

Satellite-derived Indian methane emission sources with TROPOMI retrievals and WRF-GHG modelling framework

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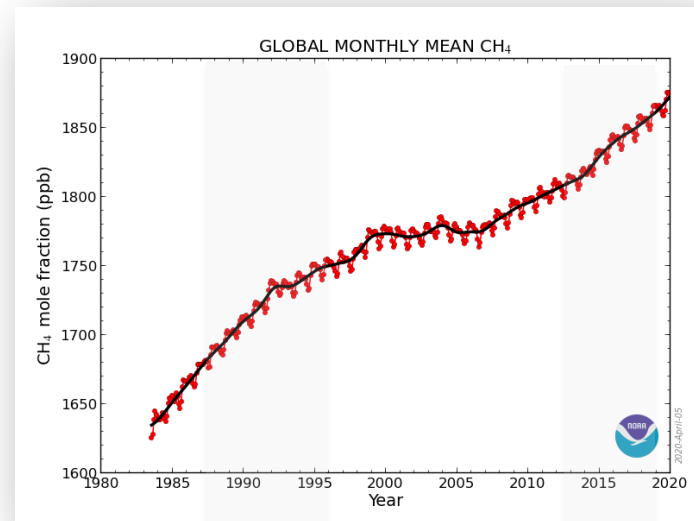
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Background and Motivation

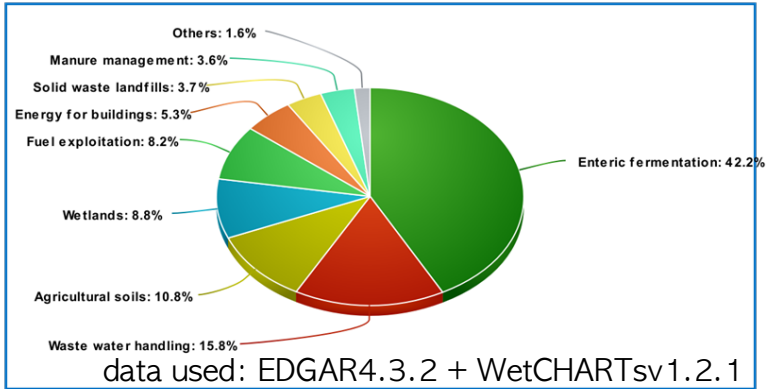
- Methane is the second largest anthropogenic contributor to climate change after CO₂.
- Complexity in CH₄ sources: Anthropogenic and natural sources are uncertain and the processes are often co-located.
- Trends in annual growth rate: To date, there is no consensus on what caused the levelling off (in 1990s), renewed growth, or accelerating rise (after 2015) of CH₄.
- Short life-time of CH₄ (~10 y) makes it an interesting target for mitigating near-term climate change (by reducing radiative forcing on fast time-scales)- “low-hanging fruit”



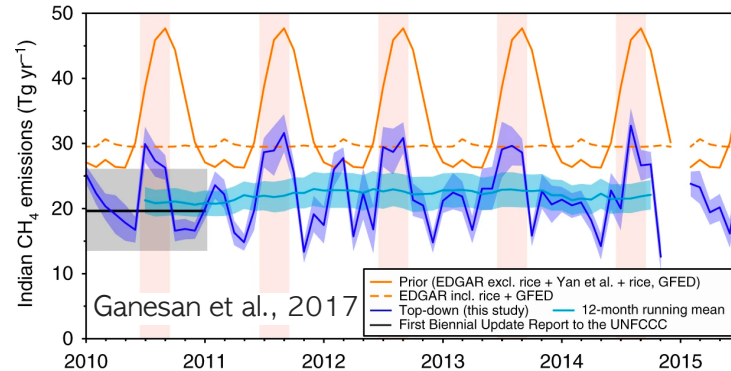
Ed Dlugokencky, NOAA/ESRL

(www.esrl.noaa.gov/gmd)

Indian Methane emissions are largely comprised by anthropogenic sources



■ Enteric fermentation ■ Waste water handling ■ Agricultural soils ■ Wetlands
■ Fuel exploitation ■ Energy for buildings ■ Solid waste landfills
■ Manure management ■ Others



