

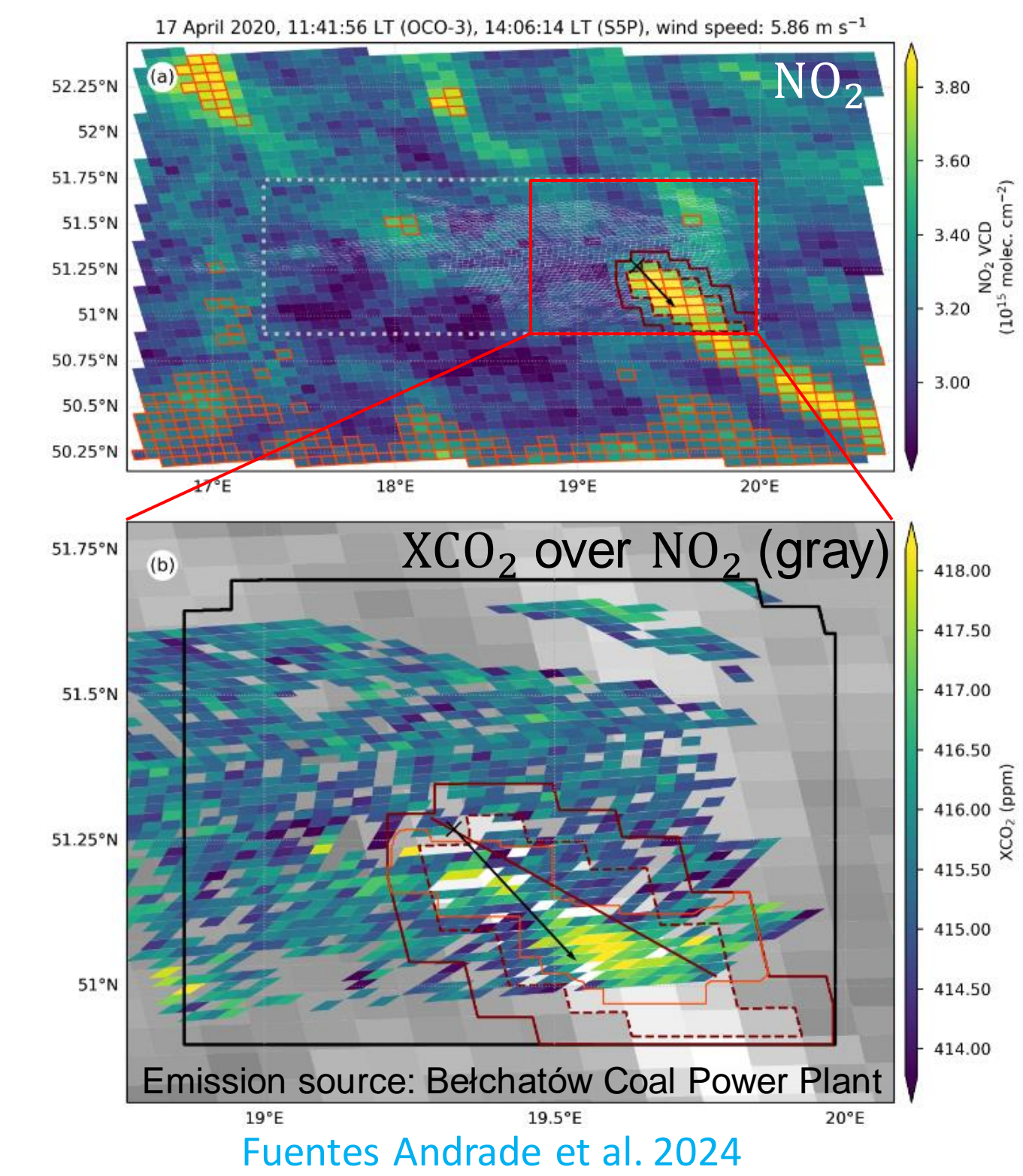
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