

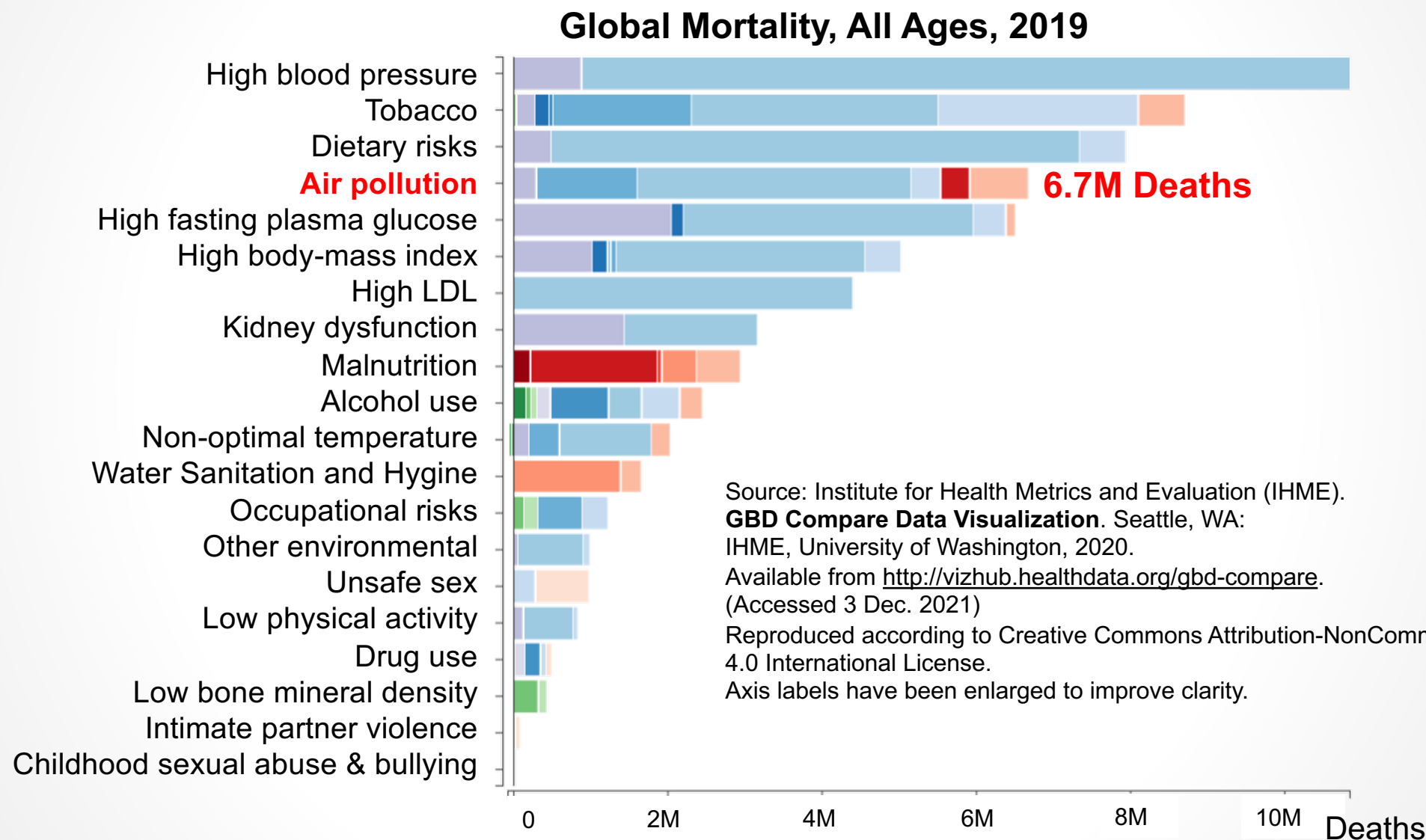


Creating & Using Open Global and Local Air Quality Data at NASA GMAO

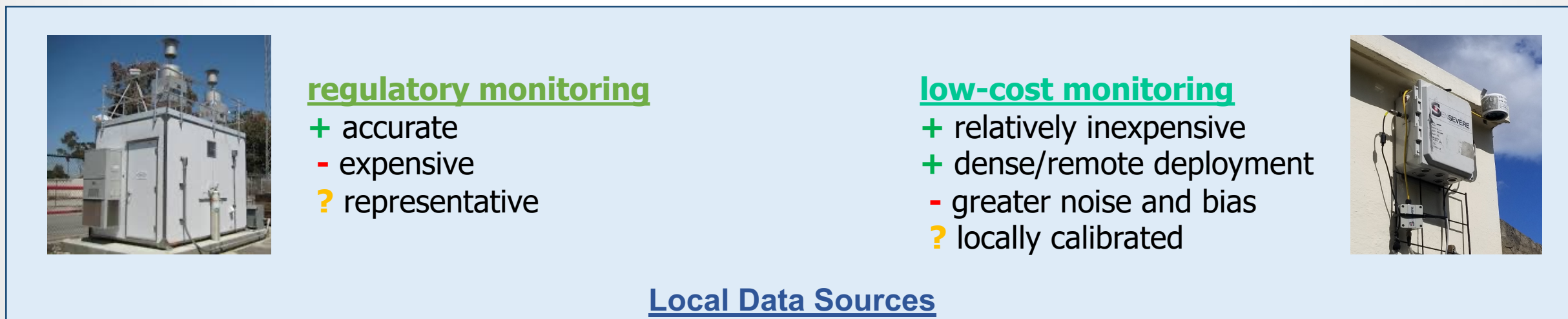
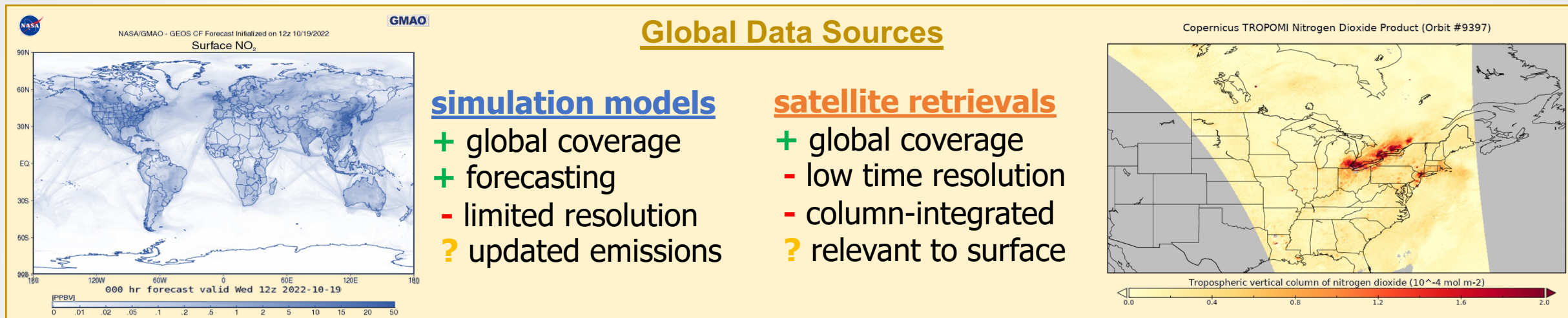
NASA GMAO: Carl Malings (Morgan State U.)
K. Emma Knowland (Morgan State U.)
Christoph Keller (Morgan State U.)
Viral Shah (SSAI)
Callum Wayman (SSAI)
Stephen Cohn
Bryan Duncan

