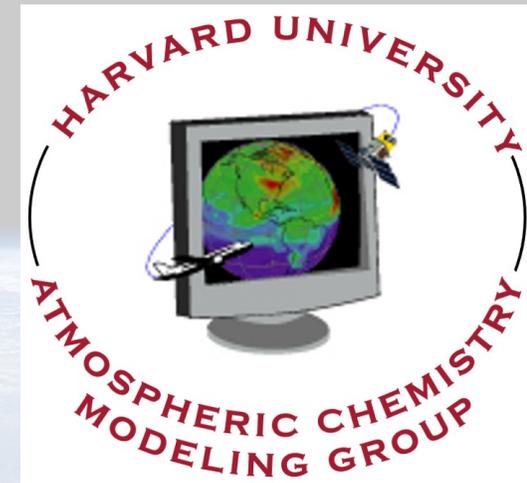


# Methane emissions from China: a high-resolution inversion of TROPOMI satellite observations



**Zichong Chen**

Chen, Z., Jacob, D., Nesser, H., Sulprizio, M., Lorente, A., Varon, D., Lu, X., Shen, L., Qu, Z., Penn, E., and Yu, X.: Methane emissions from China: a high-resolution inversion of TROPOMI satellite observations, *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2022-303>, 2022.



# China plays a large but uncertain role in global methane emissions. 46-74 Tg a<sup>-1</sup> out of a global anthropogenic source of 349-393 Tg a<sup>-1</sup>

Largest coal producer



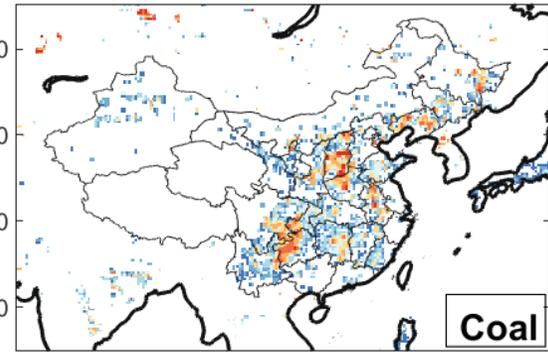
Coal-to-gas transition;  
15% energy supply by 2030



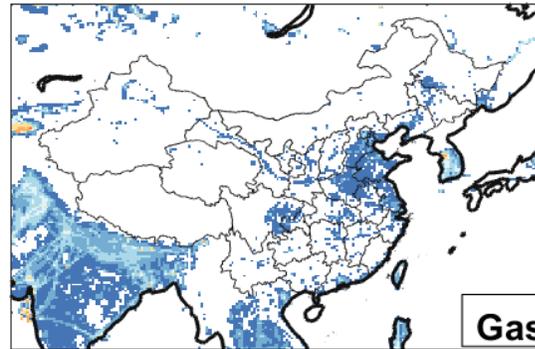
Major livestock production



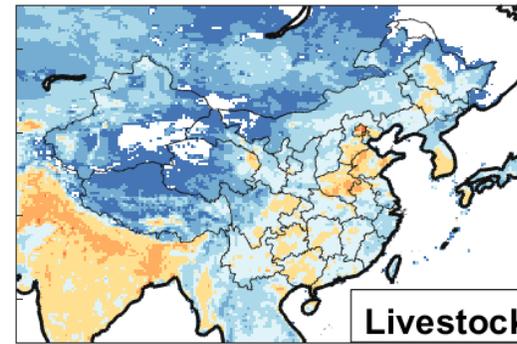
Rice paddies



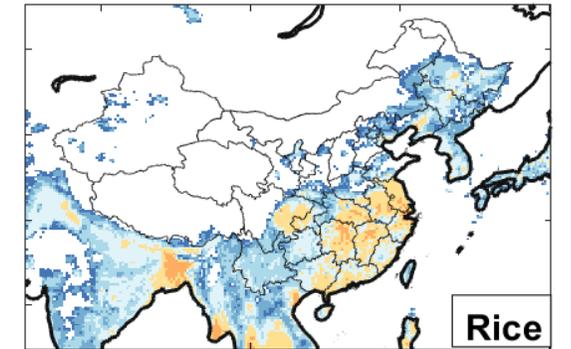
Coal



Gas

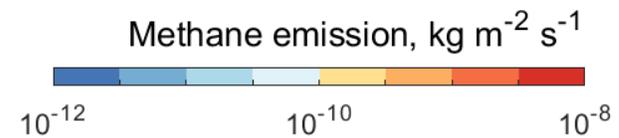


Livestock

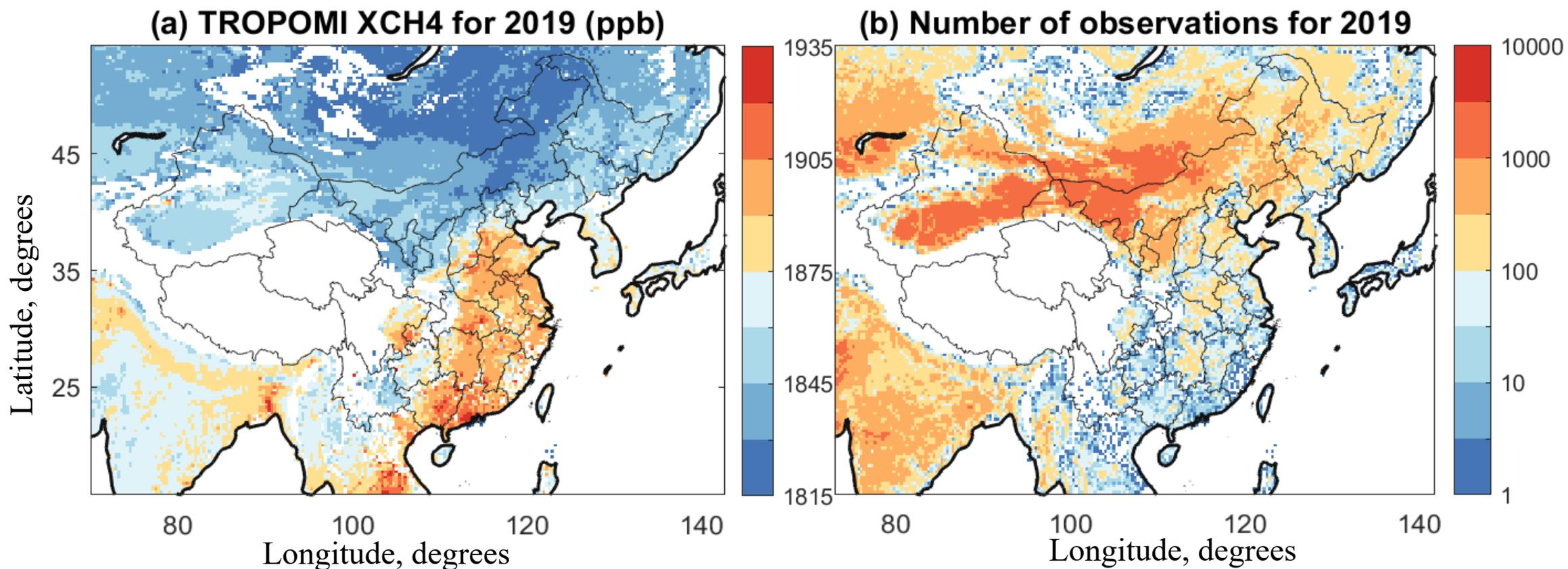


Rice

UNFCCC-based GFEI v2 (*Scarpelli et al., 2022*)