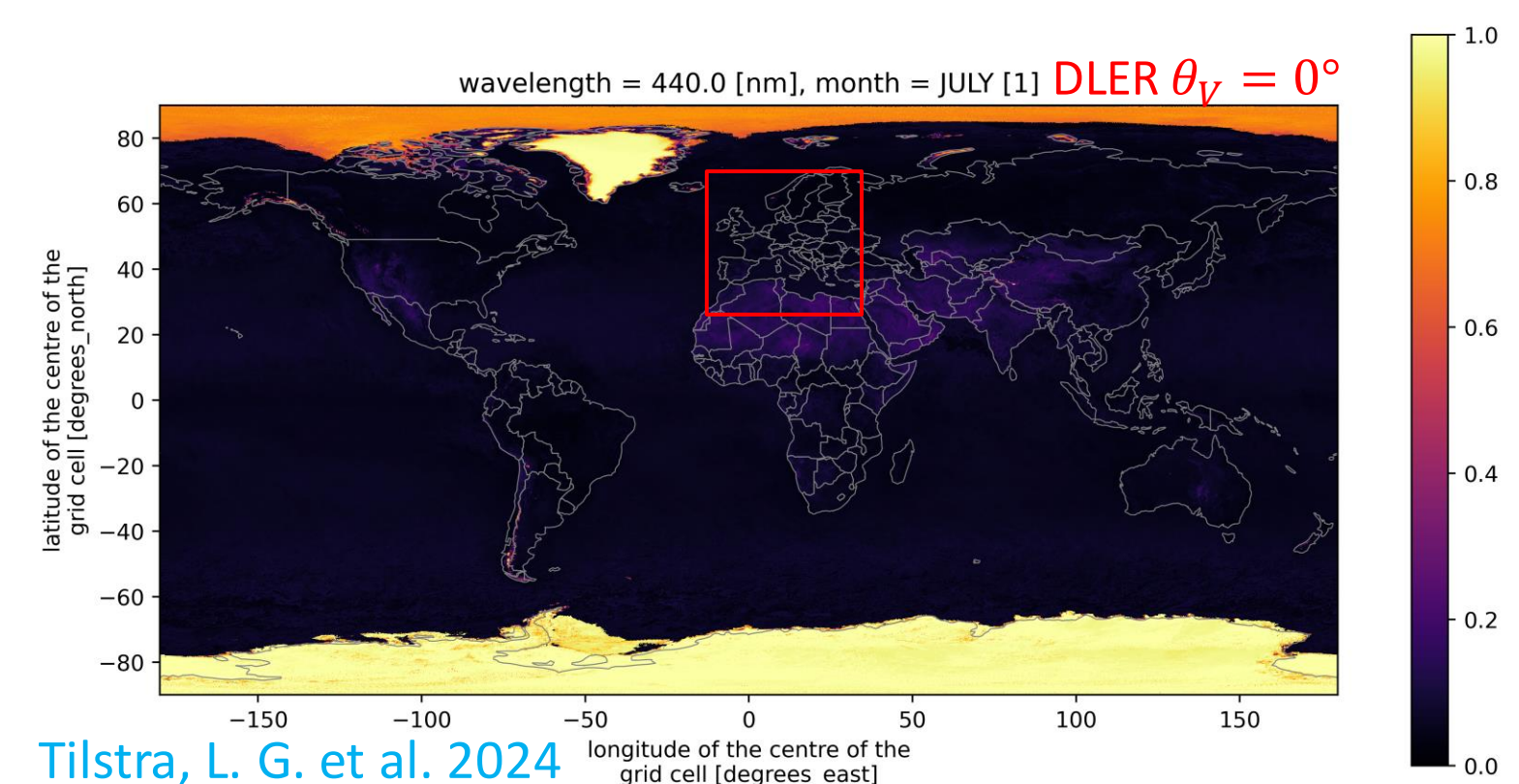
