

Can spatial properties of atmospheric CO₂ be used to verify realistic sectoral emissions trends and spatial shifts?

Nina Randazzo^{1,2}, Lesley Ott¹, Michael Long^{1,3}, Sourish Basu^{1,2}, Brad Weir^{1,4}, Tomohiro Oda^{2,5,6}

1 NASA Goddard Space Flight Center, Greenbelt, Maryland, USA; 2 University of Maryland, College Park, Maryland, USA; 3 Science Systems and Applications Incorporated, Lanham, Maryland, USA; 4 Morgan State University, Baltimore, Maryland, USA; 5 Universities Space Research Association, Washington, DC, USA; 6 Osaka University, Suita, Osaka, Japan

Summary of methods

- We focused on the Northwestern United States, a region that has experienced substantial decreases in power plant emissions since the early 2000s.
- Sector-specific emissions, biogenic fluxes, and advected CO₂ were propagated through an atmospheric model as tagged tracers (Figure 1).
- Gridded simulated atmospheric CO₂ was averaged by month and spatially processed using a high-pass spatial filter to minimize the influence of background CO₂ and isolate the signal associated with local sectoral emissions hotspots.
- The contribution of individual tracers to the temporal trends in the processed CO₂ signals associated with emissions hotspots was examined (Figure 1).
- Specifically, we tested whether trends in sectoral emissions hotspots could be inferred from total CO₂ before and after spatial processing under this “best case scenario” of perfect knowledge of atmospheric CO₂.

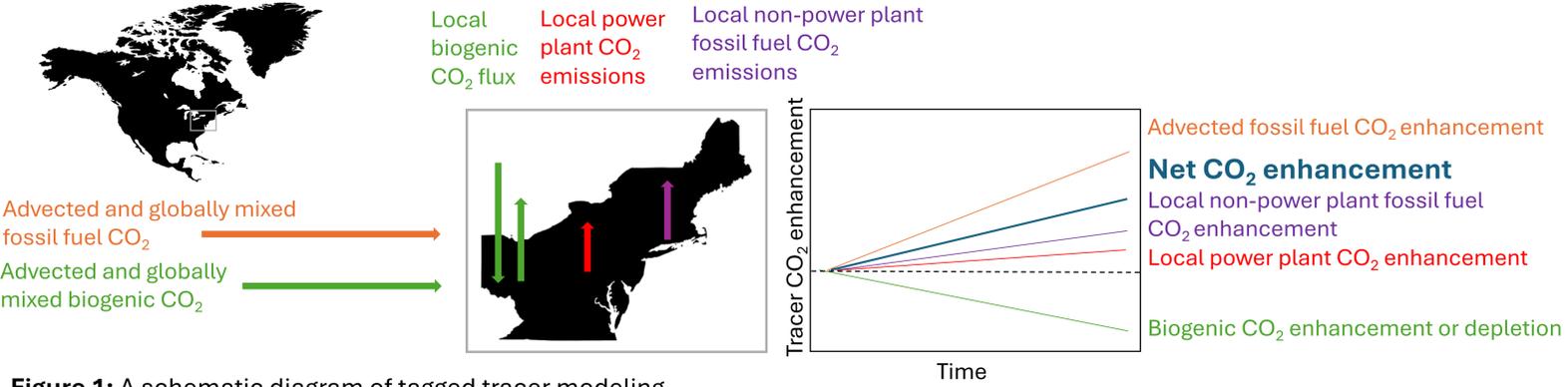


Figure 1: A schematic diagram of tagged tracer modeling

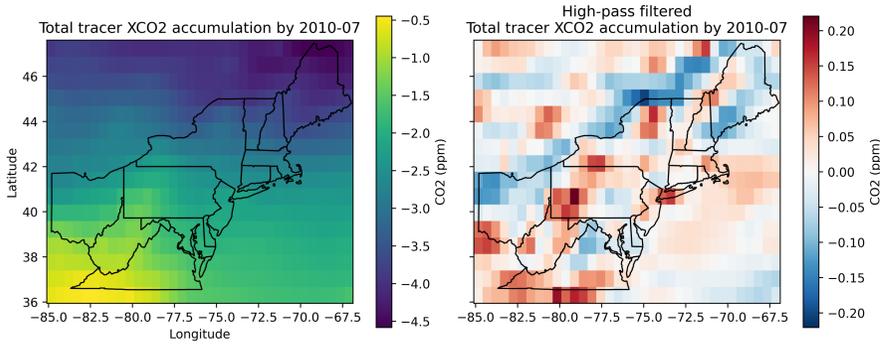


Figure 3: By July of the second year of the simulation, the accumulation of CO₂ tracers shows a spatial pattern in total column CO₂ dominated by a global latitudinal gradient (left panel), but a high-pass filter can identify more local variability (right panel).