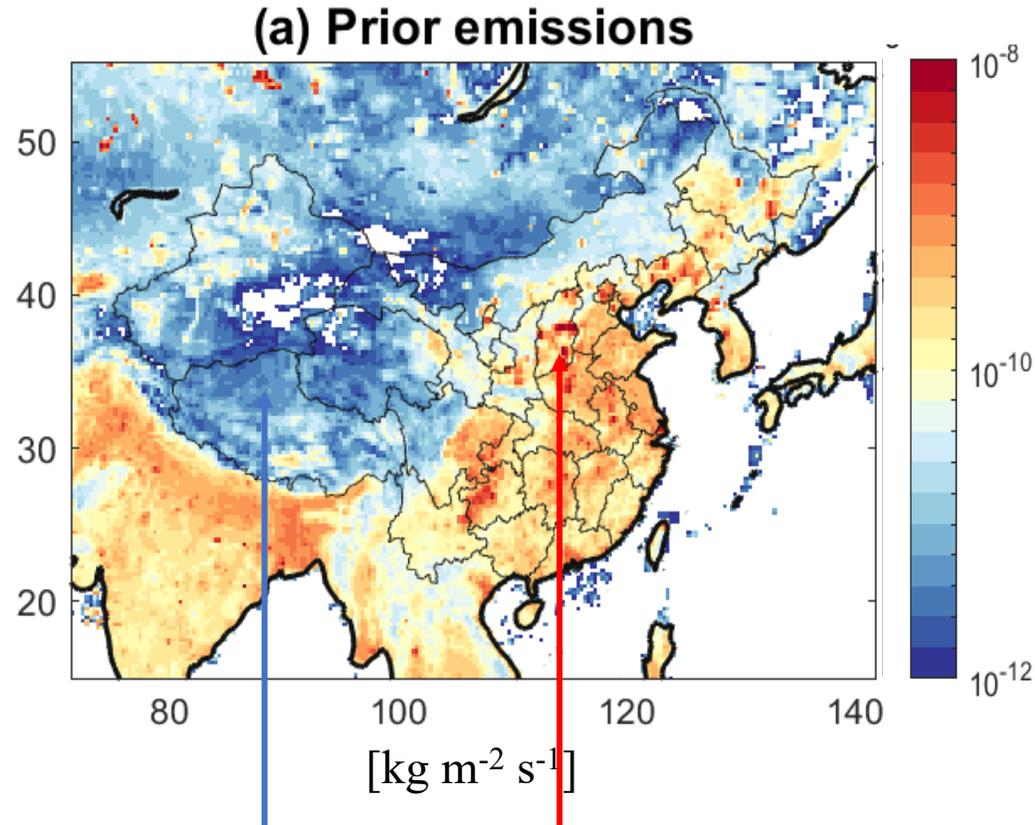
# Using TROPOMI satellite data to infer emissions



- Data available since May 2018
- Footprint  $7/5.5 \text{ km} \times 7 \text{ km}$
- 4% retrieval success limited by clouds, surface heterogeneity, aerosols

- Quality flag  $>0.5$
- Surface altitude  $<2000 \text{ m}$
- Filtering for snow-covered scenes
- Comparing with GOSAT

# Analytical inversion setups



**1. retaining native resolution for strong localized source**

**2. merging weak source regions**

1. GEOS-Chem forward model (v13.0.0), GEOS-FP  $0.25^\circ \times 0.3125^\circ$  met field.

2. **Oil, gas, and coal** are from UNFCCC-based GFEIv2, Other anthropogenic emissions (livestock, landfills, wastewater, rice) are from UNFCCC with allocation from EDGARv432; Wetland are from WetCHARTs (*Ma et al.*, 2021).

3. **Gaussian mixture model (GMM)** of *Turner and Jacob* (2015) with 600 state vector elements to find suitable dimension.

4. **Log-normally distributed errors** for high-tailed emission PDFs.

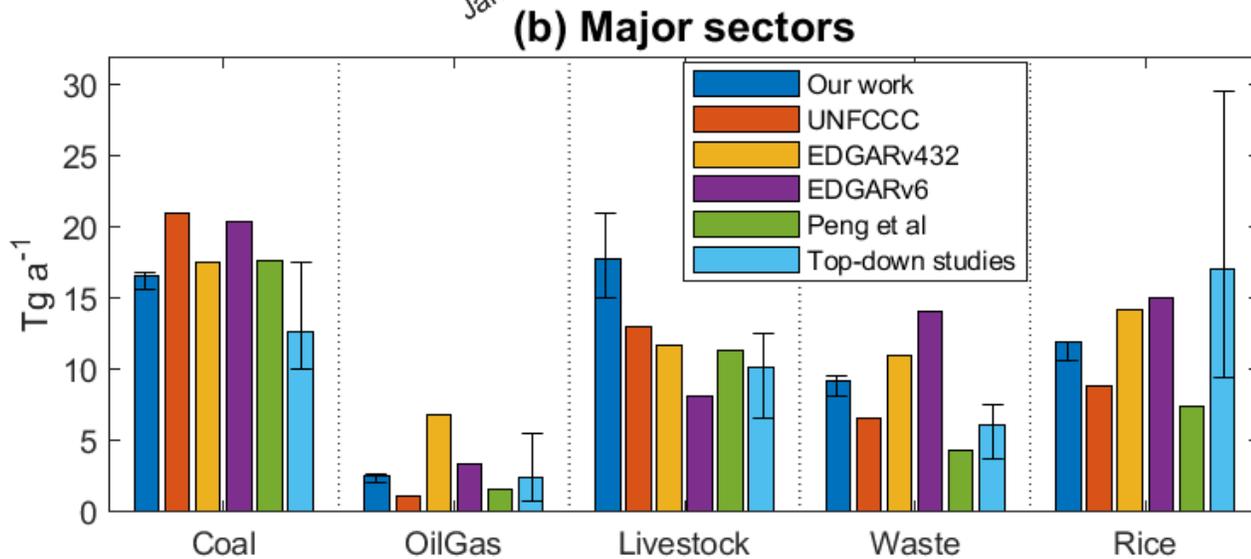
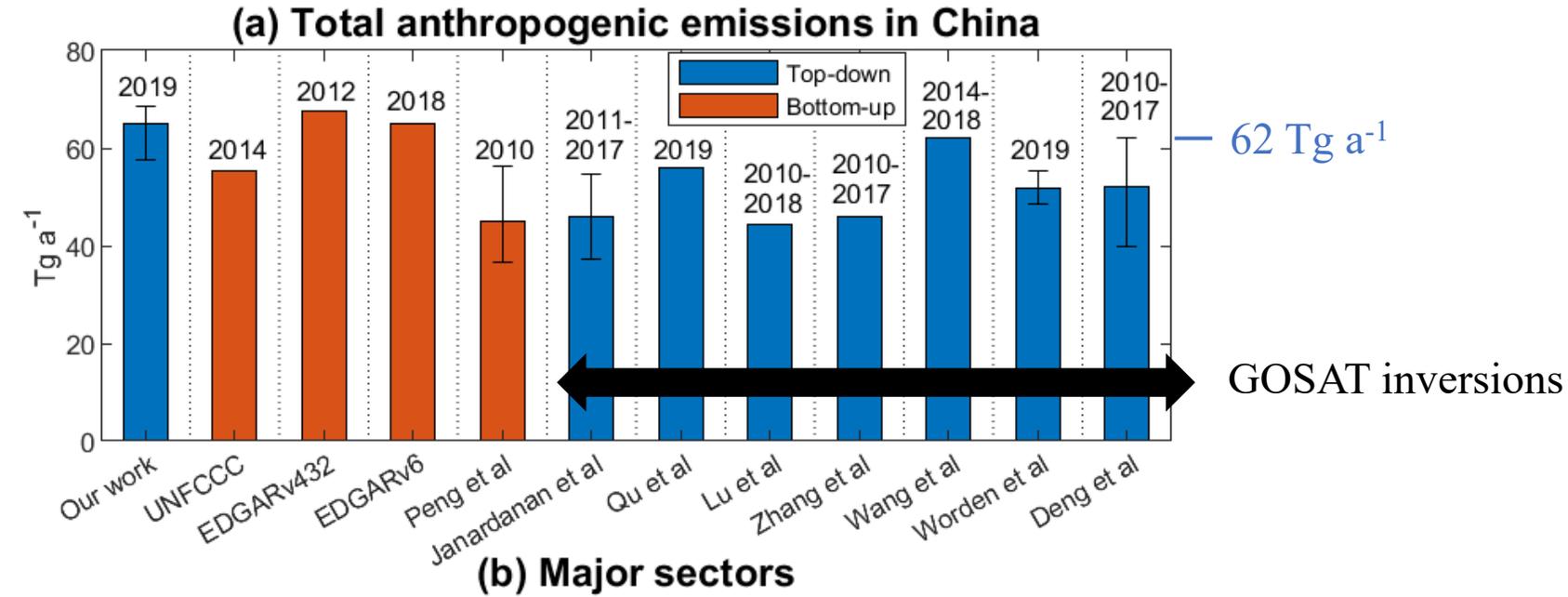
## Underestimate of anthropogenic emissions in the national government report

