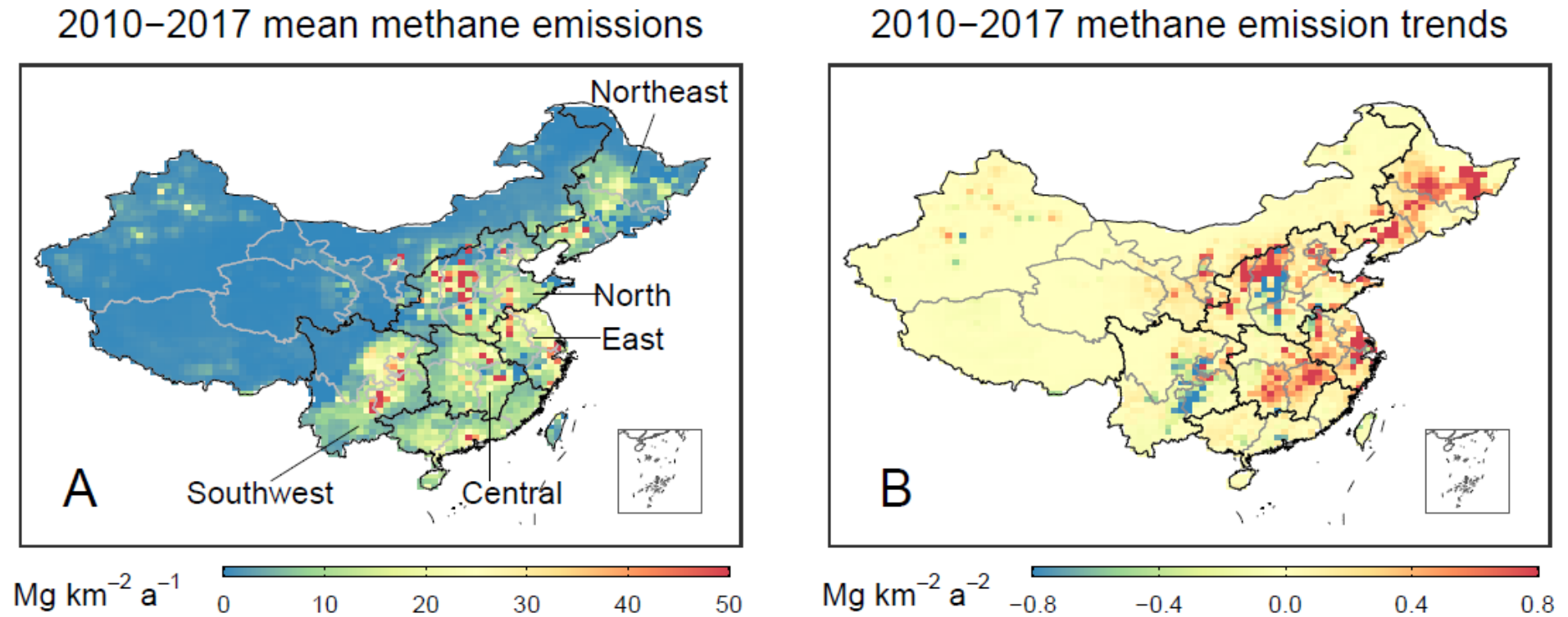
Global livestock emission trend – shown in methane and NH₃

Bottom-up inventory cannot capture the observed correlation between CH₄ and NH₃



Tang et al., in prep

China's methane emissions



- Data repository: Optimized monthly emission fluxes on 0.5x0.6 grid (2010-2017)

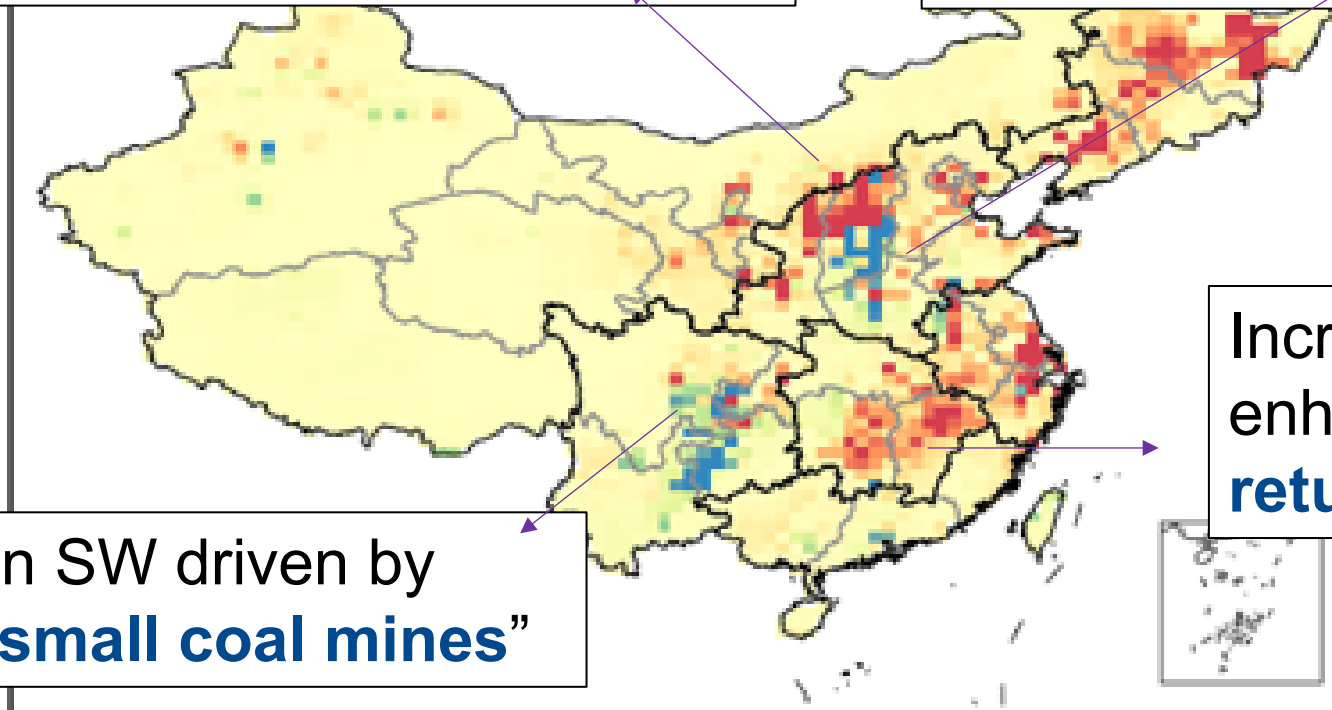
<https://doi.org/10.57760/sciencedb.02269> Zhang et al., PNAS, 2022