- Largest contribution from ruminants, followed by waste water, fossil fuels and rice paddies. Range of contributions differ significantly from inventory to inventory
- Wetland (natural source) emission contribution range from ~0.5 to 9% depending on the wetland model used.
- Rice and biomass burning emissions exhibit seasonality
- Independent emission verification: No sufficient surface measurements available



Good news is the expansion of satellite fleet for observing Methane at high precision and accuracy

2000 2005 2010 2015 2020 2025

SCIAMACHY / ENVISAT

(ESA, 30×60 km², 6d)

GOSAT

(JAXA, 10×10 km², 3d)

TROPOMI

(ESA, 7×7 km², 1d)

GOSAT 2

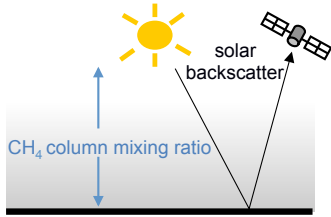
(JAXA, 10×10 km², 3d)

MERLIN

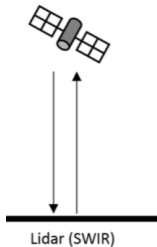
(DLR/CNES, Pencil)

2000 2005 2010 2015 2020 2025

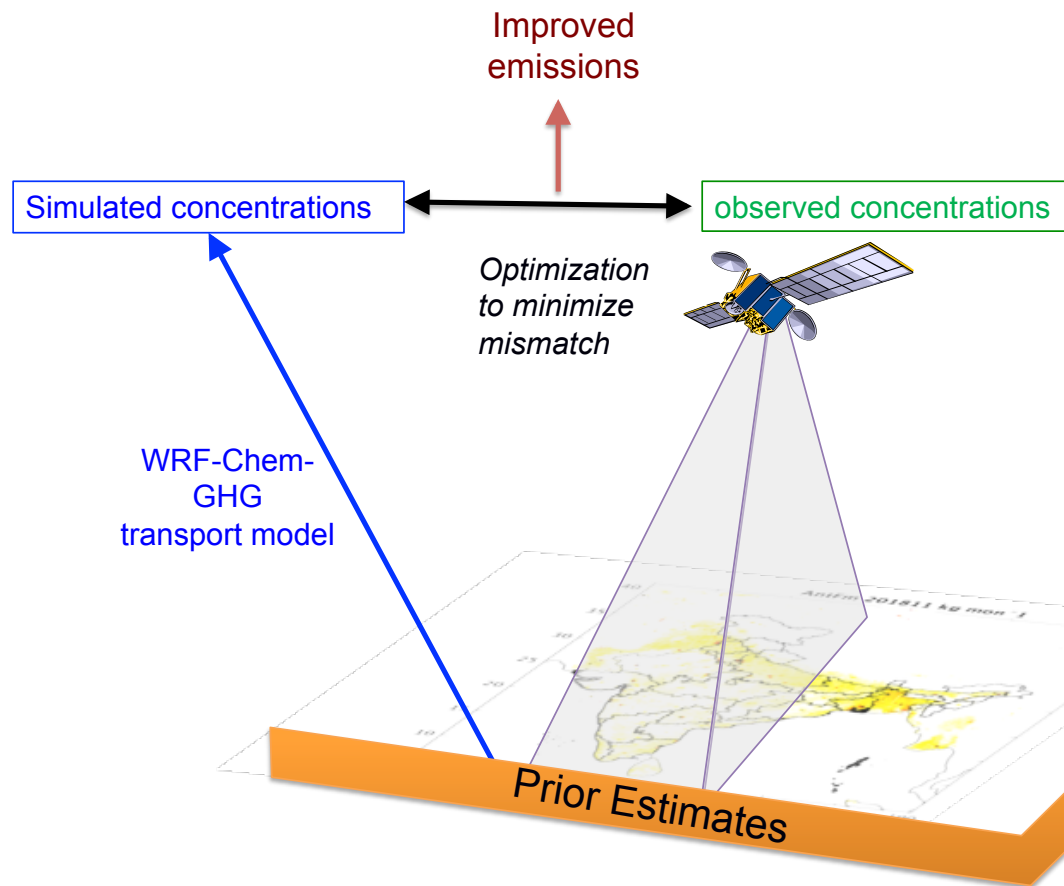
SCIAMACHY, GOSAT, TROPOMI, GOSAT2



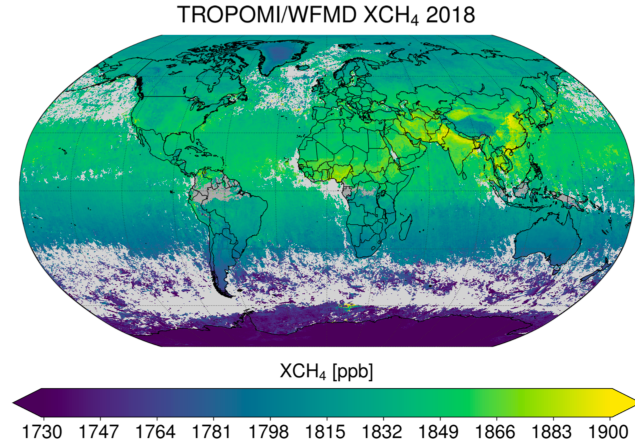
MERLIN



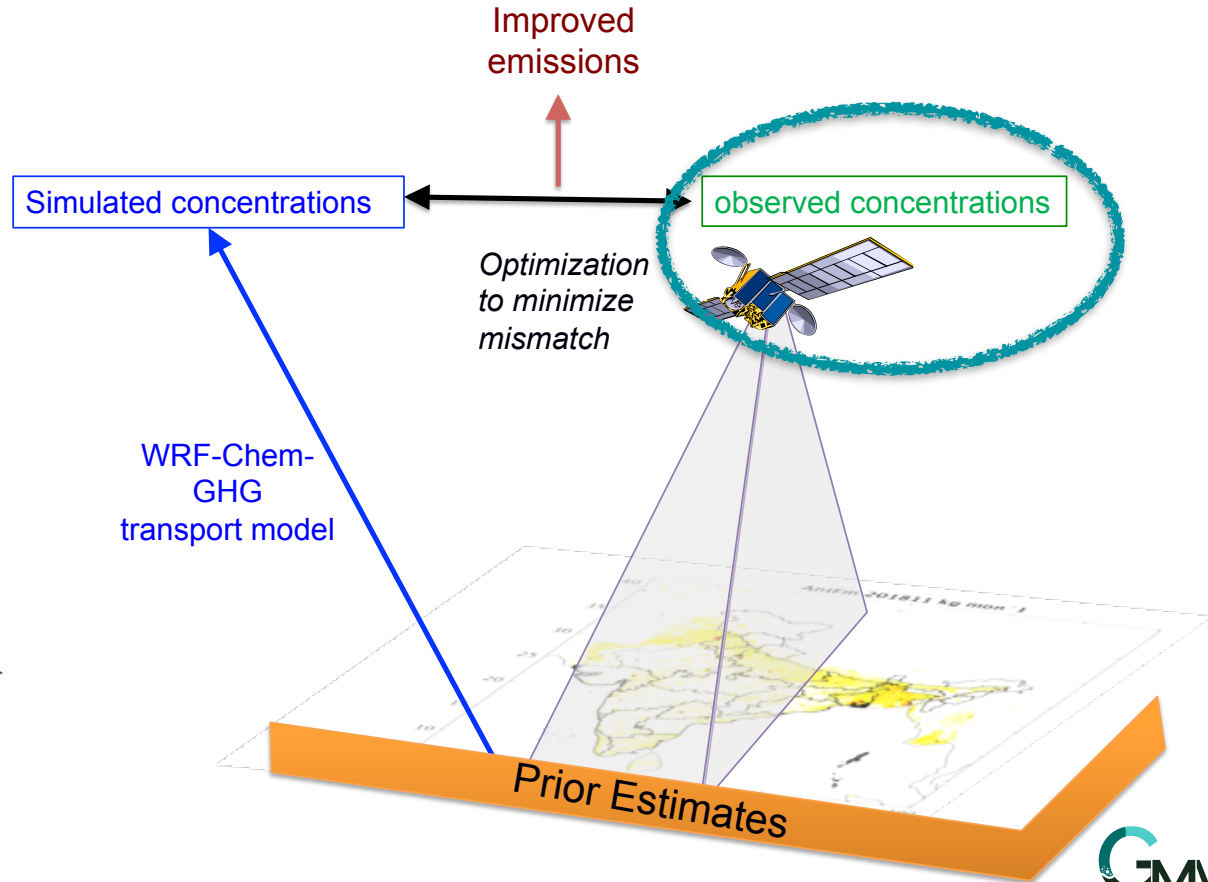
Our approach uses TROPOMI observations to improve emission inventories



