

Insights into urban CO₂ emissions from BEACO₂N



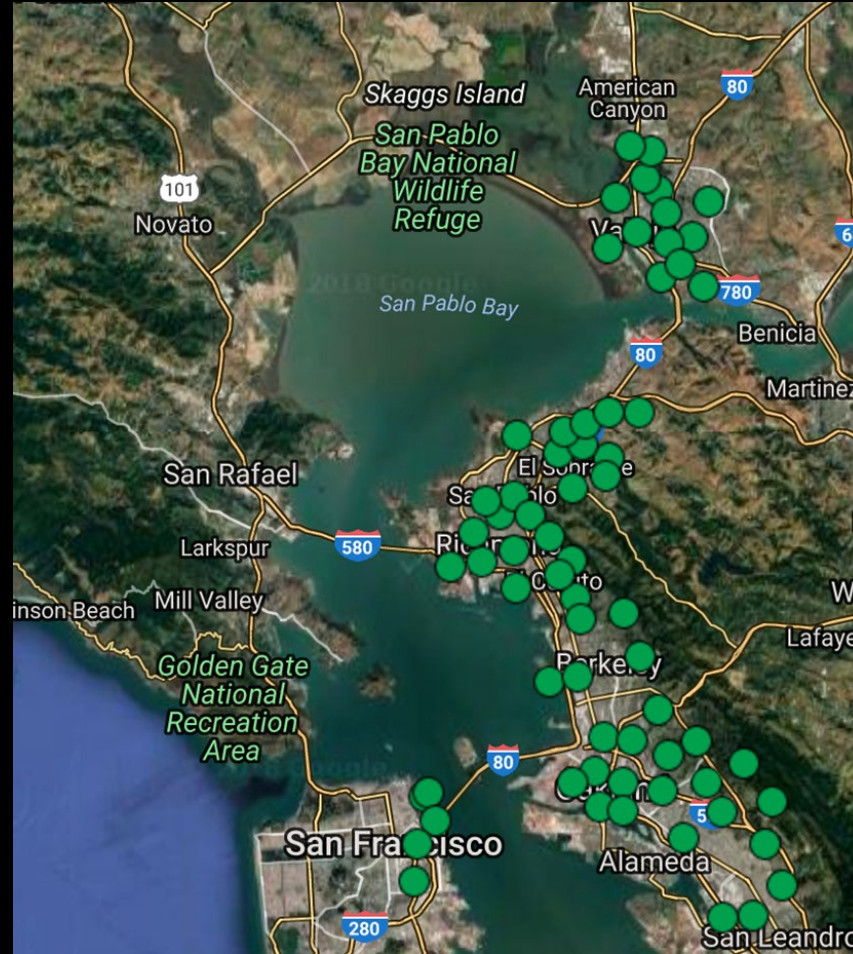
Ronald C. Cohen
UC Berkeley

What observations would support cities' evaluation of their successes in various kinds of emissions reductions?

How can observations point to win-win-wins for climate, air quality and equity of exposure.

Dense Observing systems—a new approach to understanding urban emissions

Berkeley Environmental Air Quality and CO₂ Observation Network



<http://beacon.berkeley.edu>

NO₂, NO, O₃, CO, CO₂, particles

\$8500 + shipping; Monthly fee for share of technical support and data management

Maintenance ~10% or less/yr

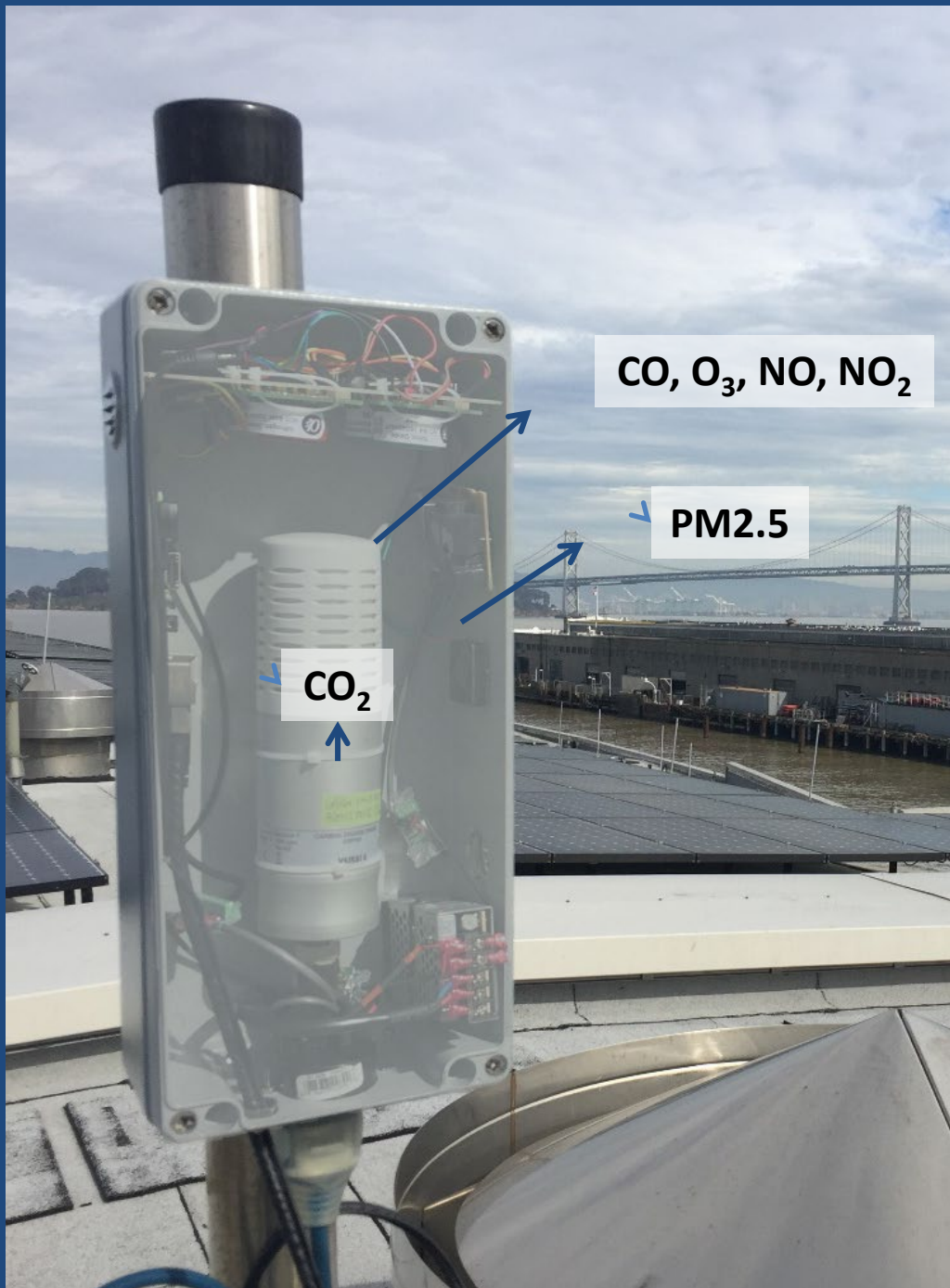
Note: Hardware costs are small compared to ongoing interpretation.

