



Copernicus
Atmosphere
Monitoring
Service
(CAMS)

Estimating lockdown-induced NO₂ changes in Europe

Jérôme Barré¹, Hervé Petetin², Augustin Colette³, Marc Guevara², **Vincent-Henri Peuch**¹, Laurence Rouil³, Richard Engelen¹, Antje Inness¹, Johannes Flemming¹, Carlos Pérez García-Pando^{2,4}, Dene Bowdalo², Frederik Meleux³, Camilla Geels⁵, Jesper H. Christensen⁵, Michael Gauss⁶, Anna Benedictow⁶, Svetlana Tsyro⁶, Elmar Friese⁷, Joanna Struzewska⁸, Jacek W. Kaminski^{8,9}, John Douros¹⁰, Renske Timmermans¹¹, Lennart Robertson¹², Mario Adani¹³, Oriol Jorba², Mathieu Joly¹⁴, and Rostislav Kouznetsov¹⁵

¹European Centre for Medium-range Weather Forecast (ECMWF), Sinfield Park, Reading, UK

²Barcelona Supercomputer Centre (BSC), Barcelona, Spain

³National Institute for Industrial Environment and Risks (INERIS), Verneuil-en-Halatte, France

⁴ICREA, Catalan Institution for Research and Advanced Studies, Barcelona, Spain

⁵Department of Environmental Science, Aarhus University, Roskilde, Denmark

⁶Norwegian Meteorological Institute, Oslo, Norway

⁷Rhenish Institute for Environmental Research at the University of Cologne, Cologne, Germany

⁸Institute of Environmental Protection – National Research Institute, Warsaw, Poland

⁹Faculty of Environmental Engineering, Warsaw University of Technology, Warsaw, Poland

¹⁰Royal Netherlands Meteorological Institute (KNMI), De Bilt, the Netherlands

¹¹Netherlands Organisation for Applied Scientific Research (TNO), Climate Air and Sustainability Unit, Utrecht, the Netherlands

¹²Swedish Meteorological and Hydrological Institute (SMHI), Norrköping, Sweden

¹³Italian National Agency for New Technologies, Energy and Sustainable Economic Development (ENEA), Bologna, Italy