Our approach uses TROPOMI observations to improve emission inventories

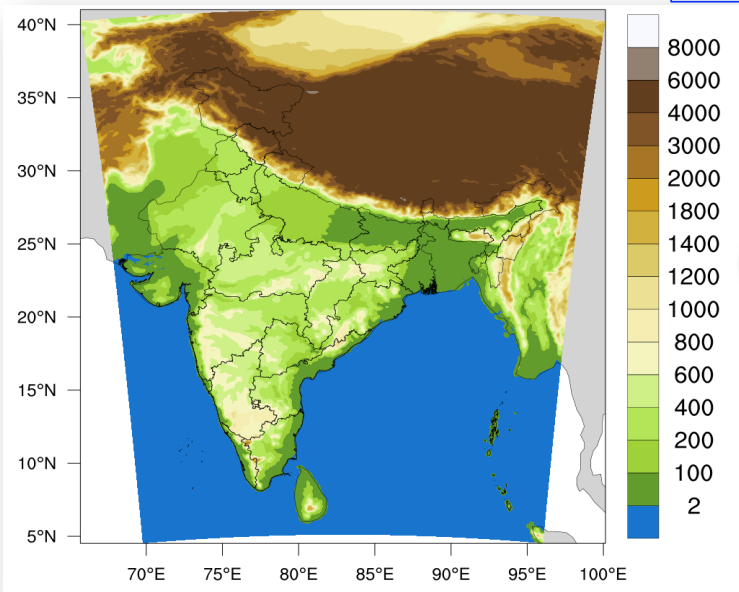


Schneising et al., 2019



Our approach uses TROPOMI observations to improve emission inventories

WRF-GHG Model domain showing terrain height



Simulated concentrations

observed concentrations

Improved emissions

Optimization to minimize mismatch

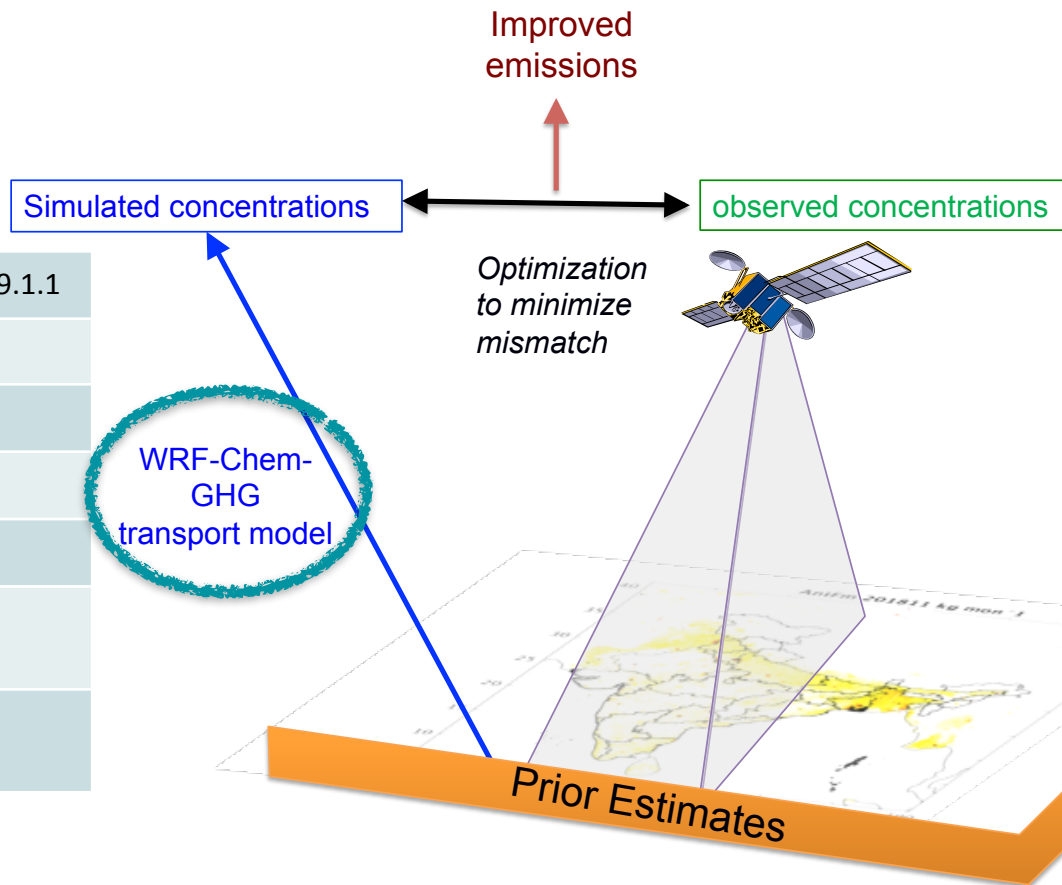
WRF-Chem-GHG transport model

Prior Estimates

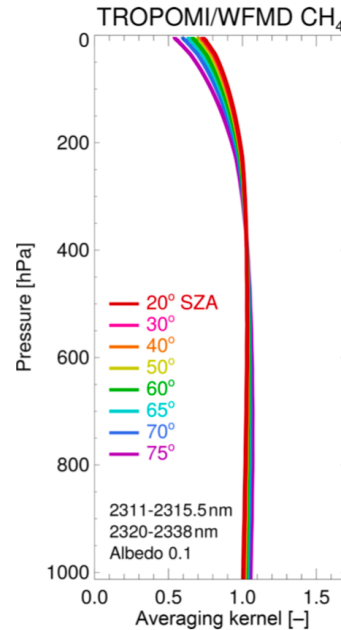
