



Carbon monoxide during pollution events in Asia: Evolution and trends from 17 years of IASI data

Selviga Sinnathamby¹, S. Safieddine¹, C. Viatte C.¹, A. Boynard^{1,2}, J. Hadji-Lazaro¹, M. George¹, D. Hurtmans³, P. Coheur³, M. Doutriaux-Boucher⁴, J. Onderwaater⁴ and C. Clerbaux^{1,3}

¹LATMOS/IPSL, Sorbonne Université, UVSQ, CNRS, Paris, France (<u>selviga.sinnathamby@latmos.ipsl.fr</u>) ²SPASCIA, Ramonville-Saint-Agne, 31520, France ³Spectroscopy, Quantum Chemistry and Atmospheric Remote Sensing (SQUARES), Université libre de Bruxelles (ULB), Brussels, Belgium ⁴EUMETSAT, Darmstadt, Germany







Introduction – Pollution events in Asia

- Asia has undergone rapid industrialization and demographic growth, leading to significant emissions of pollutants in recent years. ٠
- Some regions in Asia frequently experience severe air pollution events (Northern India and Pakistan, November 2024). ٠



The New Hork Times

https://www.nytimes.com/2024/11/19/world/asia/india-delhi-airpollution.html

Delhi Trudges Through Another Air Pollution Nightmare With No Answers

The government seems powerless to protect its citizens from this annual crisis, let alone prevent it.



Yibin, China [Dheera Venkatraman & Wang Xi]



India Gate in New Delhi was enveloped with smog on Tuesday. Anushree Fadnavis/Reuters



Introduction – IASI CO Climate Data Record

- Since 2007, IASI has been monitoring global and regional carbon monoxide (CO) concentrations, twice a day [Clerbaux et al., 2009].
- CO is mainly emitted during incomplete combustion processes (anthropogenic activities and biomass burning) and is eliminated by OH in late spring-summer.
- It is a good marker of pollution, with its lifetime of a few days to 2 months [Holloway et al., 2000].





 For the first time, EUMETSAT reprocessed the IASI CO dataset (Climate Data Record, doi:10.15770/EUM_SAF_AC_0047), providing a homogeneous record of CO which allows us to do trends studies.



Introduction – Areas of study

- We will focus on CO total column (COTC) evolution and trends as seen by IASI over 2007-2024 in 3 areas :
 - North China Plain (NCP)
 - Mainland South-East Asia (Mainland SEA)
 - Indo-Gangetic Plain (IGP)







Methods – IASI's sensitivity and importance of nighttime data

- Thermal infrared instruments are known to be sensitive in the mid-troposphere [George et al., 2015]. ٠
- However, they can have a high sensitivity close to the surface in cases of strong negative thermal contrast (temperature difference ٠ between the surface and the atmospheric layer just above the surface) [Bauduin et al., 2017].
- As high thermal contrast is associated with low COTC error, we apply a error-based filter on IASI daytime and nighttime data. ٠

Example case : IGP, 2023/11/06



 \rightarrow we need to include nighttime data.

Methods – Combining daytime and nighttime IASI data

- This filter consists of discarding all the IASI pixels for which the error is greater than a certain threshold. ٠
- We set the threshold value by comparing IASI with TROPOMI, as it is more sensitive to the surface <u>under clear-sky conditions</u> [Borsdorff et ٠ al. 2018].





IASI CO total column daily evolution from 2007 to 2024, IGP

- Highest CO concentrations are found during the pre-monsoon and the post-monsoon seasons, due to agricultural waste burning, as shown by the CO emissions from the Global Fire Emissions Database (GFED4.1s, [van der Werf et al. 2017]).
- Air pollution episodes during the post-monsoon season are highly dependent on **local meteorology**.







IASI CO total column daily evolution from 2007 to 2024, Mainland **SEA**

- CO concentrations are the highest during the dry season due to vegetation fires and agricultural waste burning.
- CO from Indonesian fires occurring in September are also detected in Mainland SEA.