Sonoma Tech: Nathan Pavlovic
Alan Chan
UNEP: Sean Khan
US EPA: John White
Columbia LDEO: Daniel Westervelt
Clarity: Sean Wihera
WUSTL: Randall Martin
NYU: Noussair Lazrak

Air quality is the leading environmental health concern

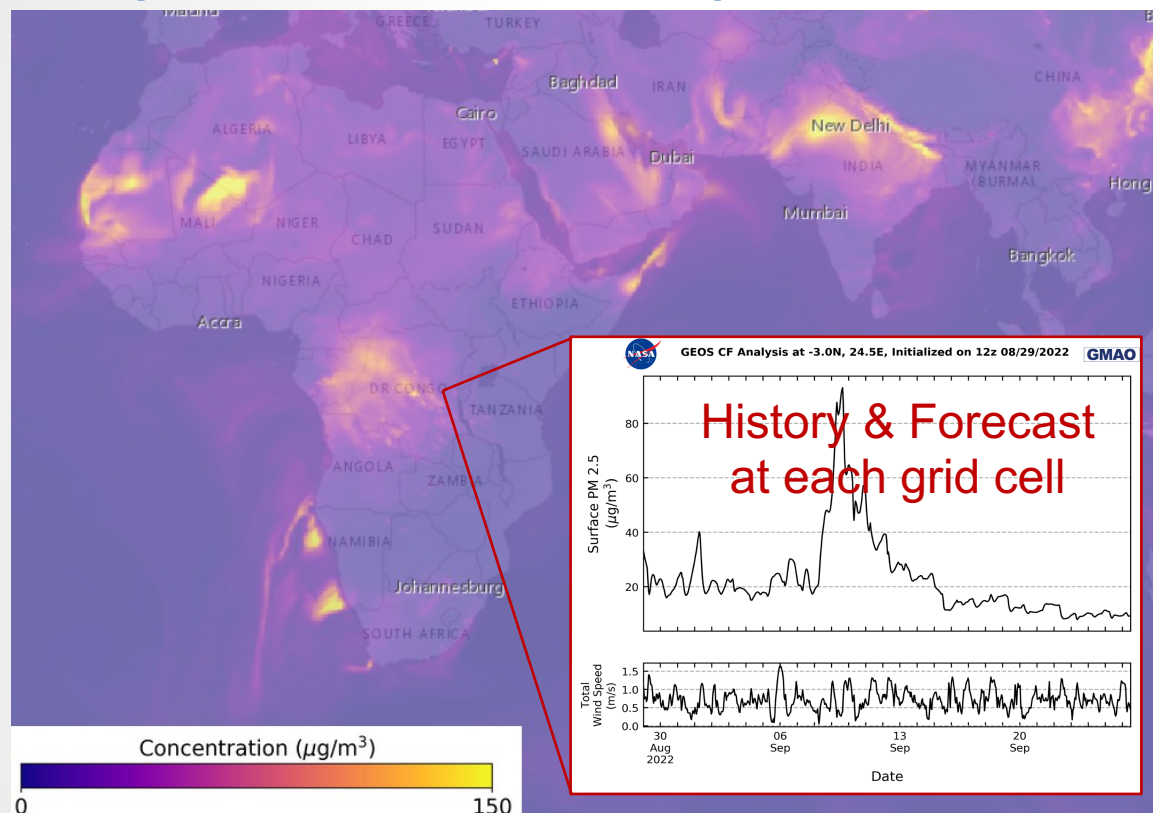


We understand air quality with both global and local data



The Big Picture: global air quality with NASA GEOS-CF

Example: interactive FLUID map for GEOS-CF PM_{2.5}



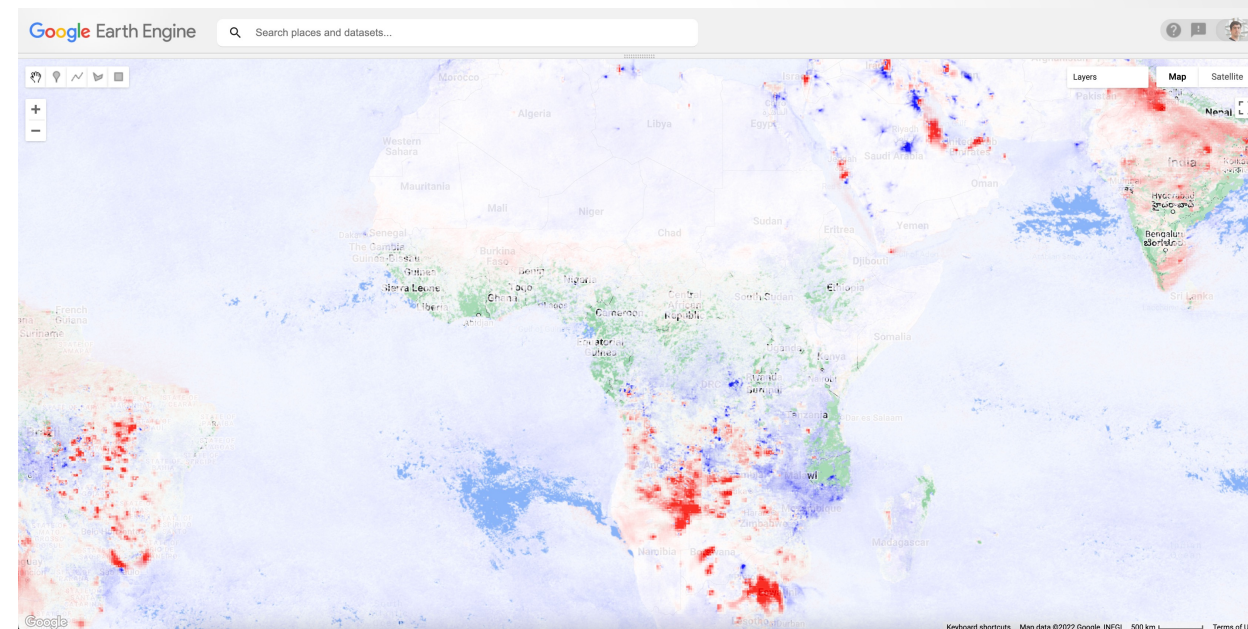
Sources: GEOS-CF FLUID interactive map

https://fluid.nccs.nasa.gov/cf_map/index

GEOS-CF on Google Earth Engine <https://code.earthengine.google.com/>

`ee.ImageCollection("NASA/GEOS-CF/v1/rpl/tavg1hr")`

Example: comparing GEOS-CF NO₂ with satellite data (ESA TROPOMI) in Google Earth Engine

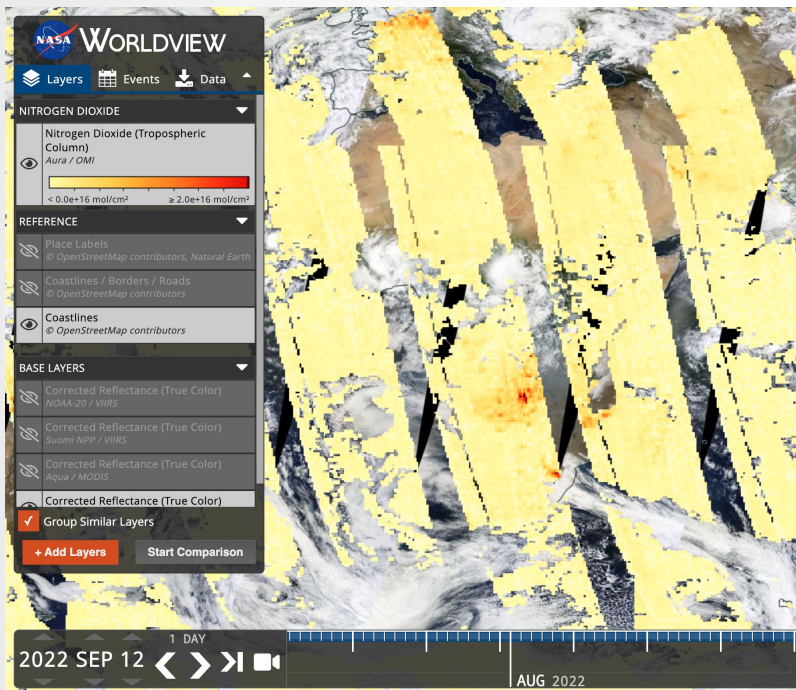


Agencies like NASA generate global air quality model analyses and forecasts & make them freely available