Our approach uses TROPOMI observations to improve emission inventories

WRF-GHG Model Configuration

Model, Version	WRF-Chem-GHG, 3.9.1.1
Horizontal resolution	10 km x 10 km
Temporal resolution	1 hour
Domain area	3080 x km ²
Vertical grid	40 layers
Initial tracer & Met. fields	CAMS, ERA5
Prior fluxes	EDGAR, GFAS, and WetCHARTs

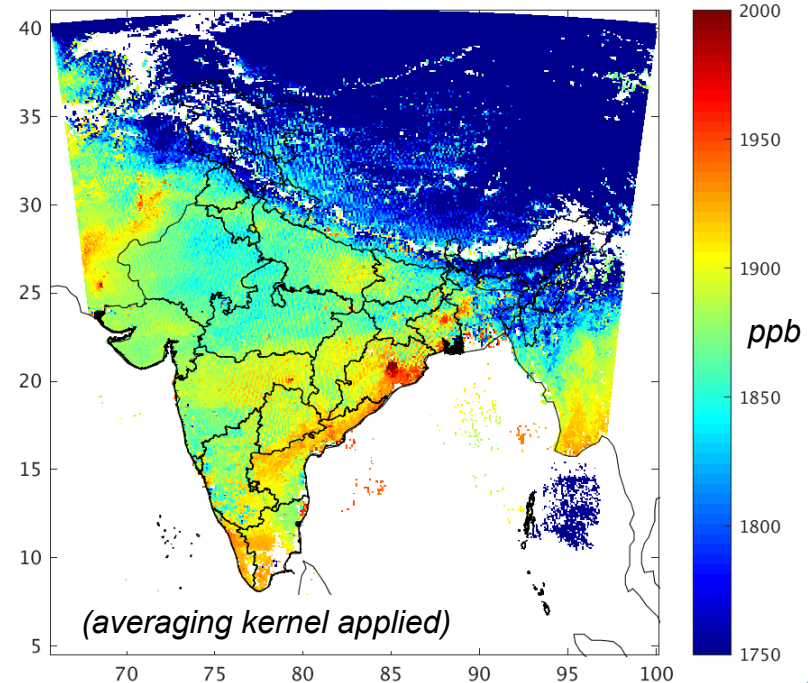


Model predictions of atmospheric XCH₄ after applying temporal sampling and averaging kernel