Observed regional emission trends linked to energy, agricultural, environmental policy

2010-2017 methane trend

Sustained increase consistent with “**consolidation of large mines**”

Decrease related to **coalbed methane development** in Qinshui Basin

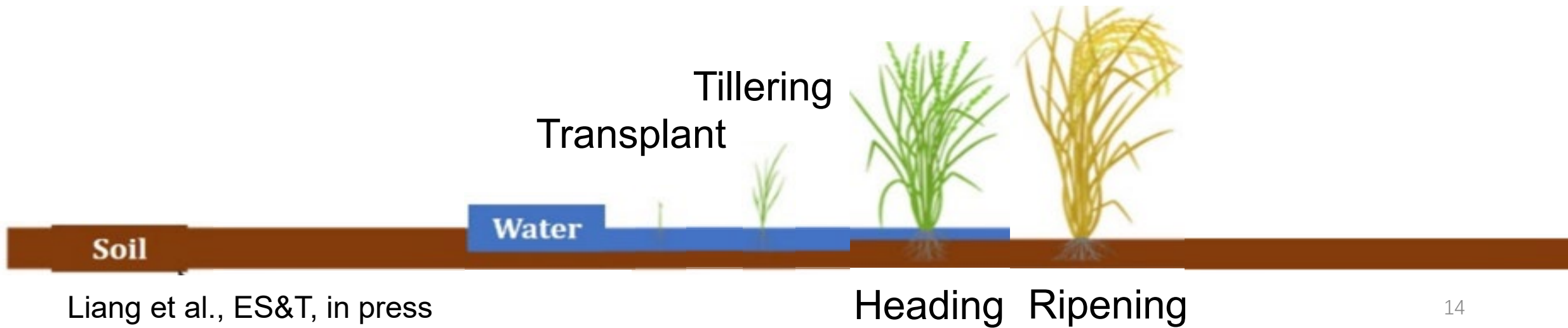
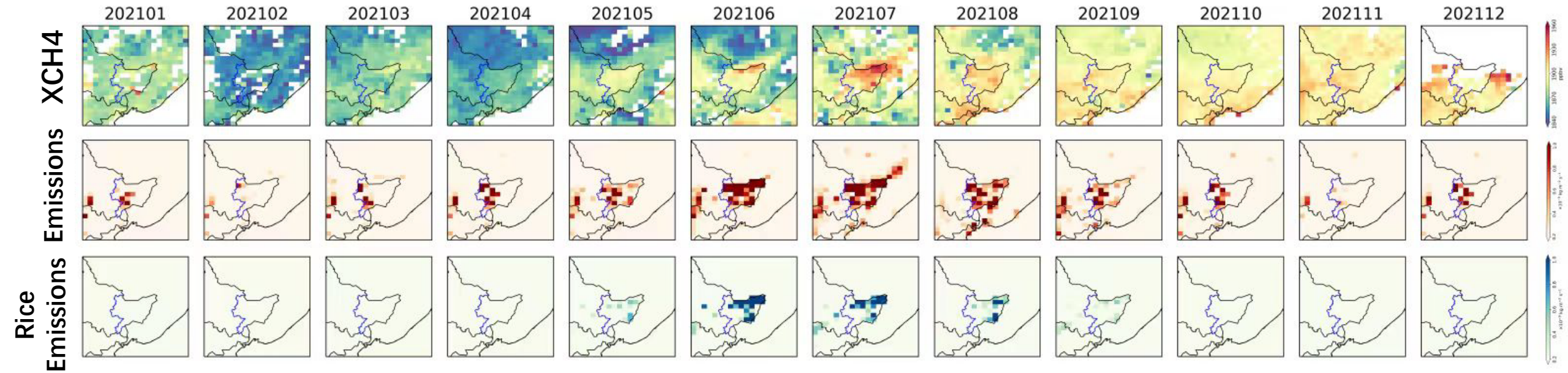


Decrease in SW driven by “**close off small coal mines**”