~20nodes = 1 person year

A. Shusterman, et al., ACP. 2016

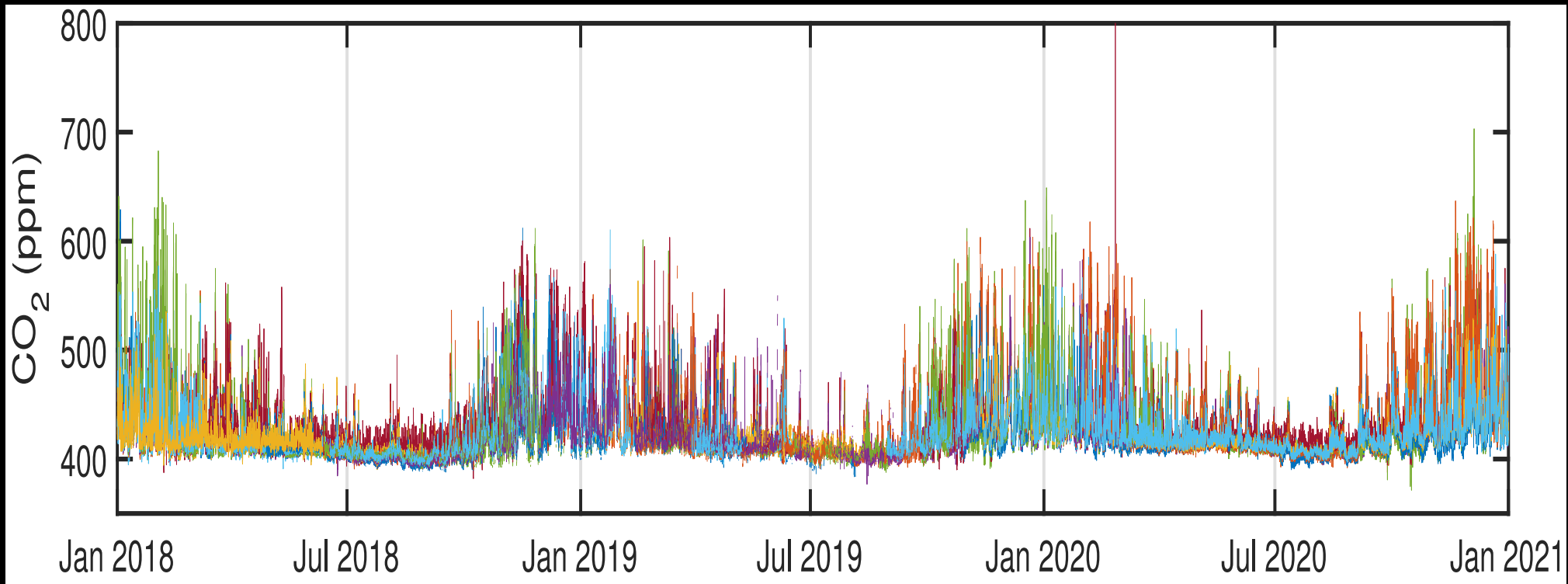
J. Kim, et al. AMT, 2018

E.R. Delaria et al. ACP. 2021

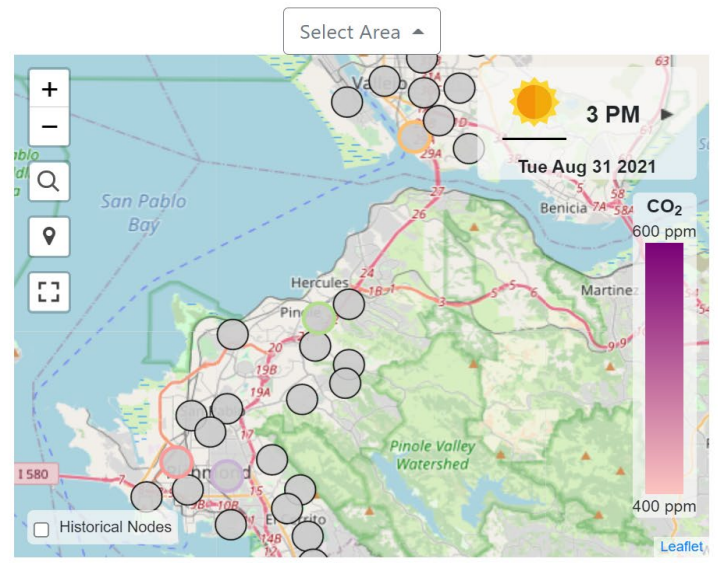




What makes BEACO₂N unique is the combination of CO₂ and AQ. This brings together siloed research (AQ, GHGs, Climate, EJ) and policy communities (NGOs, cities, regions, national government) that should be working more closely together.



climate change, and reports back to this site where the collected data is publicly available for viewing and download. The nodes also collect data on nitrogen oxides, ozone, carbon monoxide, and particulate matter which are indicators for the overall air quality of an area, and may be useful for tracing the origins of CO₂ emissions. CO₂, CO, and aerosol data are available for download via this site. Data for other species is not yet ready for public viewing, but will be added to the site soon.

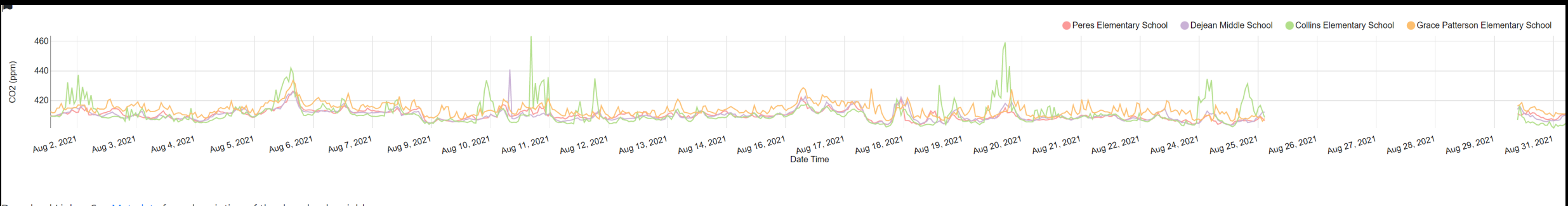


Date Range: 2021-08-01 15:00:00 to 2021-08-31 15:00:00 Real-time CO₂

- × Peres Elementary School
- × Dejean Middle School
- × Collins Elementary School
- × Grace Patterson Elementary School

Data is posted and available to the public as it comes in.

Open data, calibration and interpretation.