Top-Down View: satellite air quality data from NASA & others

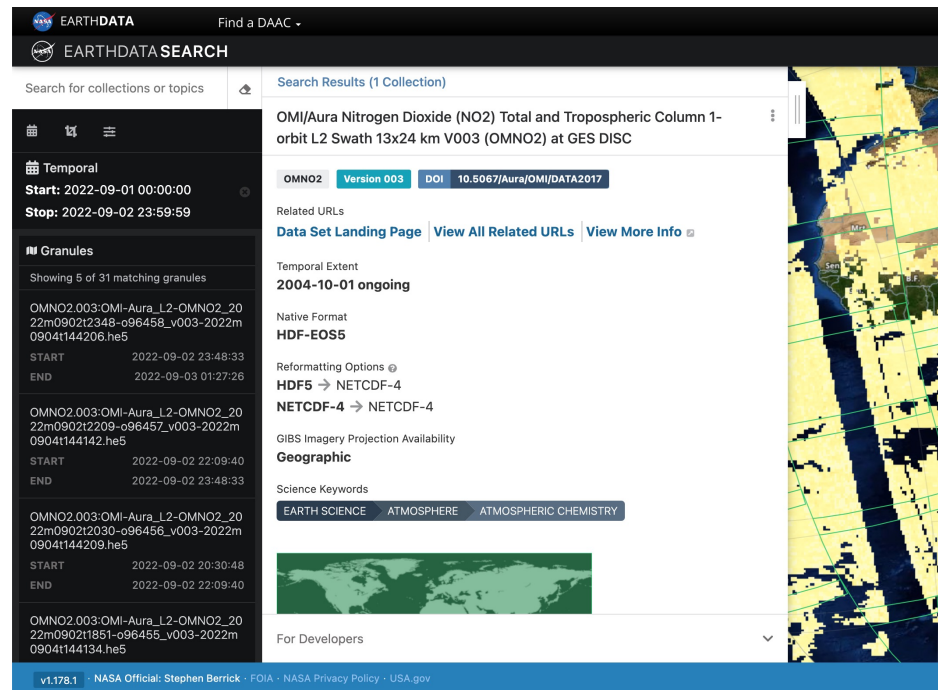
Worldview

simplified data visualization



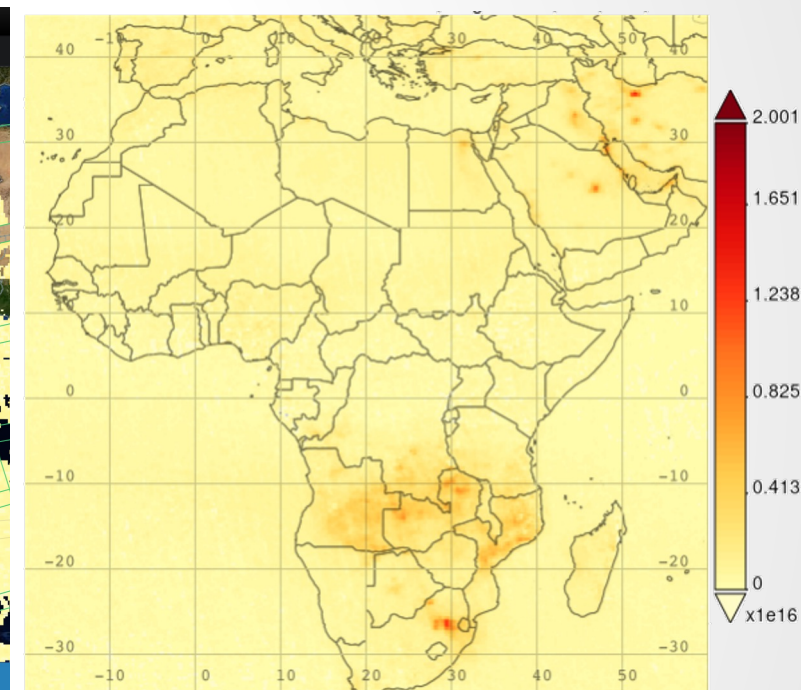
Earthdata

explore, subset, download data



Giovanni

simple online data analysis



Example: OMI Satellite Tropospheric Column NO₂

Sources: NASA Worldview. <https://worldview.earthdata.nasa.gov/>
NASA Earthdata. <https://www.earthdata.nasa.gov/>
NASA Giovanni. <https://giovanni.gsfc.nasa.gov/giovanni/>

Satellites are expensive, but provide valuable “top down” insight into air quality as they orbit above the Earth

The Ground Truth: regulatory air quality monitors and LCS

Regulatory Monitoring



These form the “backbone” of the monitoring system, but are relatively sparse (especially when taking a global view).

- + accurate
- expensive
- ? representativity

Low-Cost Sensors (LCS)