	<b>Prior estimate (Tg a<sup>-1</sup>)</b>	<b>Posterior estimate (Tg a<sup>-1</sup>)</b>	<b>Sensitivity to observations</b>
<b>Anthropogenic</b>	<b>53.6</b>	<b>65.0 (57.7-68.4)</b>	0.91
Coal mining	<b>19.5</b>	<b>16.6 (15.6-17.6)</b>	0.91
Oil	<b>0.93</b>	<b>2.3 (1.8-2.5)</b>	0.76
Natural gas	<b>0.18</b>	<b>0.29 (0.23-0.32)</b>	0.30
Livestock	13.0	17.8 (15.1-21.0)	0.75
Waste	6.6	9.3 (8.2-9.9)	0.71
Rice paddies	8.9	11.9 (10.7-12.7)	0.86
Other	4.6	6.7 (5.8-7.1)	0.81

**Downward correction** in coal (-15%).

**Upward correction** in oil (+147%), natural gas (+61%), livestock (+37%), waste (+41%), and rice (+34%).

# Also higher than recent top-down studies

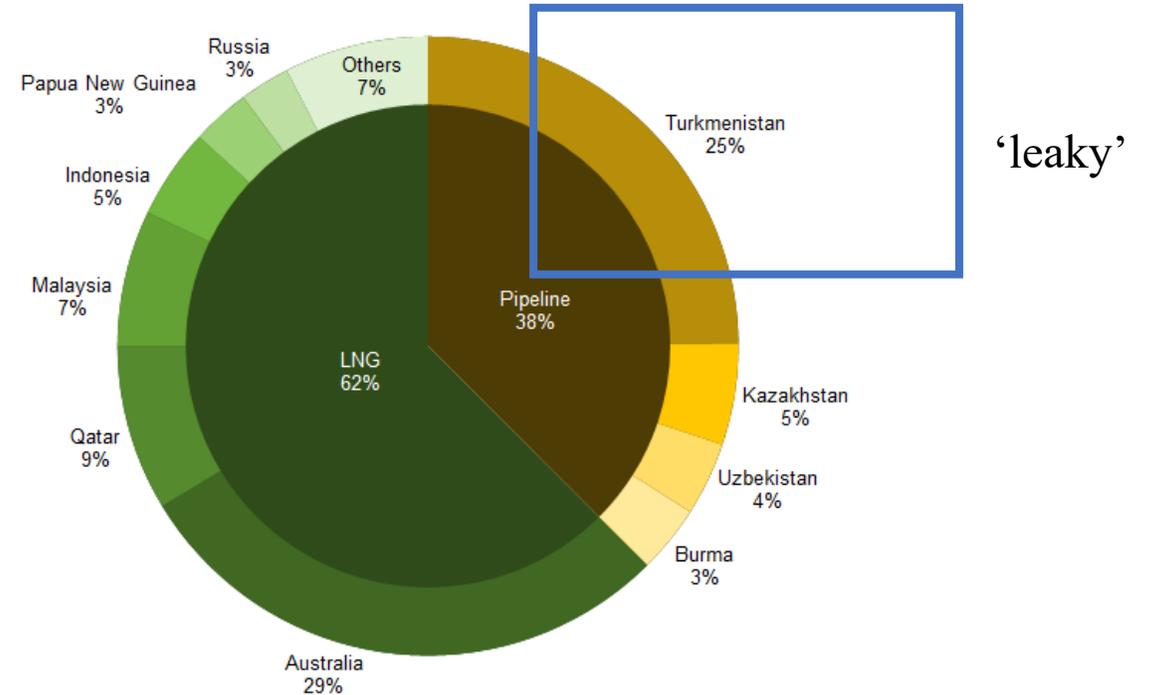


We estimate larger livestock, coal, and waste emissions.

We obtain a small loss rate, does that indicate net climate benefits from the coal-to-gas transition in China?



China imports 42% of its gas.



*Loss rate: emissions per unit of domestic gas production*

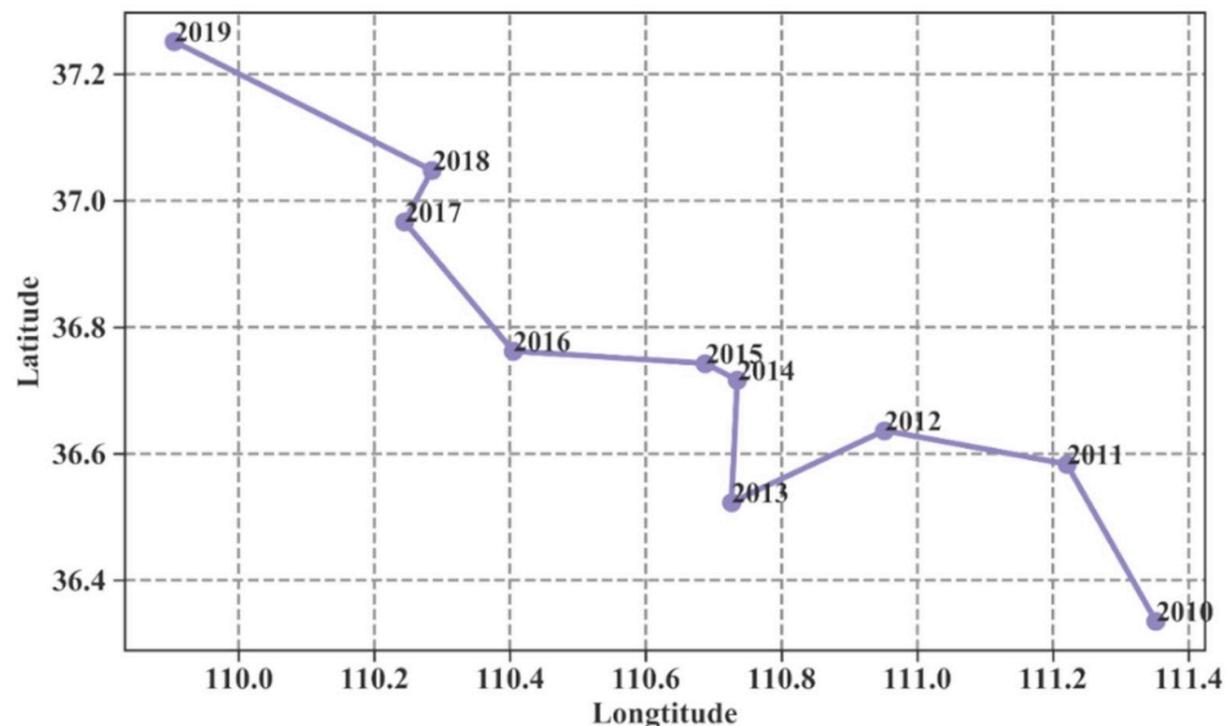
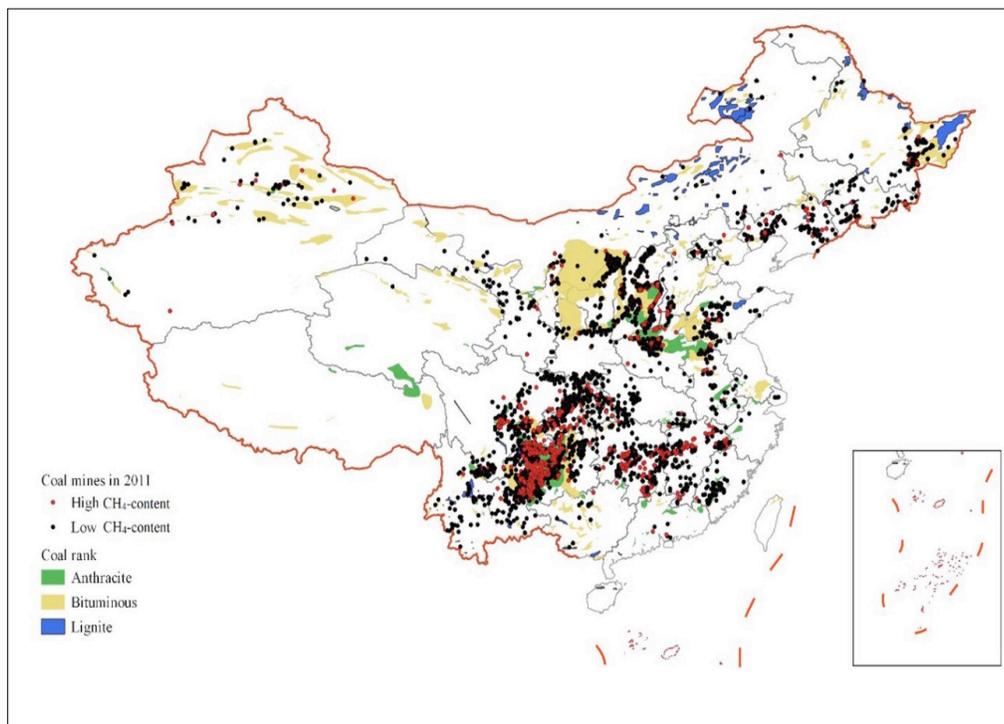
Supply chain break-even loss rate: **3.2%**  
(Alvarez *et al.*, 2012), beyond which  
climate damage of gas is worse than coal.