¹⁴Météo-France, Toulouse, France

¹⁵Finnish Meteorological Institute (FMI), Helsinki, Finland



S5P NO2 EUROPEAN LOCKDOWN OVERVIEW

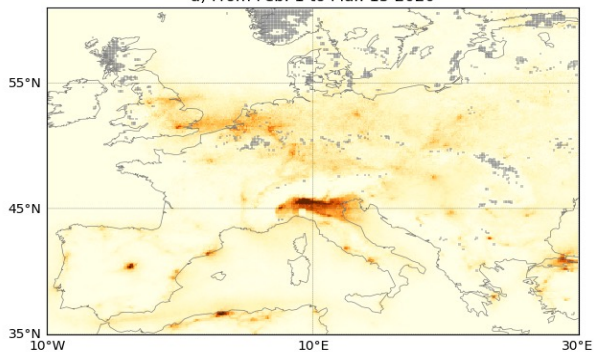
2020

Before Lockdown

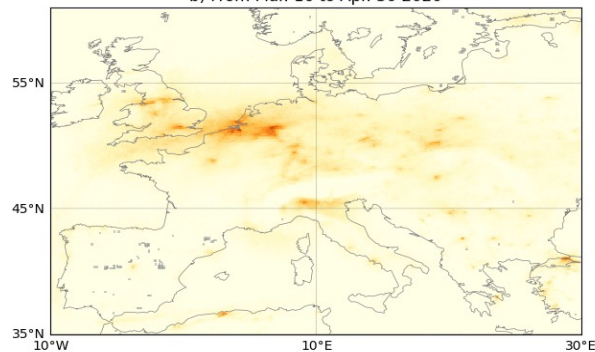
TROPOMI NO2 time averages

During Lockdown

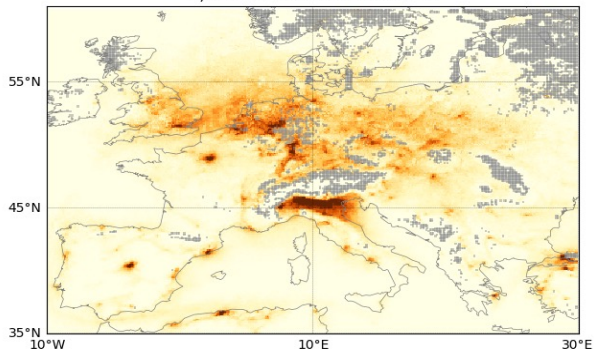
a) From Feb. 1 to Mar. 15 2020



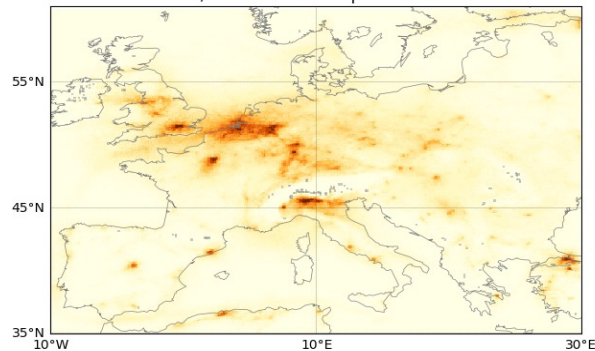
b) From Mar. 16 to Apr. 30 2020



c) From Feb. 1 to Mar. 15 2019



d) From Mar. 16 to Apr. 30 2019



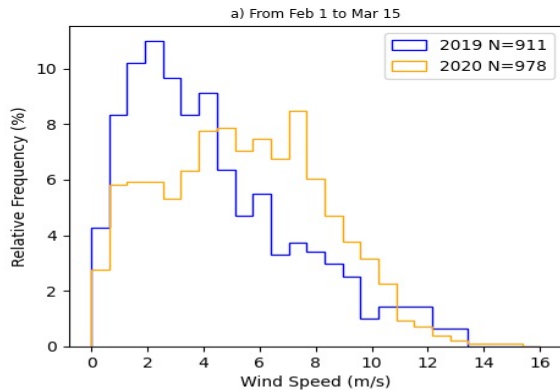
2019



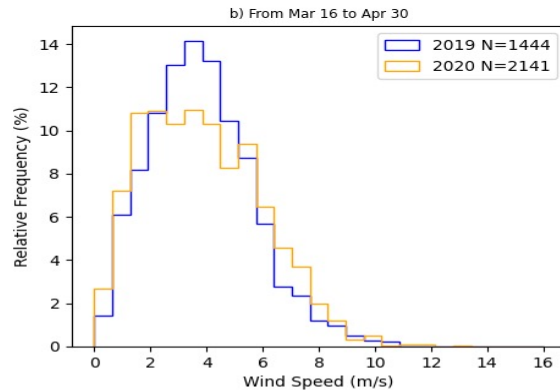
2019-2020 DISTRIBUTION OF WEATHER PARAMETERS

Windspeed

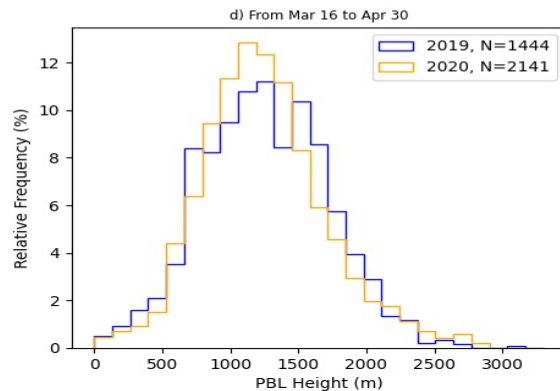
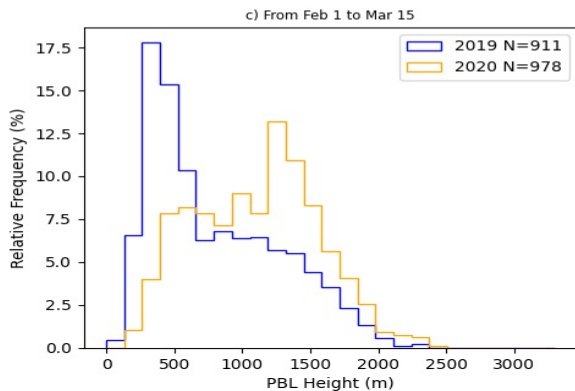
Before lockdown