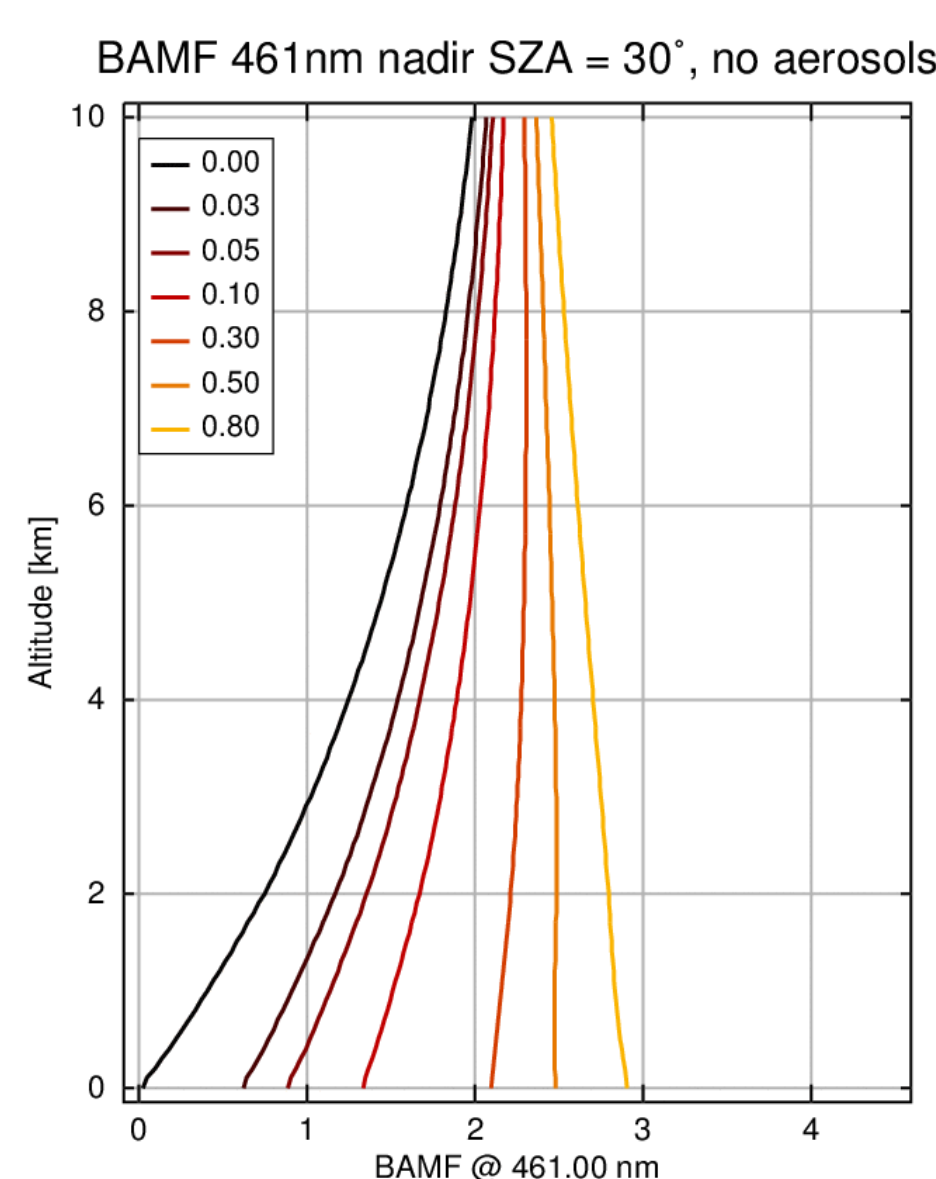
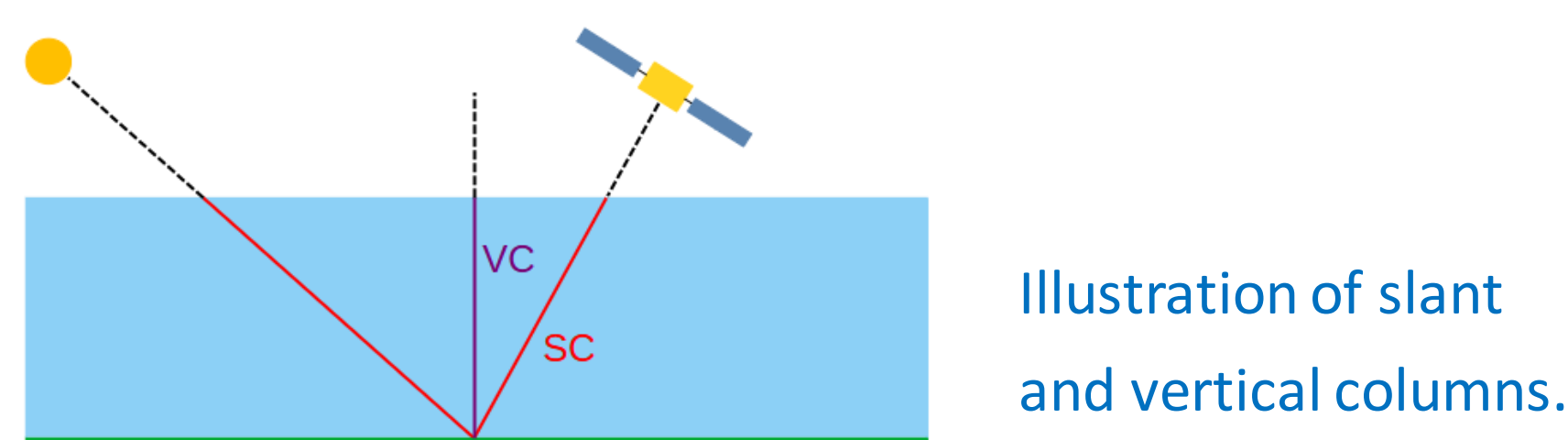
1. Motivation