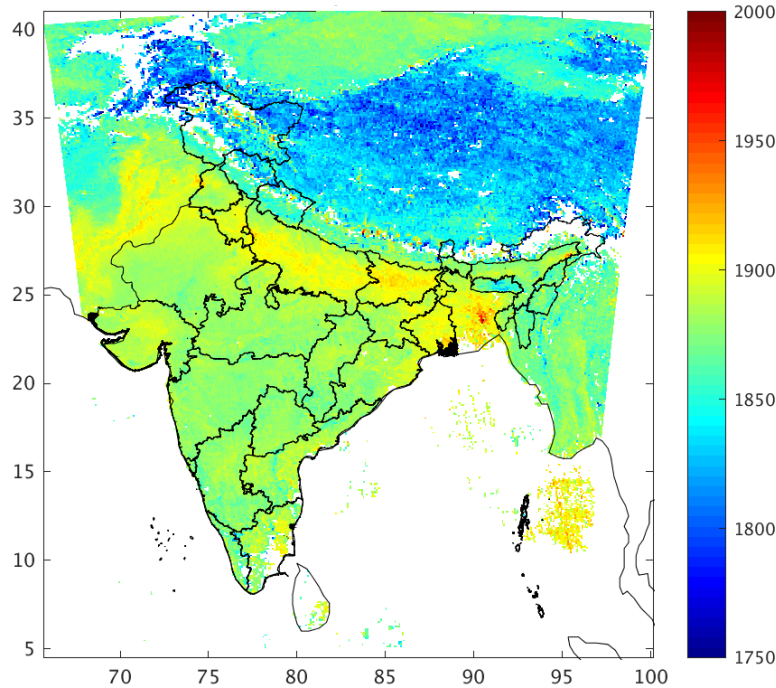
Schneising et al., 2019

WRF XCH₄ averaged for November 2018

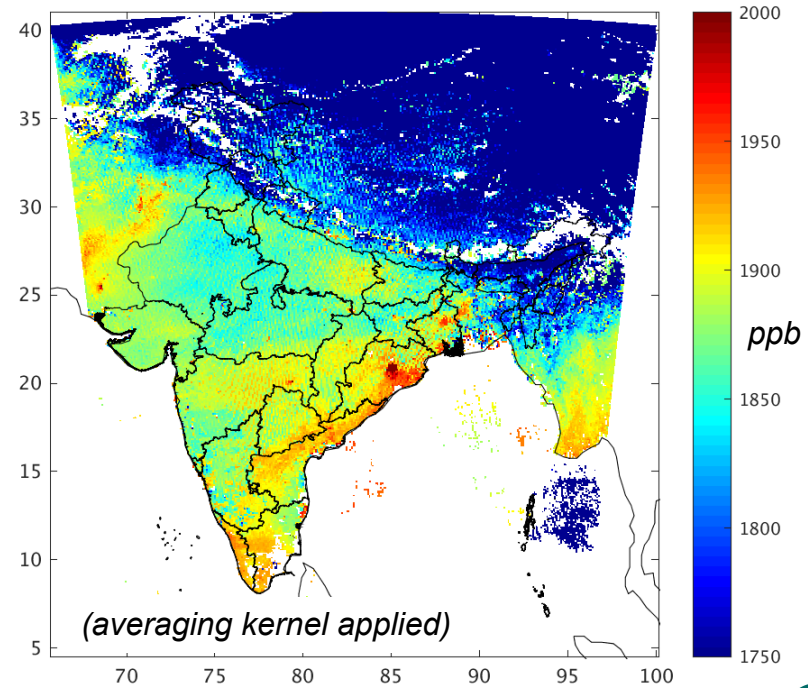


Model predictions of atmospheric XCH₄ after applying temporal sampling and averaging kernel

