

Space-based constraints on reactive nitrogen:

From emissions of NH_3 at the surface to NO_x in the upper troposphere



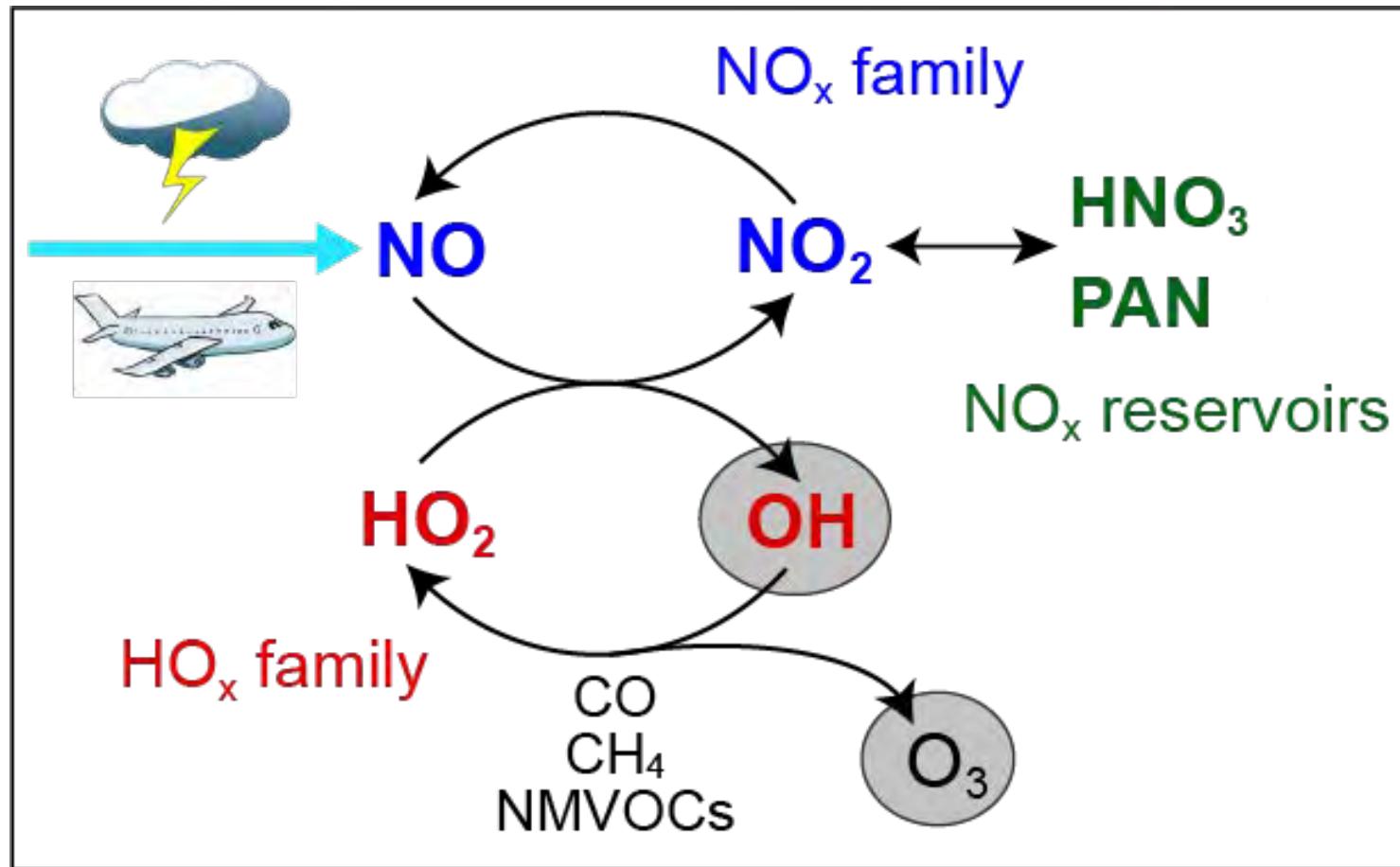
First Estimate of Upper Tropospheric NO₂ from TROPOMI



Just published: <https://doi.org/10.5194/amt-14-2389-2021>

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Nitrogen oxides (NO_x) in the Upper Troposphere (8-12 km)



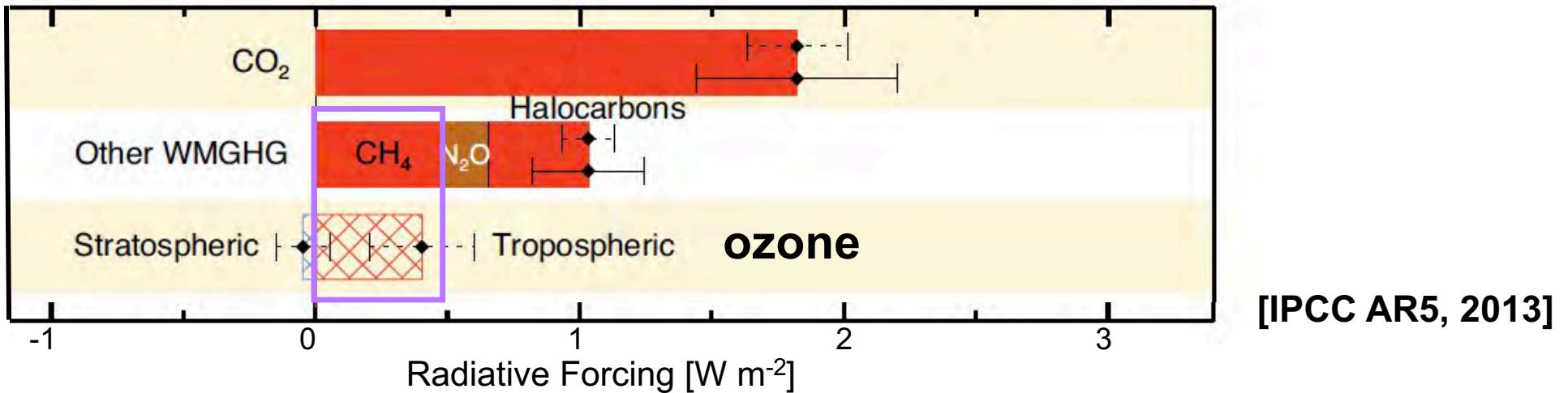
Influence atmospheric oxidants (OH, O_3) and climate (O_3 formation, methane persistence)

Other sources: injection of surface pollution, rockets (?), long-range transport

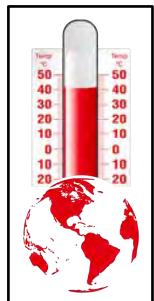
Why uncertainties in the upper troposphere matter

Warming due to tropospheric ozone is similar to that of methane (CH_4)

Tropospheric ozone and methane have near-equal climate impacts



We are 100% reliant on models to estimate pre-industrial ozone



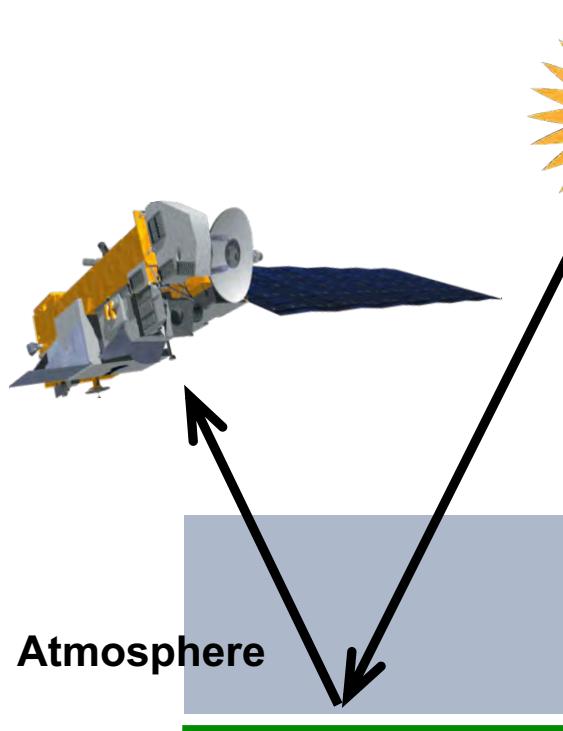
Errors in tropospheric ozone radiative forcing



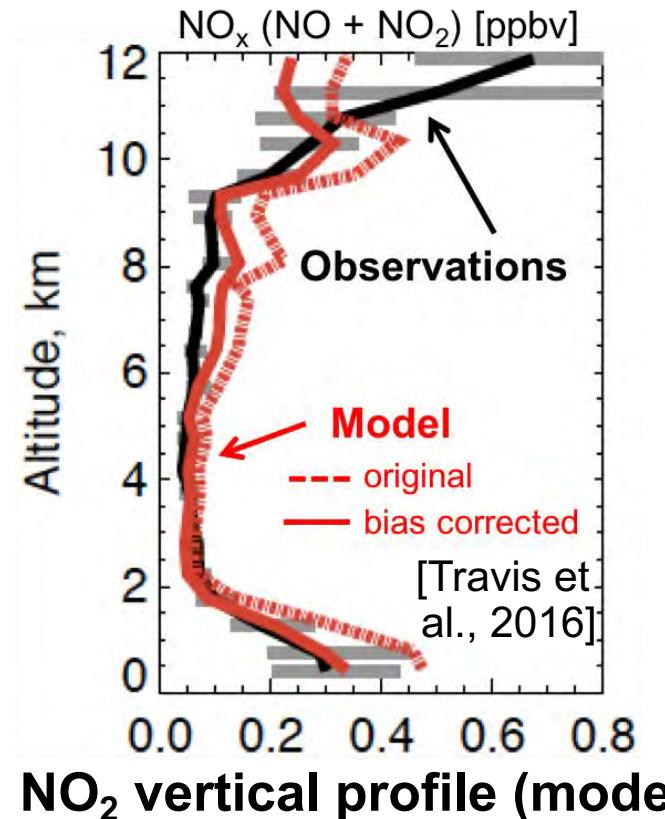
Indicator paper used to measure ozone in the mid to late 19th century

Why uncertainties in the upper troposphere matter

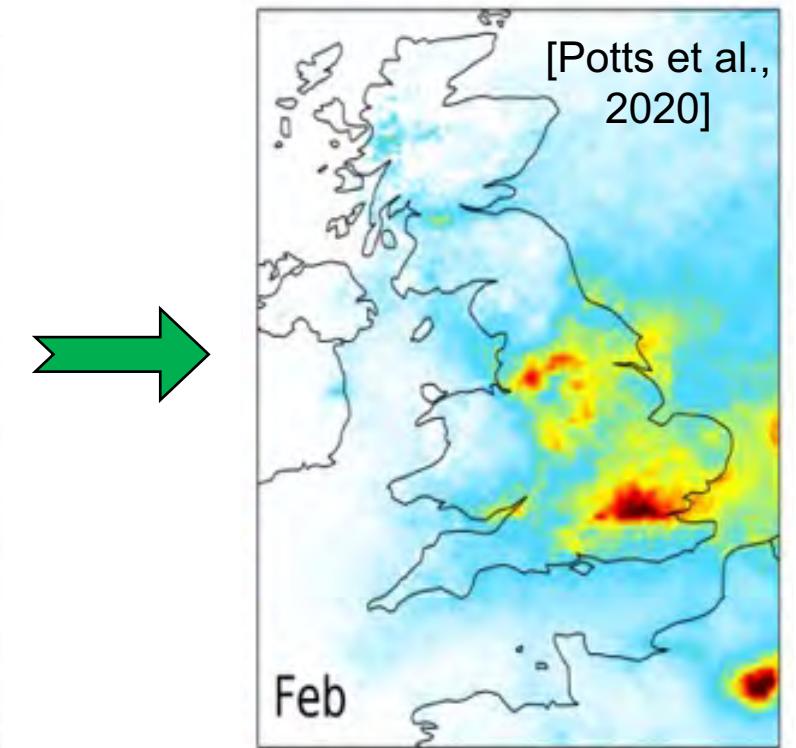
We rely on models to retrieve atmospheric composition from satellite instruments



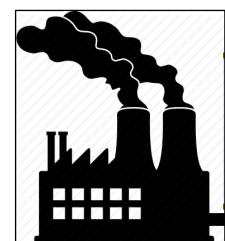
NO₂ along the viewing path



NO₂ vertical profile (model)