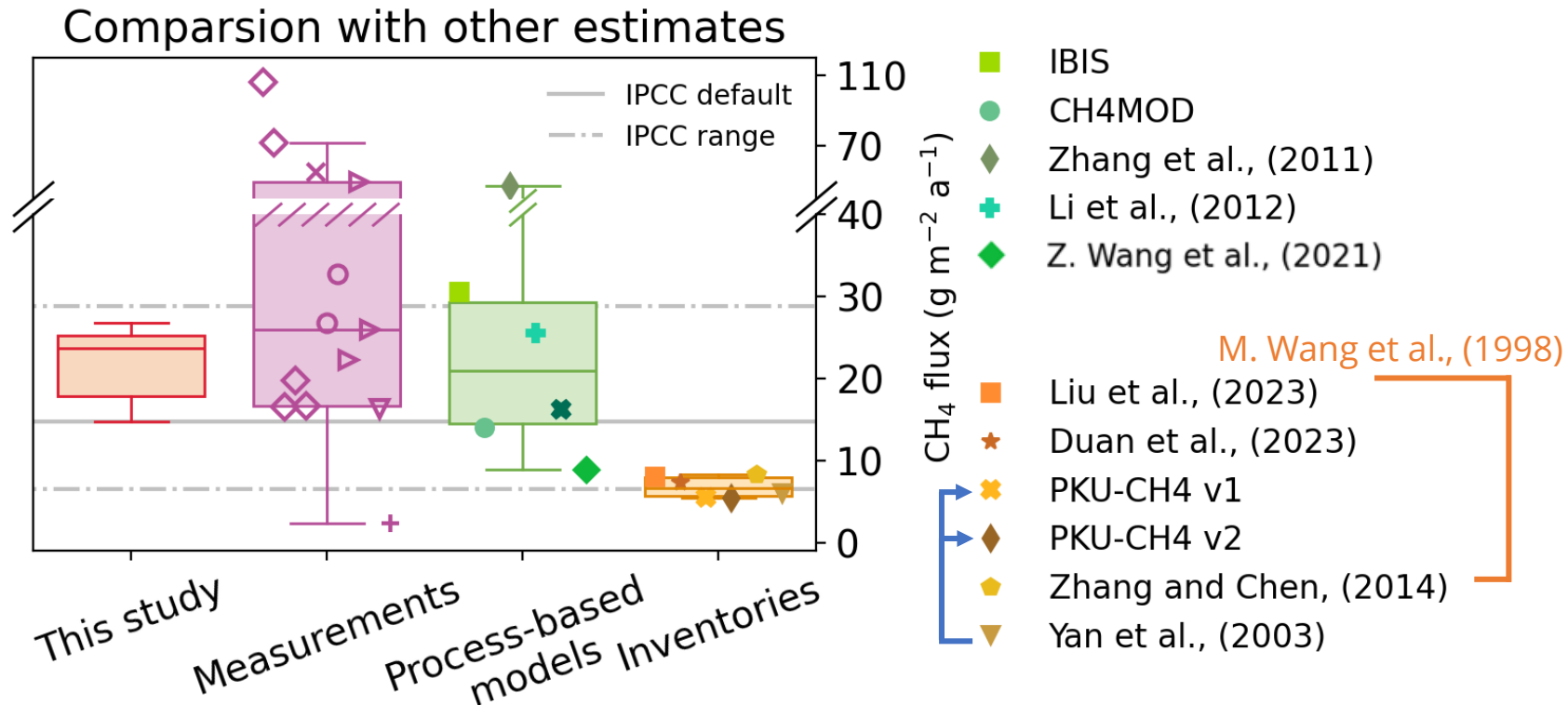
Increase explainable by enhanced **crop residue return**