Our result for China: **1.7 (1.3-1.9) %**

*EIA, 2020*

However, this low loss rate could be somewhat misleading.

# Downward correction of coal emissions



Gao et al., 2020

- Closure of mines with low coal production efficiency.
- Regional shift of coal productions, driven by de-capacity policies in China.  
[more than 6000 coal mines were closed over 2010–2019 due to the de-capacity policies]

**Under-accounting** of surface mines and coal mine methane (CMM) utilization

**Overestimate of EFs**

Coal mining industry is shifting rapidly for economic, safety, and political reasons – a new coal inventory with up-to-date coal emission locations is essential for an effective inversion.

# Conclusions

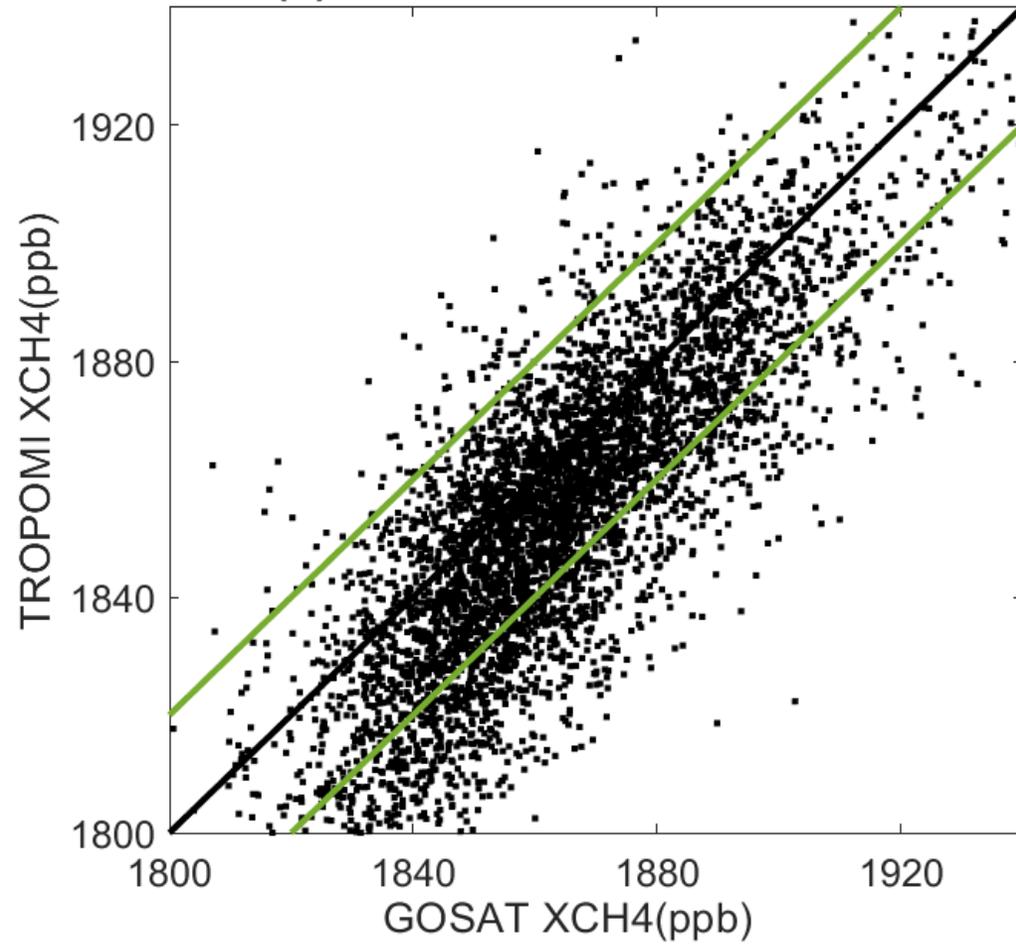
We quantify methane emissions in China and the contributions from different sectors by inverse analysis of 2019 TROPOMI satellite observations of atmospheric methane:

- Our estimate is 21% higher than the Chinese inventory reported to the UNFCCC (53.6 Tg a<sup>-1</sup>), reflecting upward corrections to emissions from oil (+147%), gas (+61%), livestock (+37%), waste (+41%), and rice paddies (+34%), but downward correction for coal (-15%).
- We are better able to separate coal and rice emissions.
- Our higher livestock emissions are attributed largely to northern China where GOSAT has little sensitivity.
- Underestimate of oil emissions in the UNFCCC report appears to reflect unaccounted super-emitting facilities.
- Our estimate of emissions per unit of domestic gas production indicates a low life-cycle loss rate of 1.7 (1.3-1.9) %, which would imply net climate benefits from the current coal-to-gas energy transition in China. However, this small loss rate is somewhat misleading considering China's high gas imports.
- Coal mining industry is shifting rapidly in China, and therefore a new coal inventory with up-to-date coal emission locations is essential for inversion.