CO₂ is the most important anthropogenic greenhouse gas. Detecting and quantifying CO₂ emission sources is an important task, but difficult to achieve, using satellite data. NO₂ is co-emitted with CO₂ during combustion processes and has short lifetime. Therefore, NO₂ plumes have a stronger gradient to surroundings, making NO₂ a useful proxy for CO₂ emissions. The objective of this study is to establish a high resolution TROPOMI NO₂ product for Europe, optimised for small scale processes.



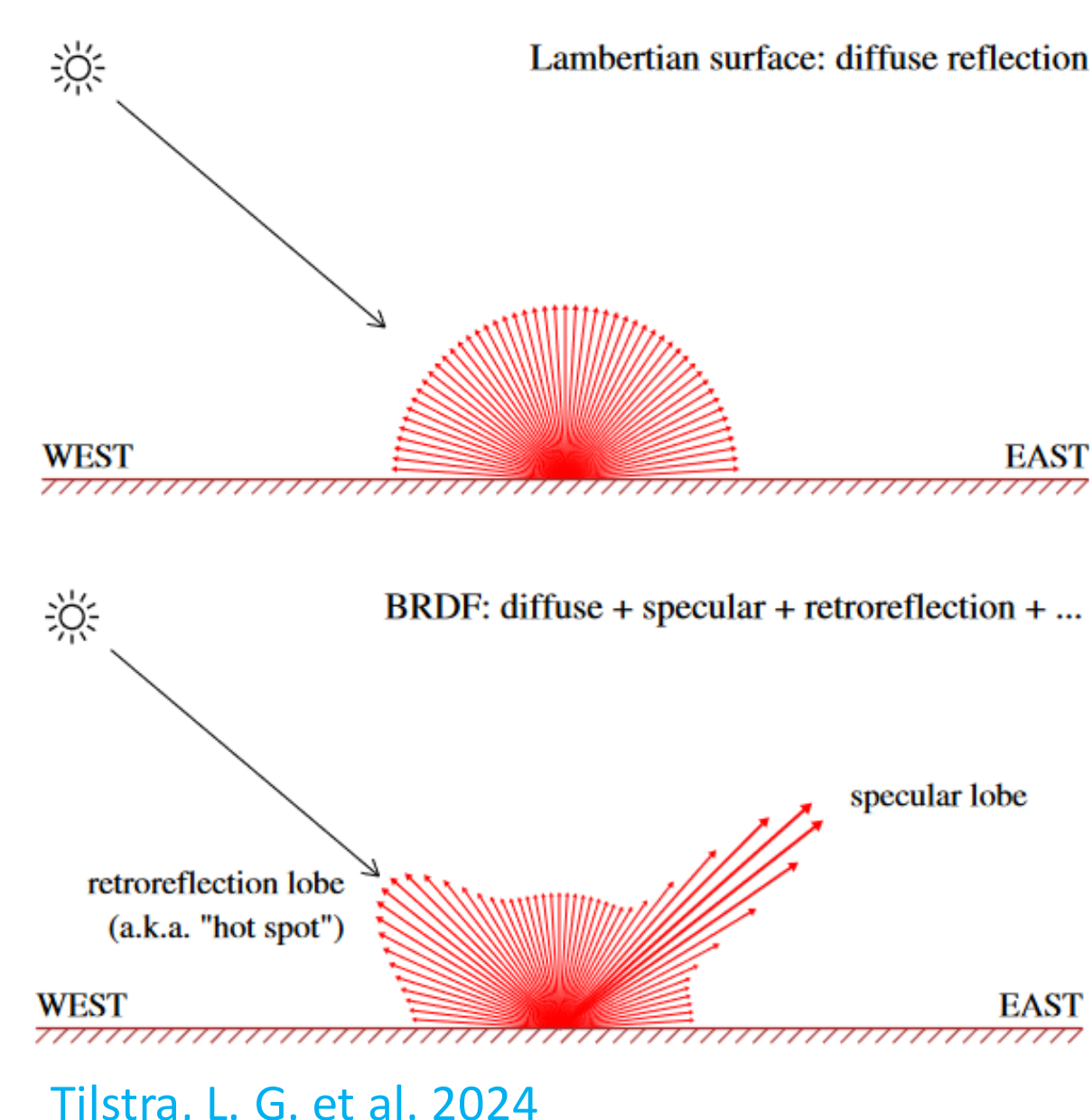
2. From spectrum to tropospheric vertical column

- Measure spectra using TROPOMI
- Fit spectra using DOAS method → total slant column (SC)
- Use STREAM to estimate strat. SC, subtract from total SC → tropospheric SC
- Transform into tropospheric VC using **Air Mass Factor (AMF)**
- AMF $\stackrel{\text{def}}{=} \frac{SC}{VC}$ depends on a priori and surface reflectivity
- Use MODIS Bi-directional Reflectance Distribution Function (BRDF) data for reflectivity (e.g. Zhou et al. 2010) to determine box-AMF (BAMF) using LUT
- Use CAMS-REG/CAMS global as NO₂ a priori to calculate AMF



Tilstra, L. G. et al. 2024
Surface reflectivity of DLER product shows low reflectivities in target area (red)