During lockdown



PBLH

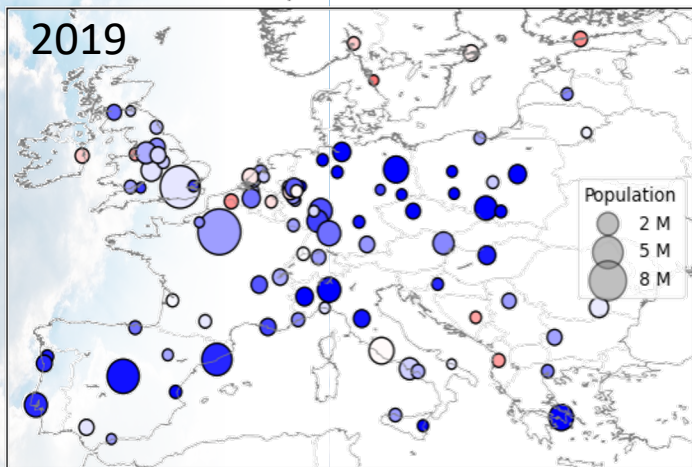




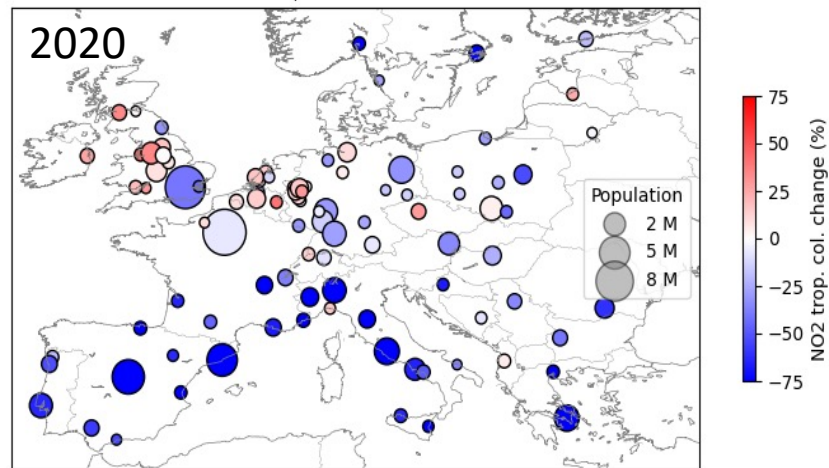
NON-WEATHER NORMALIZED ESTIMATES

- Using “before-during” estimate i.e. subtract [1Feb:15Mar] to [16Mar:30Apr]
- Select urban areas >0.5M : 100 across Europe
- **Significant reductions in 2019! Stronger than in 2020 in many areas.**

a) Satellite “Before-During” Estimates:
OBS[16Mar2019:30Apr2019] - OBS[1Feb2019:15Mar2019]



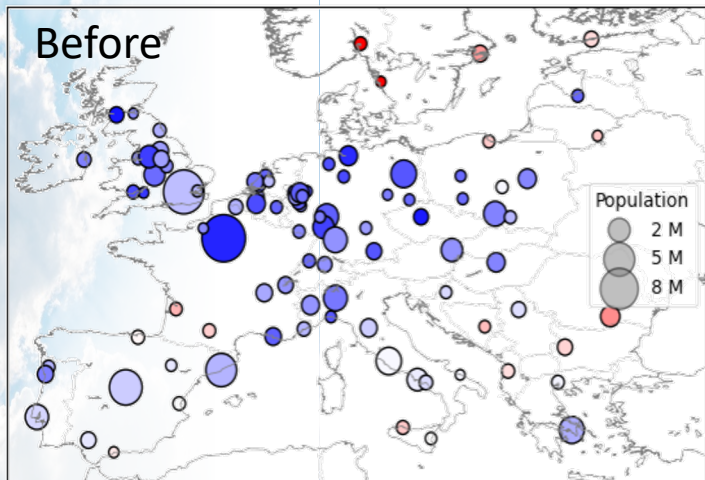
b) Satellite “Before-During” Estimates:
OBS[16Mar2020:30Apr2020] - OBS[1Feb2020:15Mar2020]