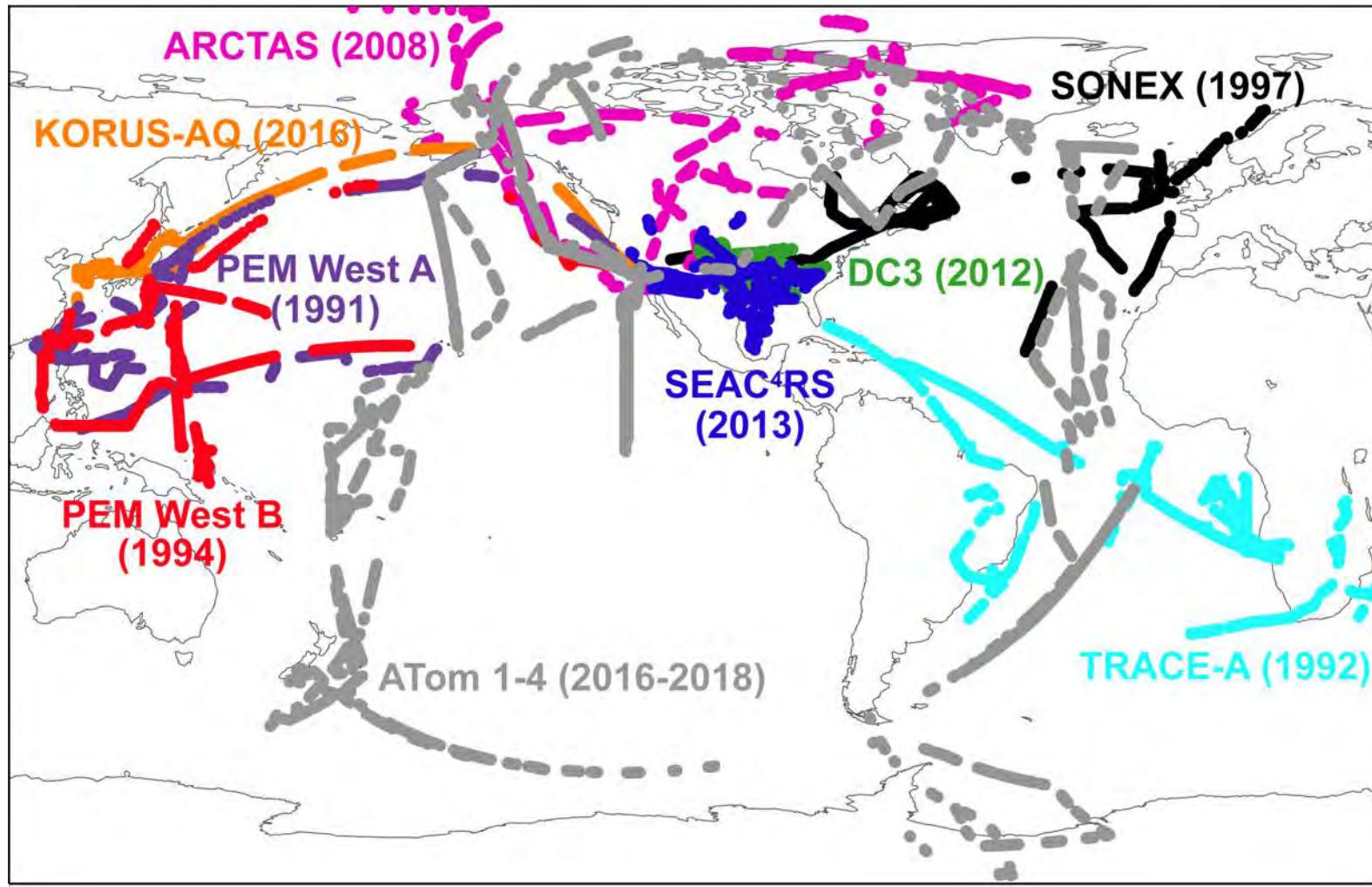
Vertical column densities



Uncertain global air quality constraints from satellites

Sampling of the upper troposphere is limited

NASA DC8 research aircraft flight tracks in the upper troposphere



There are also measurements from commercial aircraft

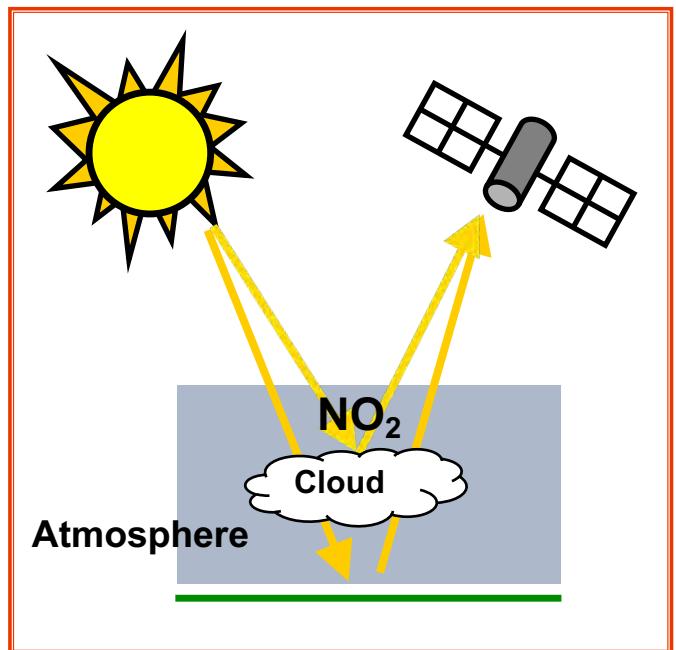
All use instruments that are susceptible to large biases in the upper troposphere

Satellite products obtained using the cloud-slicing technique

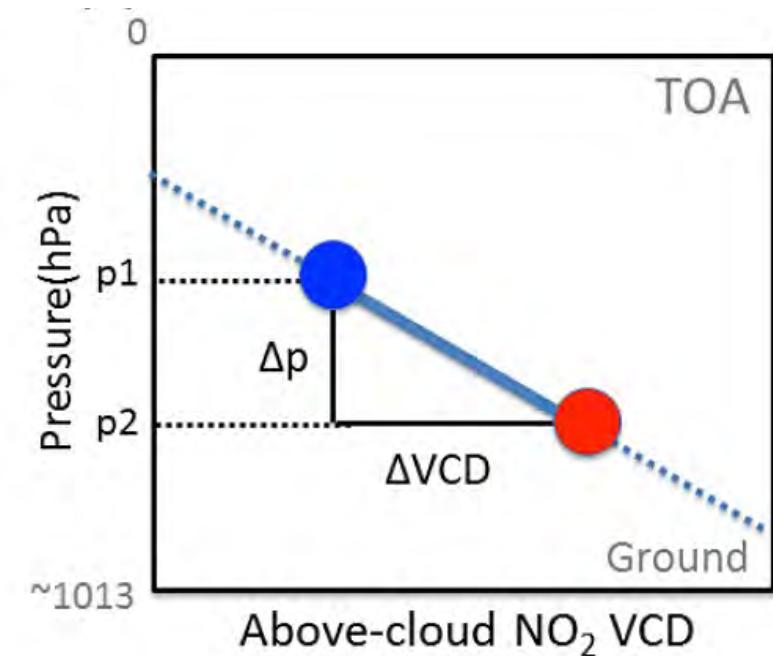
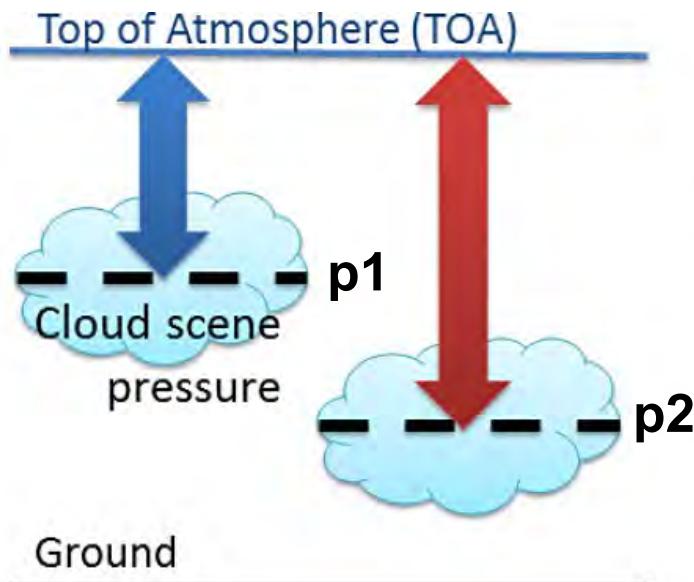
First applied by Ziemke et al. [2001] to TOMS ozone

Retrieve partial NO₂ columns over cloudy scenes at different heights

APPROACH



Use cloud height variability to derive partial columns



[adapted from Choi et al., 2014]

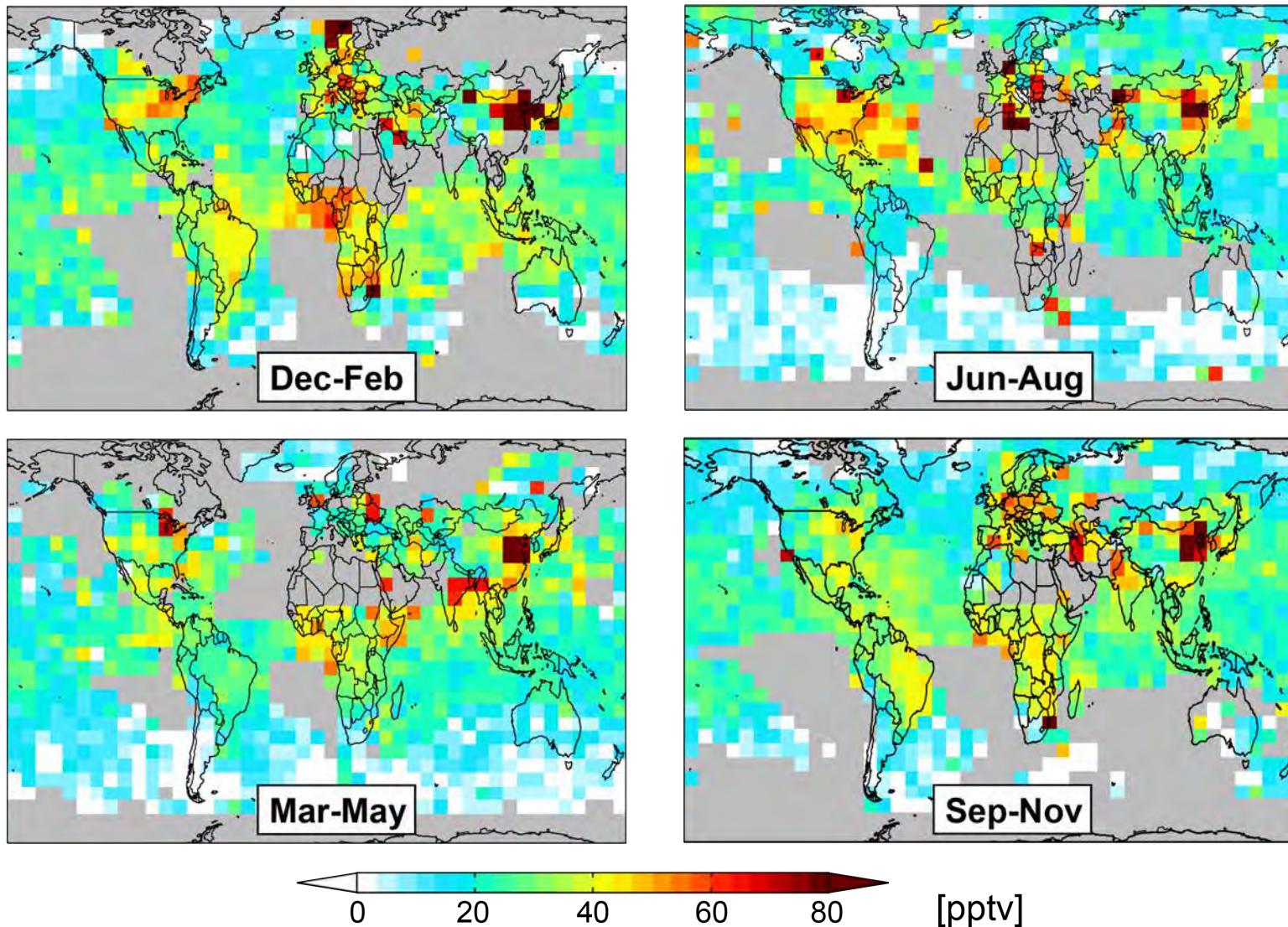
NO₂ volume mixing ratio (VMR) between clouds at p1 and p2

$$\text{NO}_2 \text{ VMR} = \frac{\Delta \text{VCD}}{\Delta p} \times \frac{k_B g}{R_{\text{air}}}$$

Products of upper tropospheric NO₂ from satellite observations

Near global spatial coverage of seasonal mean UT NO₂ at $5^\circ \times 8^\circ$ (50 km \times 80 km)

NASA OMI upper troposphere NO₂ (2005-2007)

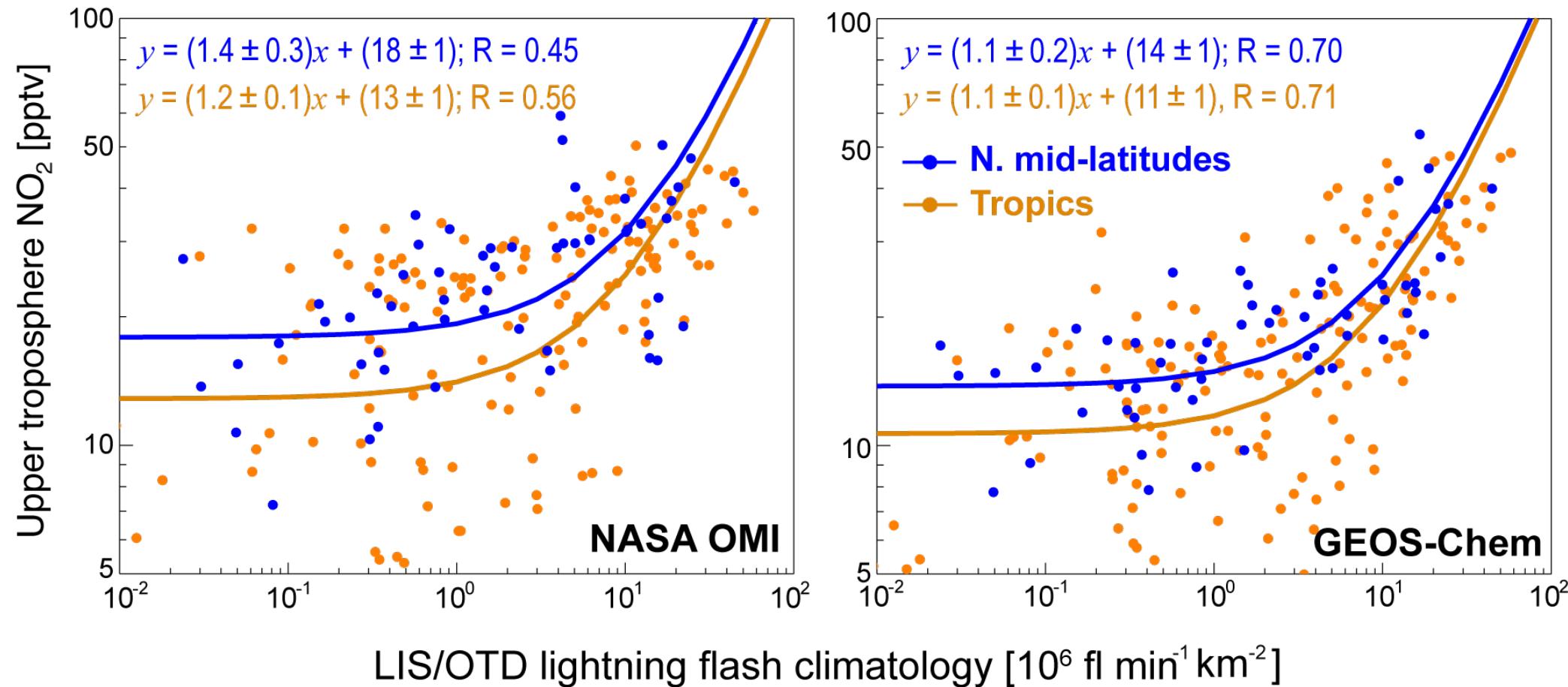


OMI:
Ozone Monitoring
Instrument

[Marais et al., 2018;
Choi et al., 2014]