Monthly average emissions from fires over 2008-2023 for different types of biomass



IASI CO total column daily evolution from 2007 to 2024, NCP

- CO concentrations are the highest during winter and spring, mainly due to high anthropogenic emissions, low OH • levels and local meteorology [Boynard et al., 2013].
- In spring, agricultural waste burning can contribute to the already elevated CO levels in the region. ۲







IASI CO total column global trends in Asia [2007/07 – 2024/06]

Trends are estimated using the Theil-Sen method (TS) [Theil, 1950; Sen, 1968] on annual averages of COTC over the period ٠ 2007/07-2024/06. Significant trends (p-value < 0.05) are stippled.







10

IASI CO total column trends during pollution seasons

IASI 17-year COTC trends for selected pollution seasons in each region are shown below. Significant trends are marked by an asterix (*). •













Explaining IASI trends with emission trends

CO IASI trends are compared to local CO emission trends from anthropogenic activities and biomass burning, estimated by • the Copernicus Atmosphere Monitoring Service (CAMS) global anthropogenic emission inventory (CAMS-GLOB-ANTv6.2, [Granier et al., 2019]) and Global Fire Assimilation System inventory (GFASv1.2, [Kaiser et al., 2012]), respectively.

During the **winter season in NCP**, the ٠ deployment of the China Air Quality Action Plan has led to a reduction of CO emissions and as such a coincident decrease in CO concentration.





-1.74 %/yr -179.5 Gg/yr²



12

Explaining IASI trends with emissions trends

CO IASI trends are compared to local CO emission trends from anthropogenic activities and biomass burning, estimated by • the Copernicus Atmosphere Monitoring Service (CAMS) global anthropogenic emission inventory (CAMS-GLOB-ANTv6.2, [Granier et al., 2019]) and Global Fire Assimilation System inventory (GFASv1.2, [Kaiser et al., 2012]), respectively.

For the **post-monsoon season in IGP**, ٠ IASI and CAMS show a positive trend, while GFAS shows a negative one, making it hard to explain concentrations trend with emission trends.





Conclusions and take home messages

- The new reprocessed IASI CO Climate Data Record allows us to conduct trend studies for the first time in heavily polluted regions in Asia.
- We show that nighttime data, usually discarded because of their supposed low sensitivity to the lower atmosphere, can be invaluable for • detecting near-surface CO.
- We developed a user-friendly method to only consider IASI pixels that are sensitive to the surface, and so improved the CO surface detection.
- In some regions, CO concentration trends from IASI can be explained by the trends as observed by local emissions inventories. •
- However, relating concentration trends with emission trends remains a challenging exercise, particularly when these ones are built on limited information. Moreover, CO concentrations also depend on transport phenomena, chemical processes and local meteorology, which need to be accounted for in the discussion of the results.