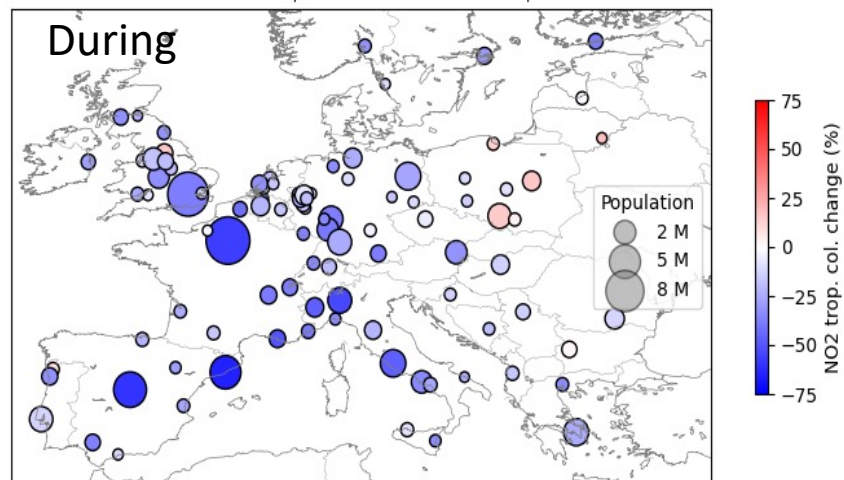


- Using “year1-vs-year2” estimate for the same period of year: i.e. subtract [2019] to [2020]
- **Significant reductions before lockdown! Stronger in certain areas as well.**

a) Satellite “Year to Year” Estimates:
OBS[1Feb2020:15Mar2020] - OBS[1Feb2019:15Mar2019]



b) Satellite “Year to Year” Estimates:
OBS[16Mar2020:30Apr2020] - OBS[16Mar2019:30Apr2019]





Machine Learning: Weather Normalization Methodology

- **Training Set**
- ECMWF NWP operational forecasts:
 - 10m wind speed and direction,
 - planetary boundary layer height
 - 2m temperature
 - surface relative humidity
 - geopotential at 500hPa
 - Date information
- CAMS regional 3D NO₂ forecasts
- S5P NO₂ columns or surface NO₂ (*target*)



Gradient boosting Regressor
Hyperparameter tuning (grid search)
with cross-validation (5 k-fold)