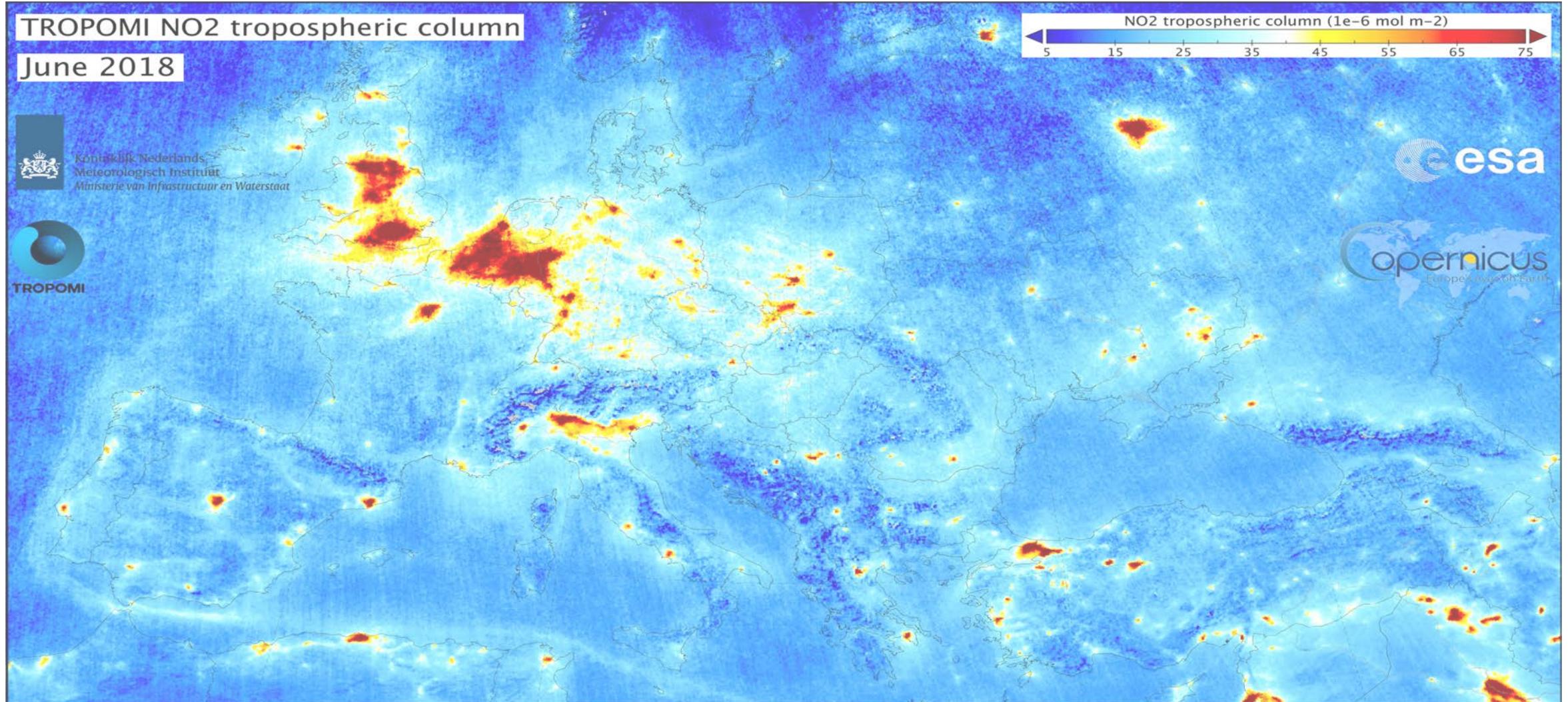
Provide improved constraints on lightning NO_x emissions

Log-log relationship between UT NO₂ from OMI and GEOS-Chem and satellite observations of lightning flashes in the **northern midlatitudes** and **tropics**



Similar slope in northern midlatitudes and tropics supports similar lightning NO_x production rates

Can we do better with high-resolution TROPOMI measurements?

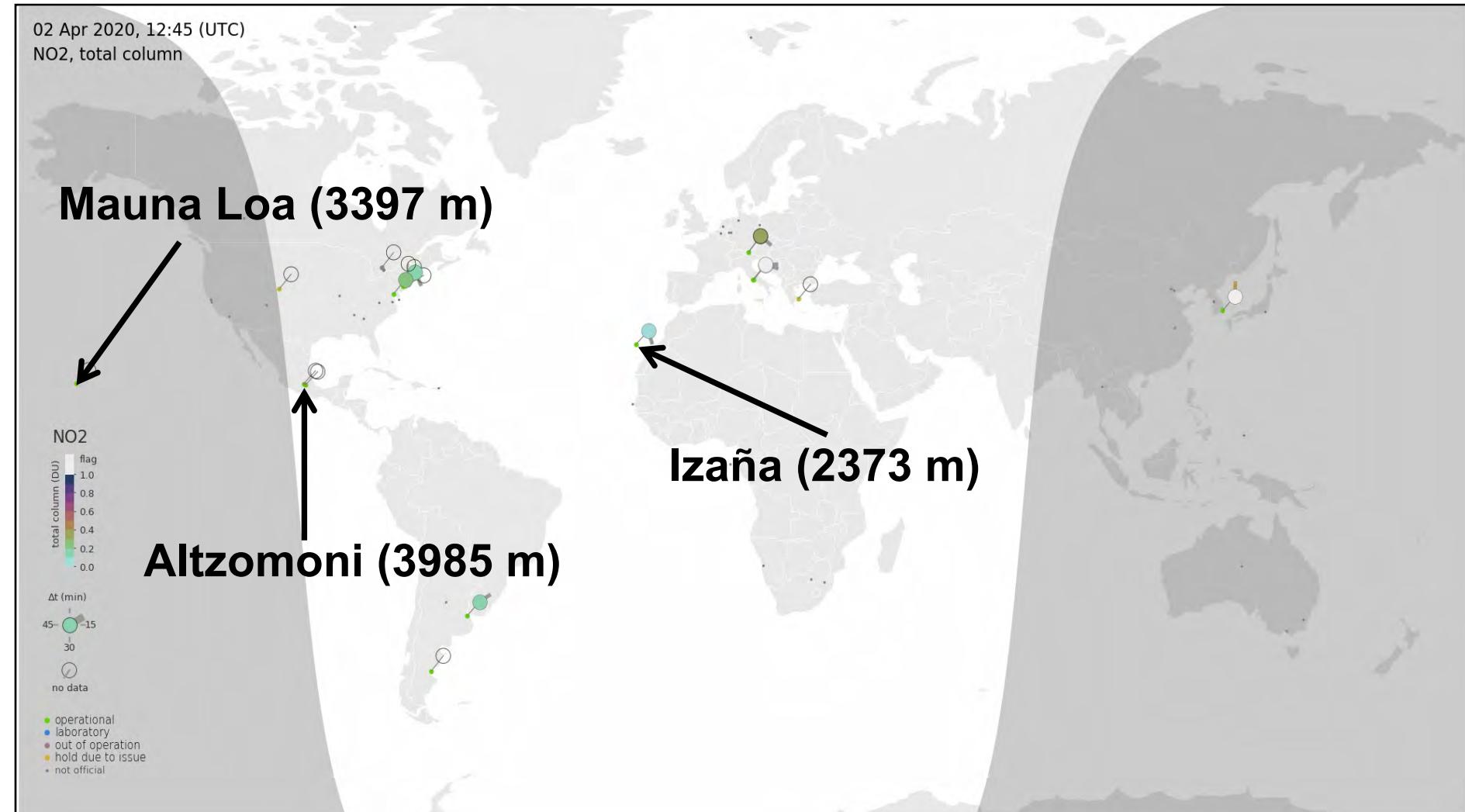
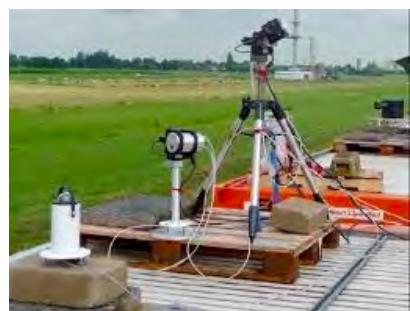


[Source: <http://www.tropomi.eu/data-products/nitrogen-dioxide>]

Nadir spatial resolutions in km (along × across): **13 × 24 (OMI)**; **5.6 × 3.5 (TROPOMI)**

Evaluate TROPOMI with ground-based measurements

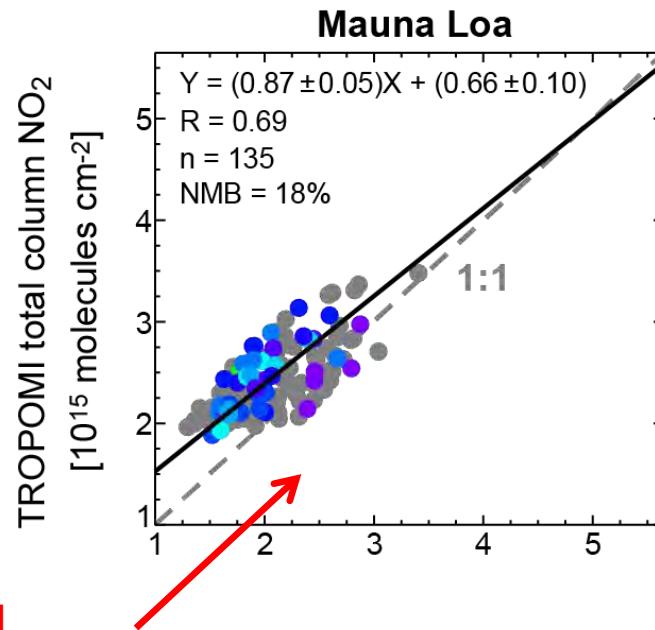
Global Pandora network, indicating locations of high-altitude sites (large relative contribution from the UT) used to evaluate TROPOMI



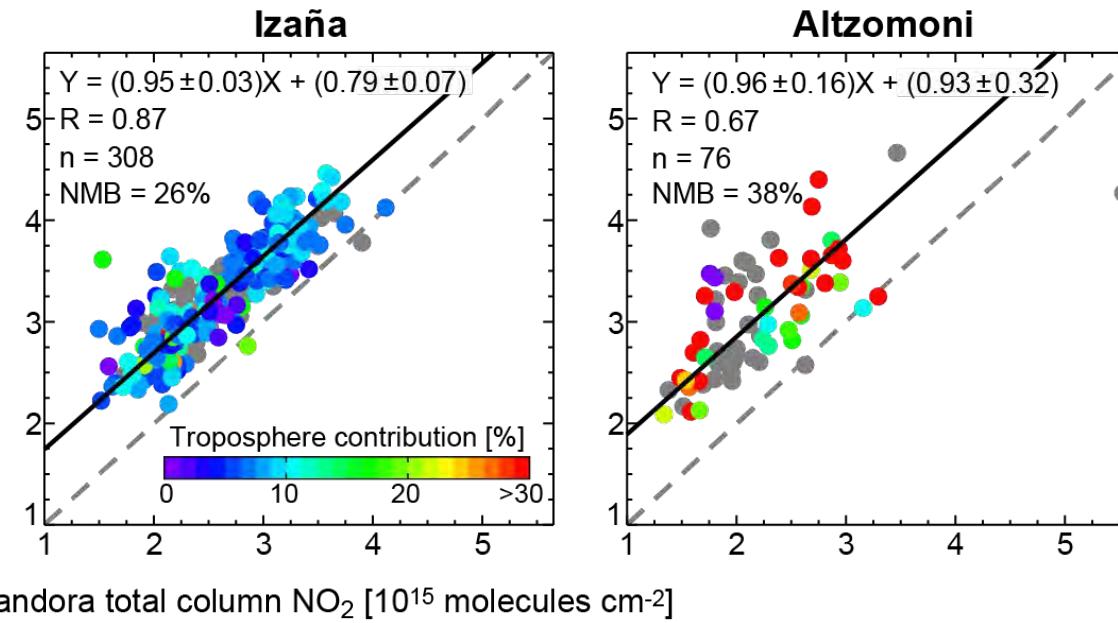
[Source: <https://www.pandonia-global-network.org/>]

Evaluate TROPOMI with ground-based measurements

Stratosphere:



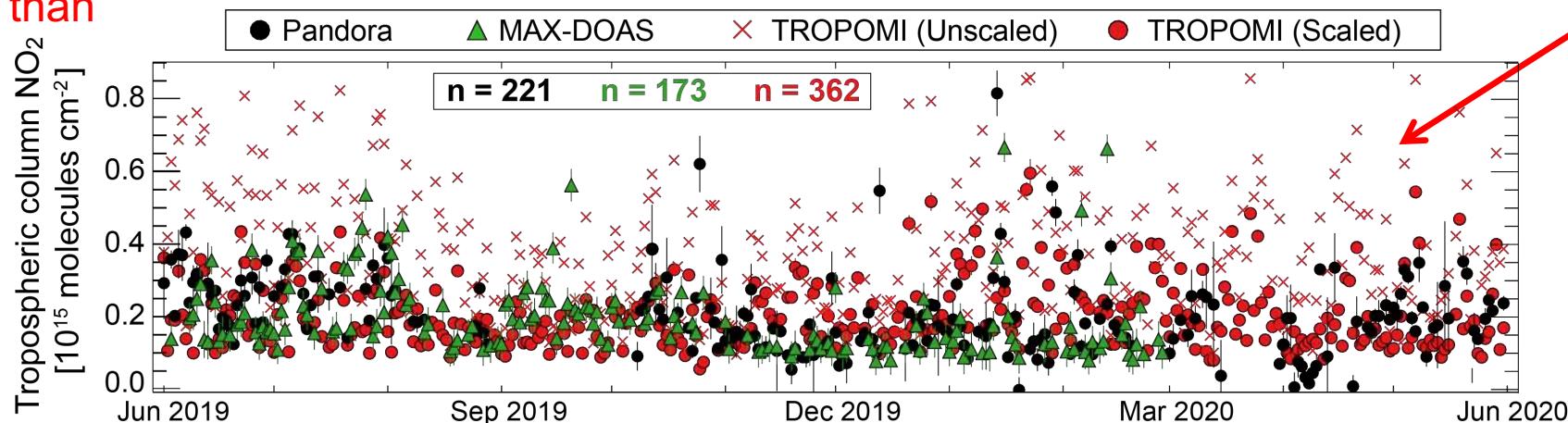
Stratosphere + free troposphere:



TROPOMI stratospheric column variance is less than Pandora

TROPOMI free tropospheric column is more than Pandora and MAX-DOAS

Free troposphere:



TROPOMI cloud-sliced upper tropospheric NO₂

Seasonal mean UT
NO₂ mixing ratios
from 2 distinct cloud
products at 1° × 1°

Range: 30-80 pptv

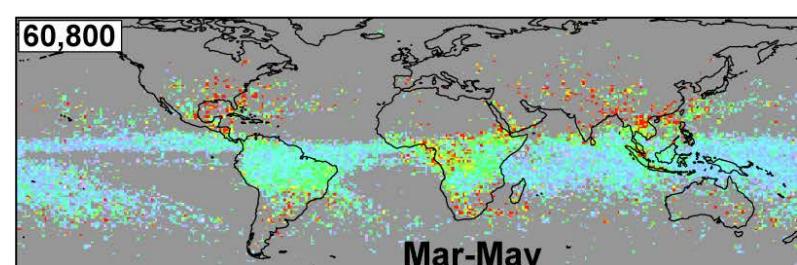
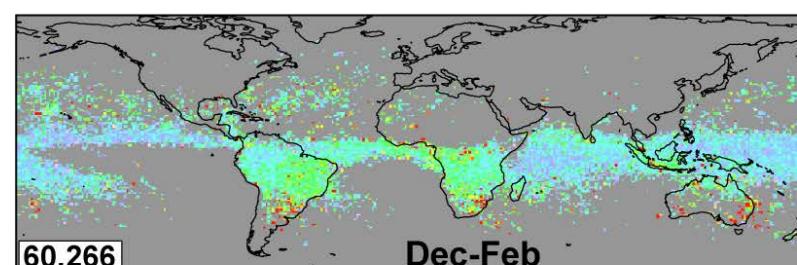
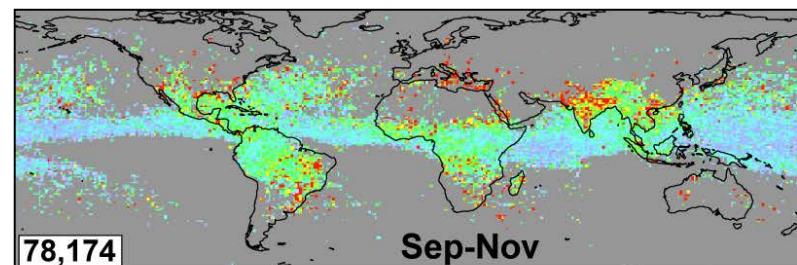
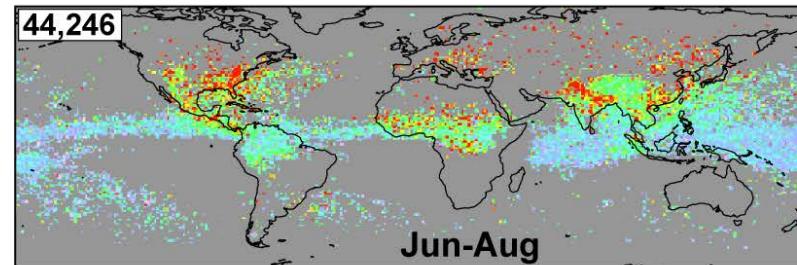
Background: ~30 pptv

Cloud products give
similar UT NO₂ in the
tropics

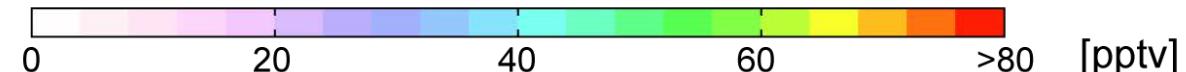
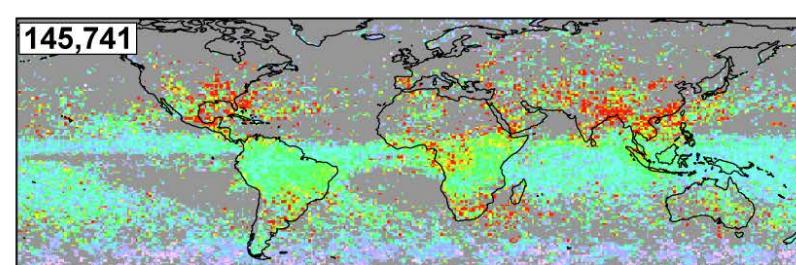
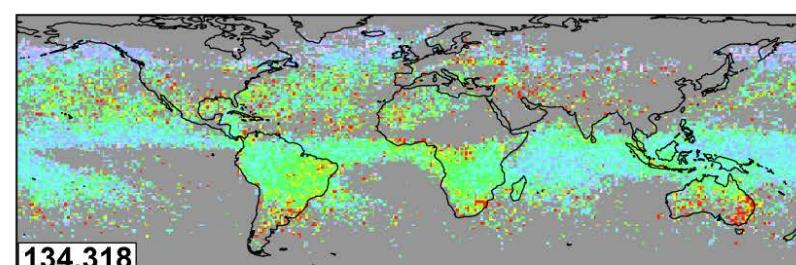
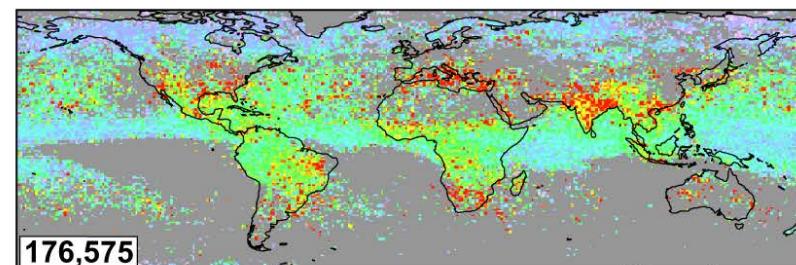
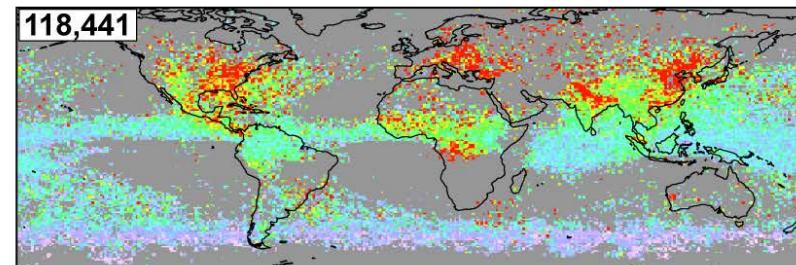
Greater coverage with
ROCINN-CAL

Some contamination:
Australia (fires), North
China (pollution)

Cloud Product One (FRESCO-S)

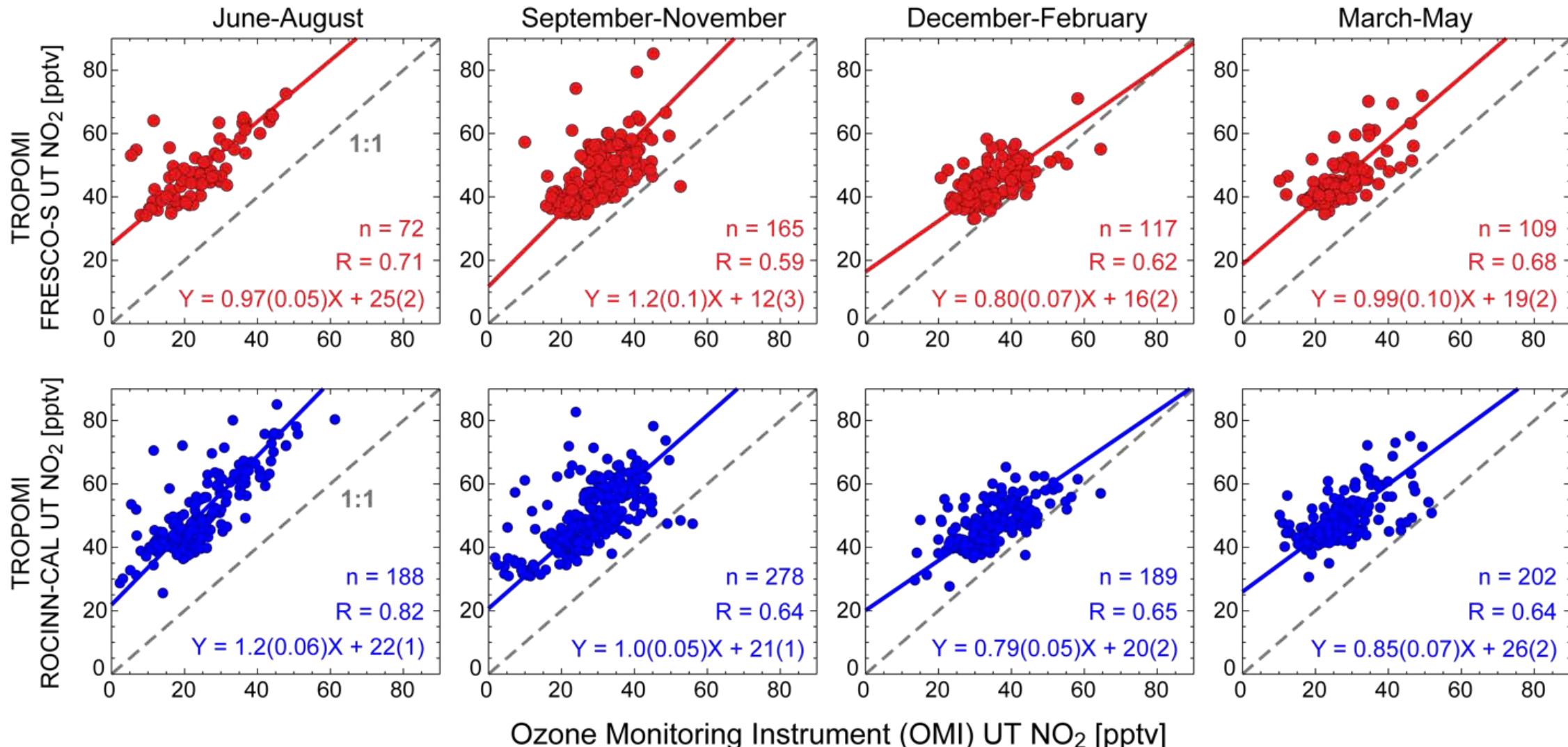


Cloud Product Two (ROCINN-CAL)



Comparison to the NASA OMI product

TROPOMI UT NO₂ obtained at 1° × 1° and gridded to the NASA product resolution (8° × 5°)

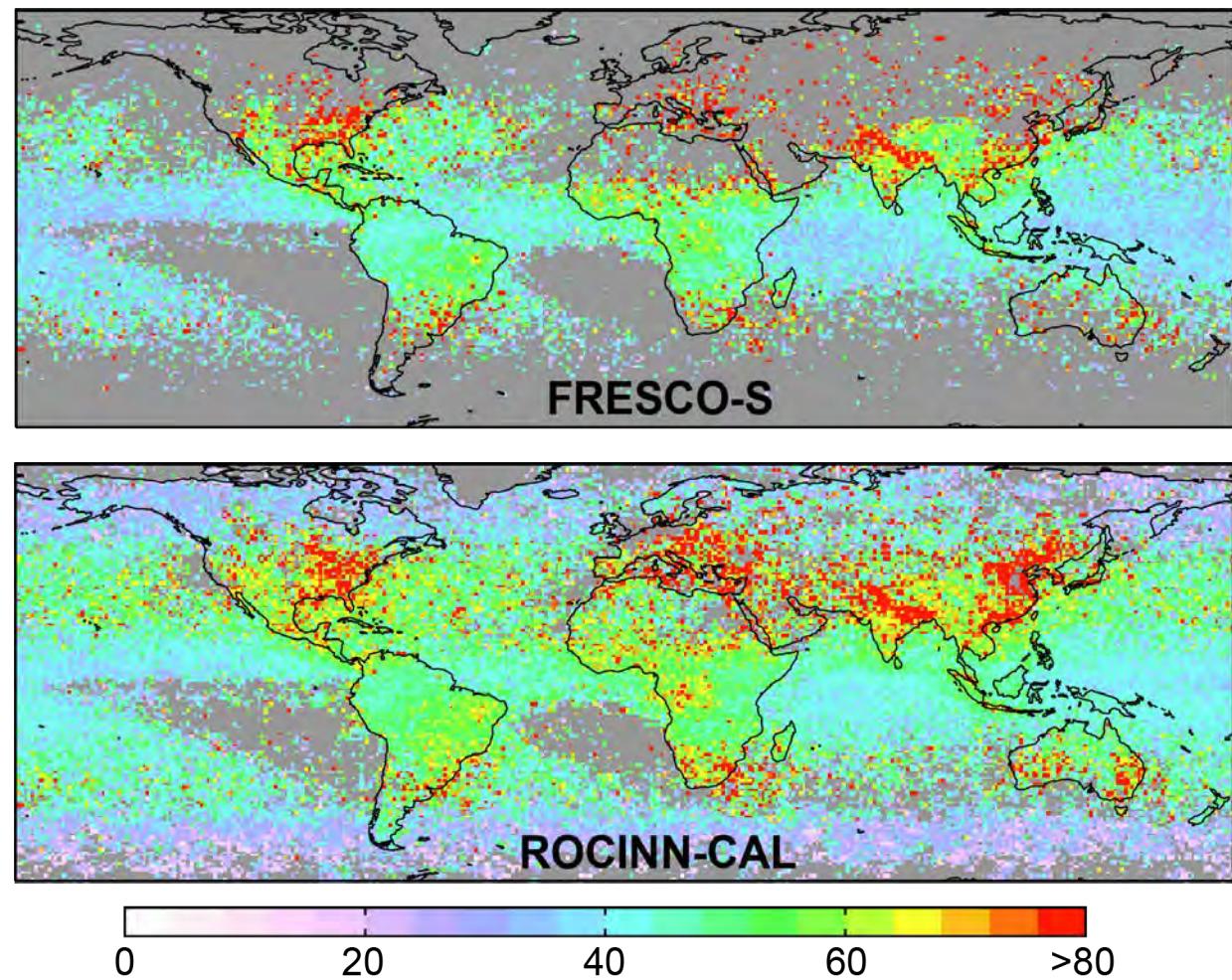


Spatial consistency. TROPOMI is 12-26 pptv more than OMI (retrieval, clouds, vertical NO₂ profile)

Concluding Remarks for TROPOMI UT NO₂

- Developed seasonal mean UT NO₂ from TROPOMI using cloud-slicing at finer resolution than ever before
- Two datasets from distinct cloud products
- Differences in cloud products, particularly in higher latitudes, lead to differences in coverage
- Consistency with existing OMI product
- Potential to address data sparsity in the UT
- We're using TROPOMI UT NO₂ to detect the influence of aircraft on UT NO₂ and quantify uncertainties in reactive nitrogen in the UT with GEOS-Chem

TROPOMI annual mean upper tropospheric NO₂ [pptv]



Interested in using the TROPOMI UT NO₂ product? Contact me: e.marais@ucl.ac.uk

Back to Earth's surface...