References

- Clerbaux, C., Boynard, A., Clarisse, L., George, M., Hadji-Lazaro, J., Herbin, H., Hurtmans, D., Pommier, M., Razavi, A., Turguety, S., Wespes, C., & Coheur, P. (2009). Monitoring of atmospheric composition using the thermal infrared IASI/MetOp sounder. Atmospheric Chemistry and Physics, 9(16), 6041–6054. https://doi.org/10.5194/acp-9-6041-2009
- Holloway, T., Levy, H., & Kasibhatla, P. (2000). Global distribution of carbon monoxide. Journal of Geophysical Research Atmospheres, 105(D10), 12123–12147. https://doi.org/10.1029/1999jd901173
- George, M., Clerbaux, C., Bouarar, I., Coheur, P., Deeter, M. N., Edwards, D. P., Francis, G., Gille, J. C., Hadji-Lazaro, J., Hurtmans, D., Inness, A., Mao, D., & Worden, H. M. (2015). An examination of the long-term CO records from MOPITT and IASI: comparison of retrieval methodology. Atmospheric Measurement Techniques, 8(10), 4313–4328. https://doi.org/10.5194/amt-8-4313-2015
- Bauduin, S., Clarisse, L., Theunissen, M., George, M., Hurtmans, D., Clerbaux, C., & Coheur, P. (2017). IASI's sensitivity to near-surface carbon monoxide (CO): Theoretical analyses and retrievals on test cases. Journal of Quantitative Spectroscopy and Radiative Transfer, 189, 428–440. https://doi.org/10.1016/j.jgsrt.2016.12.022
- Borsdorff, T., De Brugh, J. A., Hu, H., Hasekamp, O., Sussmann, R., Rettinger, M., Hase, F., Gross, J., Schneider, M., Garcia, O., Stremme, W., Grutter, M., Feist, D. G., Arnold, S. G., De Mazière, M., Sha, M. K., Pollard, D. F., Kiel, M., Roehl, C., . . . Landgraf, J. (2018). Mapping carbon monoxide pollution from space down to city scales with daily global coverage. Atmospheric Measurement Techniques, 11(10), 5507–5518. https://doi.org/10.5194/amt-11-5507-2018
- Boynard, A., Clerbaux, C., Clarisse, L., Safieddine, S., Pommier, M., Van Damme, M., Bauduin, S., Oudot, C., Hadji-Lazaro, J., Hurtmans, D., & Coheur, P. (2014). First simultaneous space measurements of atmospheric pollutants in the boundary layer from IASI: A case study in the North China Plain. Geophysical Research Letters, 41(2), 645–651. https://doi.org/10.1002/2013gl058333
- Van Der Werf, G. R., Randerson, J. T., Giglio, L., Van Leeuwen, T. T., Chen, Y., Rogers, B. M., Mu, M., Van Marle, M. J. E., Morton, D. C., Collatz, G. J., Yokelson, R. J., & Kasibhatla, P. S. (2017). Global fire emissions estimates during 1997–2016. Earth System Science Data, 9(2), 697–720. https://doi.org/10.5194/essd-9-697-2017
- Theil, H. (1950). A rank-invariant method of linear and polynomial regression analysis, 3; confidence regions for the parameters of polynomial regression equations. Indagationes Mathematicae, 1(2), 467–482. https://ir.cwi.nl/pub/18448/18448A.pdf
- Sen, P. K. (1968). Estimates of the regression coefficient based on Kendall's TAU. Journal of the American Statistical Association, 63(324), 1379–1389. https://doi.org/10.1080/01621459.1968.10480934
- Granier, C., Darras, S., Van Der Gon, H. D., Jana, D., Elguindi, N., Bo, G., Michael, G., Marc, G., Jalkanen, J., Kuenen, J., Liousse, C., Quack, B., Simpson, D., & Sindelarova, ٠ K. (2019). The Copernicus Atmosphere Monitoring Service global and regional emissions (April 2019 version). HAL (Le Centre Pour La Communication Scientifique Directe). https://doi.org/10.24380/d0bn-kx16
- Kaiser, J. W., Heil, A., Andreae, M. O., Benedetti, A., Chubarova, N., Jones, L., Morcrette, J., Razinger, M., Schultz, M. G., Suttie, M., & Van Der Werf, G. R. (2012). Biomass ٠ burning emissions estimated with a global fire assimilation system based on observed fire radiative power. Biogeosciences, 9(1), 527–554. https://doi.org/10.5194/bg-9-527-2012



Back-up - IASI COTC vs CO 0-2km







Back-up – GFAS vs GFED

Annual timeseries of CO fire emissions from GFASv1.2 and GFED4.1s and COTC concentrations from IASI. •







Back-up - IASI trends with WLS method

The Weighted Least Squares (WLS) method is a linear regression that is robust to outliers. Trend magnitudes are estimated from the monthly CO anomalies time series. We defined the monthly variance of CO concentrations as weights. Significant trends to 1 standarderror are stippled. The standard error for the WLS method is also given.







Back-up - Explaining IASI trends with emission in Mainland SEA (dry season)





+30.2 Gg/yr²

-361.2 Gg/yr² -0.20 %/yr



19