S5P XCH₄ averaged for November 2018

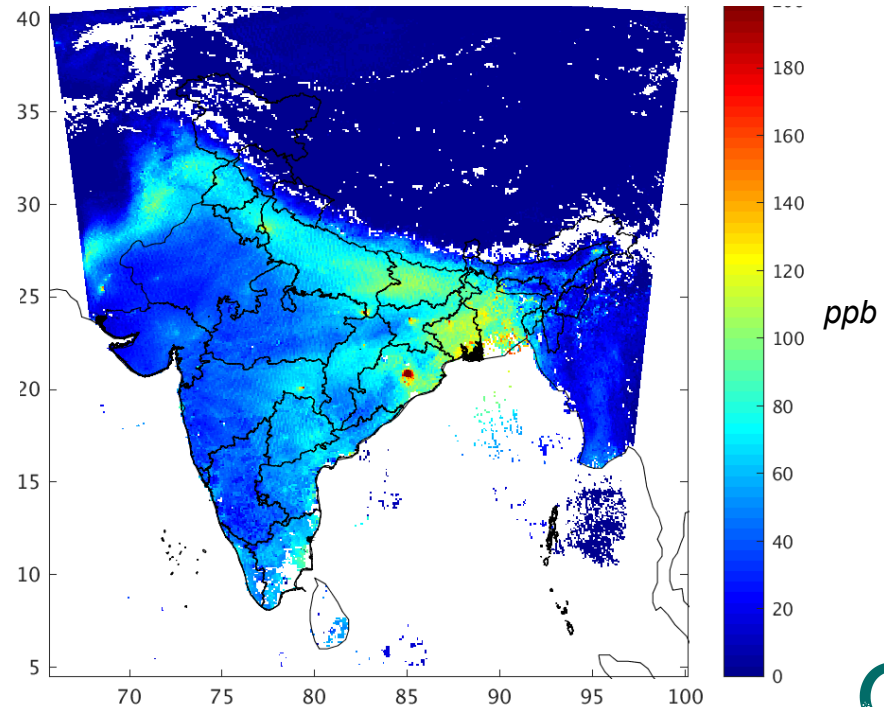


WRF XCH₄ averaged for November 2018



Separating tracers suggests possible overestimated anthropogenic contribution over Indian region

Anthropogenic contribution for November 2018

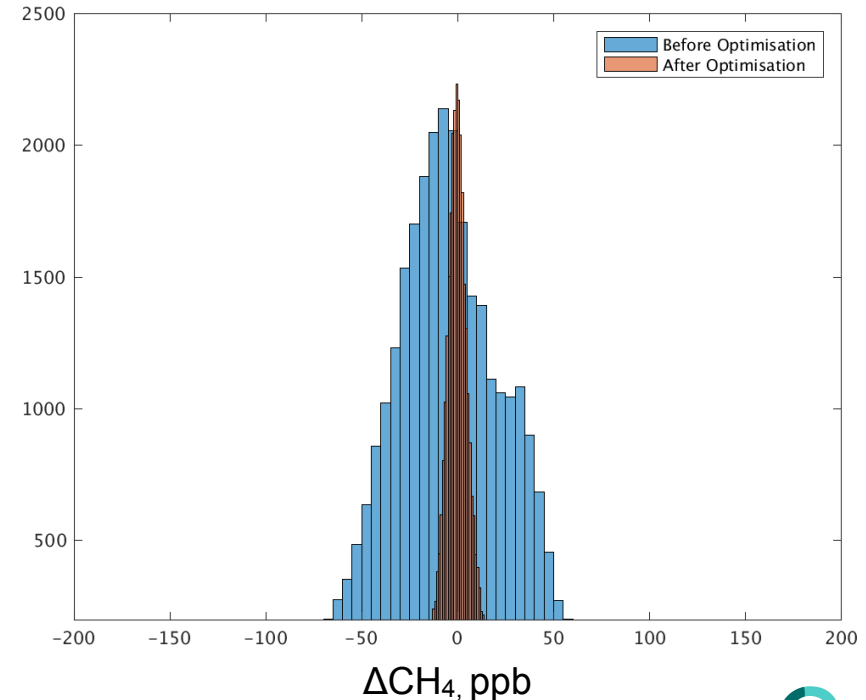


Optimisation of WRF model emission is performed by minimising the cost function

Optimisation design

- WRF tracers (contributions) are separated to understand how each one contributes independently
- Observation vector: All available good S5P column observations for November 2018
- State vector: Scaling factors for monthly averaged exchange fluxes for each political states of India
- Transport error in WRF is ignored