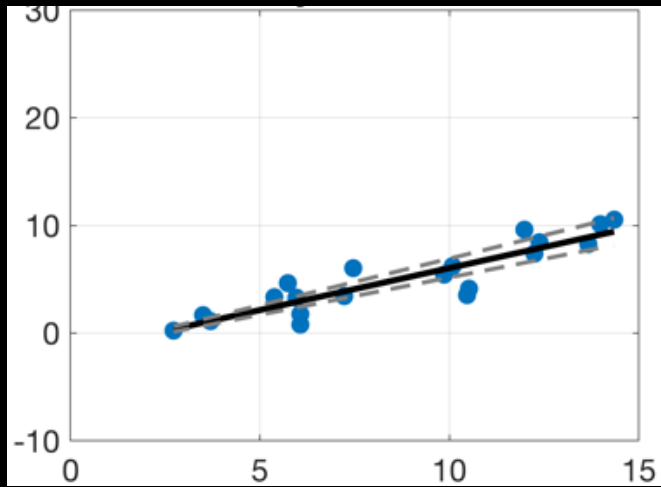
<http://beacon.berkeley.edu>

1. Trends in annual emissions per vehicle

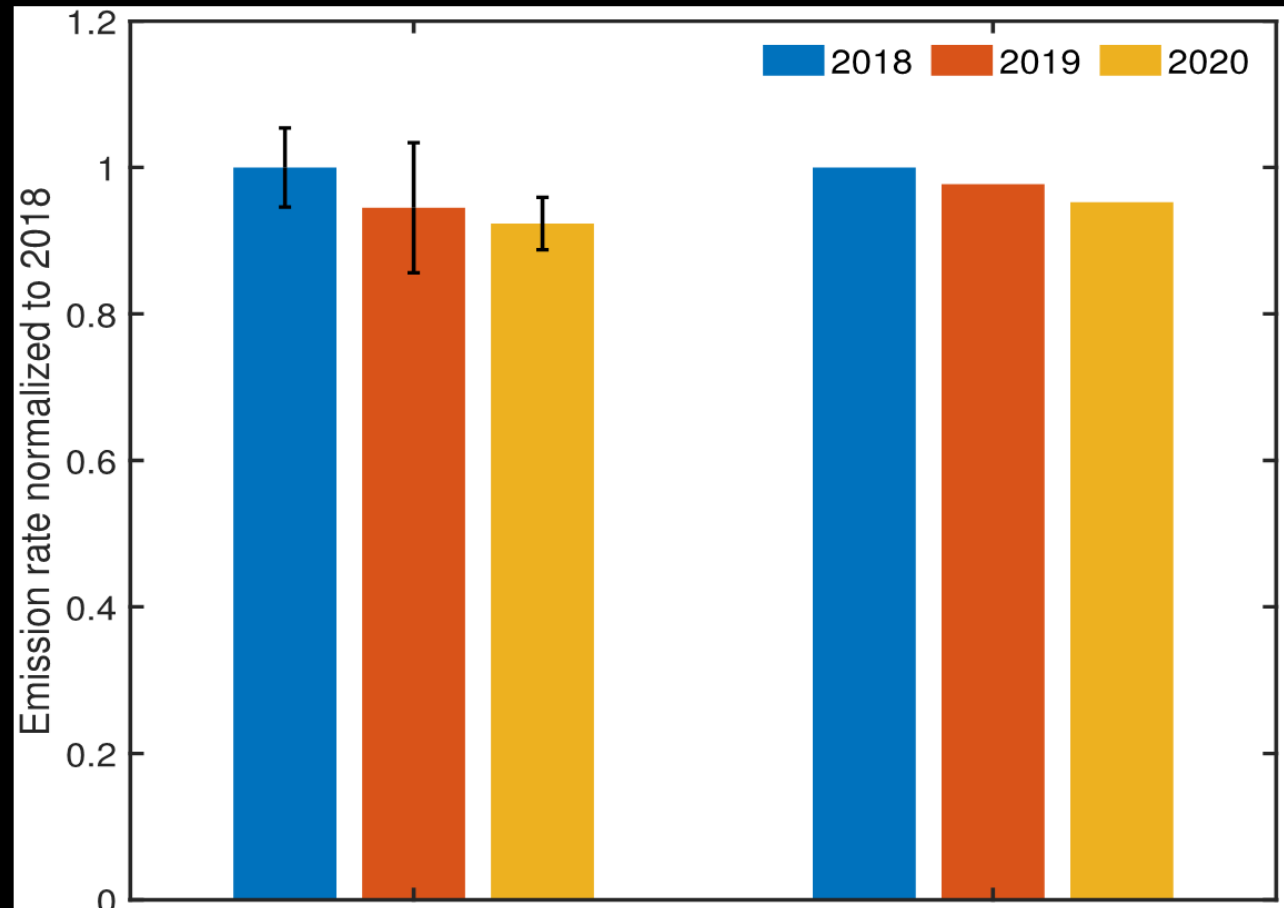


Combining data from 50 locations

Vehicle CO₂



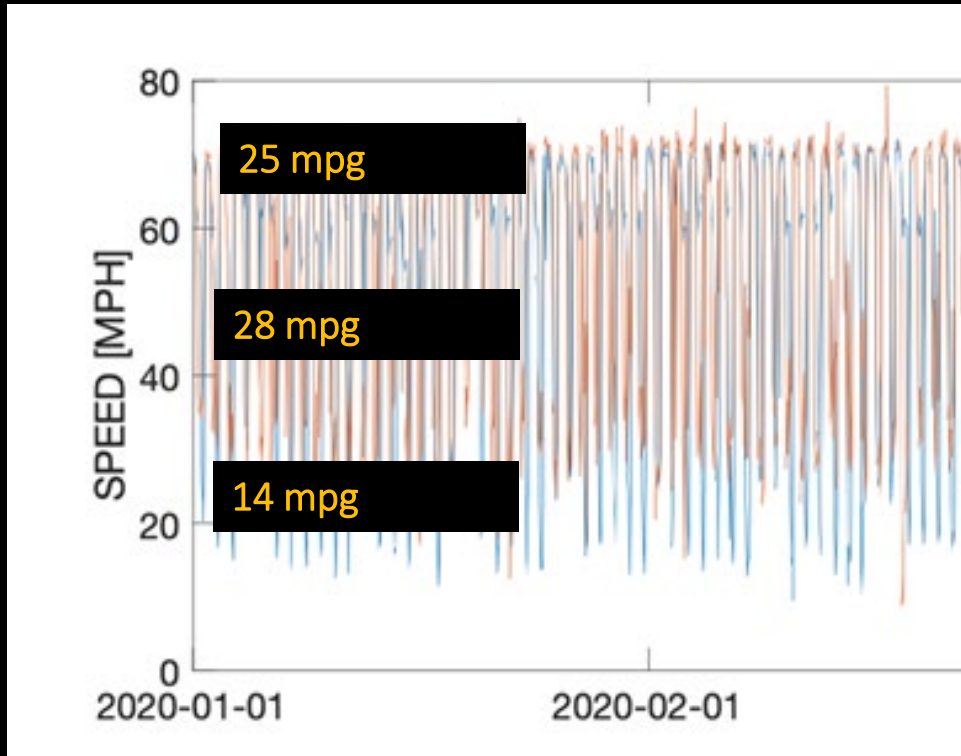
Vehicles per time



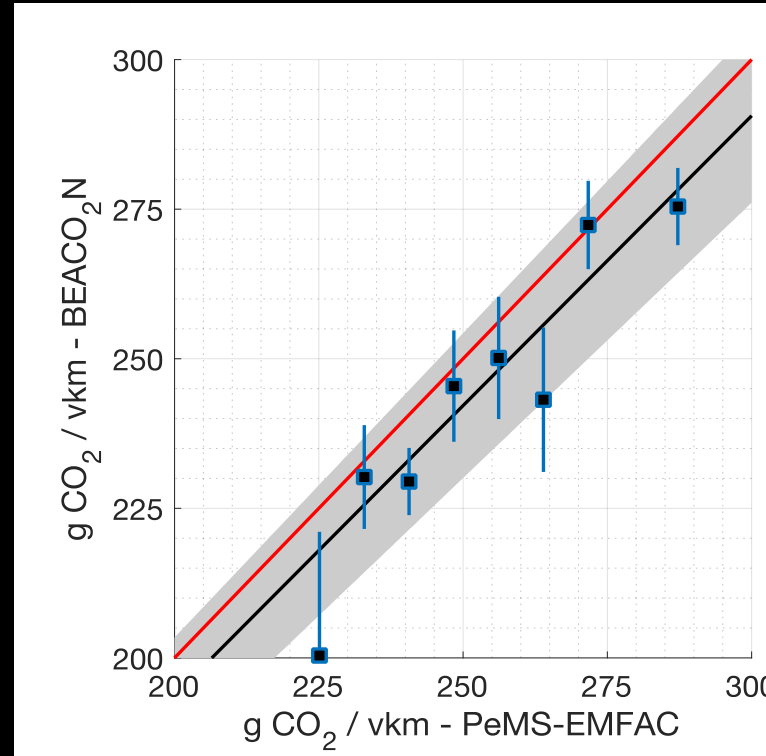
Observed 3.8%/yr

Modeled 2.3%/yr

2. Current fleet fuel efficiency vs. speed



