

S5P/TROPOMI XCH₄: current status and the path to cross-instrument harmonisation

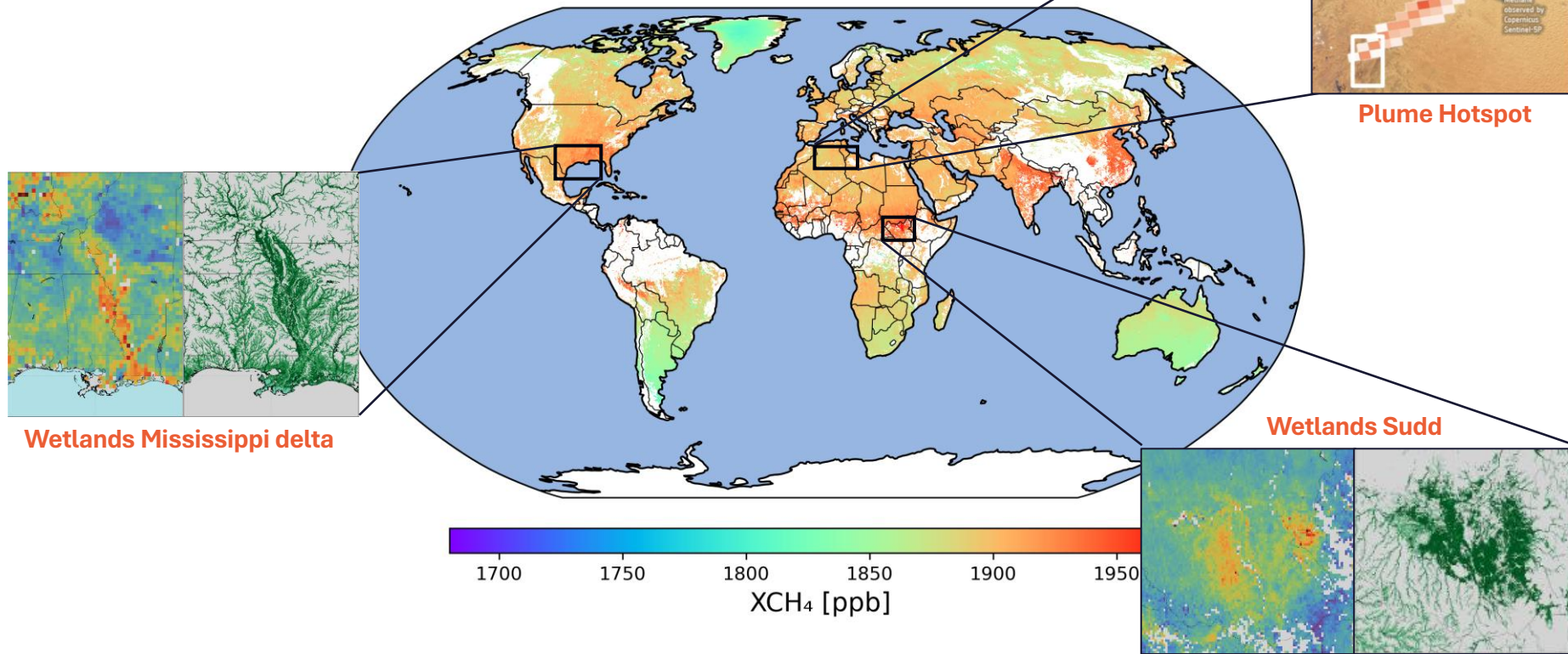
Soumyajit (Jit) Mandal, Jochen Landgraf, Andrew Barr