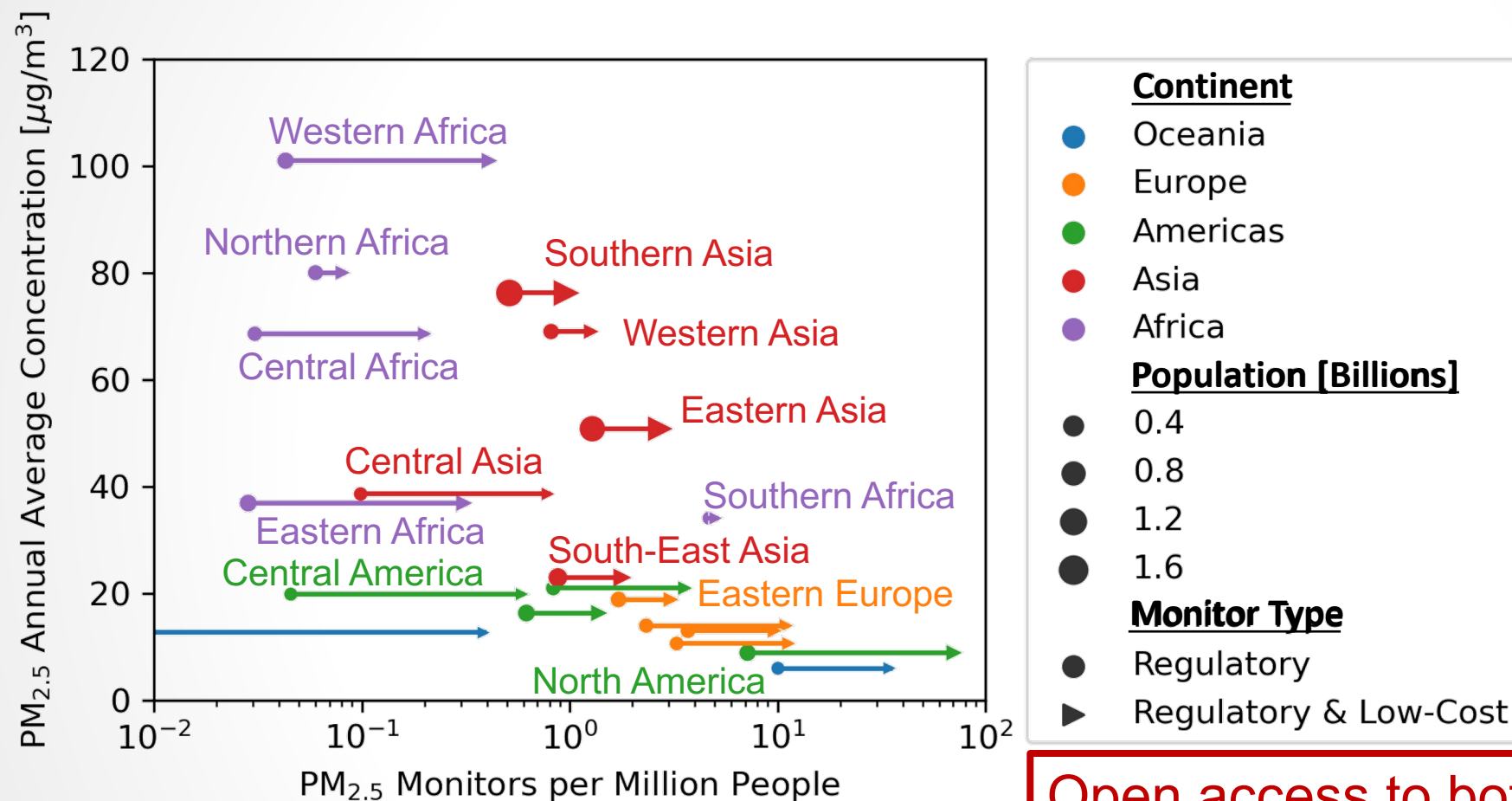


calibration is an open issue, but leveraging network density can offset some of these shortcomings, and allows greater access to air quality monitoring technologies

- + relatively inexpensive
- + dense and/or remote deployment
- greater noise and bias

Accurate but expensive regulatory monitors plus cheaper but more uncertain LCS give the “ground truth”

Global inequalities in air quality data: every datapoint counts



Many regions (especially Africa & Asia) feature high PM_{2.5} concentration but low per-capita PM_{2.5} monitor density, leading to poor AQ data coverage.

Including low-cost sensors increases per-capita AQ monitor density by up to an order of magnitude in some regions.

Source: Malings et al. (2020). "Application of low-cost fine particulate mass monitors to convert satellite AOD to surface concentrations in North America and Africa." *Atmospheric Measurement Techniques*. DOI: 10.5194/amt-13-3873-2020.
Updated analysis based on open air quality data available from openAQ.org

Open access to both regulatory and low-cost sensor datasets are key to filling in global air quality data gaps

Evaluating GEOS-CF model with Open Surface Data

NASA GMAO GEOS Composition Forecasts (GEOS-CF)

Evaluation of daily forecasts

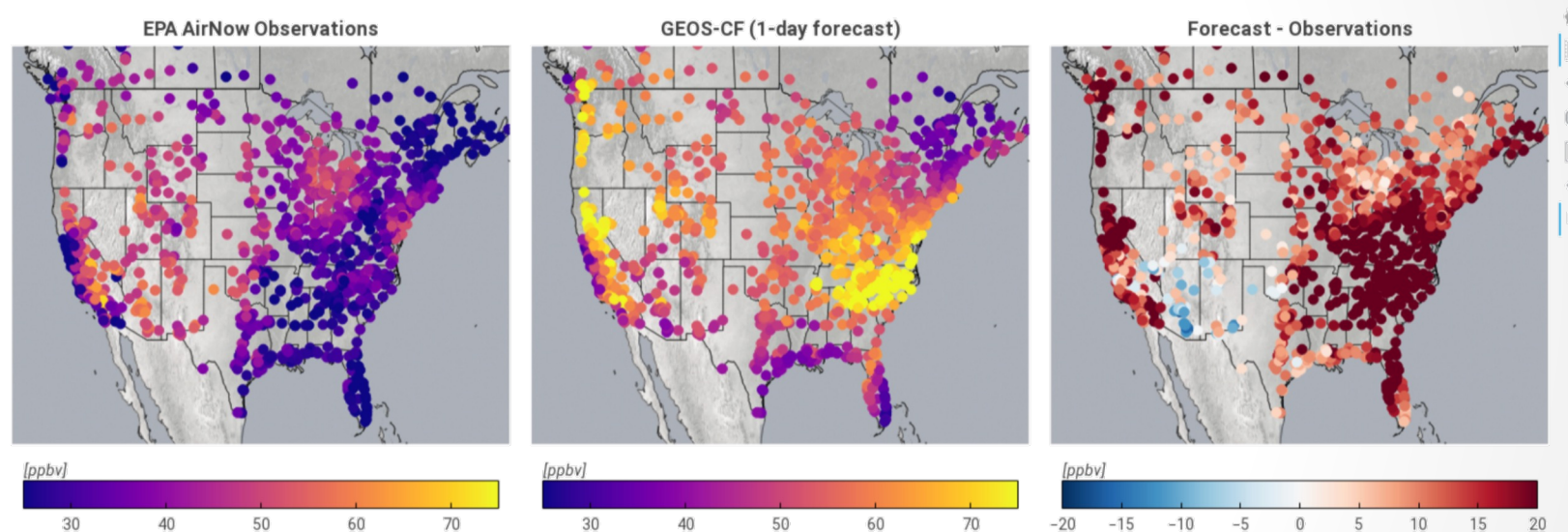
Date
31 July 2022

Region
U.S.