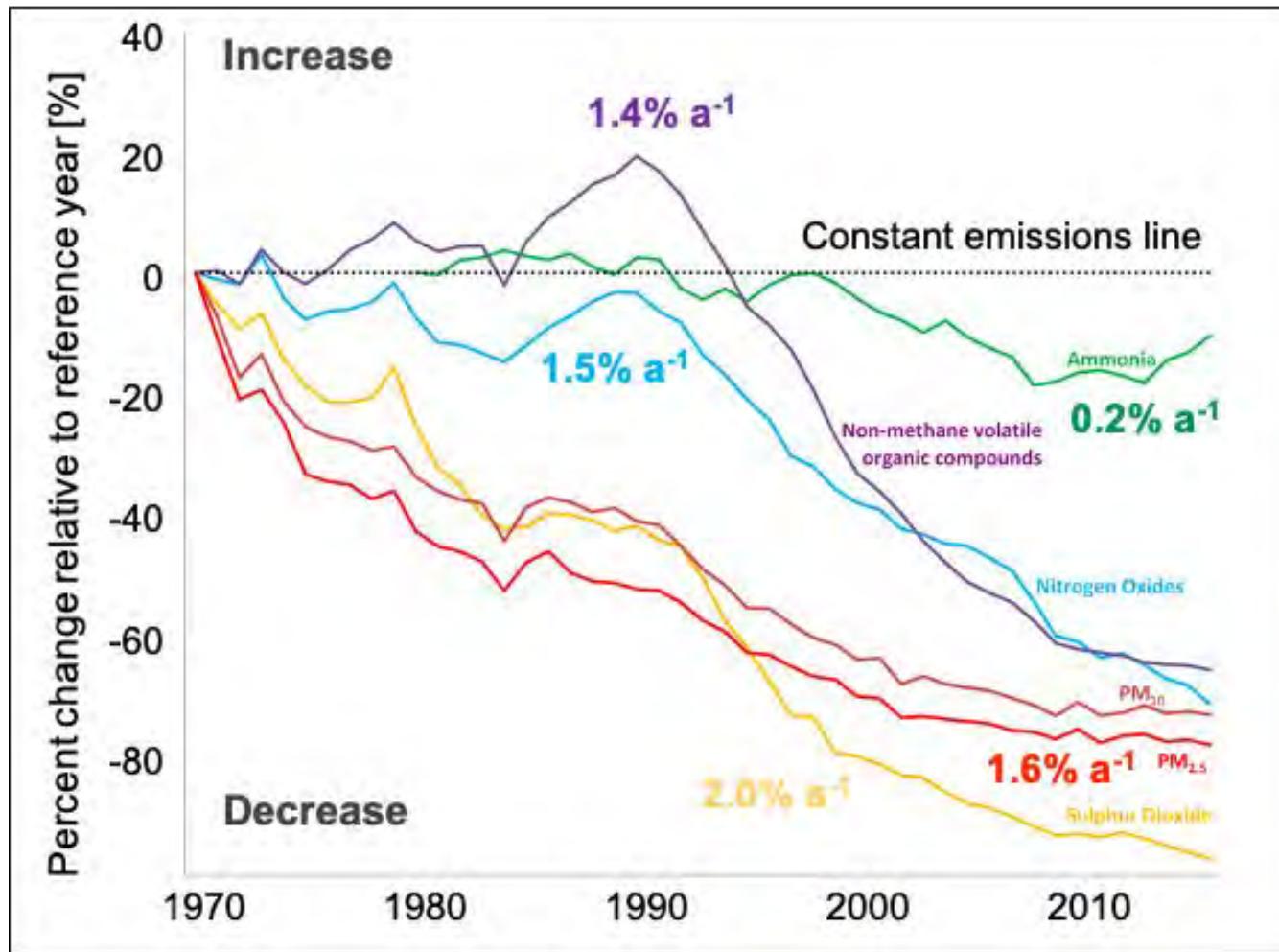
Top-down estimate of UK ammonia emissions



Contributors: A. Pandey, M. Van Damme, L. Clarisse, P. F. Coheur, M. Shephard, K. Cady-Perreira, T. Misselbrook, L. Zhu, F. Yu, G. Luo

Ammonia emissions in the UK: the bottom-up perspective

Temporal (Time) Variability in Emissions

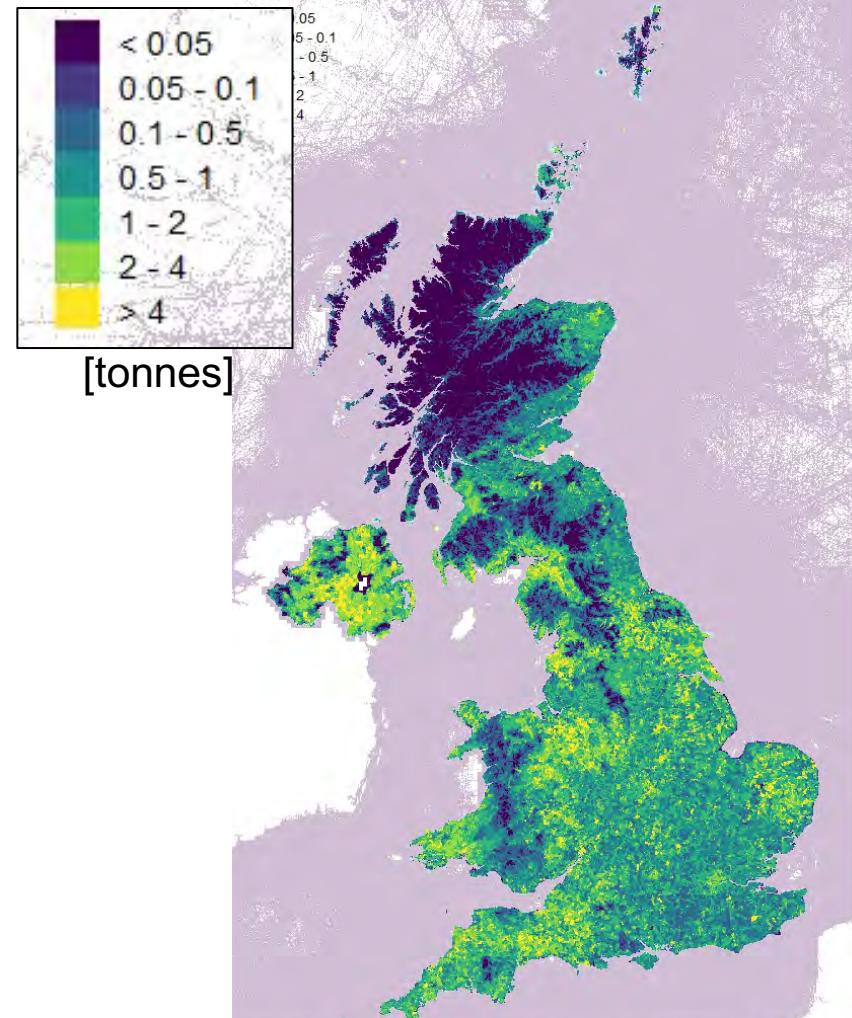


[Adapted from Defra, 2018]

Successful decline in all emissions, except ammonia (NH₃)

Spatial Variability in Emissions

NH₃ emissions for 2018 at 1 km

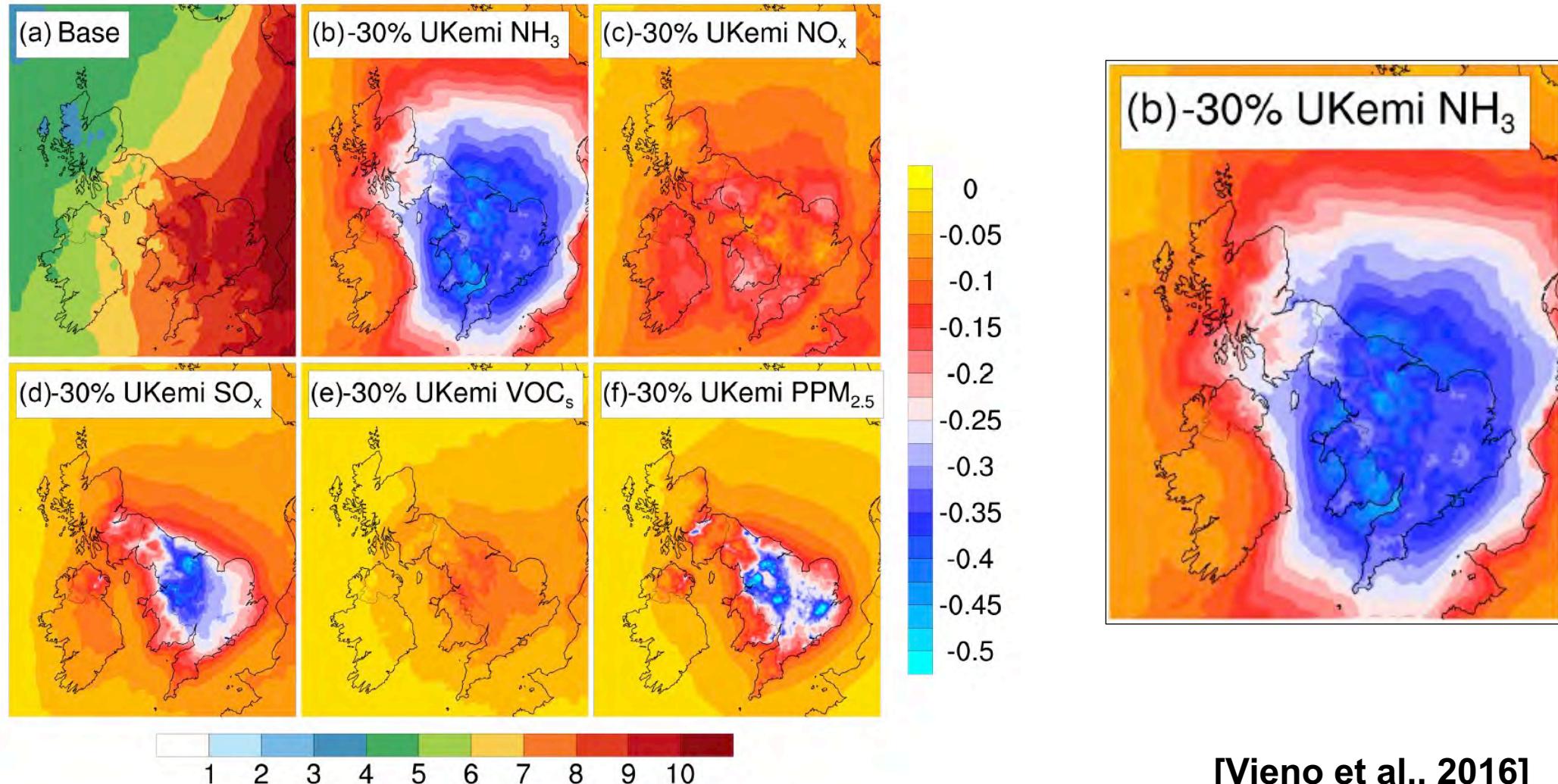


All maps © Crown copyright. All rights reserved Defra, Licence number 100022861 (2020) and BEIS, Licence number 100037028 (2020) LPS © Crown copyright and database right 2020 Licence INSP594

[Adapted from <https://naei.beis.gov.uk/data/>]

Ammonia impact on air pollutants hazardous to health

Effect of precursor emission reductions on 2010 PM_{2.5}



[Vieno et al., 2016]

Largest and most extensive decline in PM_{2.5} achieved by targeting ammonia sources

Top-down emissions estimated with satellite observations

Convert atmospheric **column concentrations** to surface **emissions** by relating the two with a **model**