CEOS AC-VC-22
SRON Leiden
04/06/2026

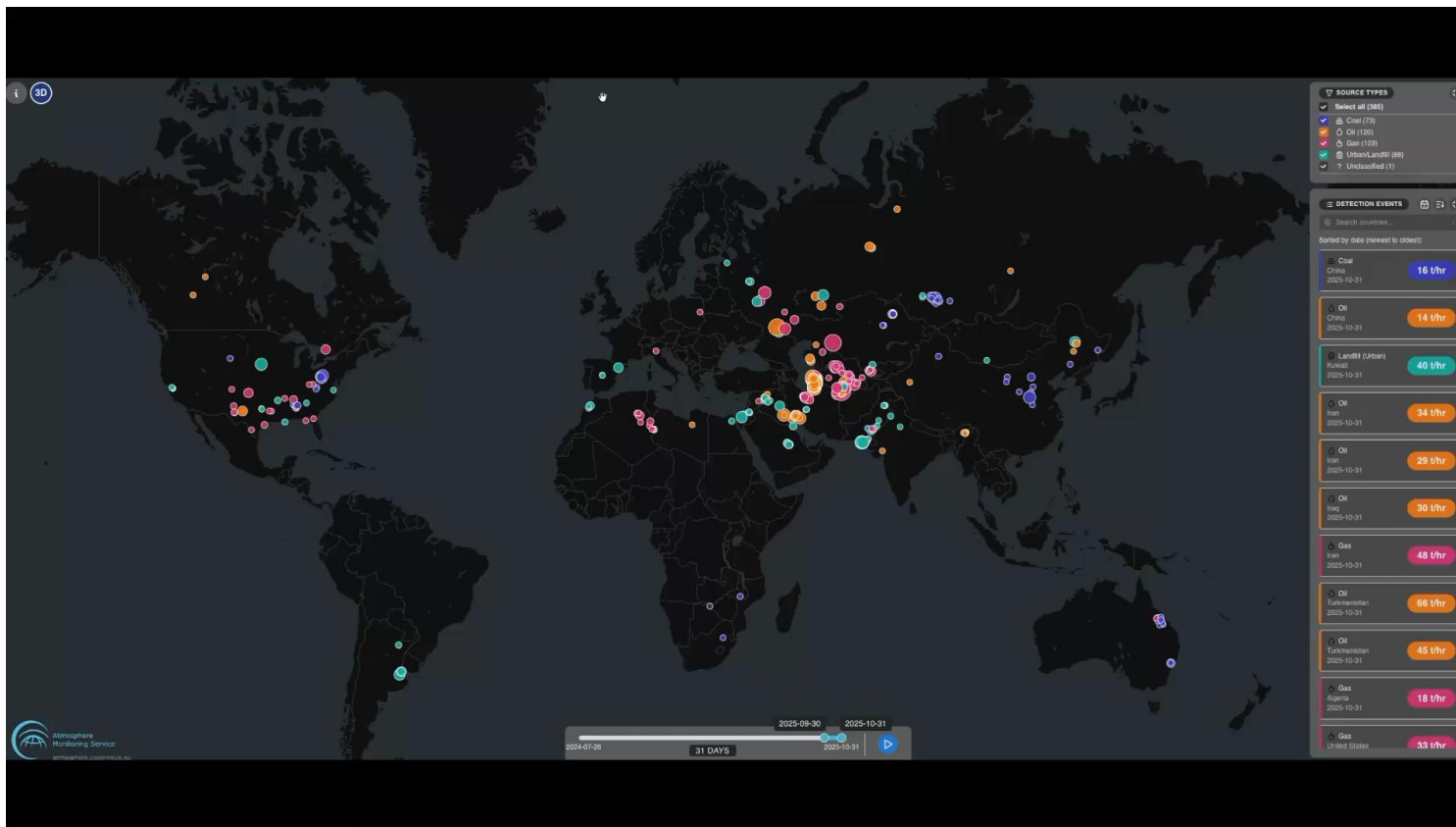


Tropomi Global Map



CAMS Methane Hotspot Explorer

<https://atmosphere.copernicus.eu/ghg-services/cams-methane-hotspot-explorer>



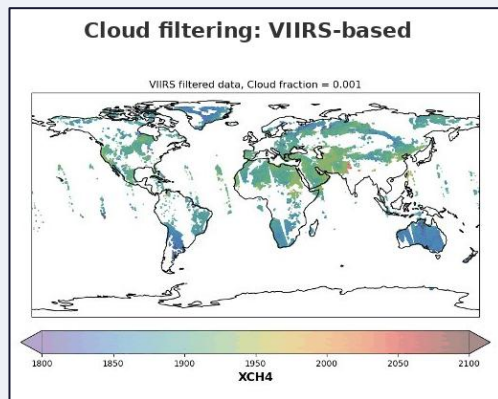
Credits: Bram Maasackers, Ilse Aben & L4 team at SRON