Parameter
MDA8 ozone

Submit

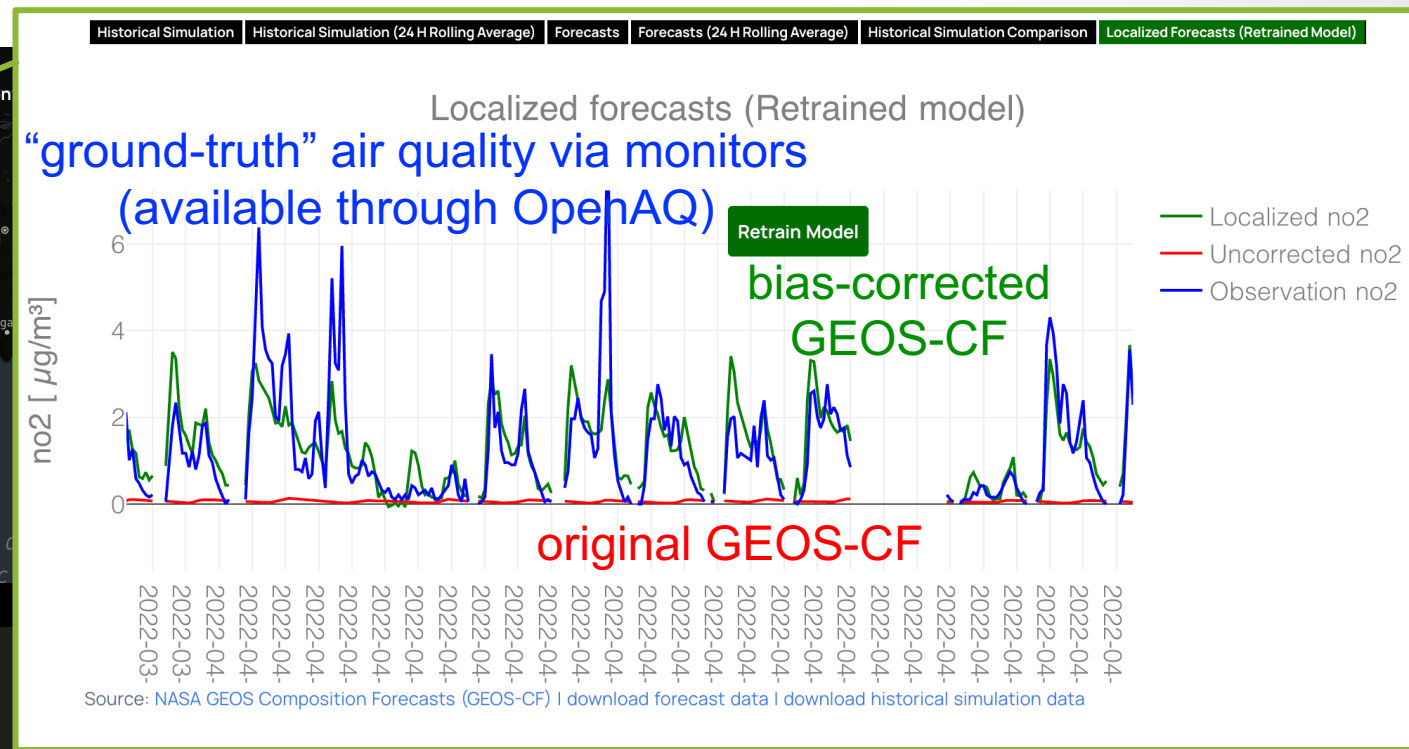
MDA8 ozone for 31 July 2022



Tool developed by Viral Shah, NASA GMAO (SSAI)

Notes and data sources: The observations are from the [EPA AirNow](#) program and were made by over 120 federal, state, local, and tribal air quality agencies. The AirNow observations are preliminary and subject to change. The [GEOS-CF](#) system is a product of the NASA Global Monitoring and Analysis Office (GMAO). It provides daily, global forecasts of atmospheric composition at a spatial resolution of 25 km, using meteorological analyses from other GEOS systems. These forecasts are intended as research datasets for use by NASA researchers and the broader community.

Localized bias-corrected GEOS-CF forecasts with CityAQ



Sources: **Project overview:** <https://www.wri.org/initiatives/cityaq>

Methodology description: Keller et al. (2021). “Global impact of COVID-19 restrictions on the surface concentrations of nitrogen dioxide and ozone.” *Atmospheric Chemistry and Physics*. DOI: 10.5194/acp-21-3555-2021.

Web-based tool: https://noussairlazrak.github.io/localized_forecasts/#

Visualizations provided by Noussair Lazrak, NYU

Bias-correcting global air quality forecasts with local data from OpenAQ monitor stations in the CityAQ project

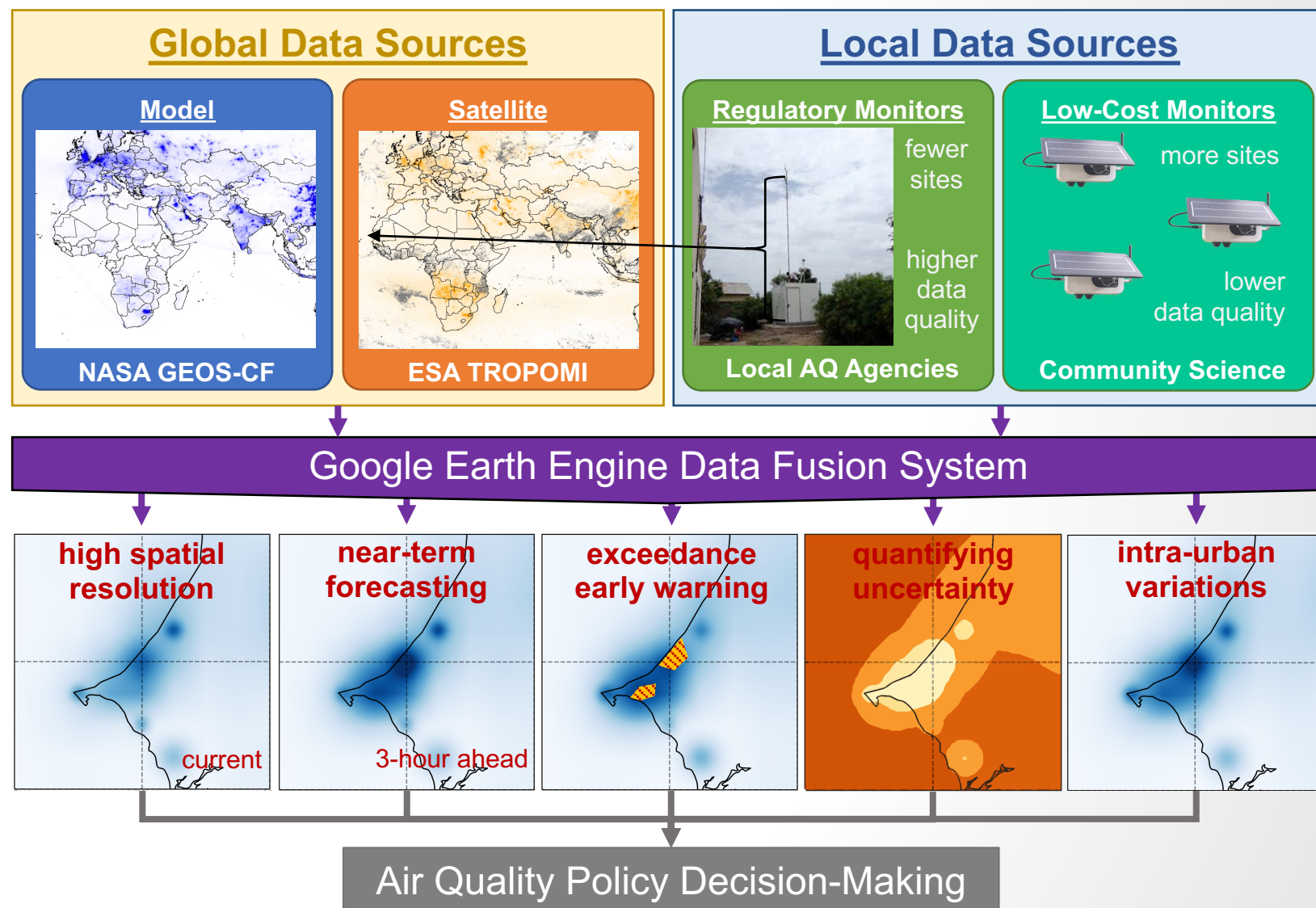
Our ongoing NASA-funded project's objective is to...

...integrate diverse **global** and **local** air quality data sources...