Simulate “business as usual” (BAU) S5P NO₂
columns or surface NO₂

Evaluation with **Test Set**

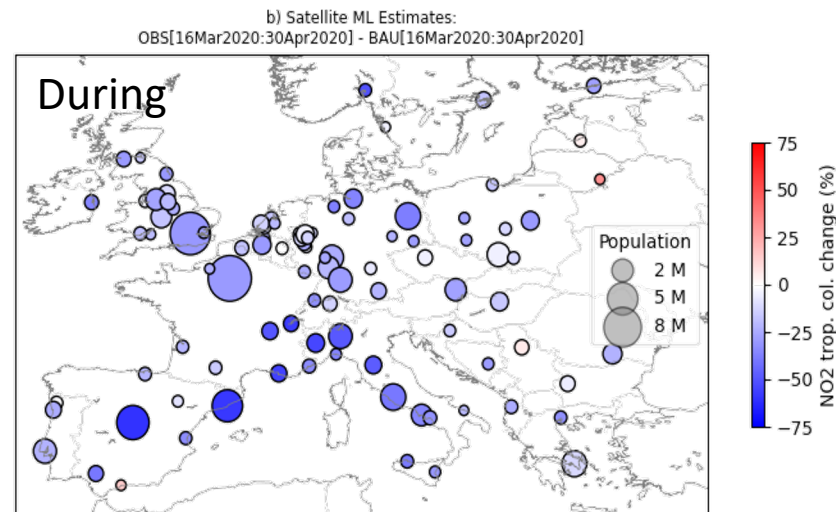
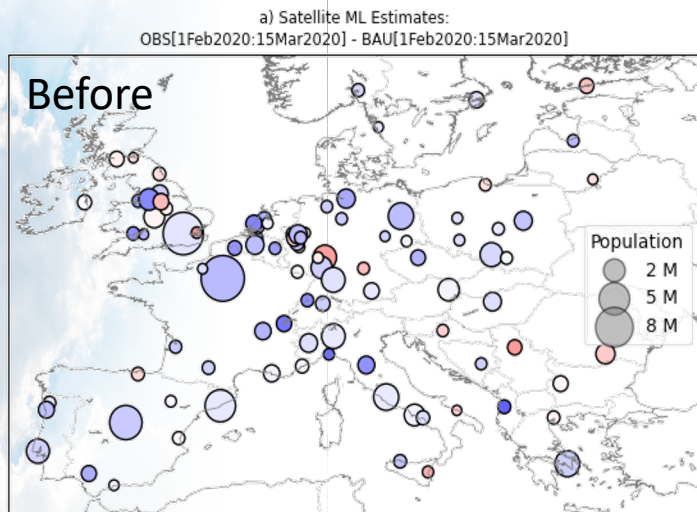


Calculate 2020 estimates:
Rea I– BAU



WEATHER NORMALIZED ESTIMATES (SATELLITE)

- Lower reductions in the period before lockdown
- Caveat: ML model is not perfect and cannot fully represent the anomaly

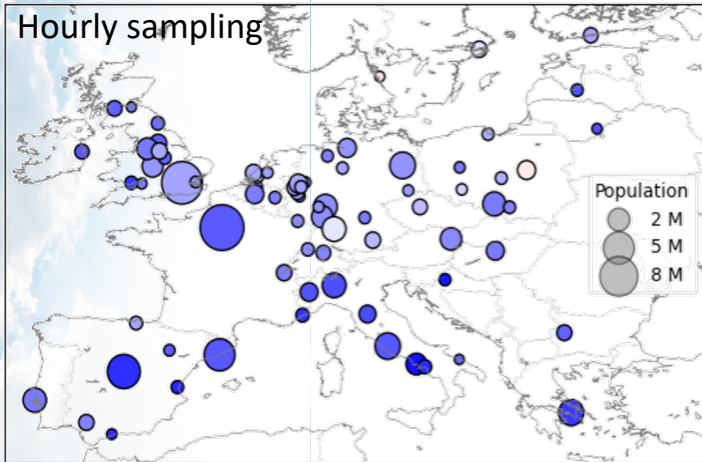




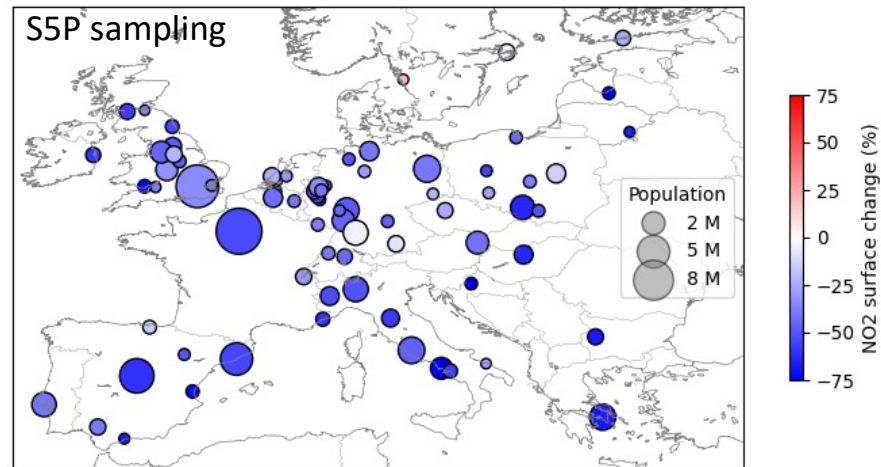
WEATHER NORMALIZED ESTIMATES (SURFACE STATIONS)