$\text{Mg km}^{-2} \text{a}^{-2}$ -0.8 -0.4 0.0 0.4 0.8

In-depth understanding of key methane sources



Reduction of uncertainty in regional rice methane emissions



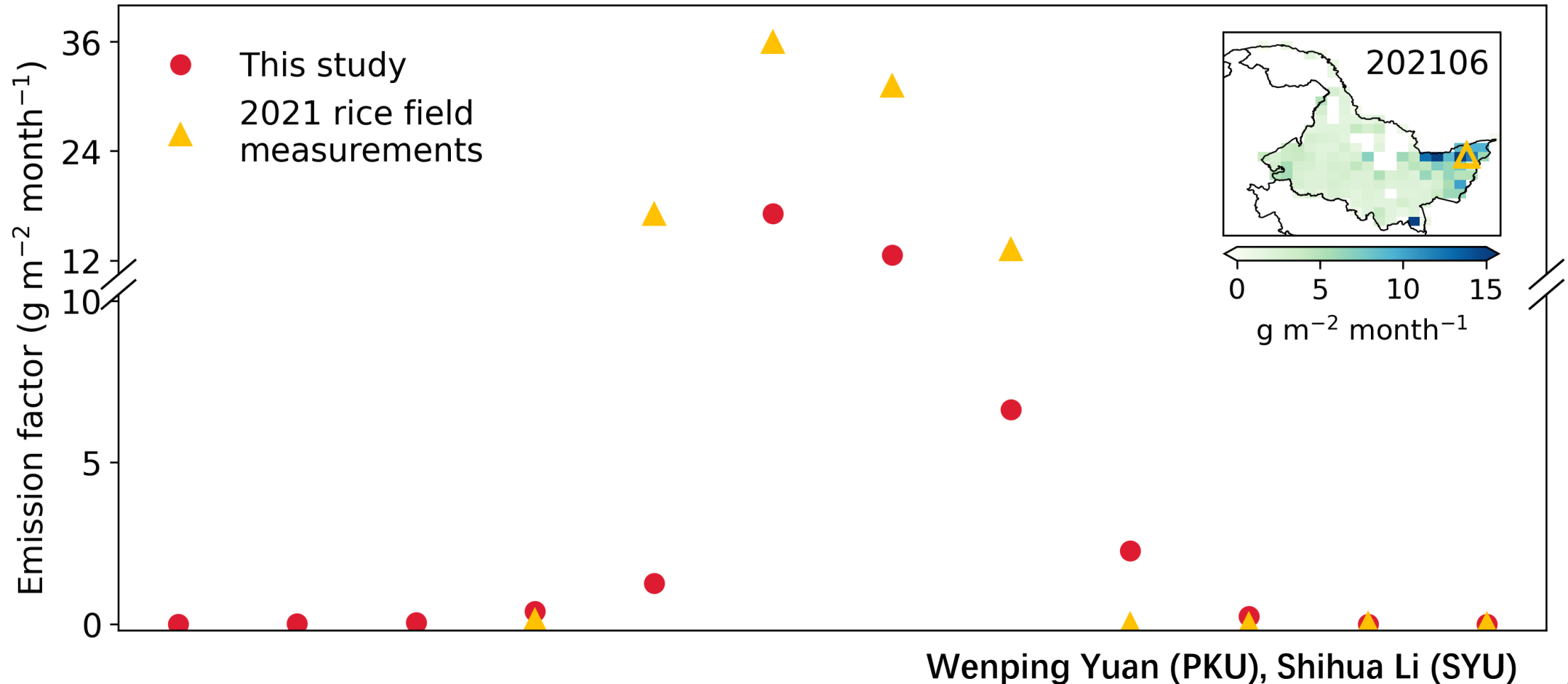
Annual averaged emission factor: 23.0
(14.7 – 26.9) $\text{g m}^{-2} \text{a}^{-1}$: Compatible with
IPCC default, **measurements** and **process
models** with **reduced uncertainty**

Emission factors used in **global/national
inventories** based on sparse and outdated
data (Yan et al, 2003; M. Wang et al., 1998)

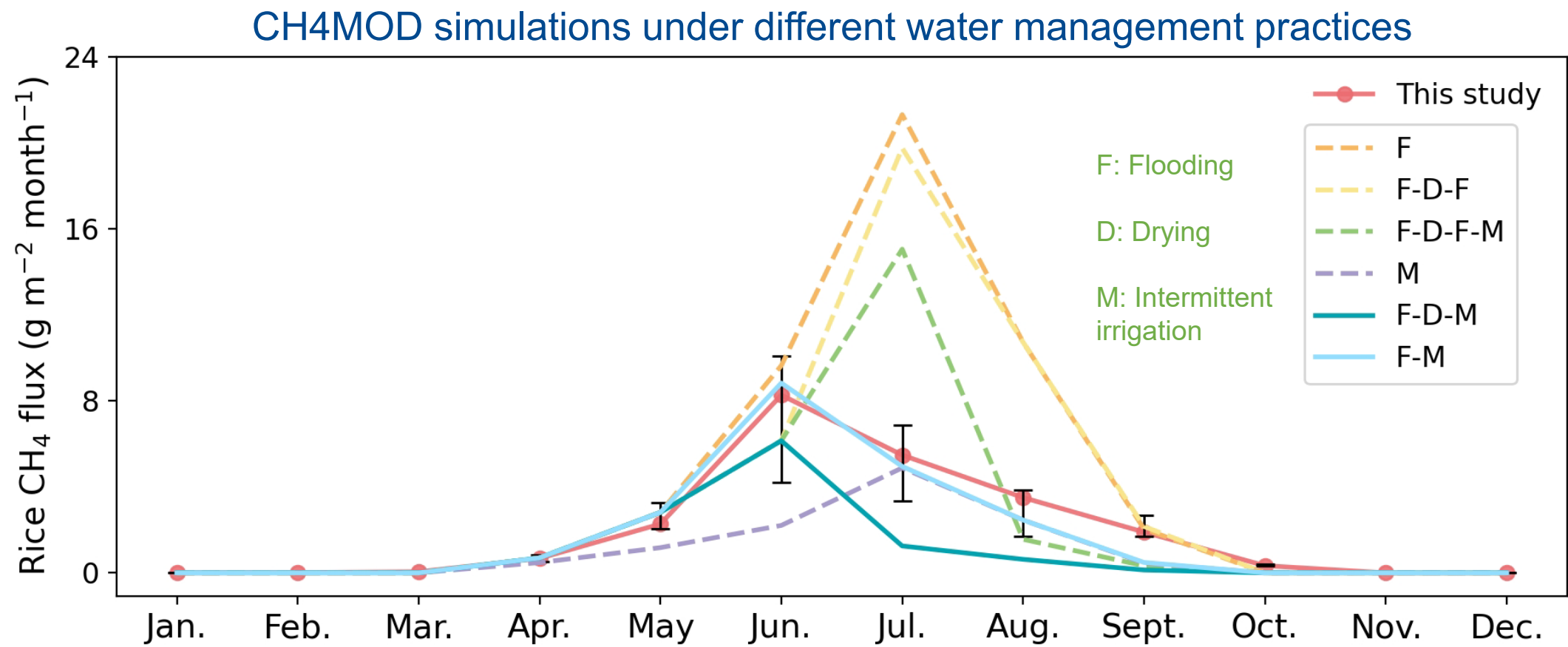
Verification of key parameters for process-based modeling

Infer regional water management mode based on seasonality

(b) Comparison of this study with rice field measurements



Water management controls emission magnitude and seasonality



Liang et al., ES&T, in press

In collaboration with Qiwen Hu (SYU), Tingting Li (CAS)

How to distinguish contributions of different urban sources?