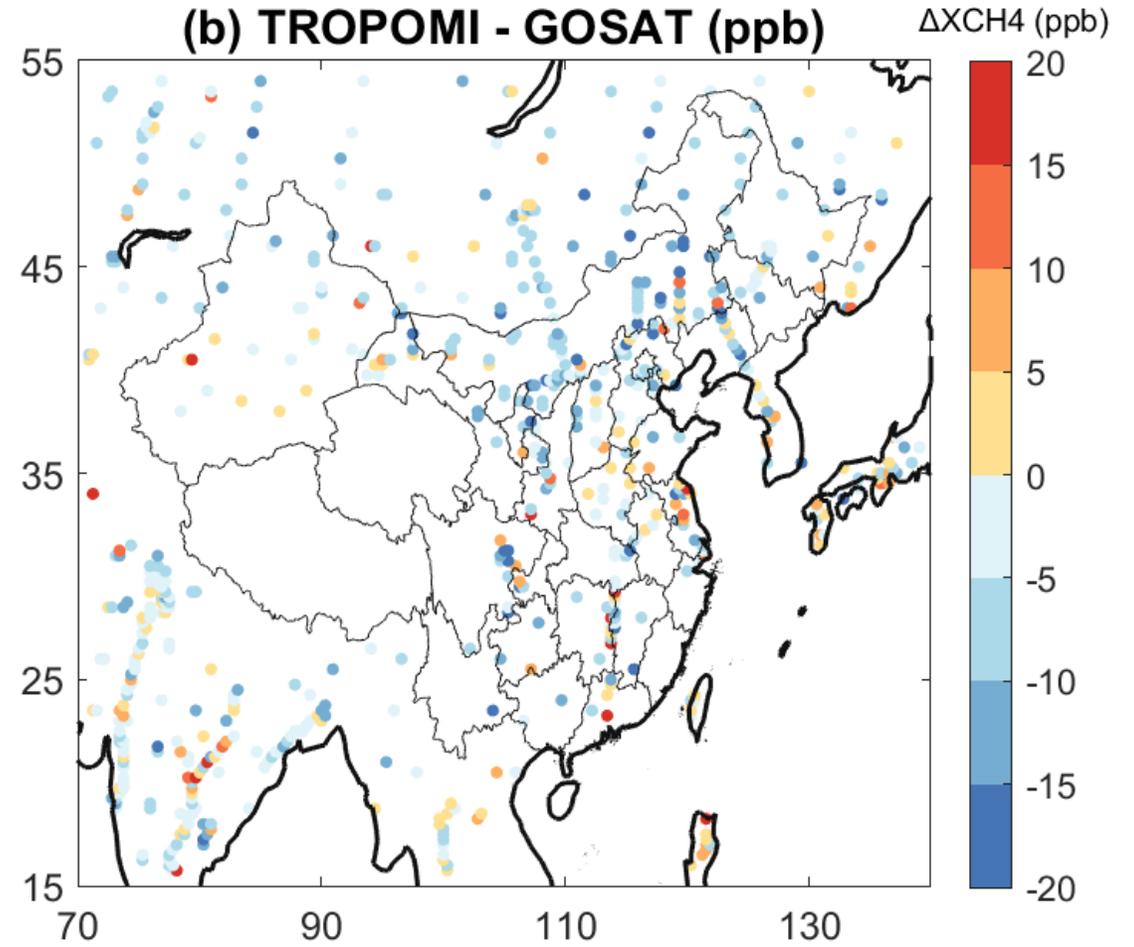
Extra slides

# TROPOMI-GOSAT comparison

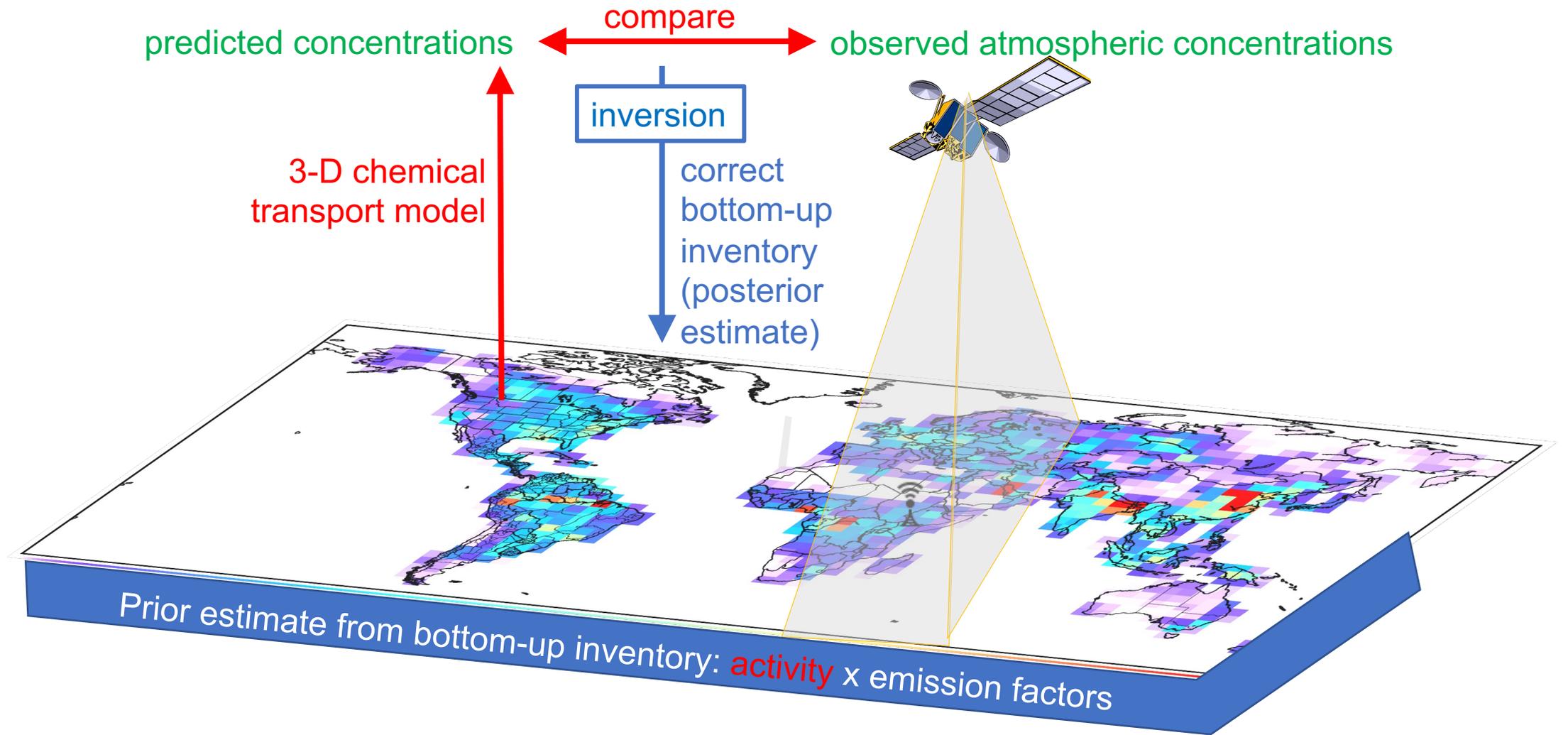
(a) TROPOMI versus GOSAT



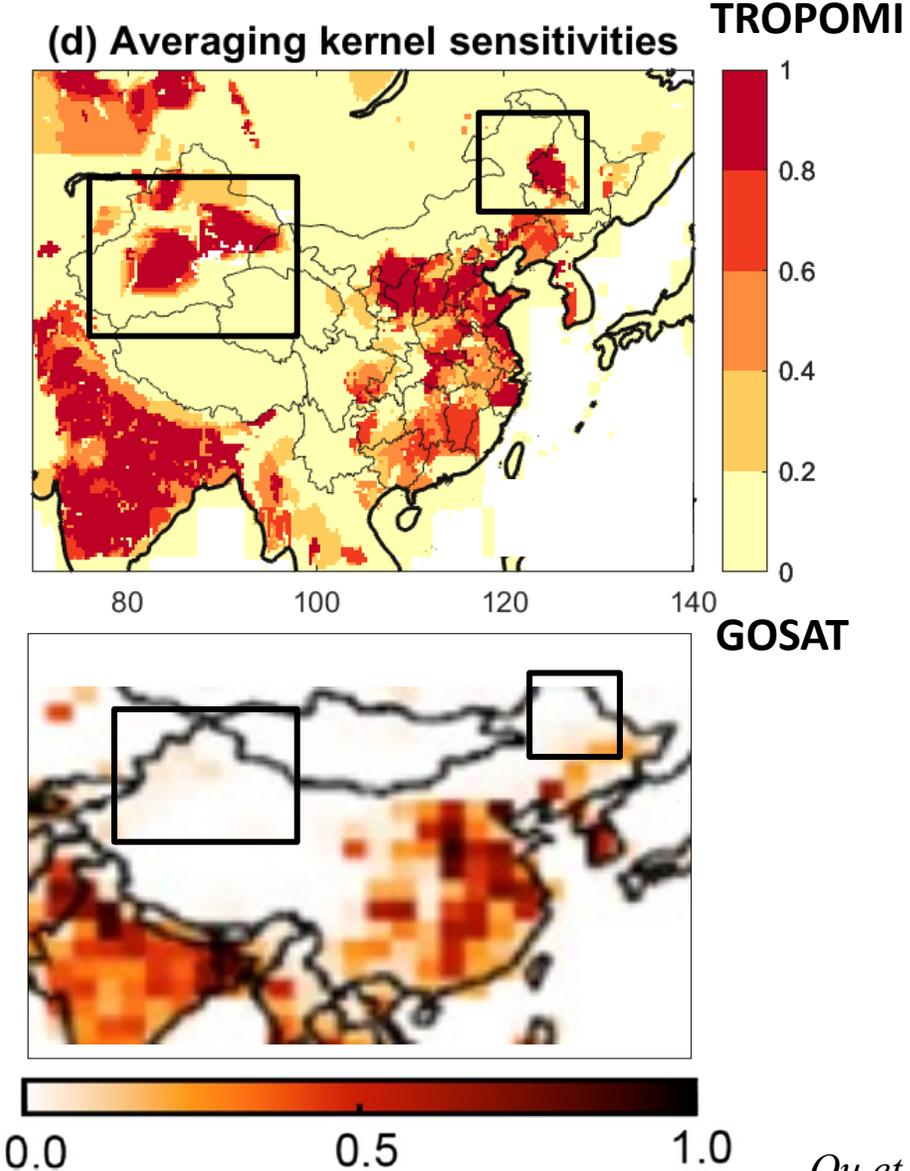
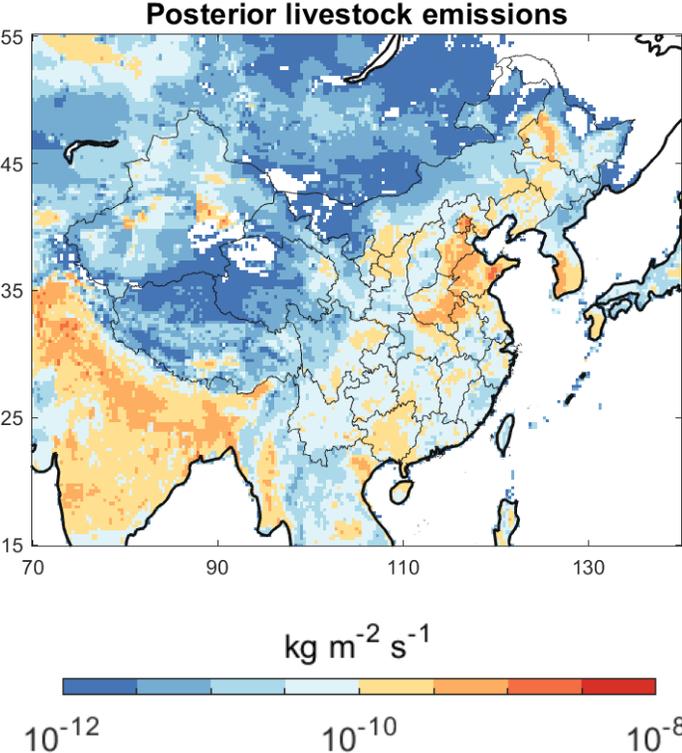
(b) TROPOMI - GOSAT (ppb)