Results

- Areas within the study region experienced particularly dramatic decreases in power plant emissions (Figure 2).
- While patterns of overall CO₂ tracer accumulation are unsurprisingly dominated by global influences (Figure 3, left panel), applying a high-pass spatial filter to time-averaged column CO₂ or surface CO₂ maps before a time series analysis allows the identification of local spatial variability (Figure 3, right panel) and local sectoral emissions trends relative to adjacent areas and background (Figure 4).
- However, for total column CO₂ (Figure 4, left panel), the signal may be too small to be detectable via remote sensing, even under a scenario of perfect spatial and temporal coverage of observations; under a realistic observation scenario of spatially and temporally sparse observations, the influence of transport variability and instrument uncertainty and drift would further obscure the signal.

Implications

- In order for these methods to be used to detect even large emissions trends and spatial shifts, we would need:
 - Temporally dense observations taken over a long time period
 - Spatially dense observations across regions of interest
- Emerging instruments may help to meet these requirements.
- Even with improved remote observational systems, extensive surface observations may be necessary to detect emissions trends and spatial shifts.
- As we decarbonize other sectors that do not have extensive bottom-up emissions data, these observational questions will become important for the verification of the mitigation of emissions from particular jurisdictional areas.

Data and model

- Emissions and fluxes
 - Energy Information Administration (EIA) unit-level fuel consumption and emissions factors for power plants within study region (Figure 2)
 - EDGAR fossil fuel emissions for other sectors and global emissions
 - CASA biosphere fluxes with land sink adjustment
- GEOS model atmospheric circulation
- High-pass Gaussian filter over monthly average maps of column CO₂ and near-surface CO₂ to isolate local influence

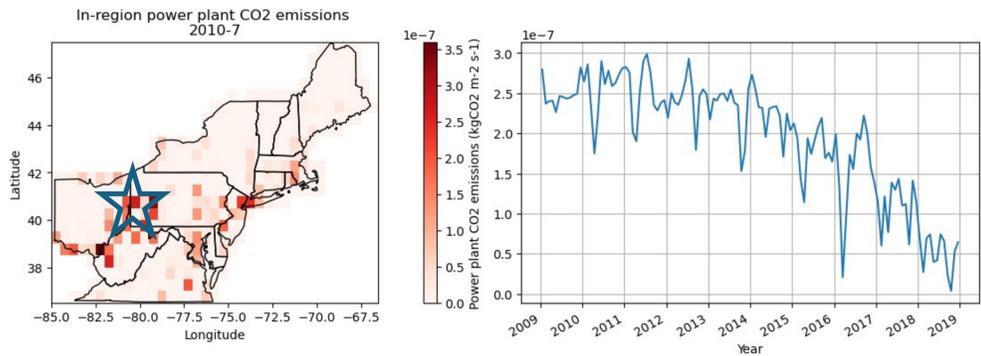


Figure 2: Power plant CO₂ emissions in the region, mapped in the left panel for July of the second year of the simulation, show dramatic decreases across the study period in key areas. In the right panel, we see one such decrease in the grid cell that is highlighted by the star in the left panel.

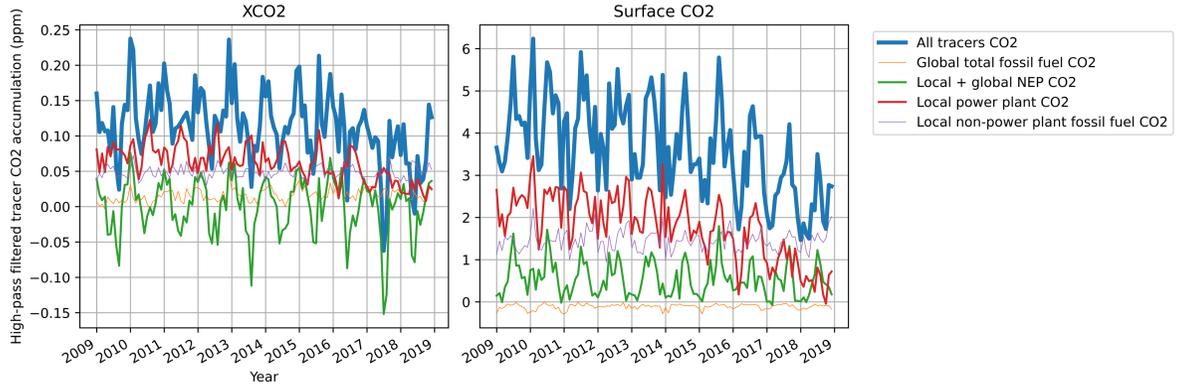


Figure 4: For both total column CO₂ (left panel) and near-surface CO₂ (right panel) in the grid cell highlighted in Figure 2, the trend in local CO₂ total tracer enhancement relative to background, as identified from the high-pass spatial filter, is dominated by local power plant CO₂. However, the total column CO₂ trend is less than a tenth of a part per million over the study period.

Ongoing work

- Maps of total column CO₂ and surface CO₂ trend magnitude and statistical significance for individual tracers, summed tracers, and total CO₂
 - Identification potential covariances between tracers’ trends
 - Quantification of noise
- Simulation of more realistic observations