ABUNDANCES

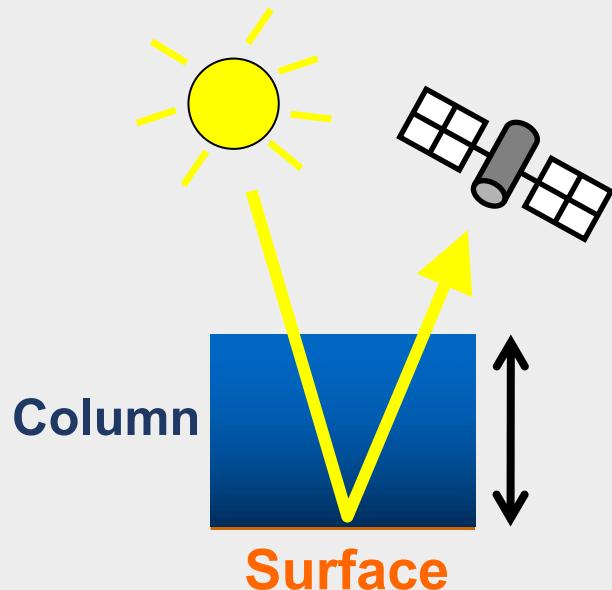


Conversion Factor

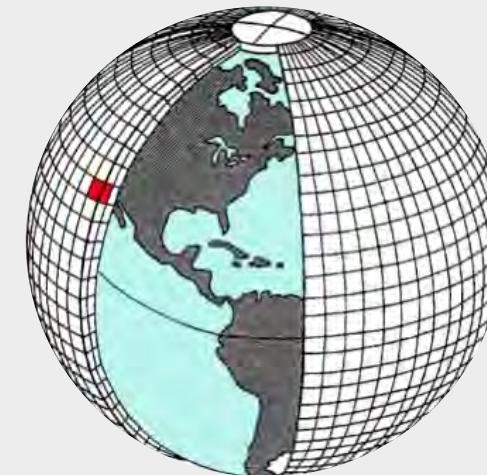


EMISSIONS

**Satellite columns
(IASI)**



**Column-to-Emission ratio
(GEOS-Chem)**



**Satellite-derived
Surface Emissions**

Emission



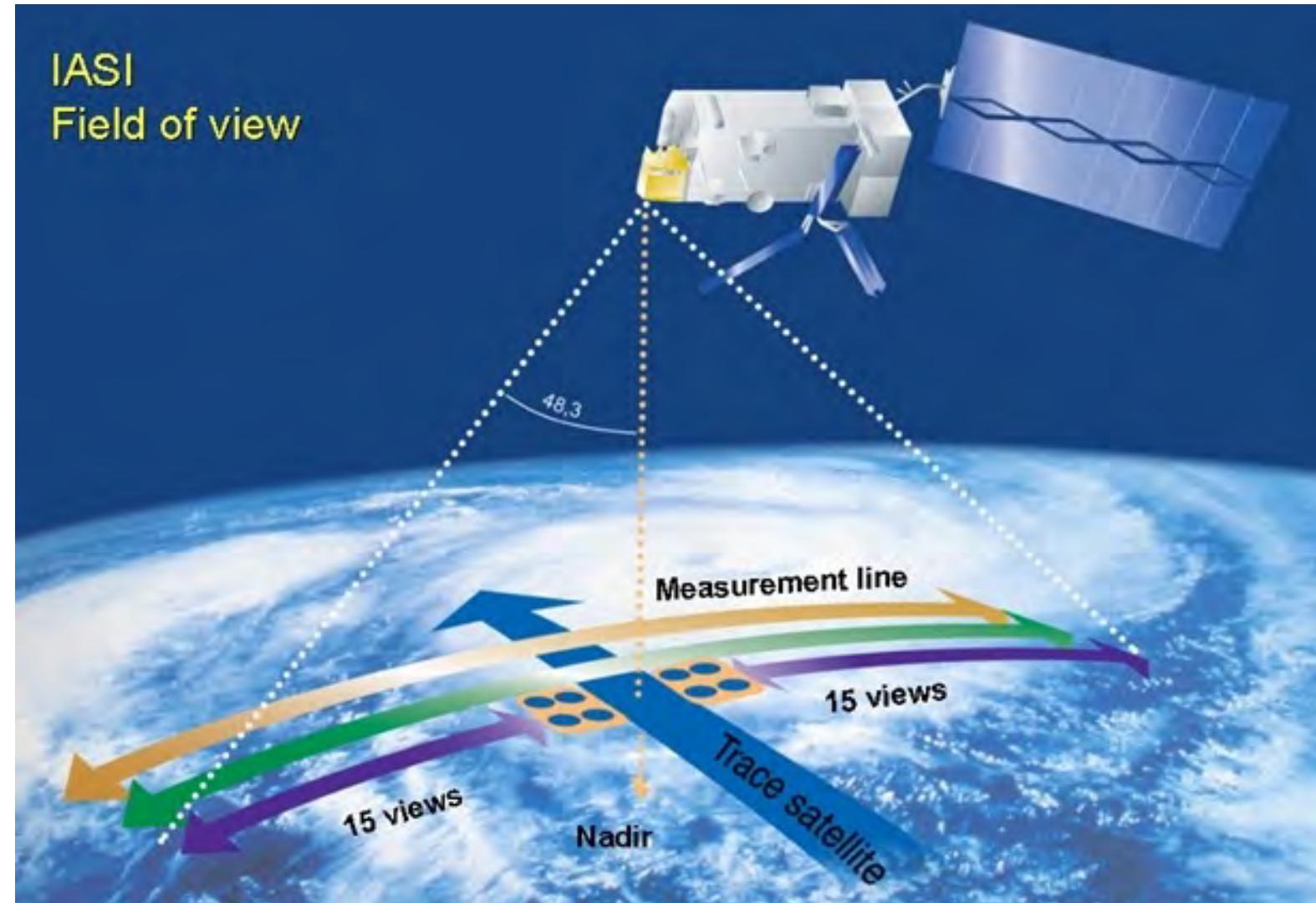
Infrared Atmospheric Sounding Interferometer (IASI) Instrument

Overpass:
9:30 local solar time

Spatial resolution:
12 km to 39 km

Swath width:
2200 km

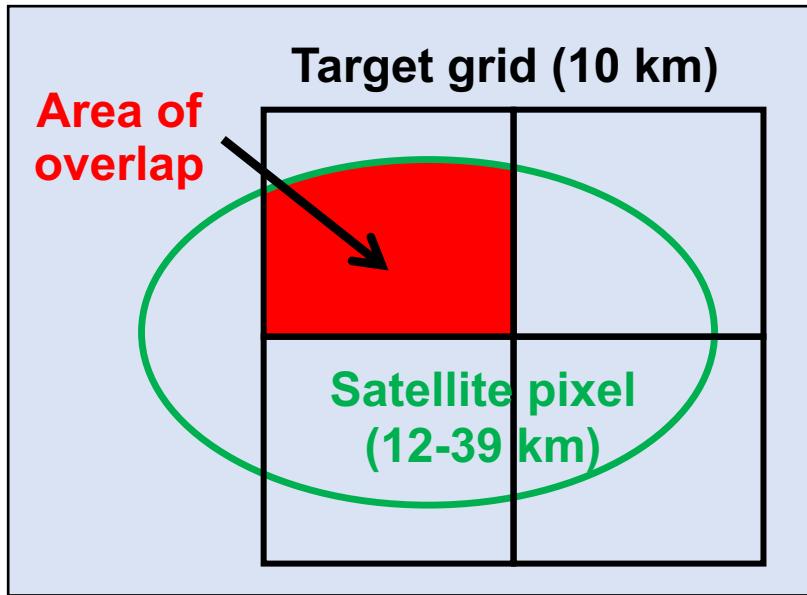
Launch date:
October 2006



Ammonia emissions in the UK: the top-down perspective

Enhance the spatial resolution relative to the native resolution of the instrument by oversampling

Oversampling Technique

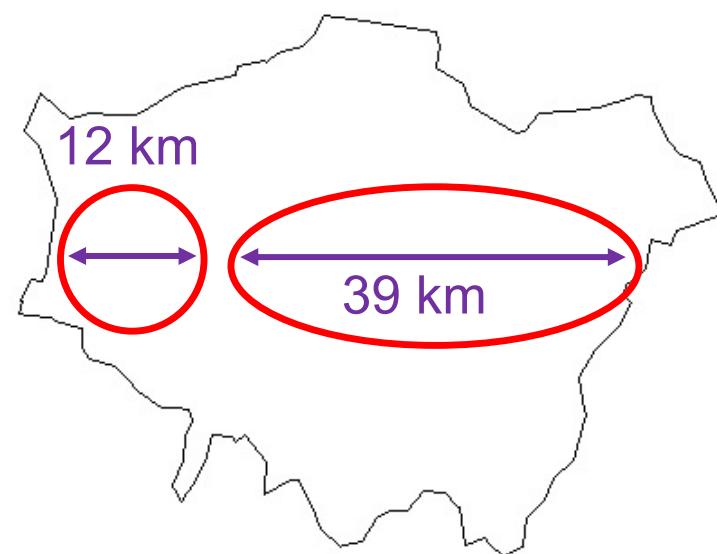


Weights each IASI NH₃ pixel by area of overlap and the reported uncertainty

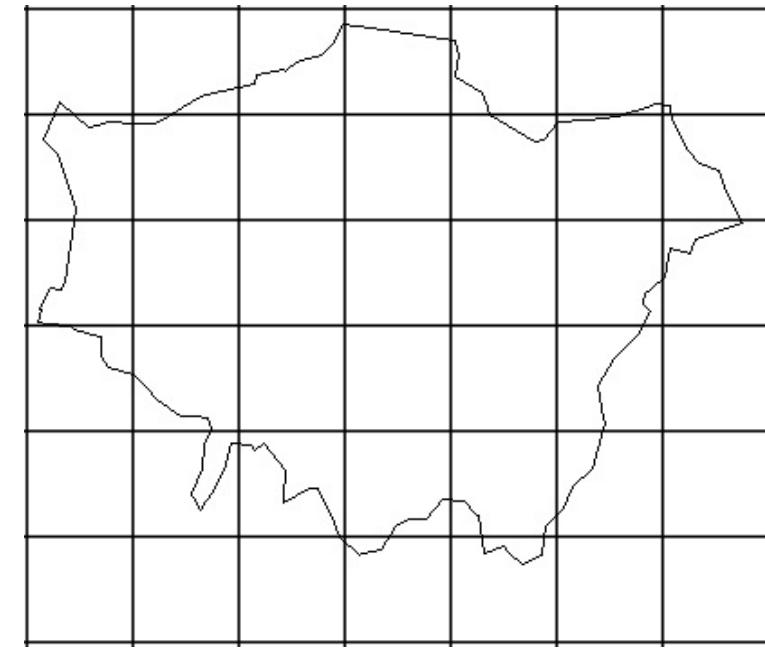
Oversampling code: L. Zhu,
SUSTech (Zhu et al., 2017)

Oversampling technique over London

IASI ground pixel



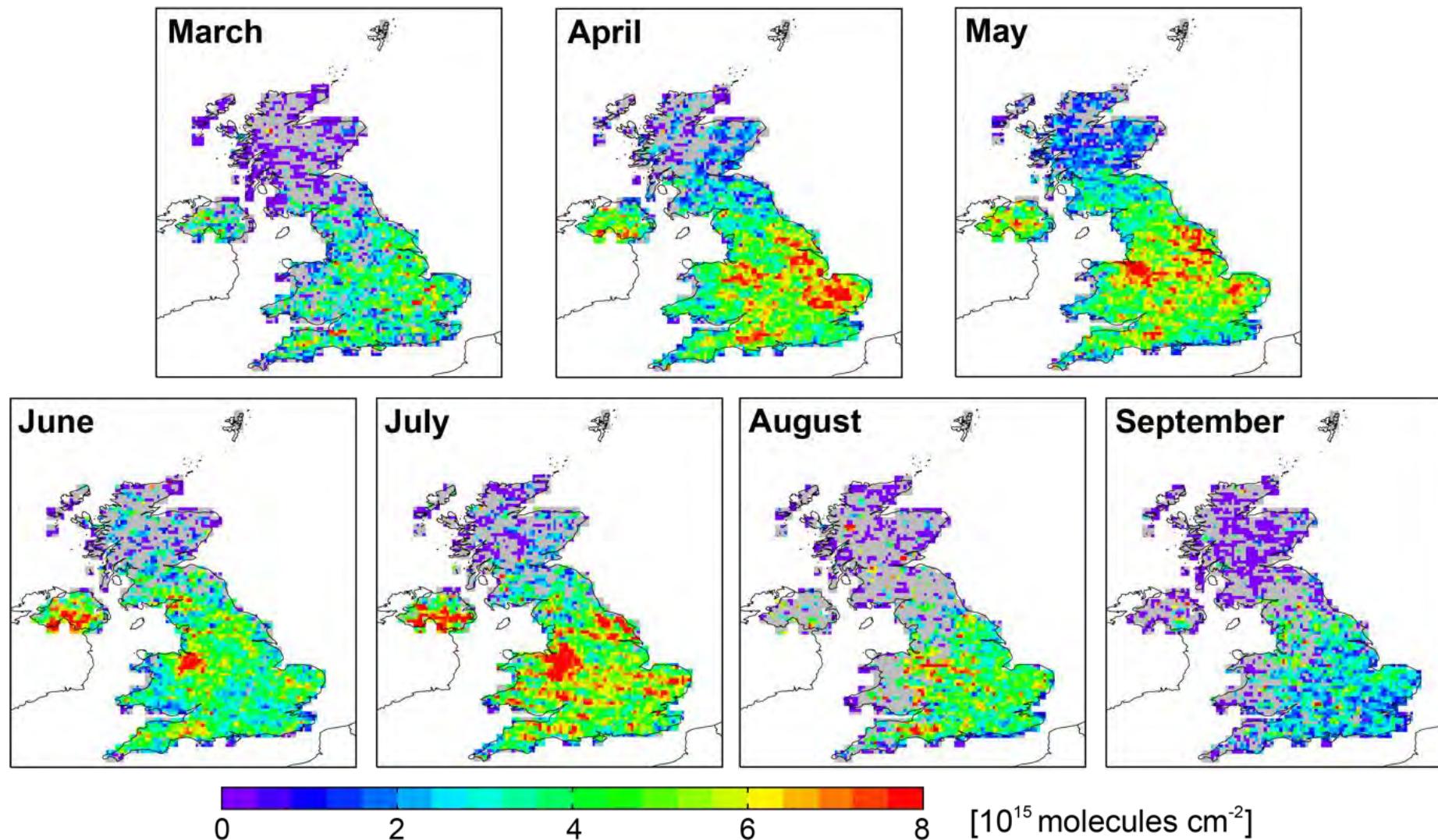
0.1° x 0.1° (~10 km) grid



Lose time (temporal) resolution; gain spatial resolution

Multiyear (2008-2018) monthly mean IASI NH₃ at 0.1° x 0.1°

IASI NH₃ retrieved using spectral enhancement due to NH₃ and machine learning (neural network)