SRON

S5P TROPOMI – Continuous Improvements

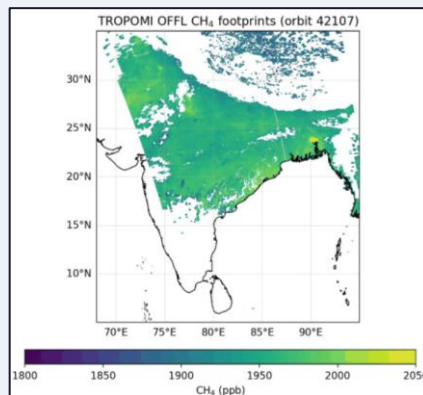
☁ Cloud Filtering

No SNPP/VIIRS dependency —
future proof ahead of VIIRS
decommissioning



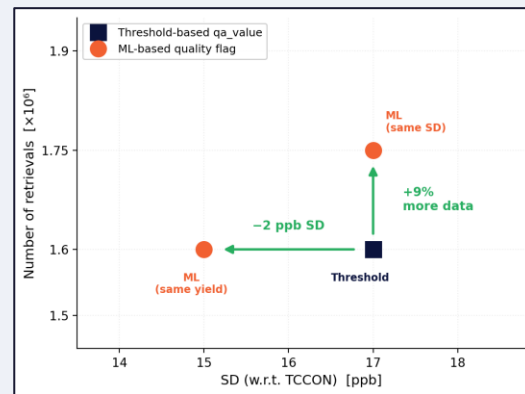
🕒 Near Real-Time Product

New cloud filtering enables a
near real-time XCH₄ data
product



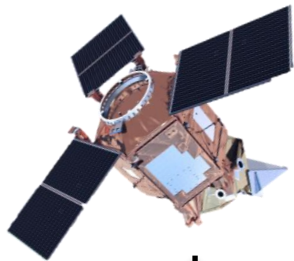
🤖 ML-based Quality Filtering

Learned quality flag replaces
the threshold-based qa_value



All described in detail in the latest CH₄ ATBD: <https://sentiwiki.copernicus.eu/web/s5p-products>

Looking ahead: cross-instrument harmonisation



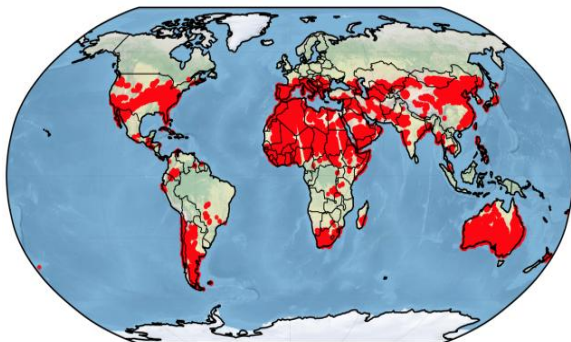
TROPOMI
2018 -



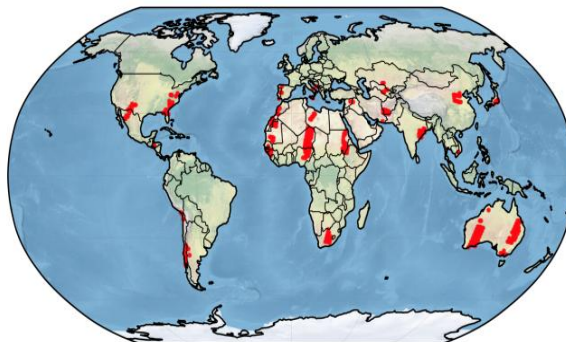
GOSAT-2
2019 -

Current approach: bias corrected as a function of surface albedo — validated against sparse TCCON sites at 300 km colocation radius