From observations



Slow speeds,
less efficient

Optimal speeds,
more efficient

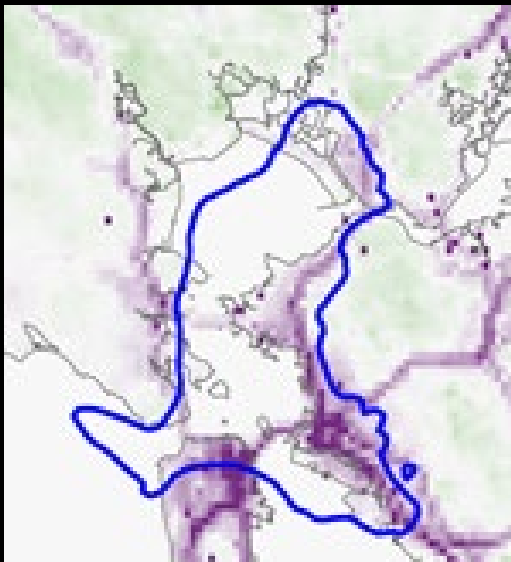
Predictions

BEACO₂N suggests entire vehicle fleet slightly more fuel efficient than predicted at all speeds. Confirms predicted speed dependence is accurate.



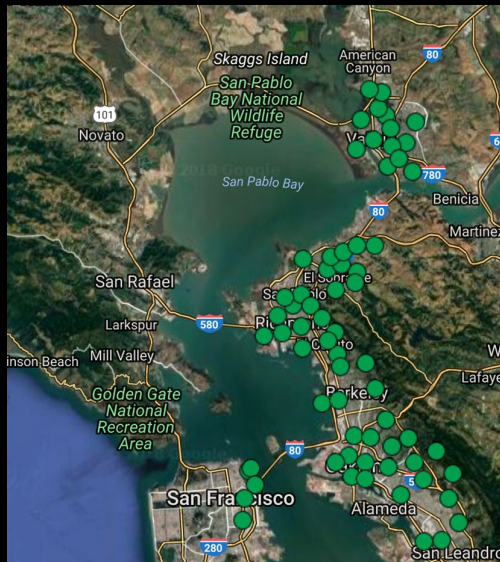