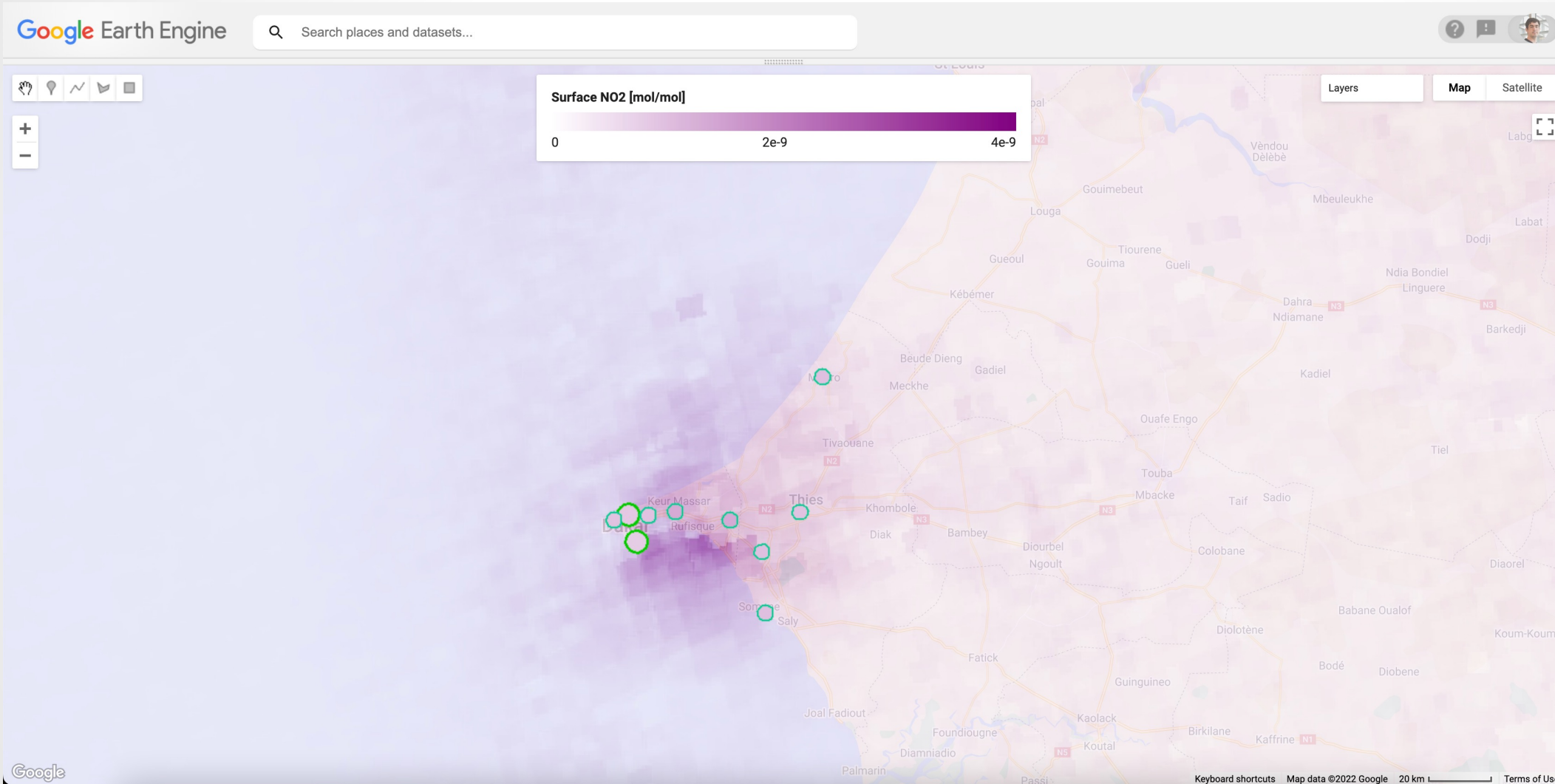
...using the cloud computing platform of **Google Earth Engine**...

...to provide synthesized **estimates** and **forecasts** of air quality at a **local scale** but with a **global scope**...

...which will be freely accessible by air quality managers worldwide, facilitating their **decision-making** processes.

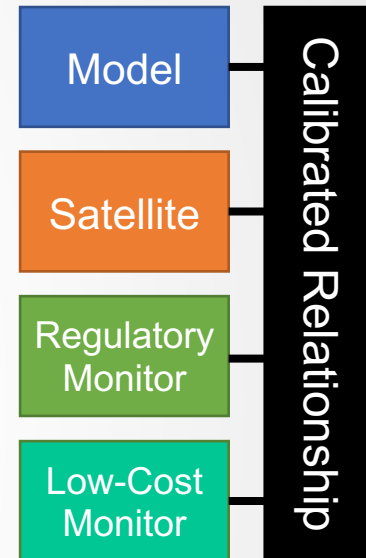


Data Fusion in GEE (preliminary demonstration)

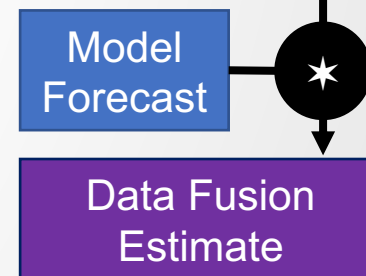


Screenshot of <https://code.earthengine.google.com/> (code by presenter)