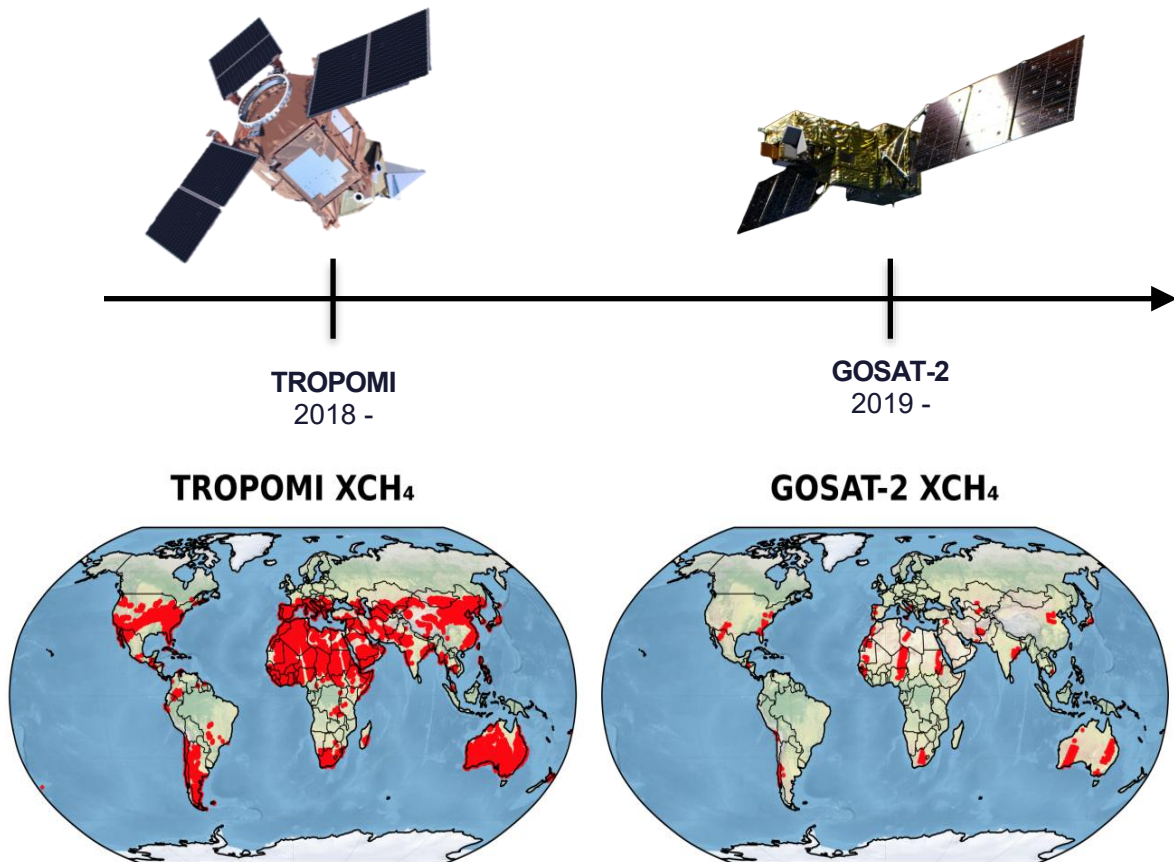
TROPOMI XCH₄



GOSAT-2 XCH₄



Looking ahead: cross-instrument harmonisation

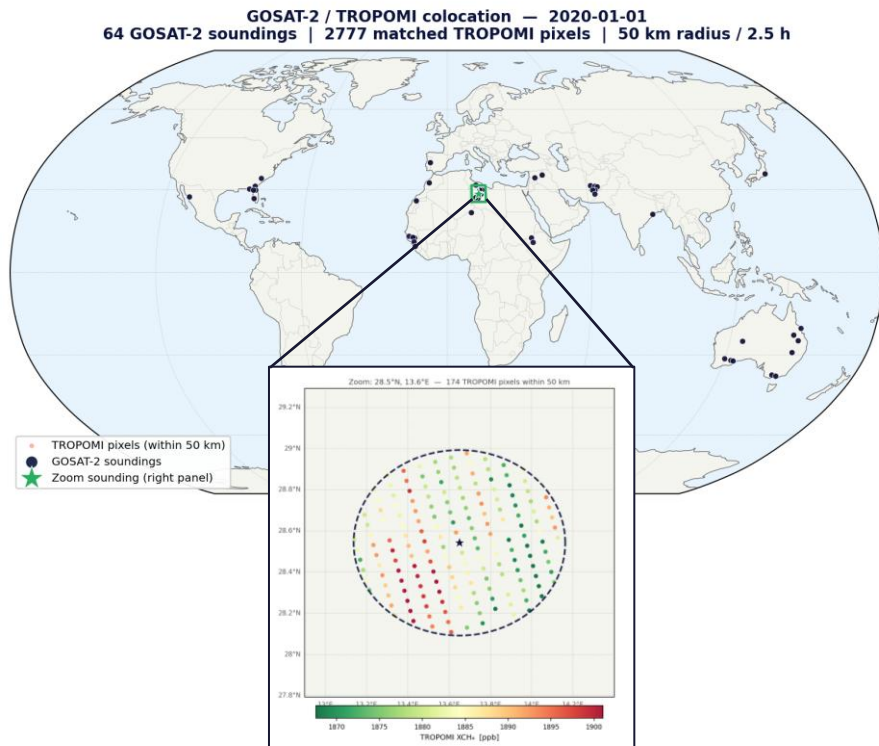


Can TROPOMI bias be corrected using GOSAT-2 as an anchor — two instruments sharing the same retrieval algorithm?