3. How much did CO₂ emissions drop in the early phase of the COVID-19 shelter-in-place and why?

3 steps in a cycle of observation driven understanding of emissions



Predictions/ hypotheses

economic,
social, and
weather



Observations

BEACO₂N

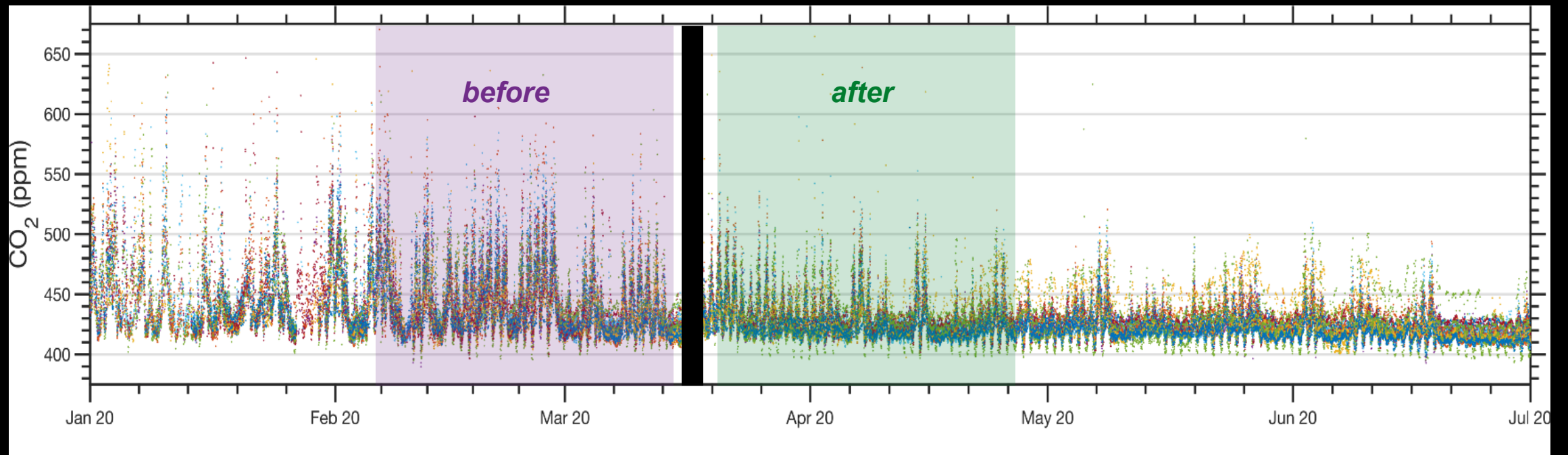
$$\hat{\mathbf{x}} = \mathbf{x}_a + (\mathbf{HB})^T (\mathbf{HBH}^T + \mathbf{R})^{-1} (\mathbf{y} - \mathbf{Hx}_a)$$