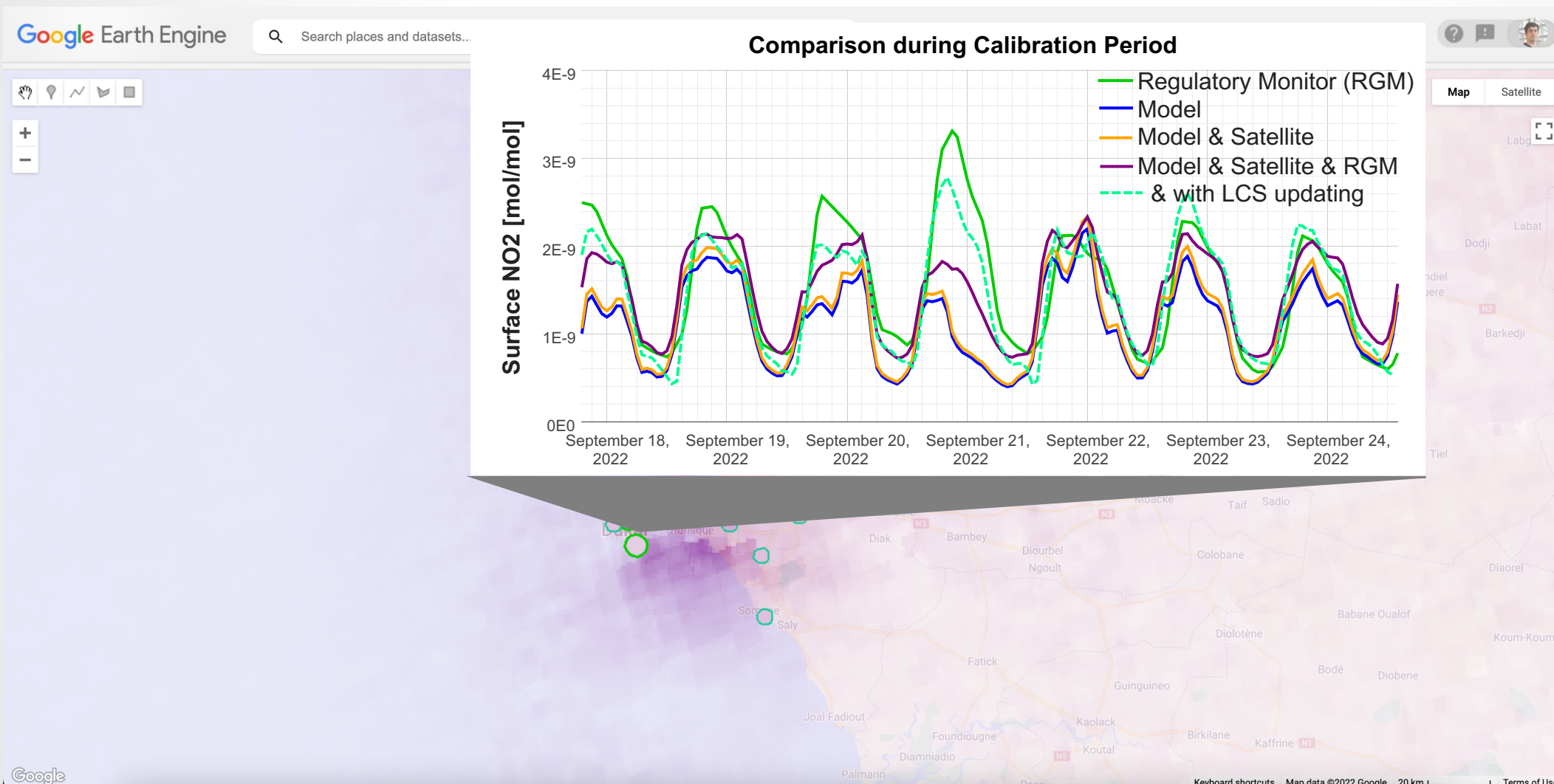
Calibration



Application

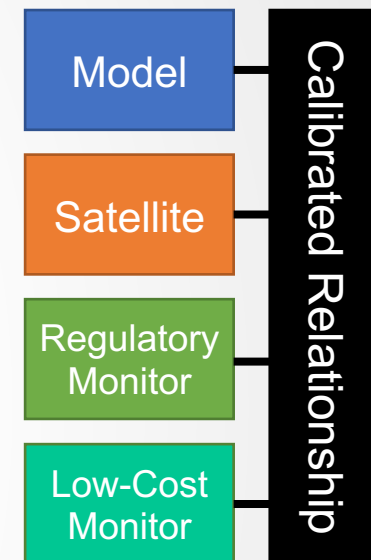


Data Fusion in GEE (preliminary demonstration)

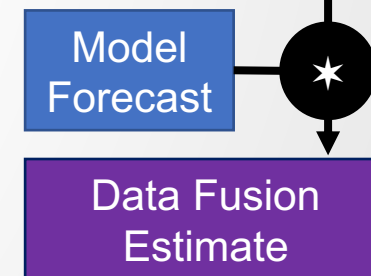


Screenshot of <https://code.earthengine.google.com/> (code by presenter)

Calibration



Application





Open data key to address the global air quality problem

Sources of Open Data

- National & International agencies (e.g., NASA): satellites & models, regulatory monitoring
- Community science and activism: local insight & data (especially low-cost sensors)

Dissemination of Open Data

- Data can't just be available, it must be accessible (e.g., API access, web tools like NASA Worldview)
- Aggregators (e.g., OpenAQ) have a big role to play here (quality filtering, data systematization, etc.)

Uses of Open Data

- Comparing different data sources boosts confidence in our air quality assessments
- Synthesizing from all available (open) data sources will give the most robust insights

Integration of Open Data

- Open geospatial data analysis platforms (e.g., Google Earth Engine) bring together many type of data with computing power to answer questions and test solutions.



Thank You!

Questions & Comments?

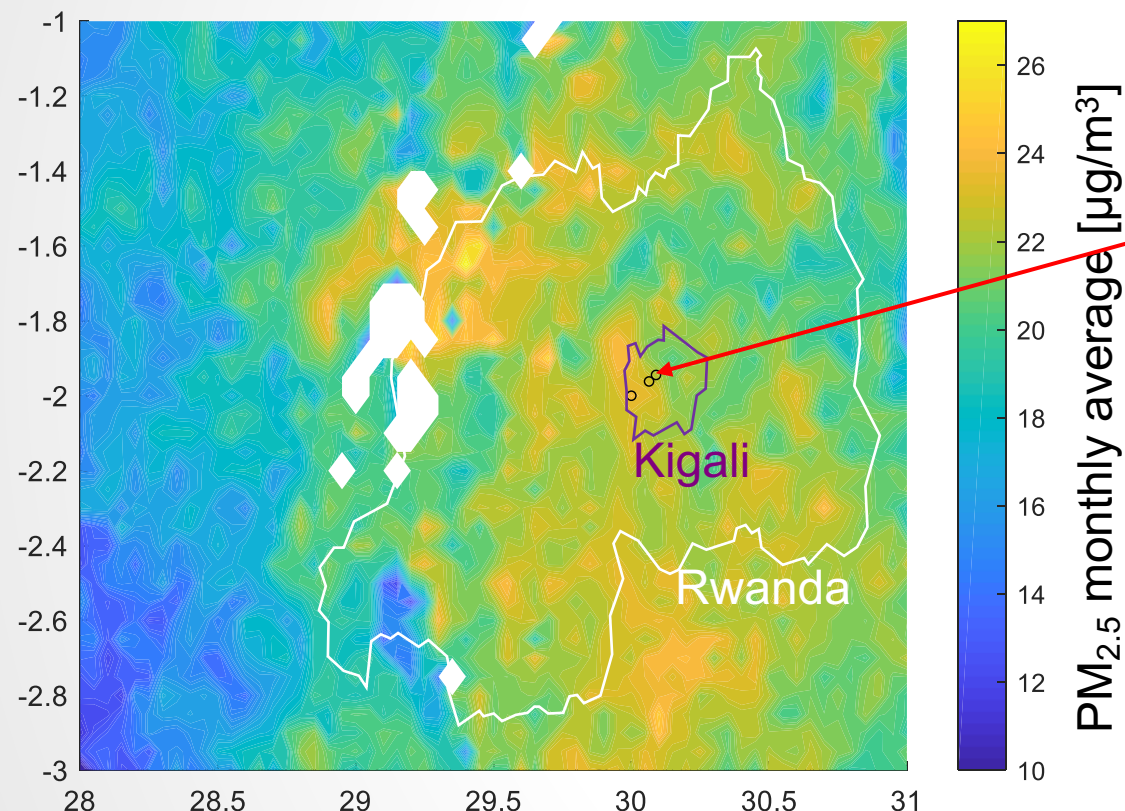
This material is based upon work supported by the National Aeronautics and Space Administration (NASA) under Grant 80NSSC22K1473 issued through the NASA Health and Air Quality Applied Sciences Program.



Backup Slides

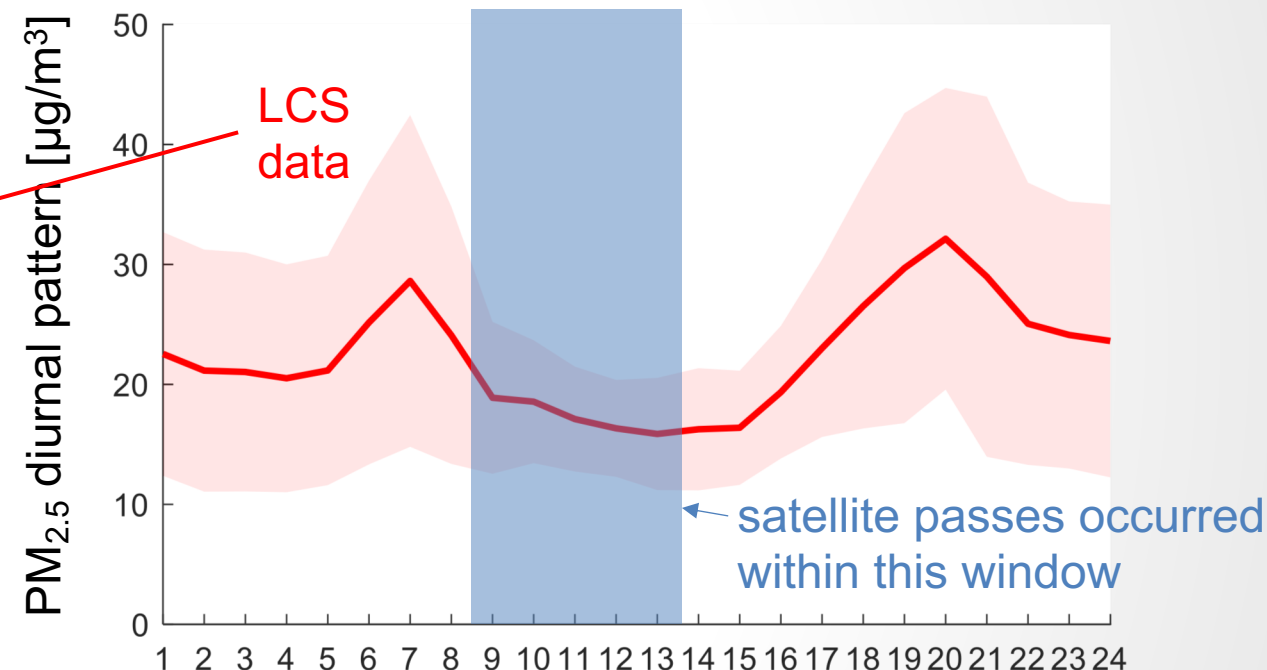
Global (Satellites) and Local (LCS) are complementary