# Using satellite observations to correct bottom-up emission inventories



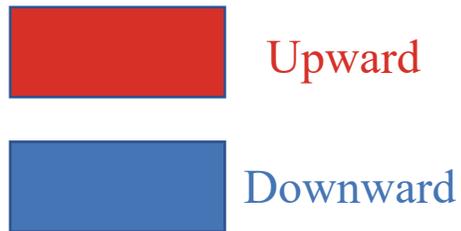
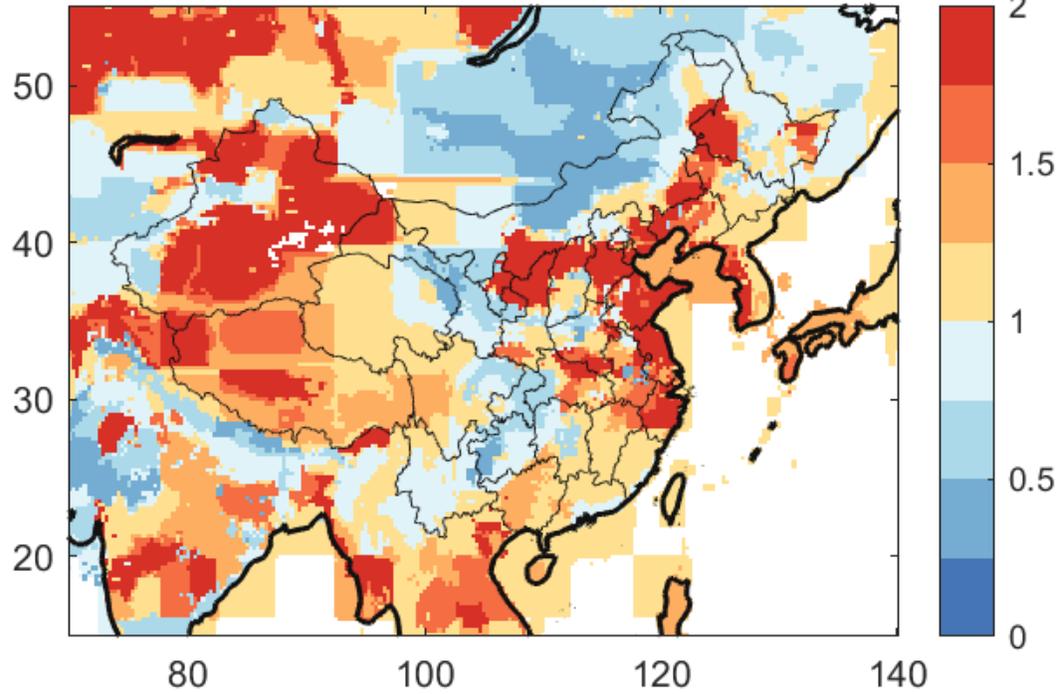
# We have higher **livestock** emissions in northern China, where GOSAT has little sensitivity



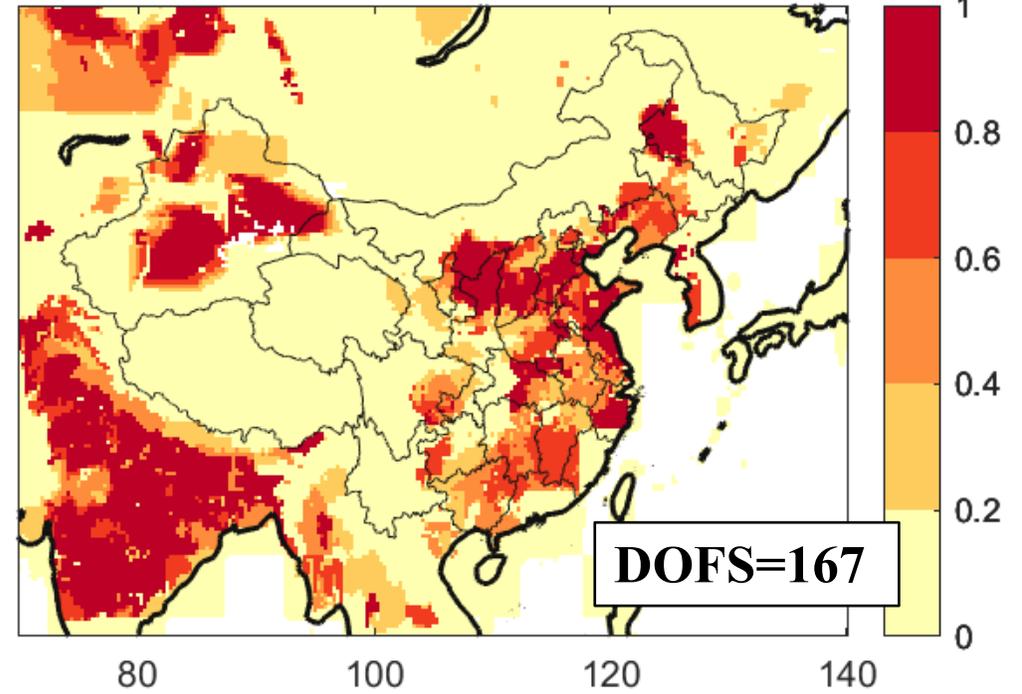
- Dense TROPOMI over **northwestern and northeastern** China enables more effective corrections than previous GOSAT inversions.
- Two regions account for **63%** of the upward adjustment for livestock.