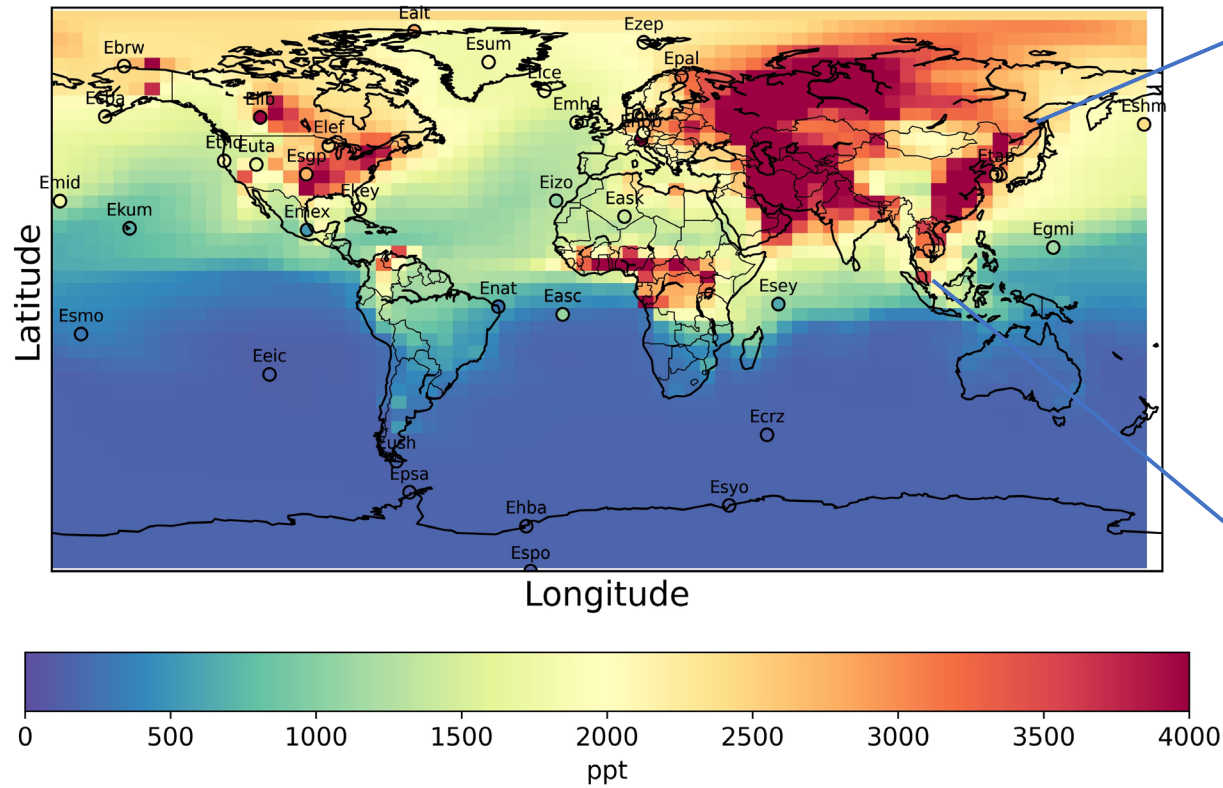


| | CH ₄ source | C ₂ H ₆ source |
|----------------------------|------------------------|--------------------------------------|
| Fossil fuel exploitation | ✓ | ✓ (Usually away from cities) |
| Natural gas end use | ✓ | ✓ |
| Rice cultivation | ✓ | |
| Landfills | ✓ | |
| Wastewater | ✓ | |
| Biomass burning | ✓ | ✓ (less in cities) |
| Livestock | ✓ | |

Joint CH₄-C₂H₆-¹³CH₄ simulation

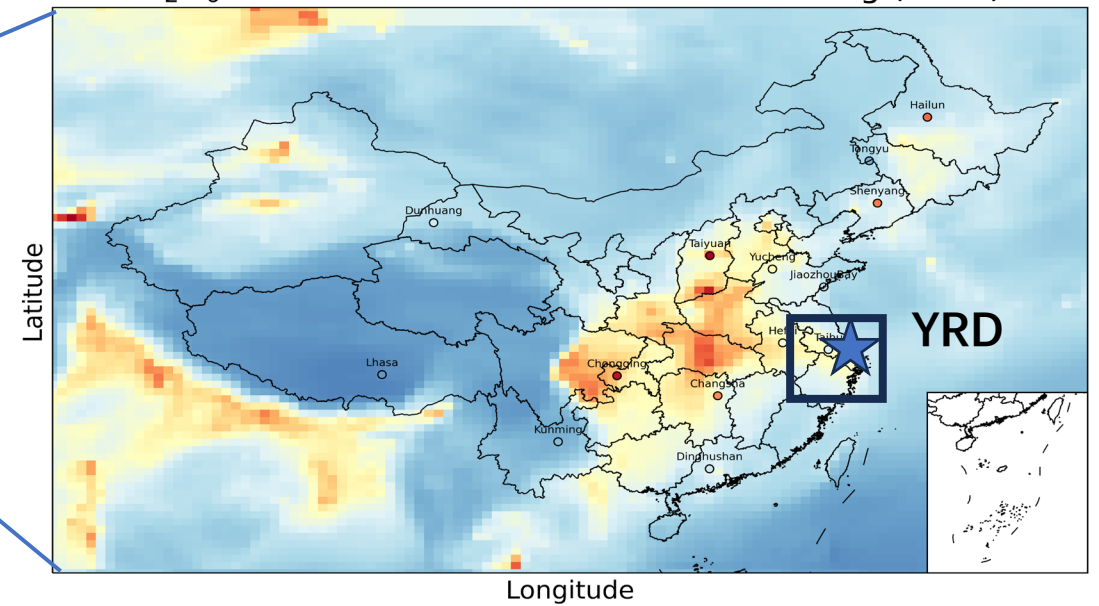
Hypothesis: Increase in methane emissions over YRD driven by natural gas consumption