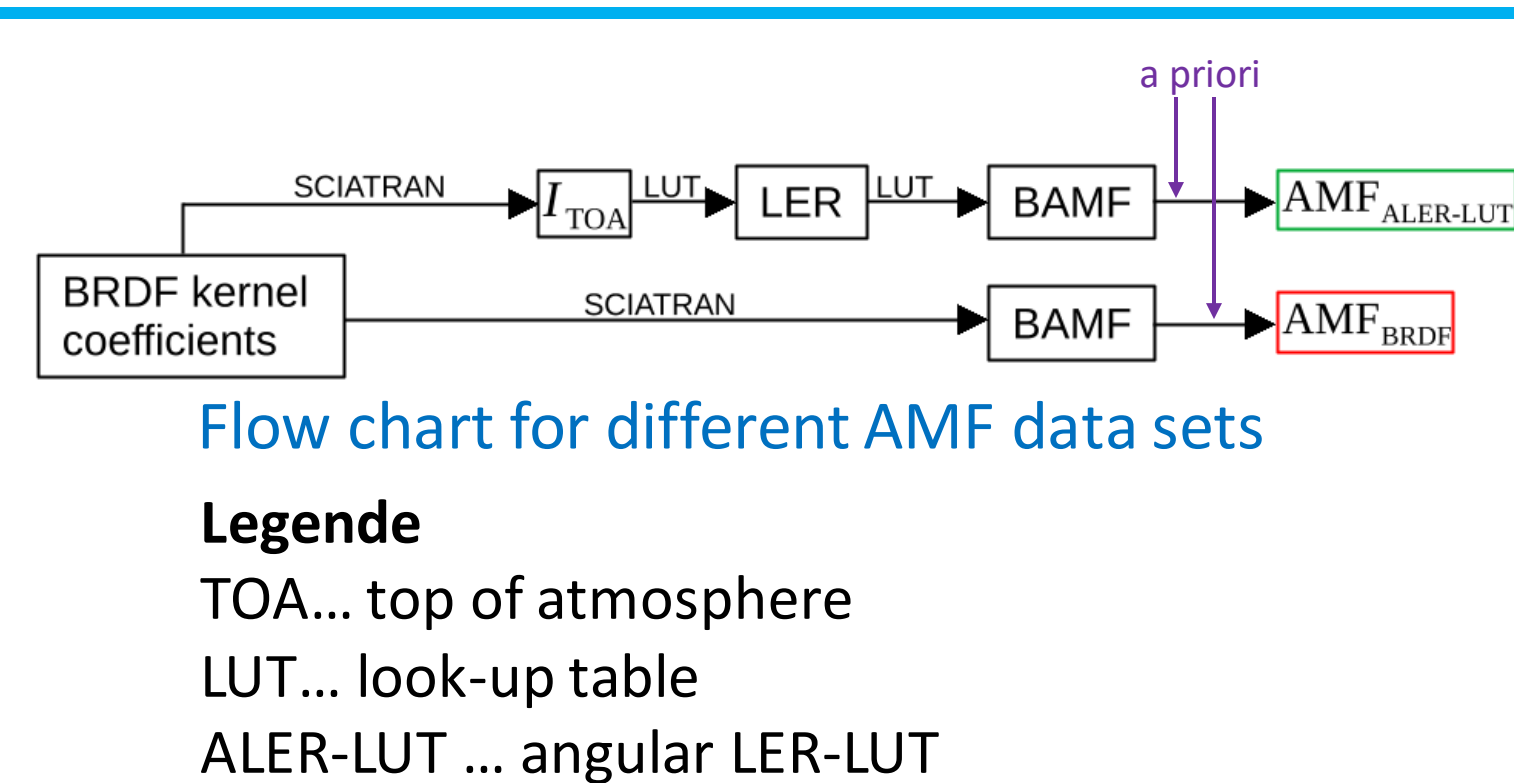
$$AMF = \frac{\sum_i BAMF_i \cdot m_i \cdot c_i}{\sum_i m_i}$$



- Lambertian equivalent reflectivity (LER) reflected radiance in all directions the same
- Bi-directional reflectance distribution function (BRDF), describes angle-dependent reflectivity of surfaces

3. AMF calculation

- AMF calculation using SCIATRAN and the BRDF for all TROPOMI pixels and orbits takes too long
- Shortcut: determine LER corresponding to TOA radiance of the BRDF case using LUT, as described by Vasilkov et al. (2017)