# Strong ability of TROPOM in determining the posterior emissions in major source regions

(c) Posterior/prior emission ratios



(d) Averaging kernel sensitivities



1: fully determined by observations  
0: fully determined by prior information

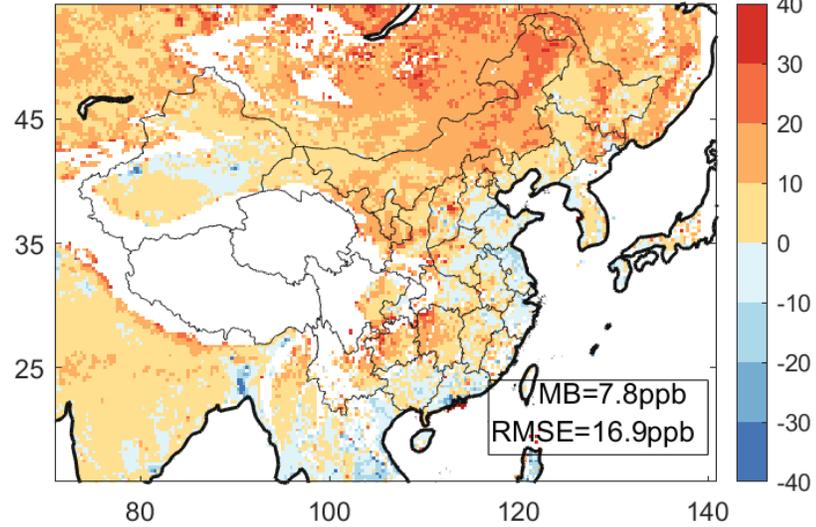
DOFS: degrees of freedom for signal  
DOFS=600 if fully constrained

*Number of independent pieces of information obtained from the observations*

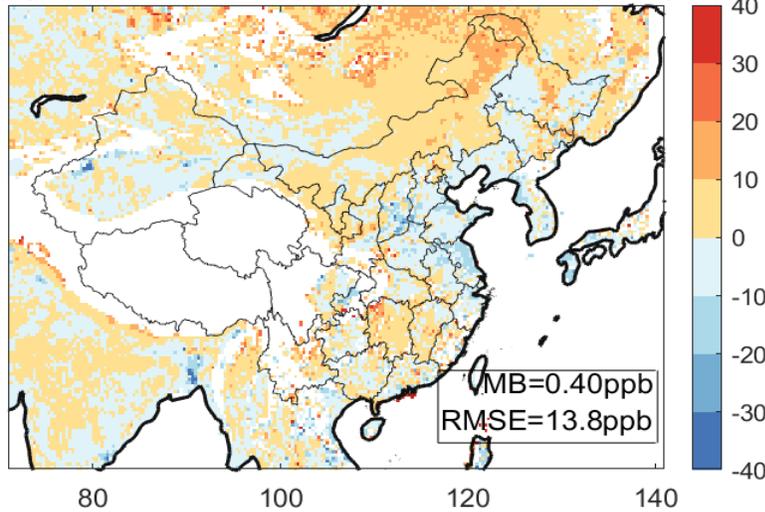
# Improved ability of posterior emissions to fit observations



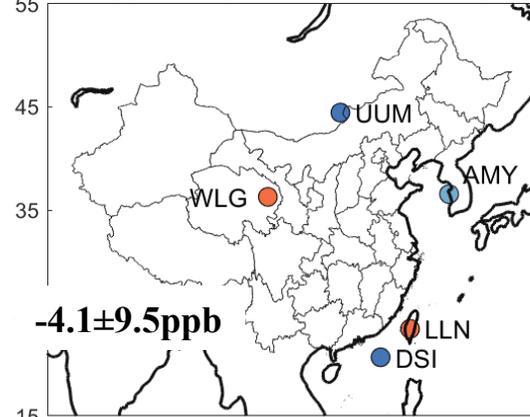
(c) Prior simulation - TROPOMI (ppb)



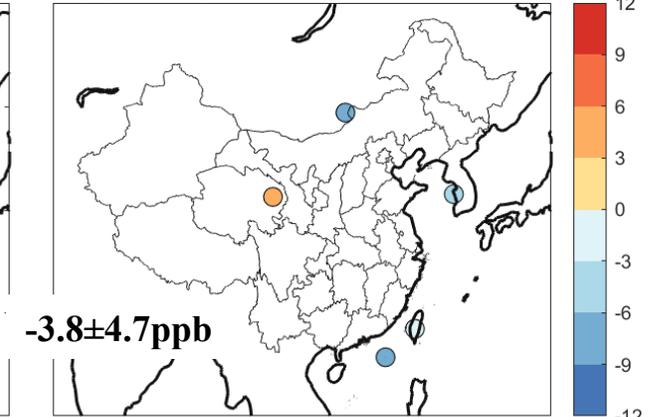
(d) Posterior simulation - TROPOMI (ppb)



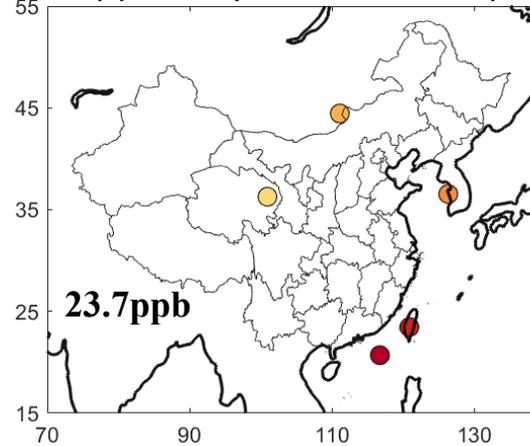
(a) Bias (Prior-Observation)



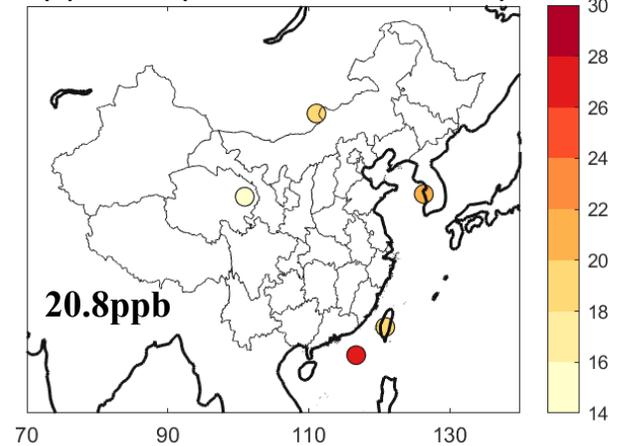
(b) Bias (Posterior-Observation)



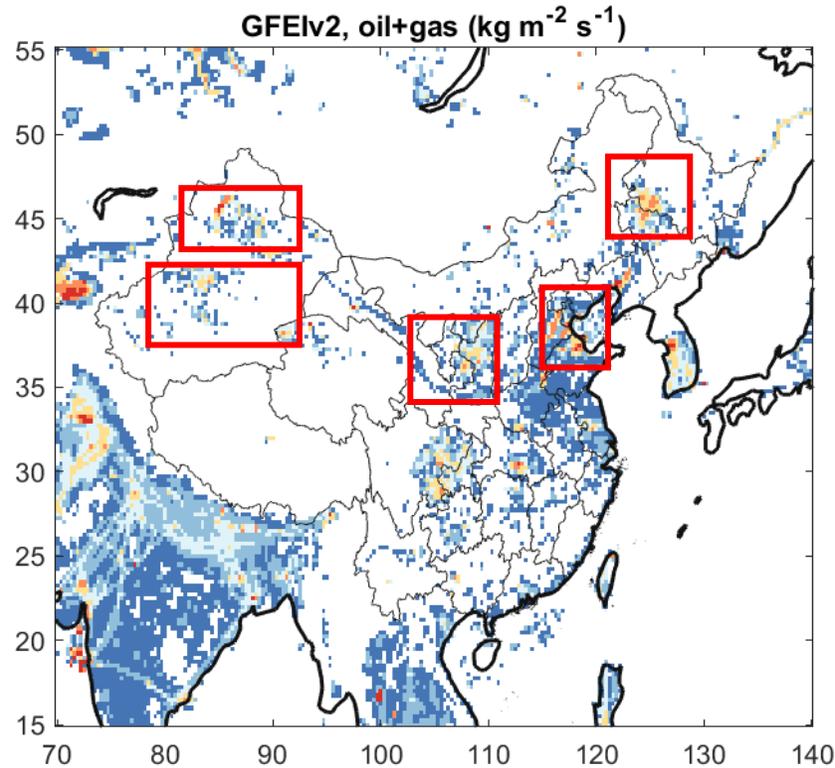
(c) RMSE (Prior-Observation)