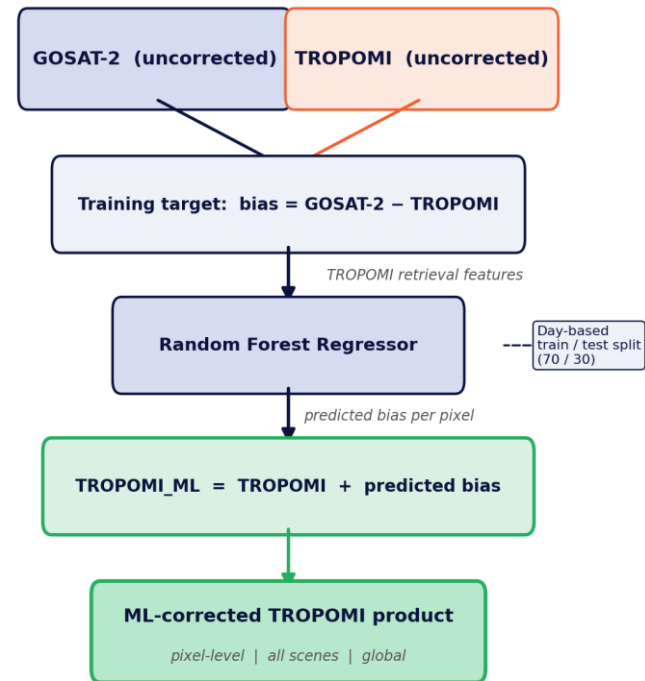
Prior work: Balasus et al. (2023, AMT) presented a blended TROPOMI+GOSAT product using a similar ML framework — a key reference for this work. Our approach is distinct: we use **GOSAT-2** (same retrieval algorithm as TROPOMI) and compare **both instruments uncorrected** — disentangling instrumental bias from retrieval corrections

GOSAT-2 / TROPOMI collocation strategy

Match criteria: spatial radius
(~50km) + temporal window (2.5h)