Global 4x5 simulation (201201)



MB: 0.12 ppb RMSE: 1.17 ppb

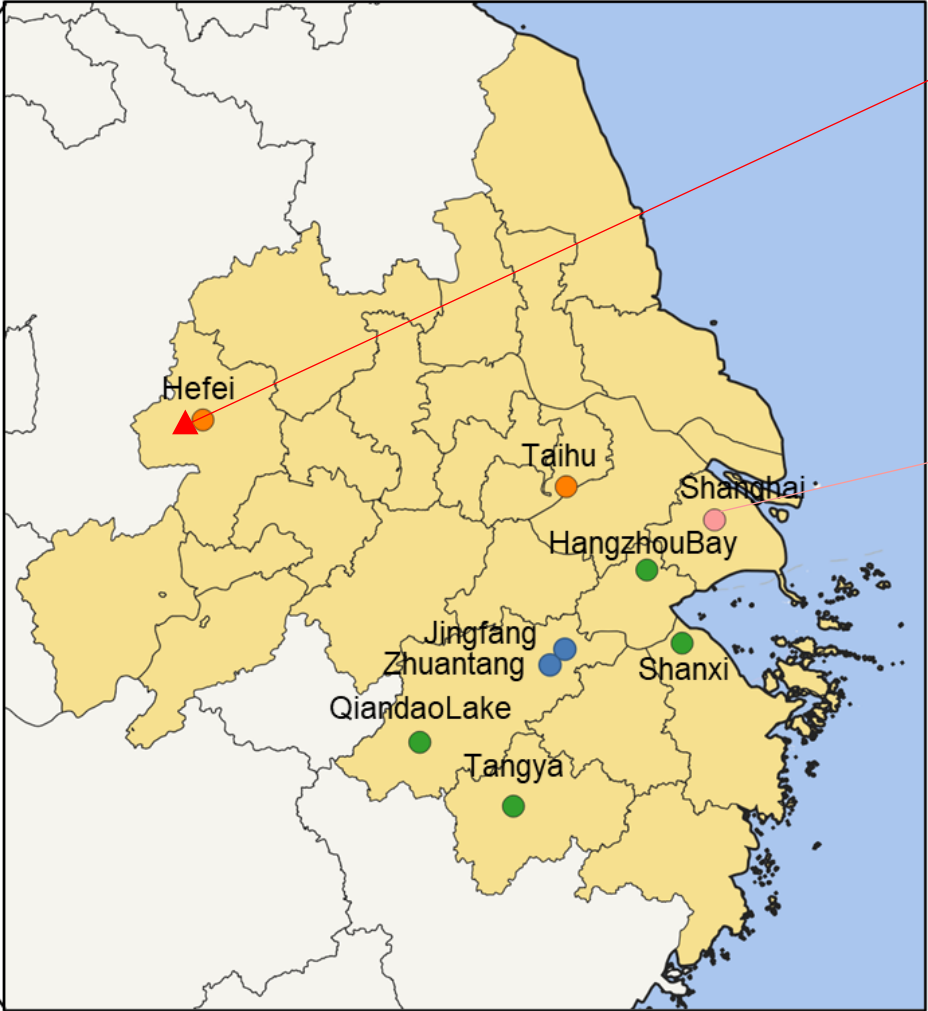
Nested 0.5x0.625 simulation (2012 heating season)



MB: 0.59 ppb RMSE: 1.15 ppb

In collaboration with Yanli Zhang

Atmospheric ethane observations



2015-2022
中科大
(remote sensing)