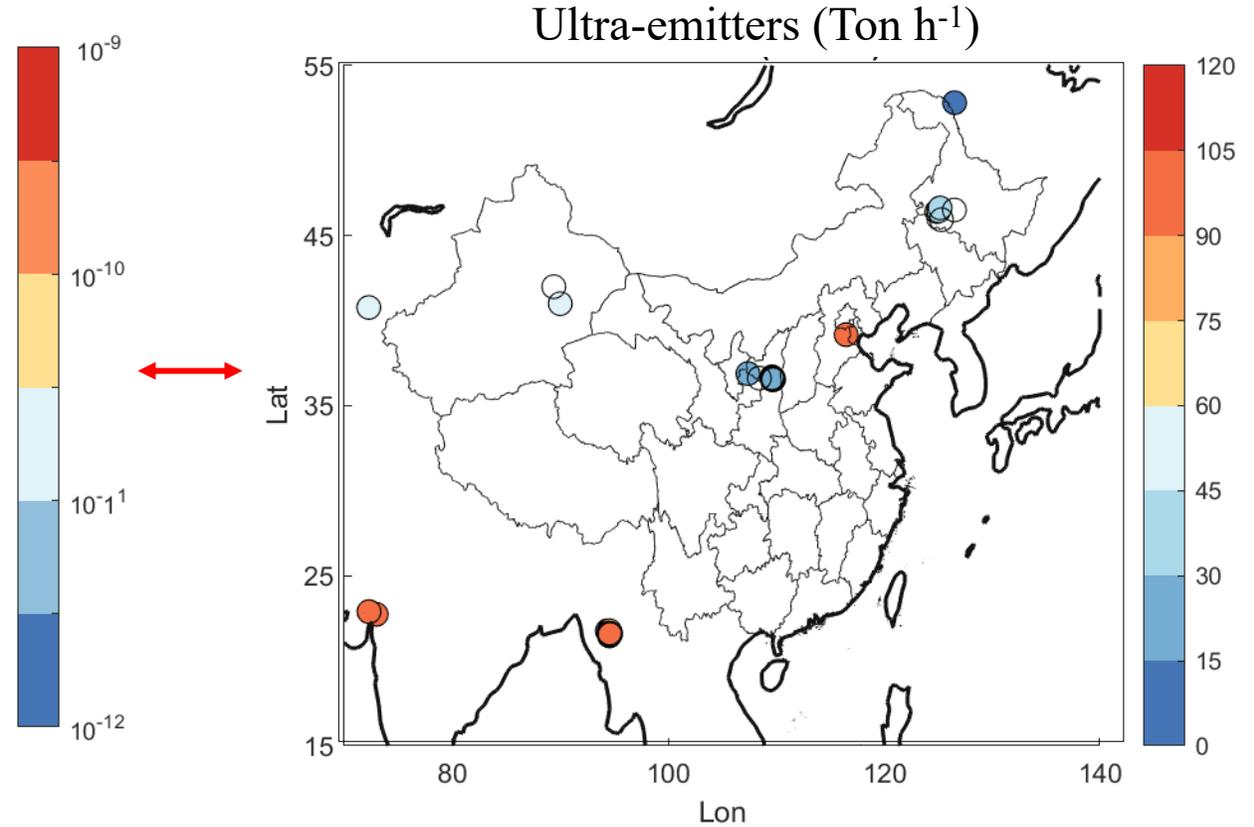
(d) RMSE (Posterior-Observation)



# Large upward correction in oil and gas system



 Where we find large upward adjustment of oil/gas.

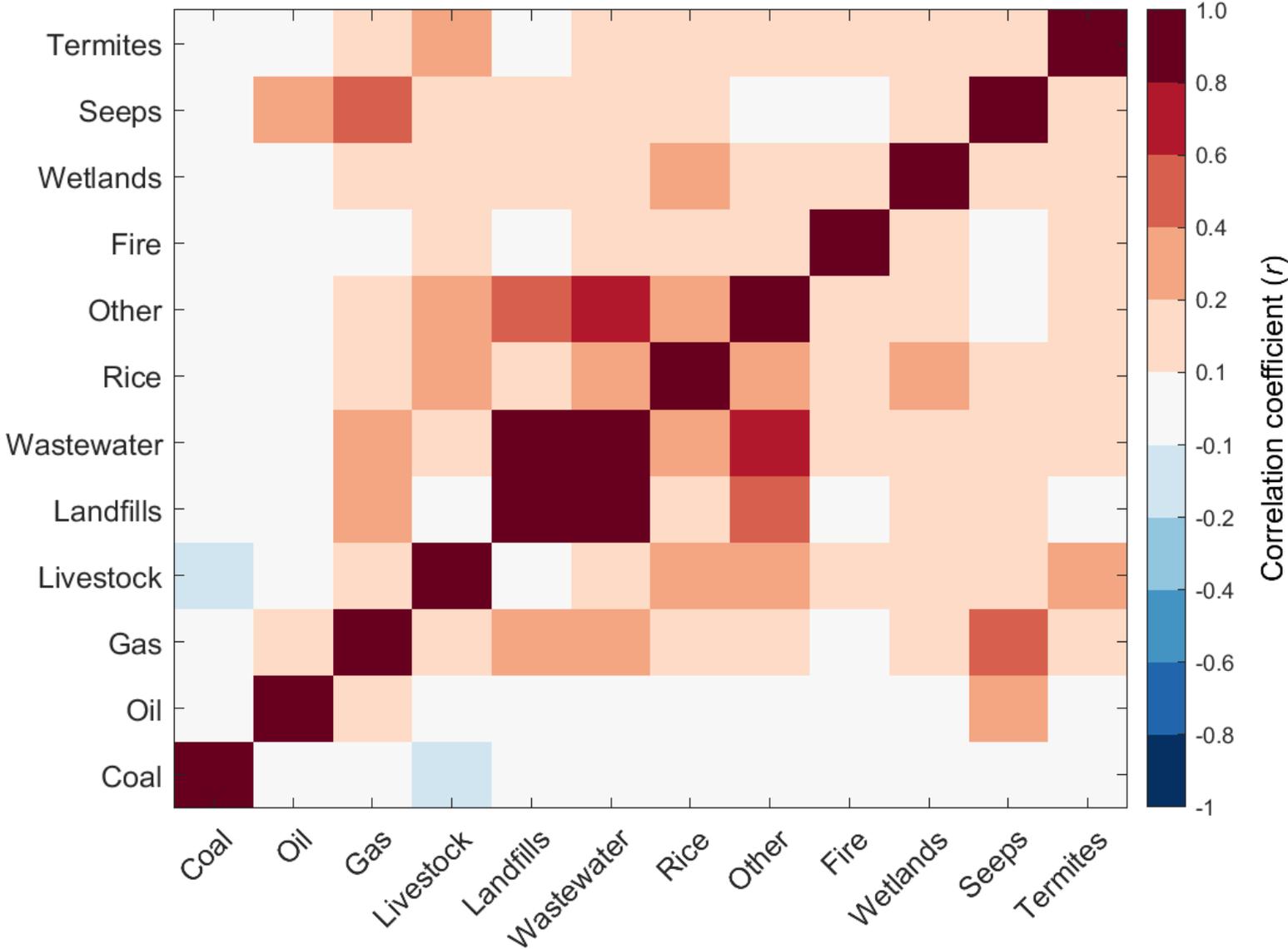


Oil/gas, Data from *Lauvaux et al., 2022*

Ultra-emitters are not fully included in inventories.

# How well are posterior emissions between sectors separated?

Posterior error correlations between sectors



- Landfills and wastewater are highly correlated, associated with urban areas; we then combine them.
- Other sectors are successfully separated ( $<0.2$ ).

0: success  
 $\pm 1$ : fail

# Prior emissions by sector