Synthesis/iterative improvement

inverse model

BEACO₂N CO₂ observations SF Bay Area January – July 2020

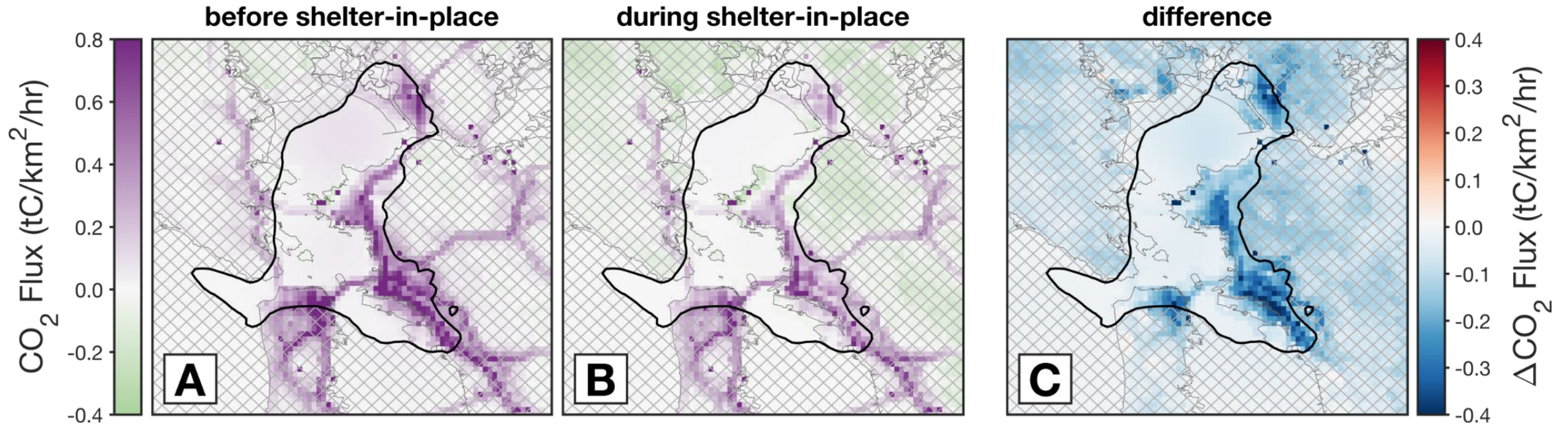


Inverse model leverages synthesizes prior knowledge with observations

$$\hat{\mathbf{x}} = \mathbf{x}_a + (\mathbf{HB})^T \left(\mathbf{HBH}^T + \mathbf{R} \right)^{-1} (\mathbf{y} - \mathbf{Hx}_a)$$