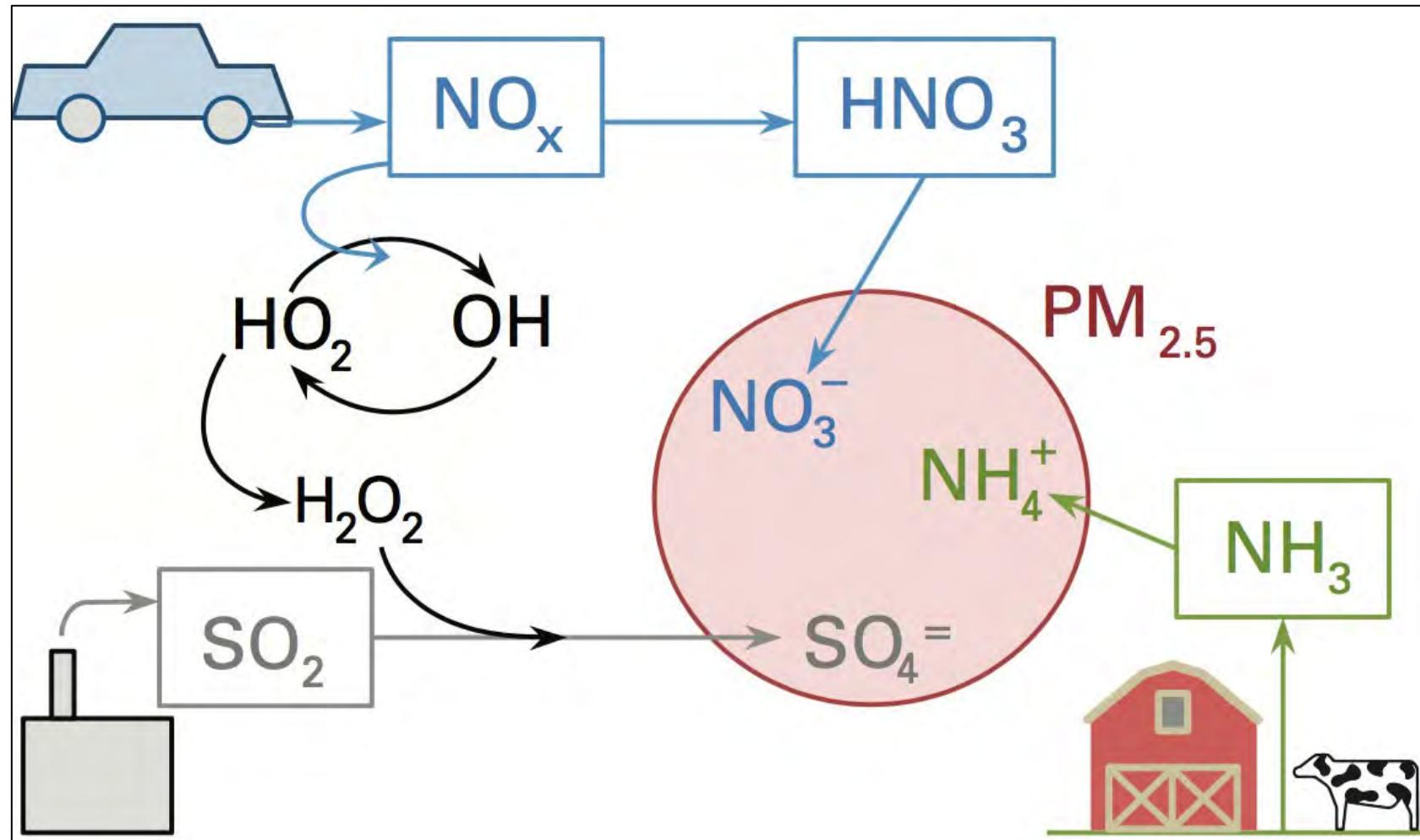


Data for Oct-Feb and over Scotland have low signal and large relative retrieval error (> 100%)

IASI data providers: M. Van Damme, L. Clarisse, P.-F. Coheur, ULB, Belgium

Ammonia abundance depends on numerous factors

Ammonia partitions to aerosols to form PM_{2.5}



[<http://climate-science.mit.edu/>]

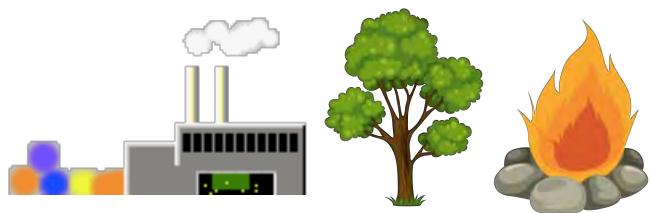
Partitioning of ammonia (NH₃) to pre-existing aerosols depends on abundance of NO_x and SO₂

Surface SO₂ concentrations calculated with GEOS-Chem



3D Atmospheric Chemistry Transport Model

Emissions
(natural/human)

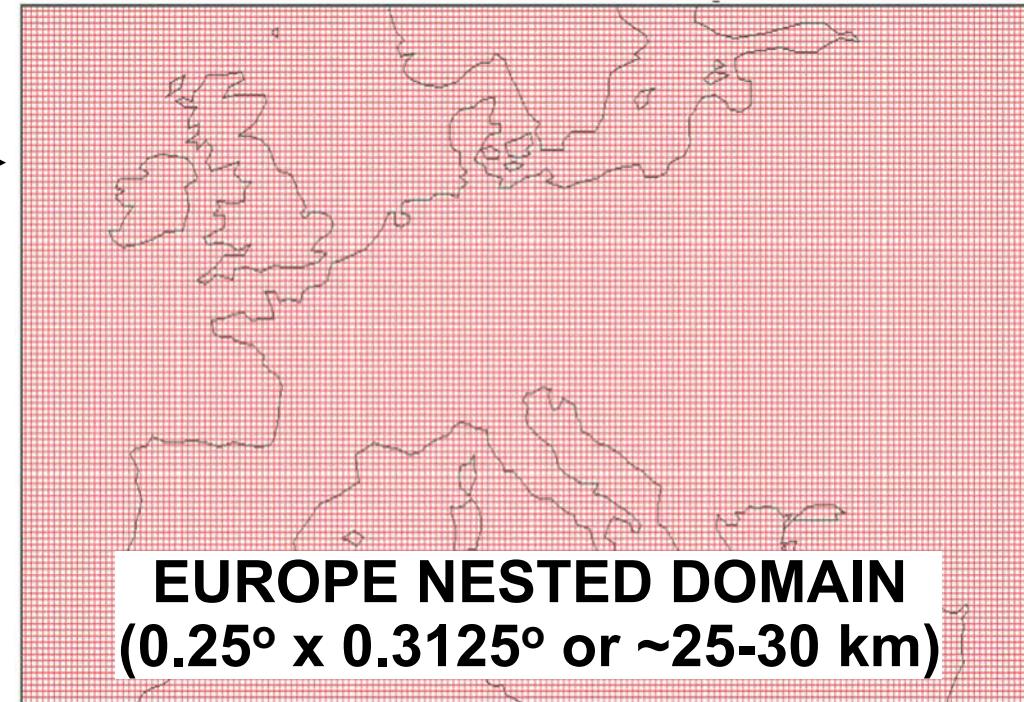


UK NAEI emissions
(with temporal information)

Offline assimilated
meteorology

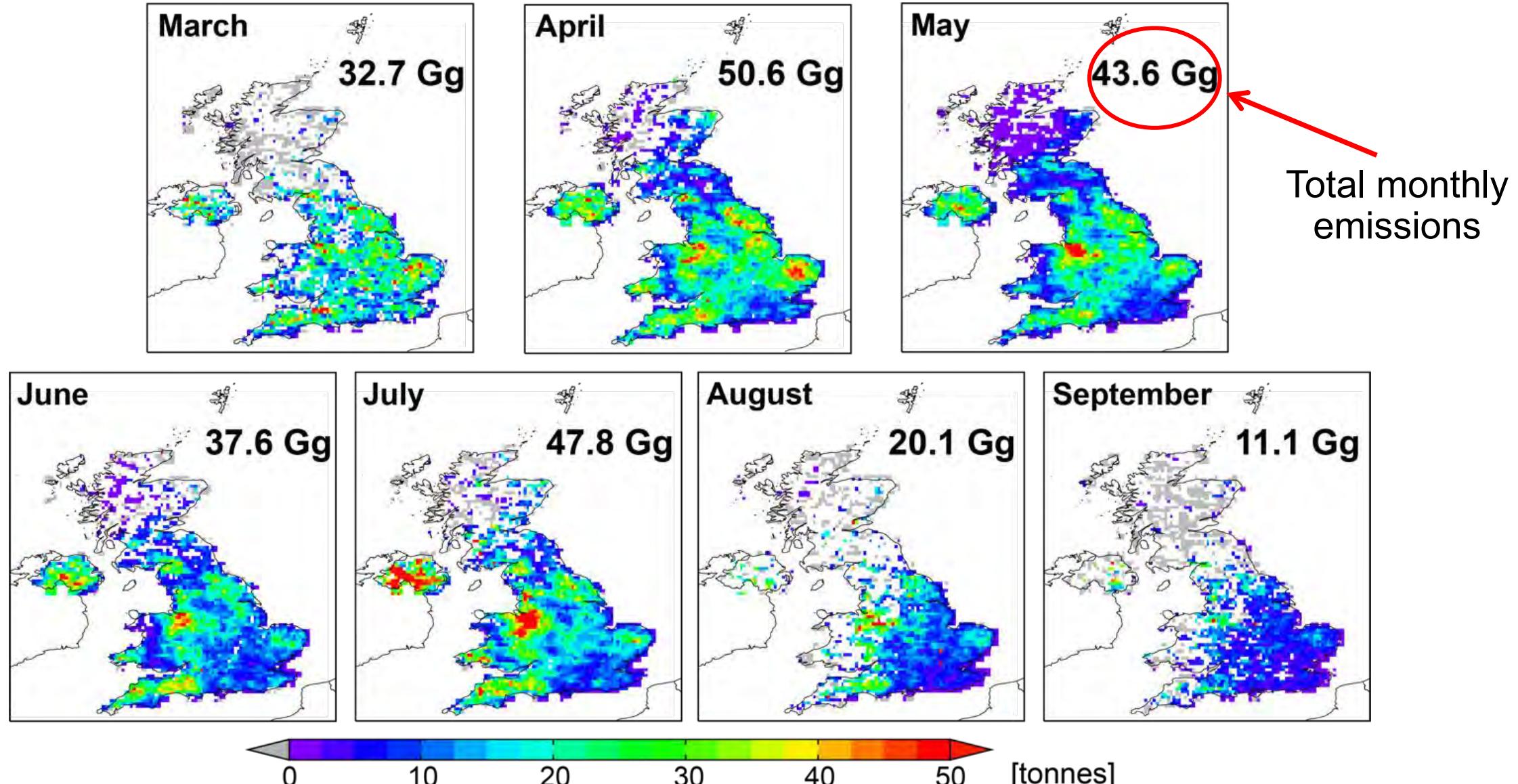


NASA GEOS-FP for 2016



Gas phase and heterogeneous chemistry
Transport
Dry/wet deposition

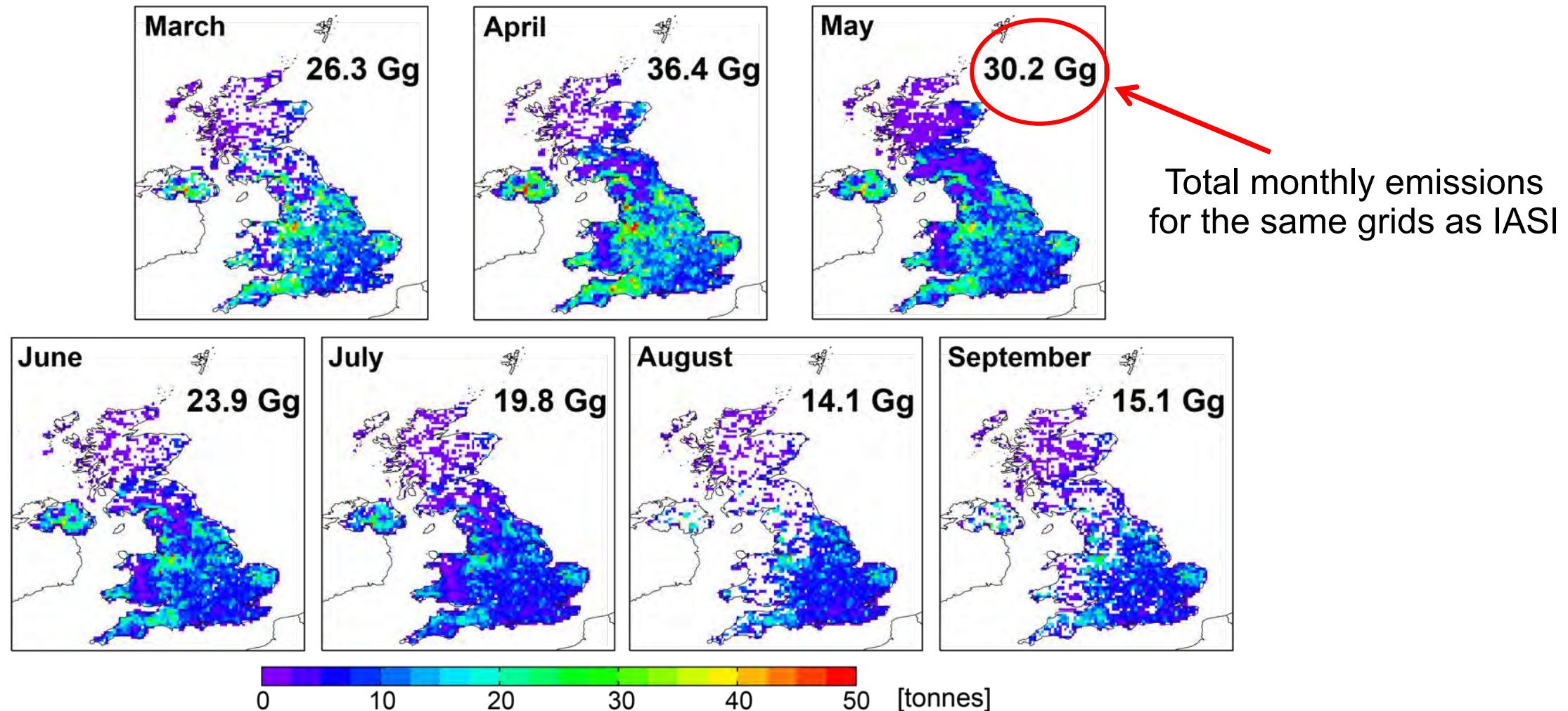
IASI-derived NH_3 emissions at $0.1^\circ \times 0.1^\circ$



Sum of IASI-derived emissions for retained grids: **243.5 Gg**

Monthly emissions from NAEI and GEOS-Chem at $0.1^\circ \times 0.1^\circ$

Obtained by multiplying NAEI annual emissions by GEOS-Chem emissions seasonality



Sum of NAEI emissions: **165.8 Gg** (56% of annual total, 32% less than IASI)