- Testing on the effect of S5P sampling versus hourly sampling
- Estimates tend to be stronger at satellite revisit times -> dependence on reduction factor and time of the day

a) Surface Obs. ML Estimates:
[OBS] - [BAU], Hourly sample (16Mar2020:30Apr2020)



b) Surface Obs. ML Estimates:
[OBS] - [BAU], S5P overpass sample (16Mar2020:30Apr2020)





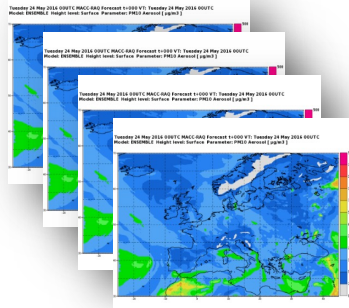
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CAMS: MULTI-MODEL LOCKDOWN HINDCAST

BAU Emissions



11 models hindcasts



BAU Ensemble median

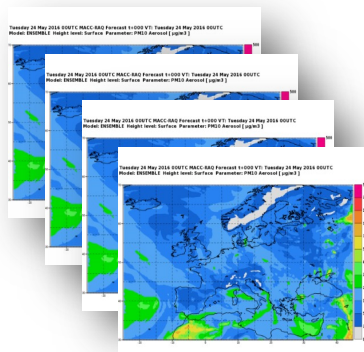
15 March – 30 April 2020

Surface pollution changes
estimates



Factual Emissions

Using recent activity
data (and ML, see
Guevara et al.,



Factual Ensemble median

11 models hindcasts

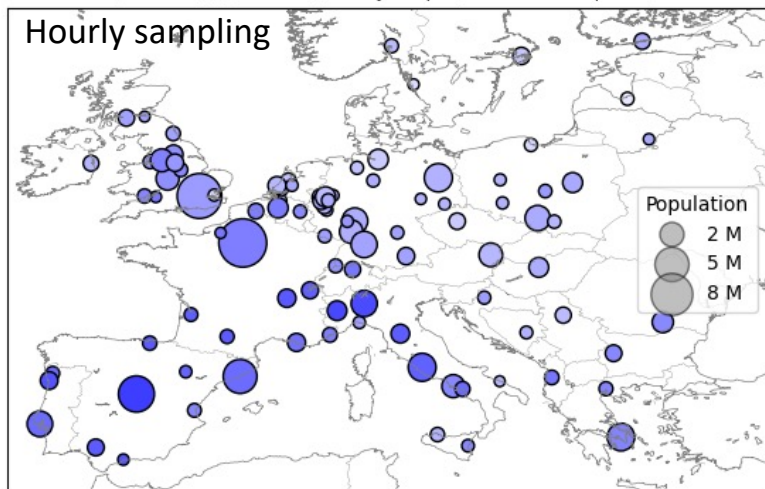


CAMS REGIONAL MODEL ENSEMBLE ESTIMATES

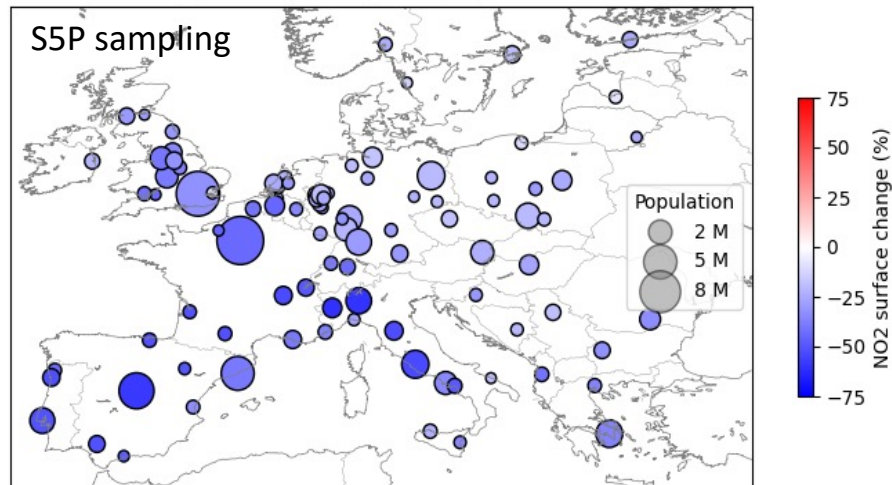
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- Early version of Guevara et al.: scaling factor per sector (road, industry and aviation) per country constant for the whole period.
- No significant dependence on time sampling as emission scaling is constant over time in these simulations.

a) Model Estimates:
[Hindcasts] - [Reference], Hourly sample (16Mar2020:30Apr2020)



b) Model Estimates:
[Hindcasts] - [Reference], S5P overpass sample (16Mar2020:30Apr2020)





SUMMARY OF RESULTS

	Average Changes	Inter city Standard Deviation
	(%)	(%)
Surface <u>stations</u> [hourly]	-36.74	15.09
Surface <u>stations</u> [S5P sampling 1pm]	-43.06	18.82
CAMS <u>model</u> ensemble [hourly]	-30.35	10.79
CAMS <u>model</u> ensemble [S5P sampling 1pm]	-31.82	11.97
<u>Satellite</u> S5P	-22.72	15.51

- Horizontal **representativeness** issues between pixel based and stations
- **Satellite overpass sampling** tests suggest a time of the day dependence on emissions reductions (model vs surface sites)