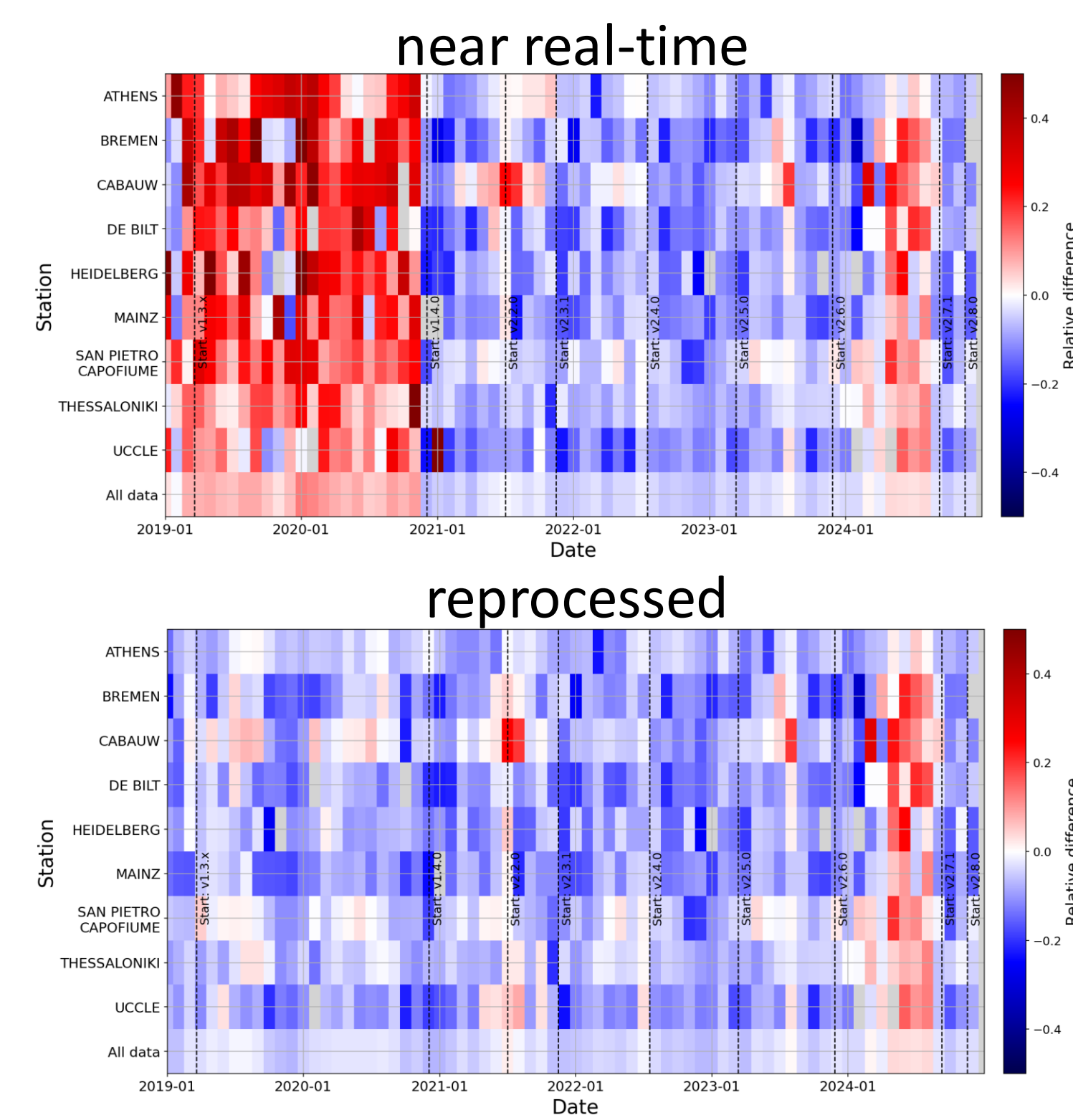


Legende
TOA... top of atmosphere
LUT... look-up table
ALER-LUT ... angular LER-LUT

4. Results

4.1 TROPOMI cloud data NRT vs RPRO

- Operational cloud data used in retrieval
 - Near real-time (NRT) cloud data reprocessed (RPRO) until 25.7.2022 → influence on retrieval?
 - Calculate relative difference in VC
- $$\Delta VC = \frac{VC^{ncc} - VC^{cc}}{VC^{cc}}$$
- ncc – no cloud correction used
cc – cloud correction used
- Large differences for NRT → gone for RPRO
 - v2.6.0 difference due to change in FRESKO leading to unintended consequences → reverted
 - Use RPRO cloud data in addition to NRT