Suggests annual total IASI emissions of **435 Gg** (> NAEI annual total of 298 Gg ceiling)

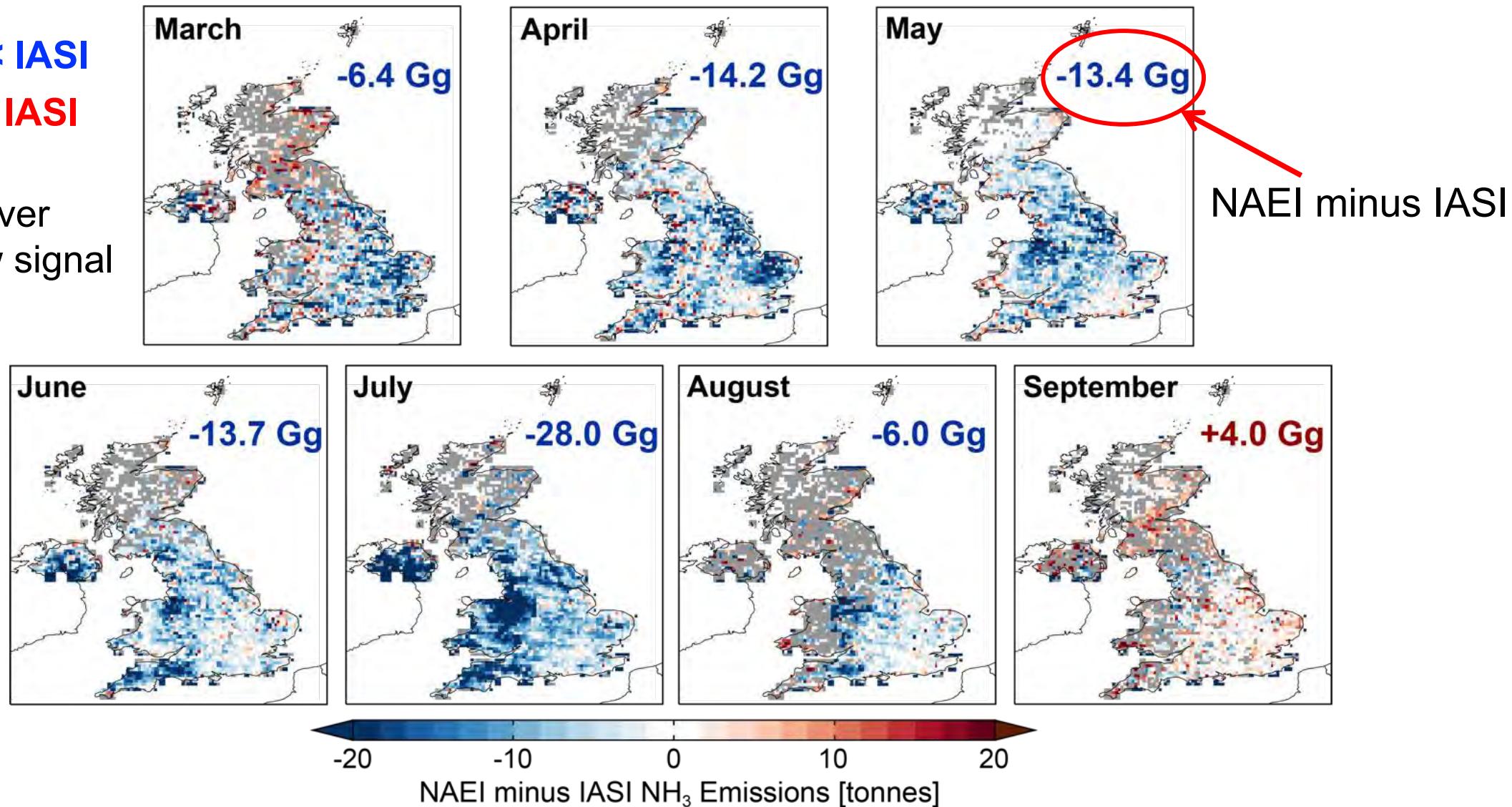
Differences in spatial distribution of IASI and NAEI

IASI-derived emissions minus NAEI-GEOS-Chem emissions

Blue: NAEI < IASI

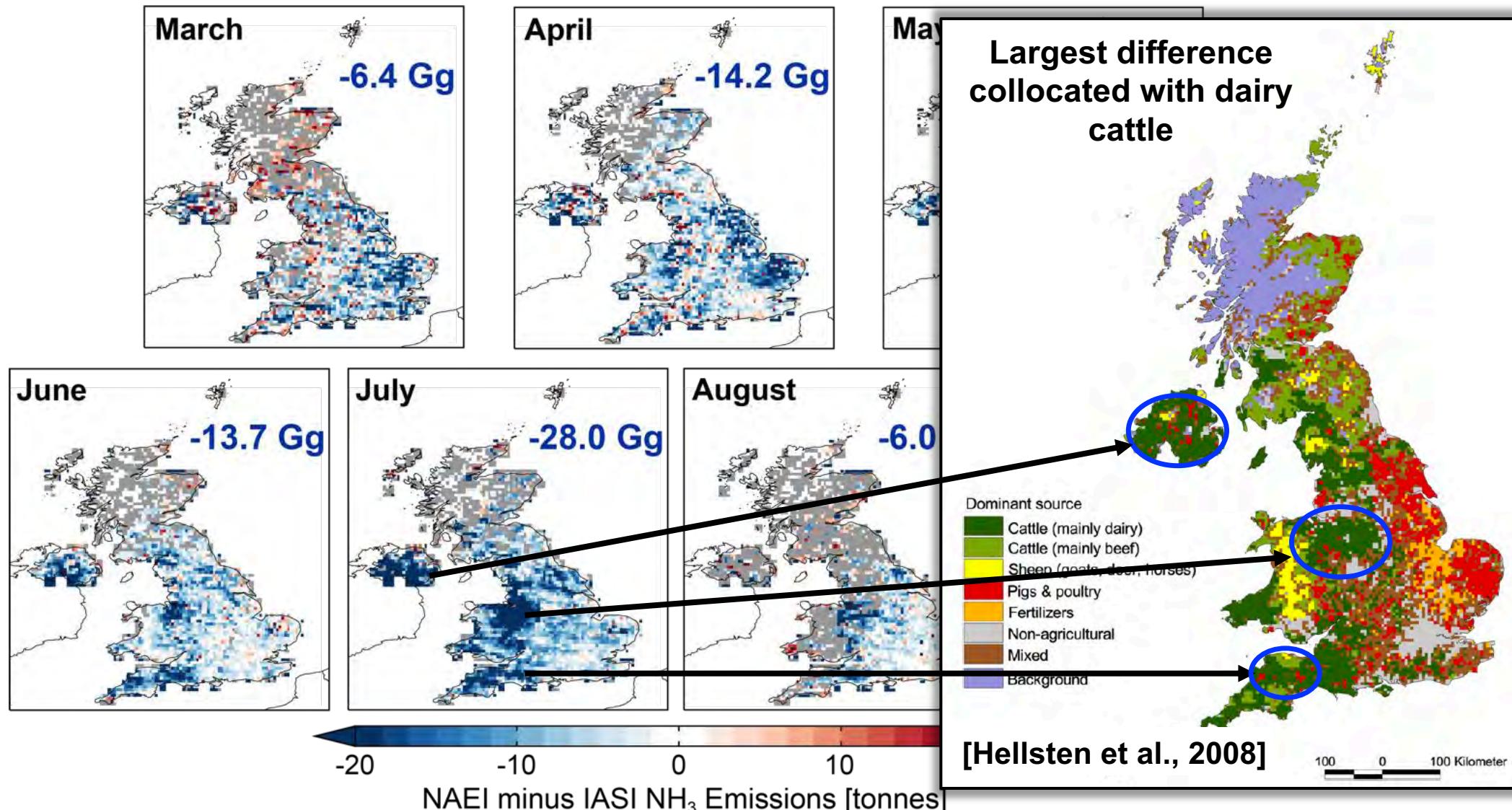
Red: NAEI > IASI

Red mostly over
grids with low signal



Differences in spatial distribution of IASI and NAEI

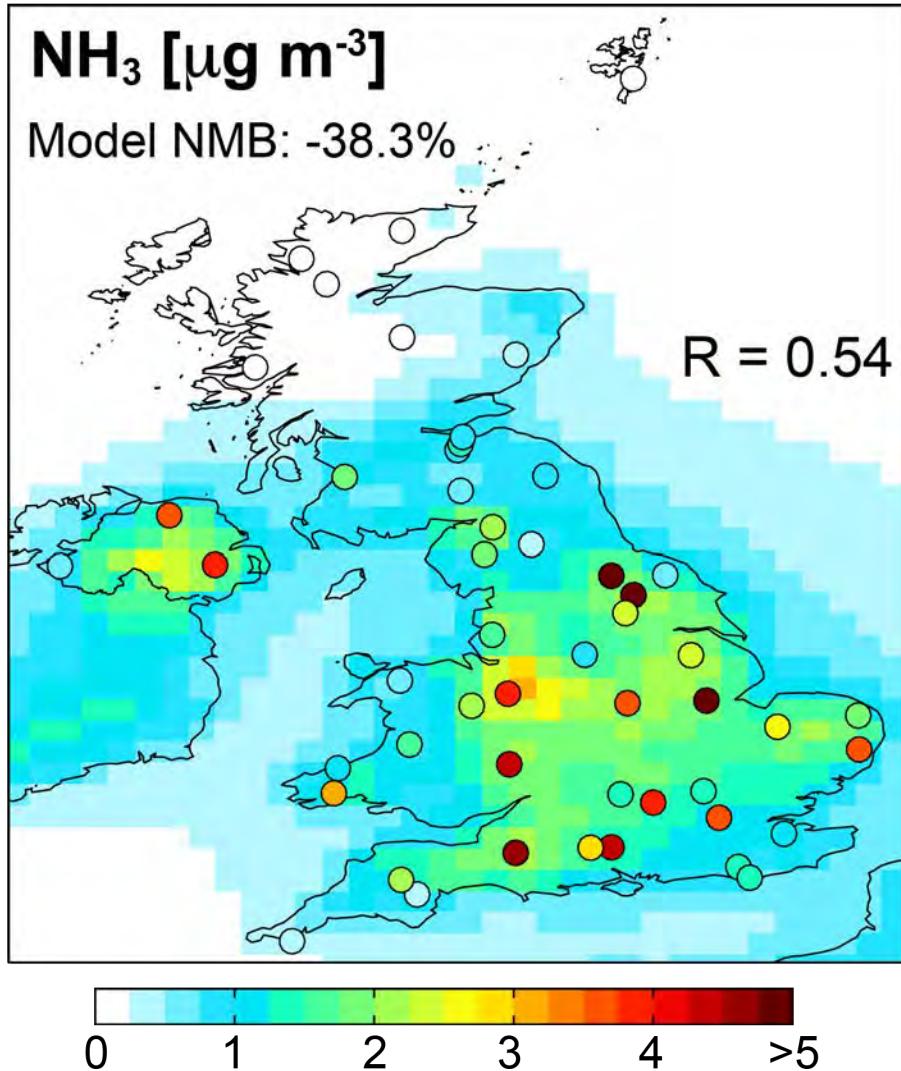
IASI-derived emissions minus NAEI-GEOS-Chem emissions



Largest discrepancy over locations dominated by **dairy farms**

Surface network observations corroborate top-down results

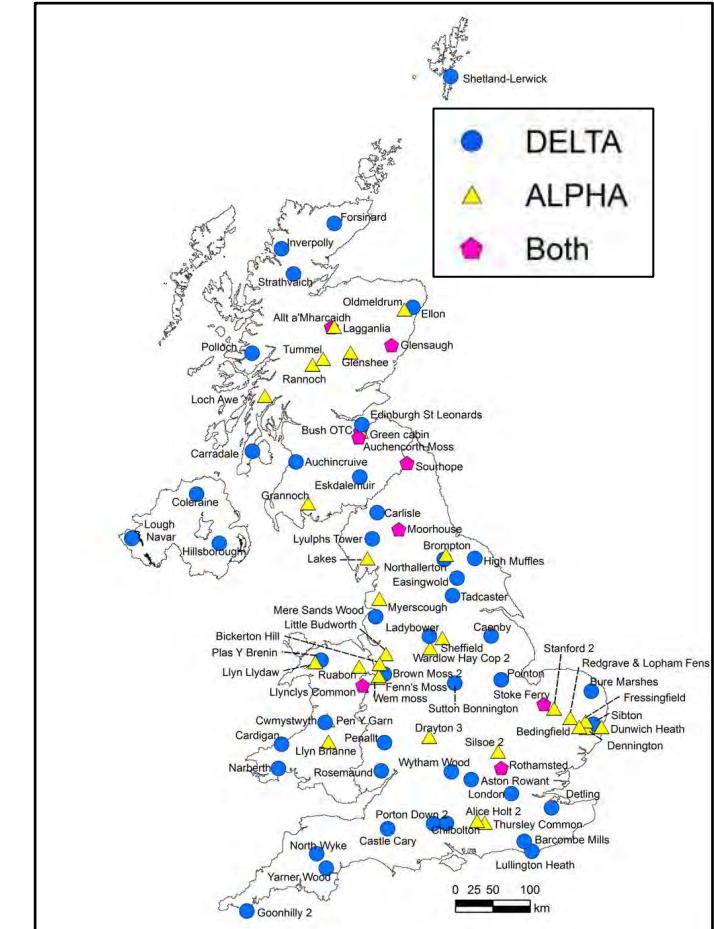
Network (points) and model (background)
surface NH_3 in Mar-Sep



Points are for DELTA
instruments (blue circles)

DELTA instruments support
model underestimate
(NMB = -38%)

So do passive low-cost
ALPHA instruments (yellow
triangles)
(NMB = -41.5%)



GEOS-Chem underestimate in surface NH_3 driven with the NAEI corroborates results from IASI

Concluding Remarks for top-down NH₃ emissions

- IASI-derived emissions are 34-46% more than those from the UK National Atmospheric Emission Inventory (NAEI).
- IASI-derived emissions random error is 4-35%.
- Largest discrepancy between top-down and bottom-up inventory is in July in locations dominated by dairy farms, but IASI data may be susceptible to a clear-sky bias in July.
- NAEI-IASI difference is similar to reported errors for the NAEI (31%), but problematic for policy development and estimating environmental impact.