- **Vertical sampling**: column vs surface, weaker reduction seen columns (satellite)



CONCLUDING REMARKS

- Effects of lockdown on surface air pollution cannot be assessed correctly if weather variability and BAU emissions variability are not considered.
- When normalized with actual weather conditions, satellite-based, surface stations -based and model-based estimates provide similar country dependent variations linked with stringency of measures.
- In average the NO₂ reductions are around -30% over Europe during the most stringent phase of the 1st lockdown. Lower surface reduction are found on other pollutants (e.g. around -10% for PM), that do not have as large a contribution from transport and industrial emissions.
- Differences still exist between the type of estimates.



REFERENCES

Barré, J., Petetin, H., Colette, A., Guevara, M., Peuch, V.-H., Rouil, L., Engelen, R., Inness, A., Flemming, J., Pérez García-Pando, C., Bowdalo, D., Meleux, F., Geels, C., Christensen, J. H., Gauss, M., Benedictow, A., Tsyro, S., Friese, E., Struzewska, J., Kaminski, J. W., Douros, J., Timmermans, R., Robertson, L., Adani, M., Jorba, O., Joly, M., and Kouznetsov, R.: Estimating lockdown induced European NO₂ changes, *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2020-995>, in review, 2020.

Guevara, M., Jorba, O., Soret, A., Petetin, H., Bowdalo, D., Serradell, K., Tena, C., Denier van der Gon, H., Kuenen, J., Peuch, V.-H., and Pérez García-Pando, C.: Time-resolved emission reductions for atmospheric chemistry modelling in Europe during the COVID-19 lockdowns, *Atmos. Chem. Phys. Discuss.*, <https://doi.org/10.5194/acp-2020-686>, in review, 2020.

Petetin, H., Bowdalo, D., Soret, A., Guevara, M., Jorba, O., Serradell, K., and Pérez García-Pando, C.: Meteorology-normalized impact of the COVID-19 lockdown upon NO₂ pollution in Spain, *Atmos. Chem. Phys.*, 20, 11119–11141, <https://doi.org/10.5194/acp-20-11119-2020>, 2020.