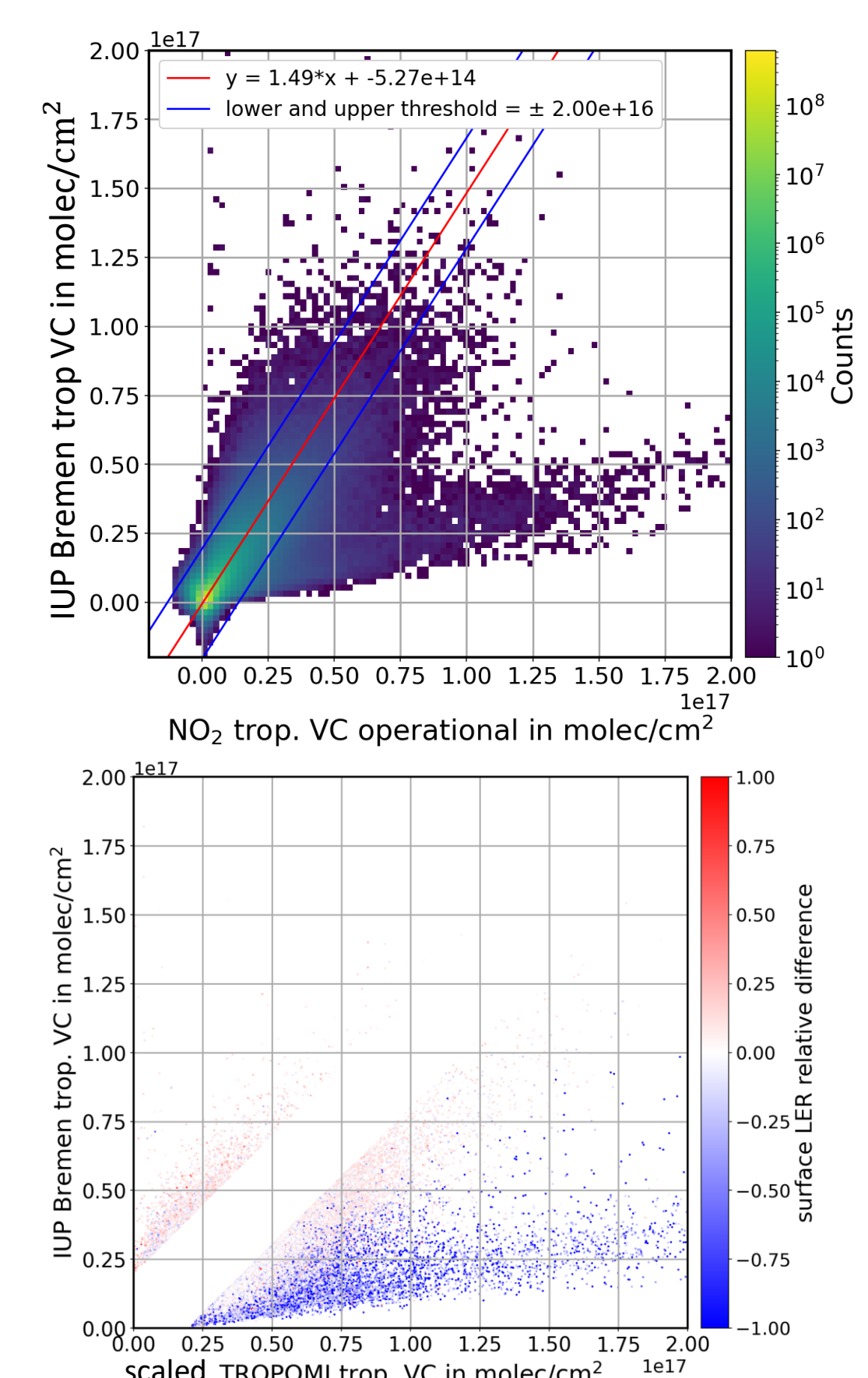
4.2 Operational vs IUP Bremen trop. VC

- Correlation plot using 6 years of data
- IUP Bremen product ~50% larger trop. VC
- Large additional branch → Why?
- Filter data using threshold of 2×10^{16} molec/cm²
- Calculate relative difference in scene LER used in retrieval