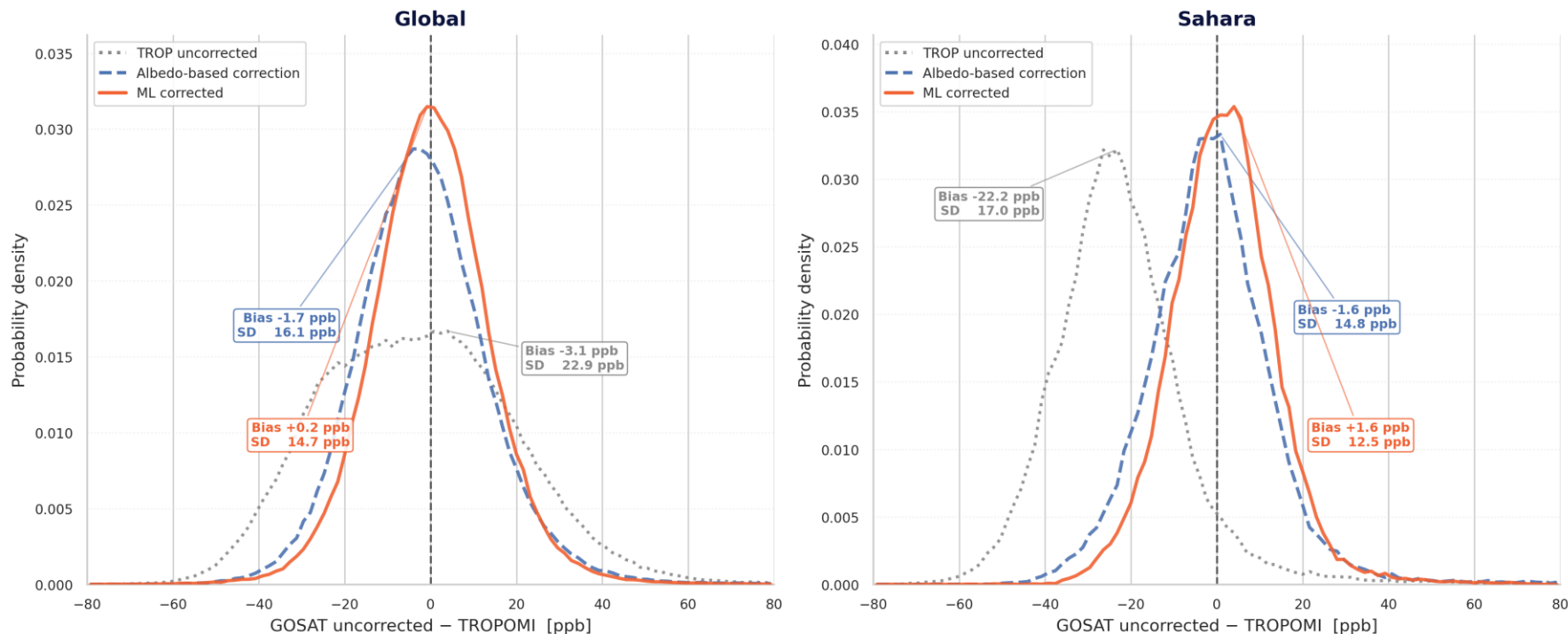
ML bias-correction pipeline





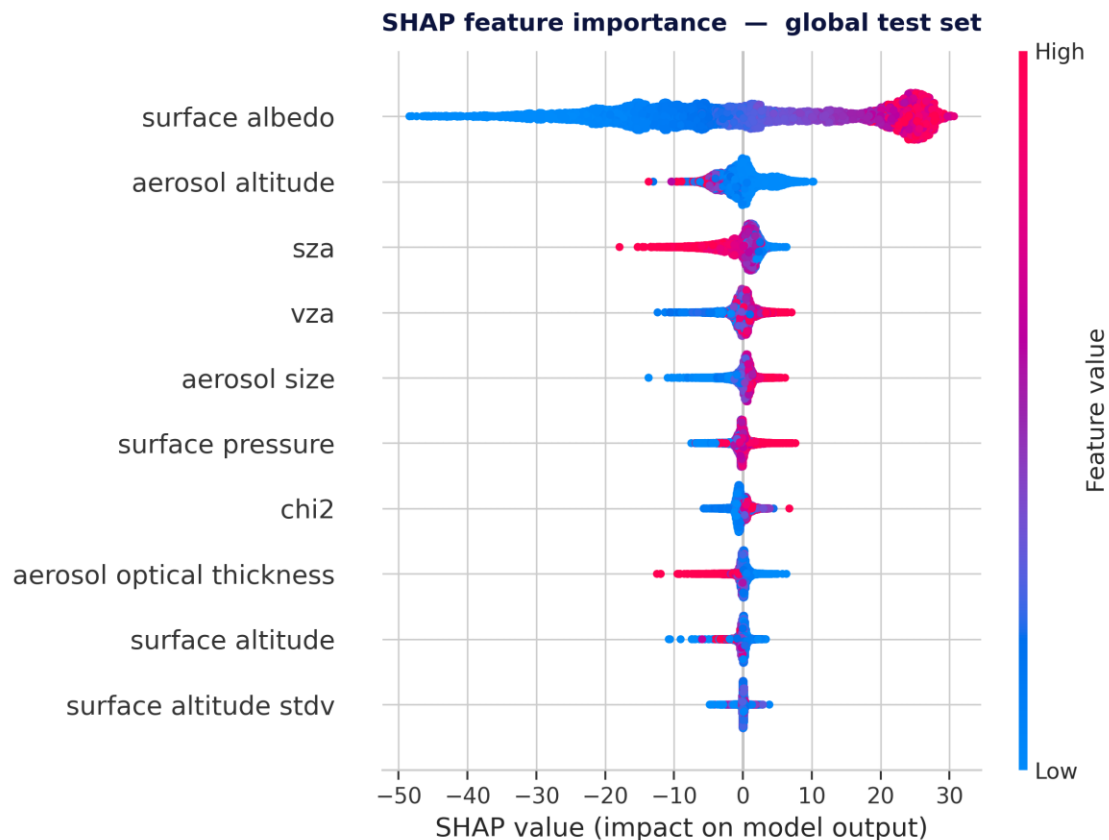
Results

Residual distributions (GOSAT uncorrected – TROPOMI)