2012-2013
广州地化所

2011-2019
上海环科院

2016-2020
浙江监测站/浙大

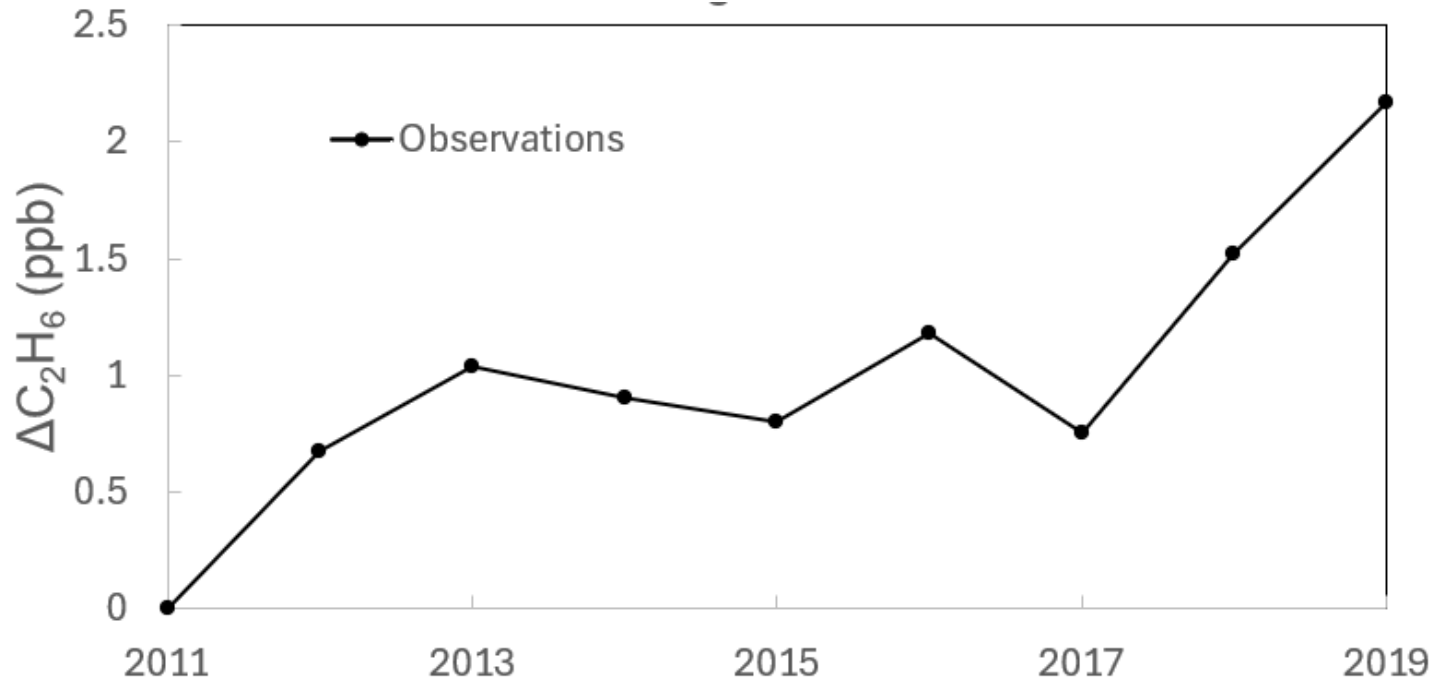
2022
杭州监测站

Increase in urban C₂H₆ concentrations

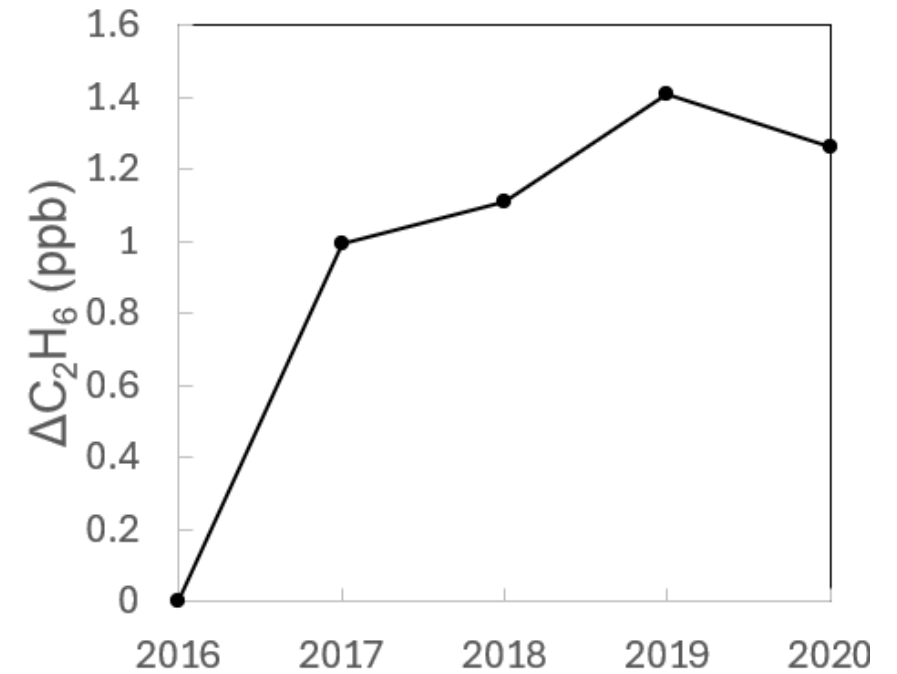
Preliminary

In collaboration with Cheng Huang (Shanghai), Zhenning Xu (Hangzhou)

Shanghai (cold season average)



Hangzhou (cold season average)