$$\Delta LER = \frac{LER_{oper} - LER_{UB}}{LER_{oper}}$$

LER_{oper} – scene LER operational
LER_{UB} – scene LER IUP Bremen

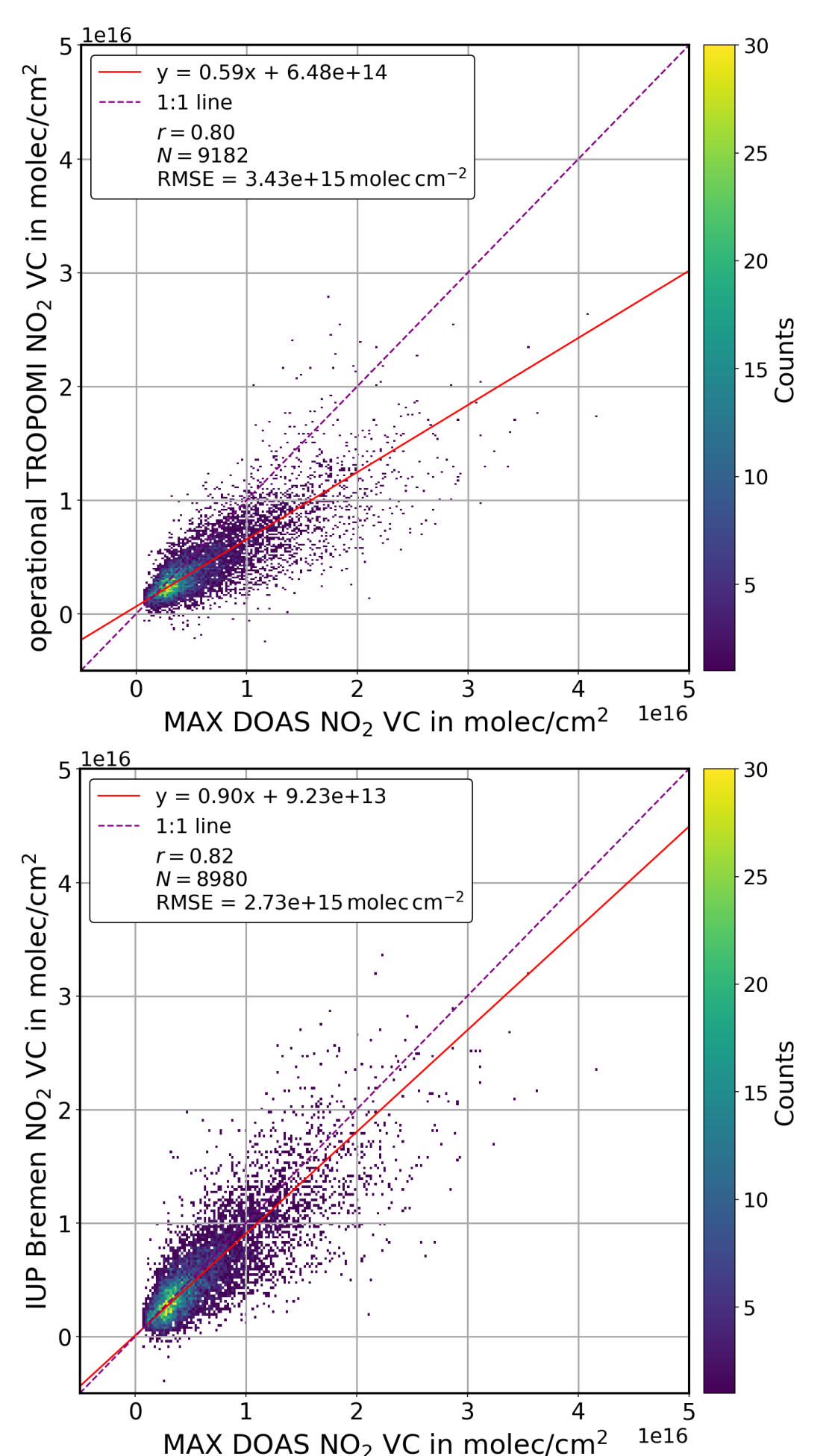
- Large differences correlate with additional branch
- Operational TROPOMI modifies scene LER, if cloud fraction would lie outside [0, 1]



4.3 Validation using MAX-DOAS data

- MAX-DOAS data (VIS) of FRM4DOAS project used for validation
- Average MAX-DOAS data for ±30 minutes around TROPOMI overpass
- For IUP Bremen and operational product, pixel above MAX-DOAS station used only
- IUP Bremen product shows clear improvement over operational product

Product	IUP Bremen			operational		
	year	points	slope	corr.	points	slope
2019	1237	0.86	0.81	1283	0.57	0.81
2020	1579	0.86	0.81	1639	0.55	0.79
2021	1532	0.92	0.81	1538	0.63	0.79
2022	1697	0.94	0.82	1775	0.62	0.80
2023	1549	0.96	0.82	1583	0.59	0.78
2024	1386	0.83	0.80	1364	0.56	0.76
all	8980	0.90	0.82	9182	0.59	0.80



5. Conclusions and outlook

- IUP Bremen TROPOMI NO₂ VC product created for 2019 – 2025, (2025 not shown)
- Importance of using RPRO cloud data confirmed
- Additional branch found → operational product adjusts scene LER
- Validation with MAX-DOAS clear improvements over operational product
- $s = 0.59 \rightarrow s = 0.90$
- $r = 0.80 \rightarrow r = 0.82$

References

Fuentes Andrade et al. (2024), A method for estimating localized CO₂ emissions from co-located satellite XCO₂ and NO₂ images, <https://doi.org/10.5194/amt-17-1145-2024>
Tilstra, L.G. et al. (2024), TROPOMI ATBD of the directionally dependent surface Lambertian-equivalent reflectivity
Vasilkov et al. (2017), Accounting for the effects of surface BRDF on satellite cloud and trace-gas retrievals: a new approach based on geometry-dependent Lambertian equivalent reflectivity applied to OMI algorithms, <https://doi.org/10.5194/amt-10-333-2017>
Zhou et al. (2010), Accounting for surface reflectance anisotropy in satellite retrievals of tropospheric NO₂, <https://doi.org/10.5194/amt-3-1185-2010>
Beirle et al. (2016), The STRatospheric Estimation Algorithm from Mainz (STREAM): estimating stratospheric NO₂ from nadir-viewing satellites by weighting convolution

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