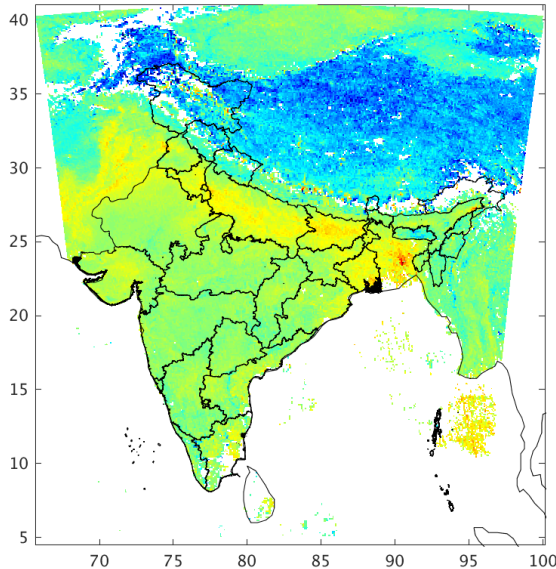
Model-Observations November 2018



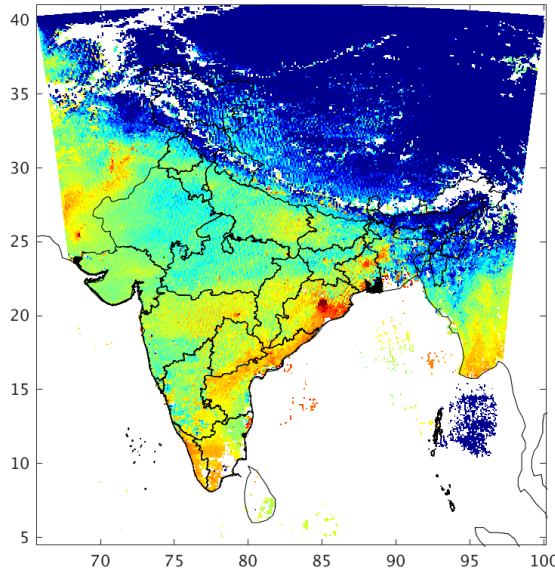
Optimisation of WRF model emission is performed by minimising the cost function



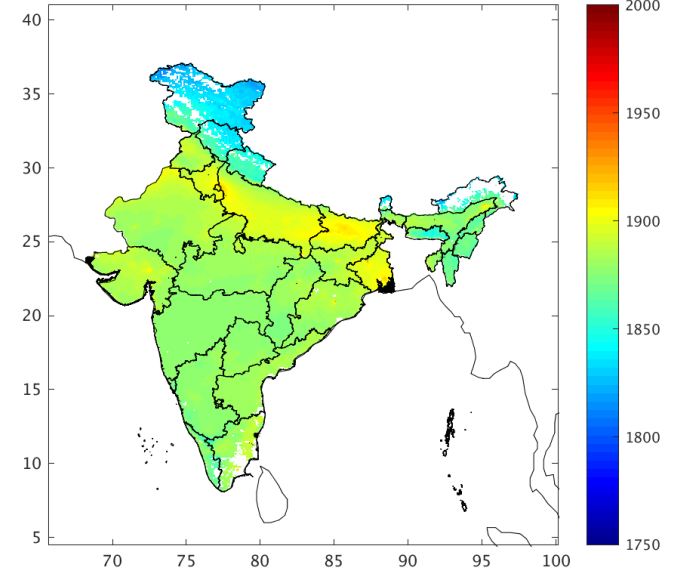
Observations



Simulations, before opt.



Simulations, after opt.



TROPOMI derived anthropogenic Methane emissions in comparison with other estimates

Top-down quantification using satellite observations

- Our estimation of Indian methane emission is consistent with Ganesan et al., 2017.

Bottom-up quantification

- Our current estimations are **~40%** lower than that of EDGAR and **~10%** higher than 2010 BUR report submitted to UNFCCC.

		Indian Methane emissions, Tg yr ⁻¹
EDGAR v4.3.2, 2012	Bottom-up approach	31.0
India's BUR (Biennial Update Report to its National Communications), 2010	Bottom-up approach	19.8
Ganesan et al., Nature Communications, 2017	Top-down approach, GOSAT, surface and aircraft observations	22.0*
This study	Top-down approach, TROPOMI observations	22.0** 22.9** (including biomass burning emissions)

*over the period 2010–2015

**November 2018

Takeaways

- Robust evaluation of national Methane emissions is necessary to improve the confidence in India's CH₄ inventory.
- TROPOMI XCH₄ observations pave the way for improving knowledge of methane emissions over India through inverse analyses. However, surface measurements are indispensable.
- A high-resolution atmospheric transport modeling system is required to simulate data from these platforms and to infer fluxes.
- By minimising the cost function, we find a better agreement by scaling down the EDGAR emissions by a factor of ~40%
- Future task: further refinement of inverse method by better taking into account of error characterisation (e.g transport error) of the solution

*For questions/comments,
please feel free to contact*



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**Greenhouse gas Modeling
and Applications Group**

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