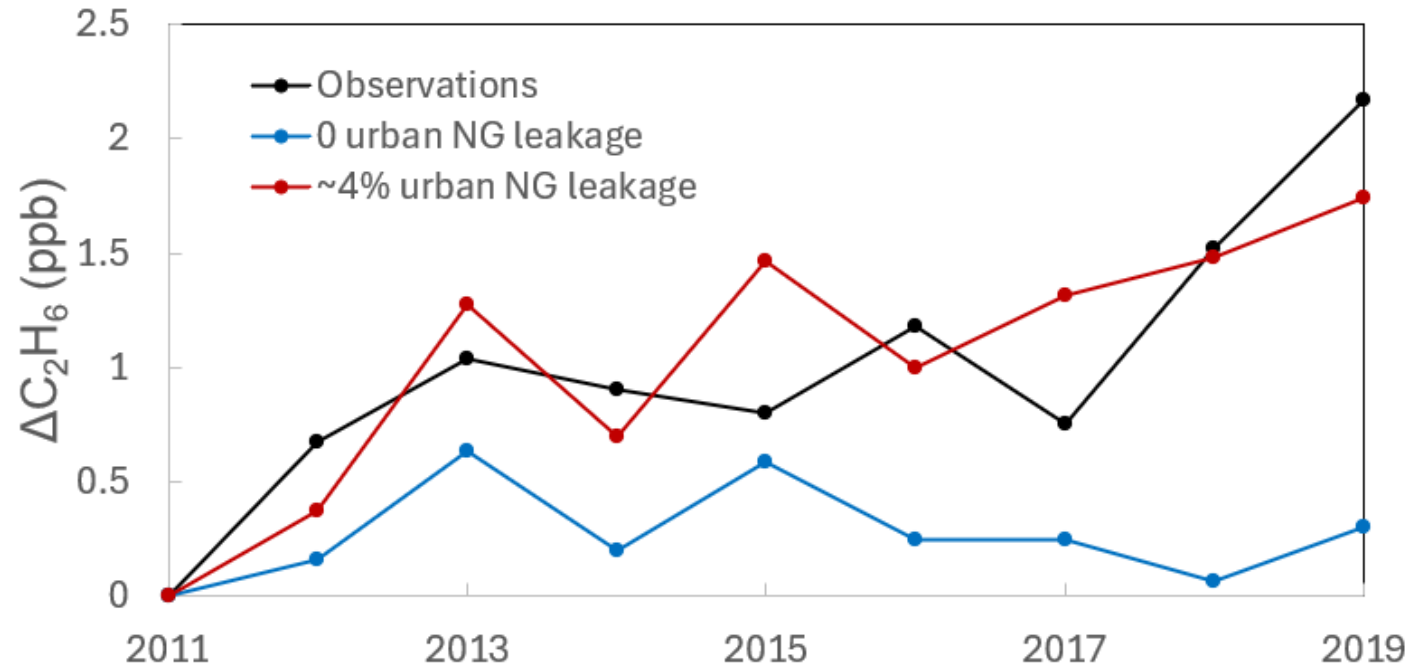
The trend is best matched with **4% leakage rate** from NG consumption

Increase in urban C_2H_6 concentrations

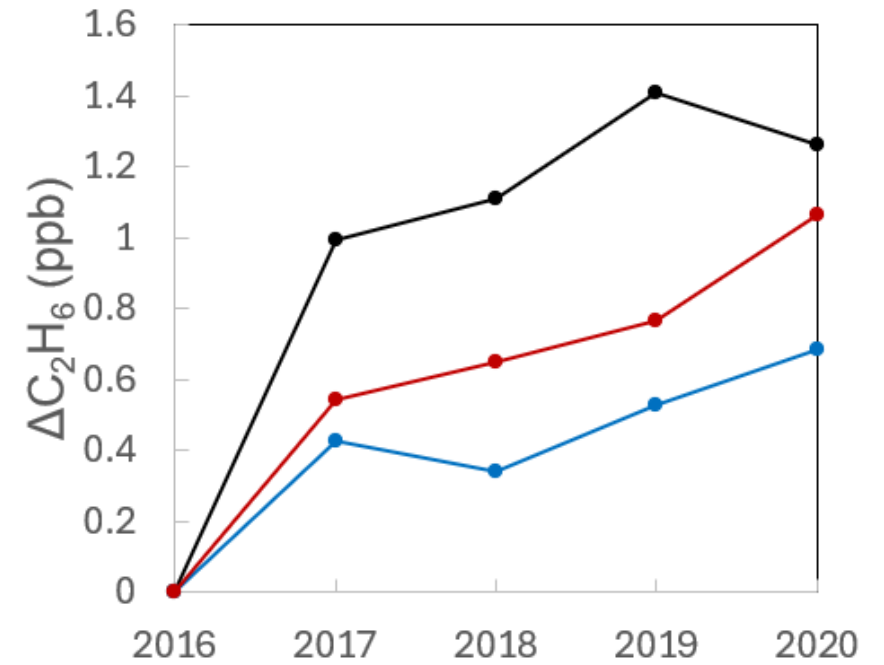
Preliminary

In collaboration with Cheng Huang (Shanghai), Zhenning Xu (Hangzhou)

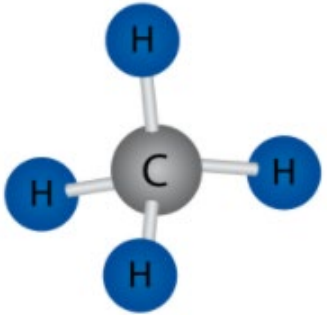
Shanghai (cold season average)



Hangzhou (cold season average)



The trend is best matched with **4% leakage rate** from NG consumption



Summary

- Satellite methane observations are useful for quantifying **global and regional methane emissions**.
- Combining methane with other species, such as **NH₃ (livestock)** and **ethane (natural gas)** are helpful for **source attribution**.
- Combining observation-based inversion with **process-based modeling provides** insight into key factors controlling methane emissions