Spatial Coverage (satellite)



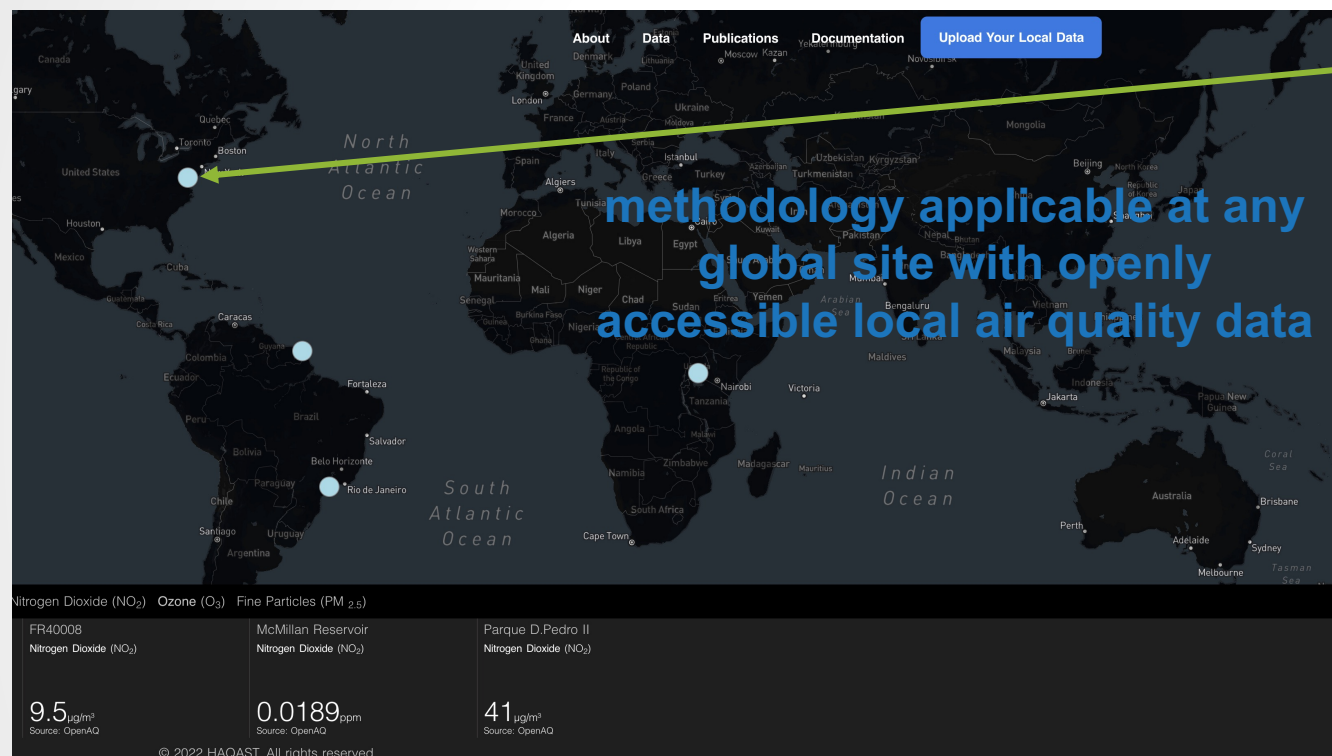
Source: Malings et al. (2020), "Application of low-cost fine particulate mass monitors to convert satellite AOD to surface concentrations in North America and Africa" *Atmospheric Measurement Techniques*. DOI: 10.5194/amt-13-3873-2020

Temporal Coverage (LCS)



Satellite gives wide-area map
Satellite only observed near mid-day
LCS measures daily cycle (and peaks)

Localized bias-corrected global forecasts with CityAQ



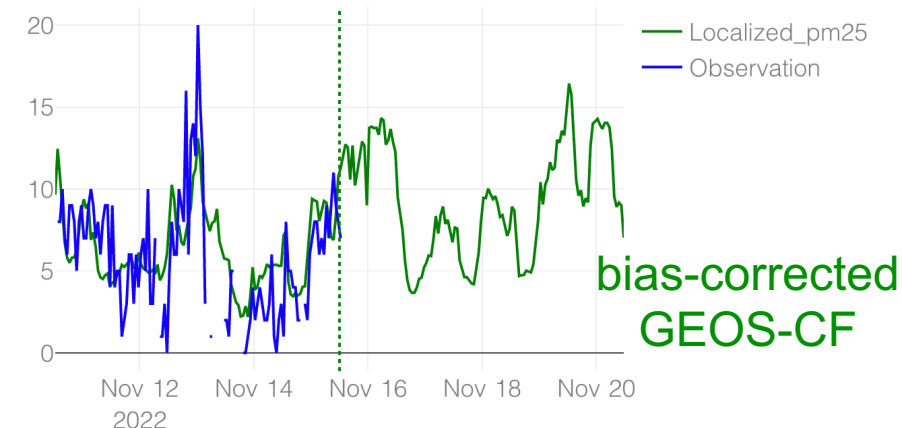
LOCATION: MCMILLAN_RESERVOIR

Observations

Forecasts

Intervals

“ground-truth” air quality via monitors
(available through OpenAQ)



Sources: **Project overview:** <https://www.wri.org/initiatives/cityaq>

Methodology description: Keller et al. (2021). “Global impact of COVID-19 restrictions on the surface concentrations of nitrogen dioxide and ozone.”

Atmospheric Chemistry and Physics. DOI: 10.5194/acp-21-3555-2021.

Web-based tool: https://noussairlazrak.github.io/localized_forecasts/#

Bias-correcting global air quality
forecasts with local data from OpenAQ
monitor stations in the CityAQ project

Challenges to be Confronted

Surface/Column Mismatch

Column quantities derived from satellite data may poorly reflect surface conditions for certain pollutants.

Different Local Data Availability

Many nations worldwide lack regulatory AQ monitoring; low-cost sensors can fill this gap, but with lower data quality.

Quantifying Uncertainty

Many sources of uncertainty (model bias, scale differences, local sensor miscalibration) confound data fusion.

User Acceptance and Adoption

End-users need to have confidence in the utility, accuracy, and transparency of the tool and underlying methods.