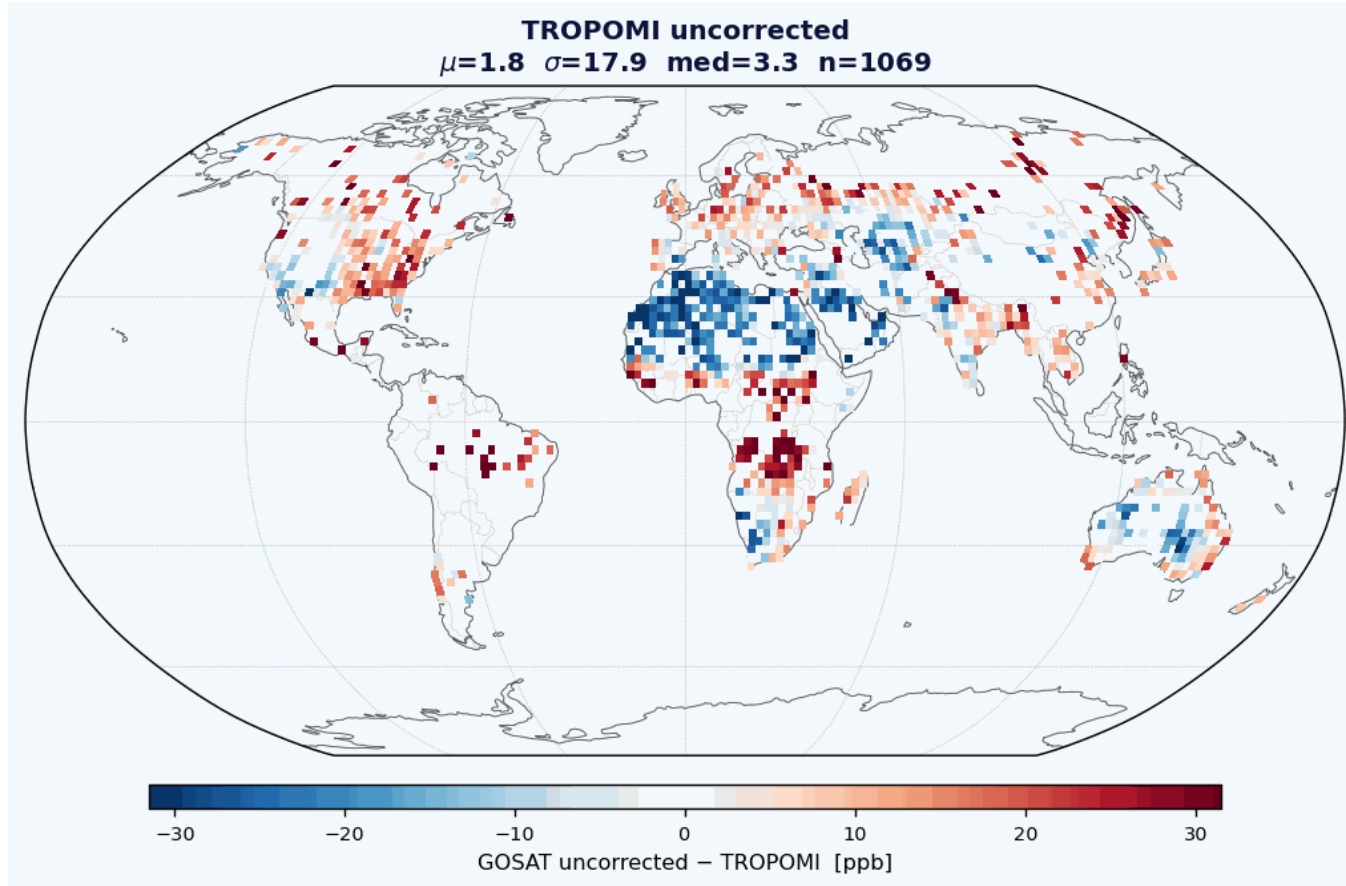
Model interpretability — SHAP analysis



- Surface albedo dominates — high albedo pushes bias up, low albedo pushes it down
- Aerosol altitude, geometry and scene properties all contribute — beyond a single-variable correction



Spatial diagnostics — $2^\circ \times 2^\circ$ grid, test set



Conclusions

Summary

- ML bias correction **outperforms** the albedo-based approach
- Model **learns physically interpretable** corrections

Next steps

- **Validate** the learned correction **against an independent reference (TCCON)**
- Investigate simulation-based correction **to separate instrumental from model-induced bias** — enabling cleaner cross-instrument harmonisation [ongoing]