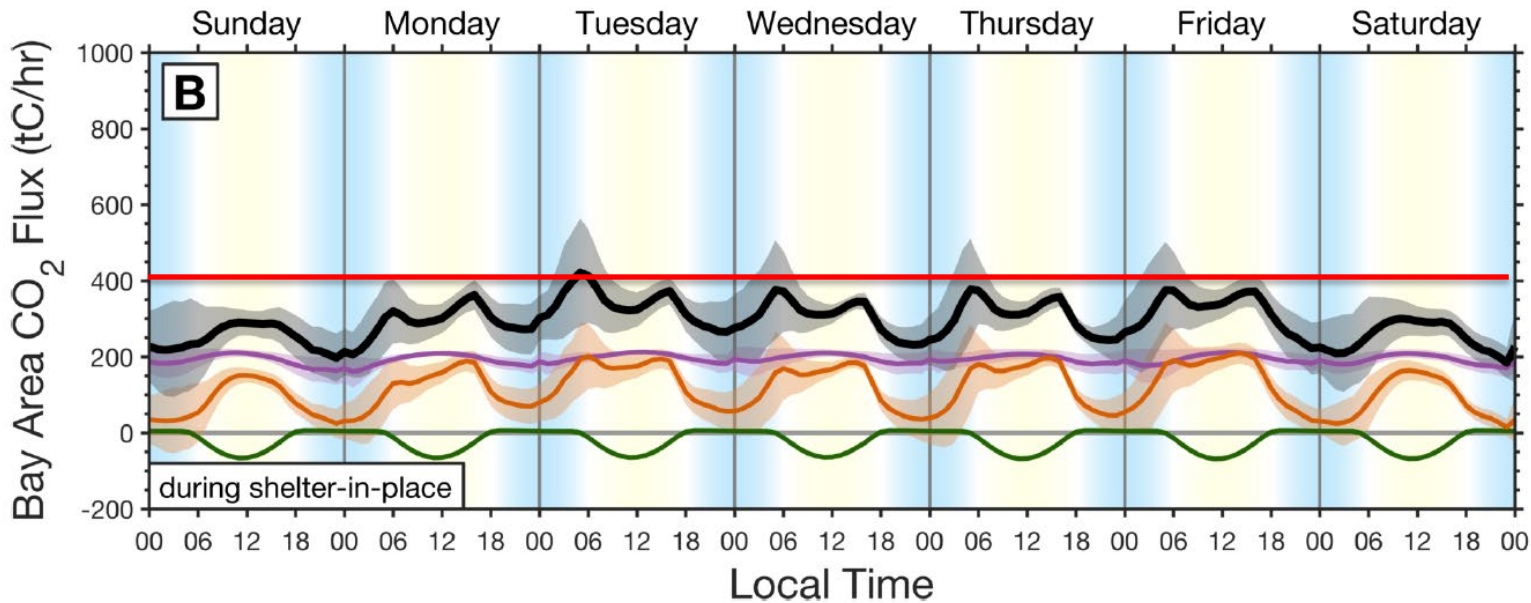
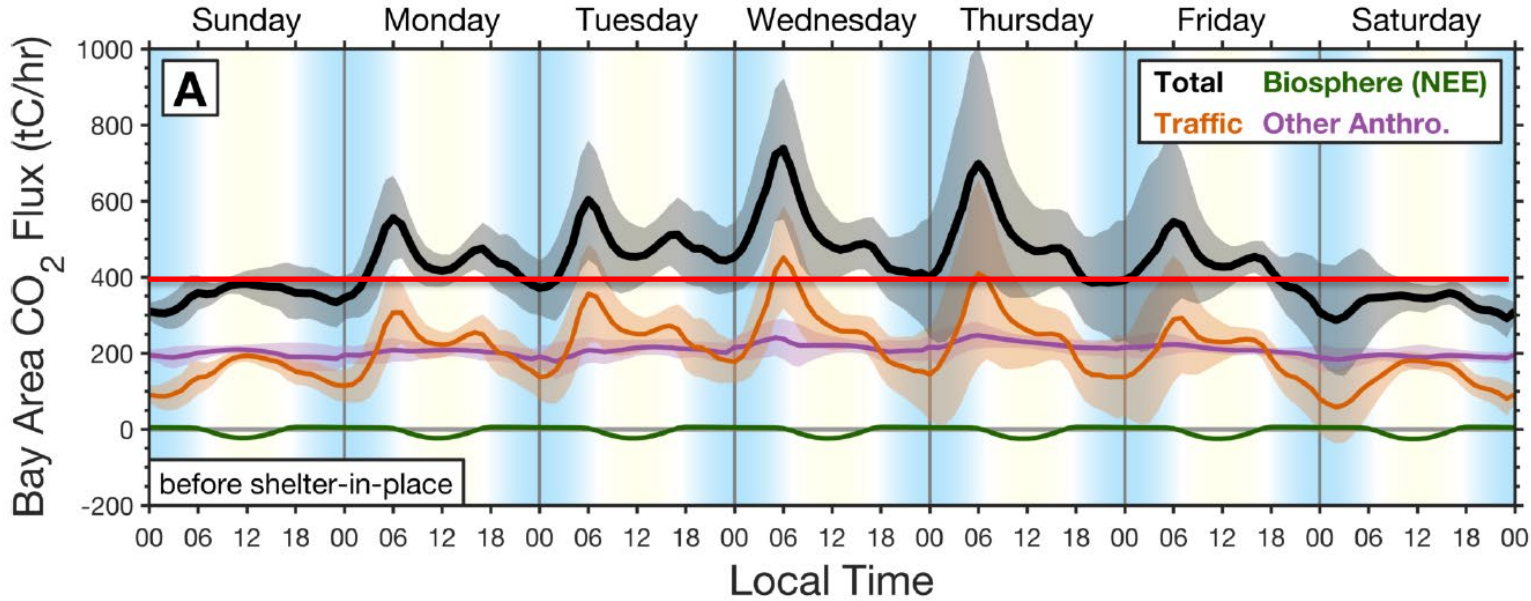


Results on a Map



- ▶ The emissions are best constrained inside the black line
- ▶ Largest changes on freeways; implied emissions on city streets changed

