

A32A-03

# Assessing China's Methane Emissions with Surface and Satellite Observations

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Group Website

# Acknowledgement

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#### Funding

Ministry of Science and Technology of China National Key Research and Development Program of China (2020YFA0607502) National Natural Science Foundation of China

Project 42007198



## **Methane emissions in China**

Country with the largest anthropogenic emissions





## **Trends in China's methane emissions**



Bottom-up inventory



#### Provincial-level activity data

Changes may not be accounted for in the bottom-up inventories

Coarse resolution Inaccurate prior sectoral distribution



# **Uncertainty in sectoral distribution**

Affect inference of methane fluxes as well as sector attribution

#### Distribution of coal mine emissions from different bottom-up inventories



Sheng et al. (2021) derived a trend (0.4 Tg a<sup>-2</sup>) smaller than Miller et al. (2019) (1.1 Tg a<sup>-2</sup>)

## Joint inversion of satellite & surface observations

#### **Observations** – satellite + surface network (2010-2017)

- GOSAT CO<sub>2</sub> proxy retrieval from University of Leicester
- Surface observation
  Ann
- Mainland China CMA 7 sites
  - ◆ Surrounding area WDCGG 6 sites



Better spatial coverage

Better measurement precision Better sensitivity to surface emissions

Zhang et al. in prep

#### **Bottom-up estimates**

Multiple anthropogenic emission inventories are used as prior for (sensitivity) inversions to explore uncertainties

Name	Anthropogenic emissions	Emissions (Tg a <sup>-1</sup> )	Trend (Tg a <sup>-2</sup> )
E1	EDGAR v4.3.2 for 2012, except for coal (Sheng et al., 2019)	59	0
E2	PKU 2010-2017	49	-0.3
E3	EDGAR v5.0 2010-2015	60	+0.4
E4	CEDS v2021-04-21	51	+0.2

#### Natural emissions are not perturbed in the ensemble

Natural	Wetland	WETCHARTs
Emissions	Biomass burning	GFED4
	Geological seeps	Maasakkers et al.
	Termites	Fung et al.



## China's methane emissions and emission trends

#### 2010-2017 mean methane emissions



#### 2010-2017 methane emission trends

Zhang et al. in prep



## Added value of surface observations

Surface observations provide additional constraints, particularly over South, East and Northeast China



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## Sectoral attribution of inferred trends, 2010-2017



# **Compare inversion of GOSAT and TROPOMI data**

Do TROPOMI and GOSAT provide consistent inversion results? How does the TROPOMI bias correction affect emission inference? Does TROPOMI help resolve finer spatial distribution?



Ruosi Liang



Mean bias corrected TROPOMI methane in 2019







### **Inversion TROPOMI**

## **Inversion GOSAT**



Consistent: NEC, CSC, BAN

Inconsistent: KAZ, XJC, IND, EC

# Impact of post bias correction for TROPOMI inversion

### **Inversion T1 (bias-corrected)**





## **Inversion T4 (un-corrected)**





kg/m2/s



## **Comparison of inversion results: KAZ**



There appears to be other factors that needs to be included in the bias correction

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- Joint inversion of surface and satellite data to infer China's methane emissions during 2010-2017.
- Surface observations add more information to satellite data, especially over southern and northeastern China.
- Unexpected increase in rice cultivation regions; uncertain & divergent trends from the coal sector.
- TROPOMI and GOSAT inversions show generally consistent correction patterns though with varied details