GHG (CO₂) Emissions from the observations



2020

CO₂ emissions before and after COVID shelter-in-place

25% overall reduction

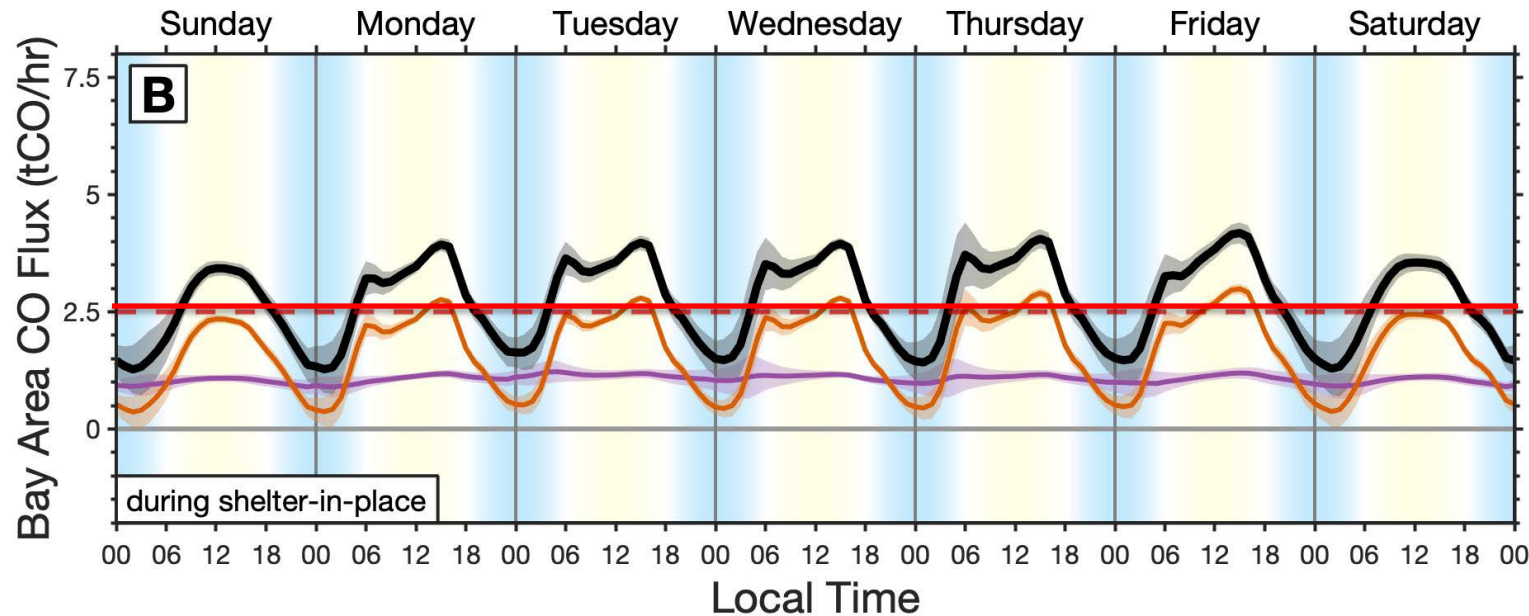
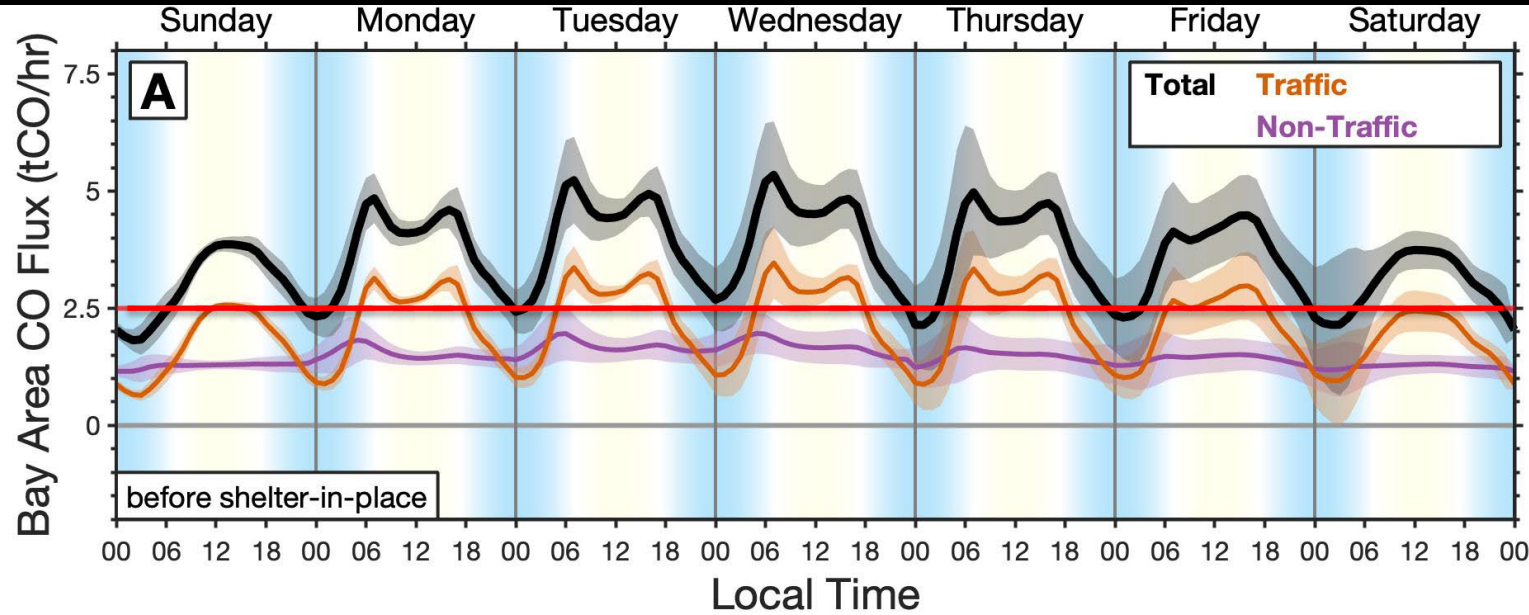
45% vehicle reduction

Air Quality (Carbon Monoxide) Emissions from the observations



2020

CO fluxes before and after COVID shelter-in-place



New locations with BEACO₂N hardware

Glasgow, Scotland — 12, soon 25 nodes
UC Berkeley and Univ. Strathclyde

Los Angeles — 12 nodes
USC, Will Berelson

Leicester, UK — delivered 10 nodes
Univ. Leicester, Hartmut Boesch

Providence — 25 nodes
Brown Univ., Meredith Hastings

In discussions with several other cities, please join us

Conclusions

Demonstrated rapid (within months) reporting of CO₂ emissions and changes, attribution of emissions by sector, testing of processes within emission models (e.g. speed), annual changes in vehicle emissions.

Our tools are ready to support these and other key elements of urban GHG reduction plans.

**AQ results emerging; tests of cold start models.
Connection to CO₂.**



Thank you!

Catherine Newman



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Jinsol Kim



Kevin Worthington



Alex Turner



Paul Wooldridge



Helen Fitzmaurice



Pietro Vannucci



Naomi Asimow



Yishu Zhu

