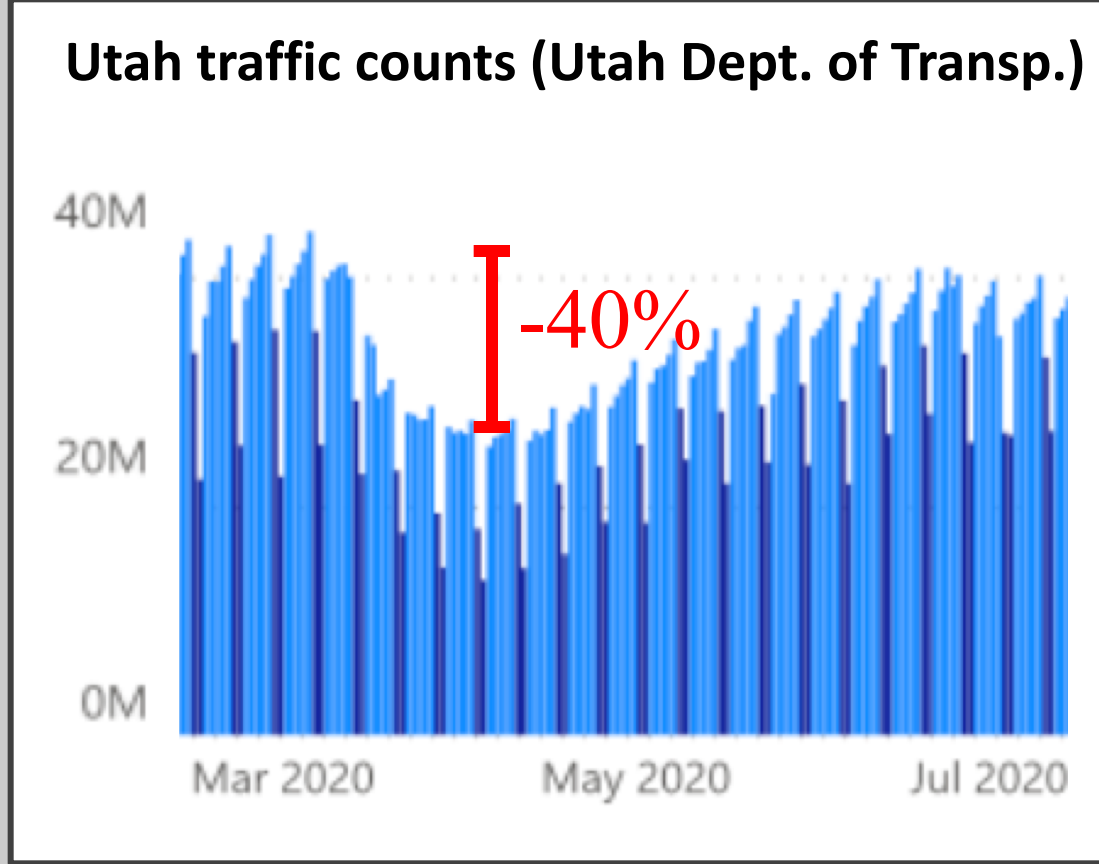


# Quantifying the impacts of the COVID-19 lockdown on urban emissions across the Salt Lake Valley

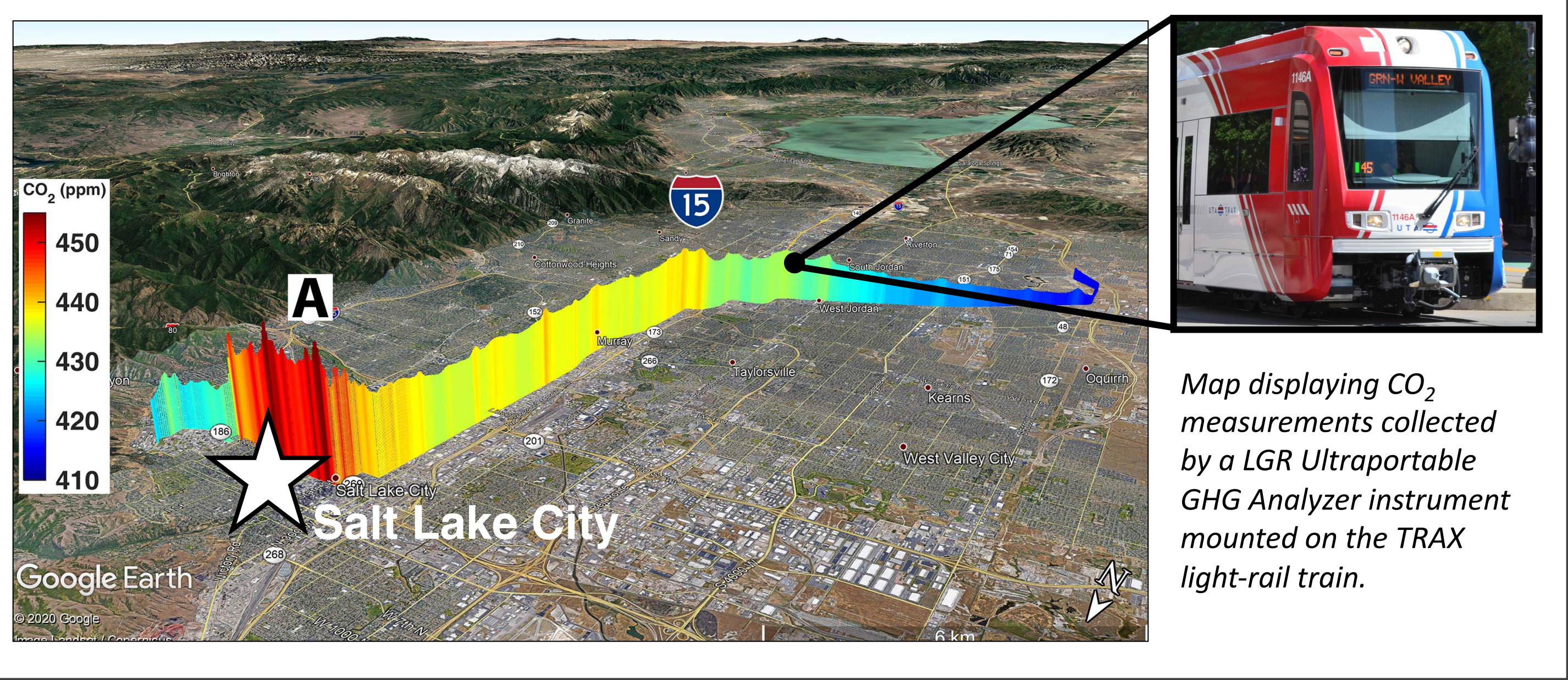
Derek V. Mallia ([Derek.Mallia@utah.edu](mailto:Derek.Mallia@utah.edu)), Logan E. Mitchell, and John C. Lin

## 1. Motivation:

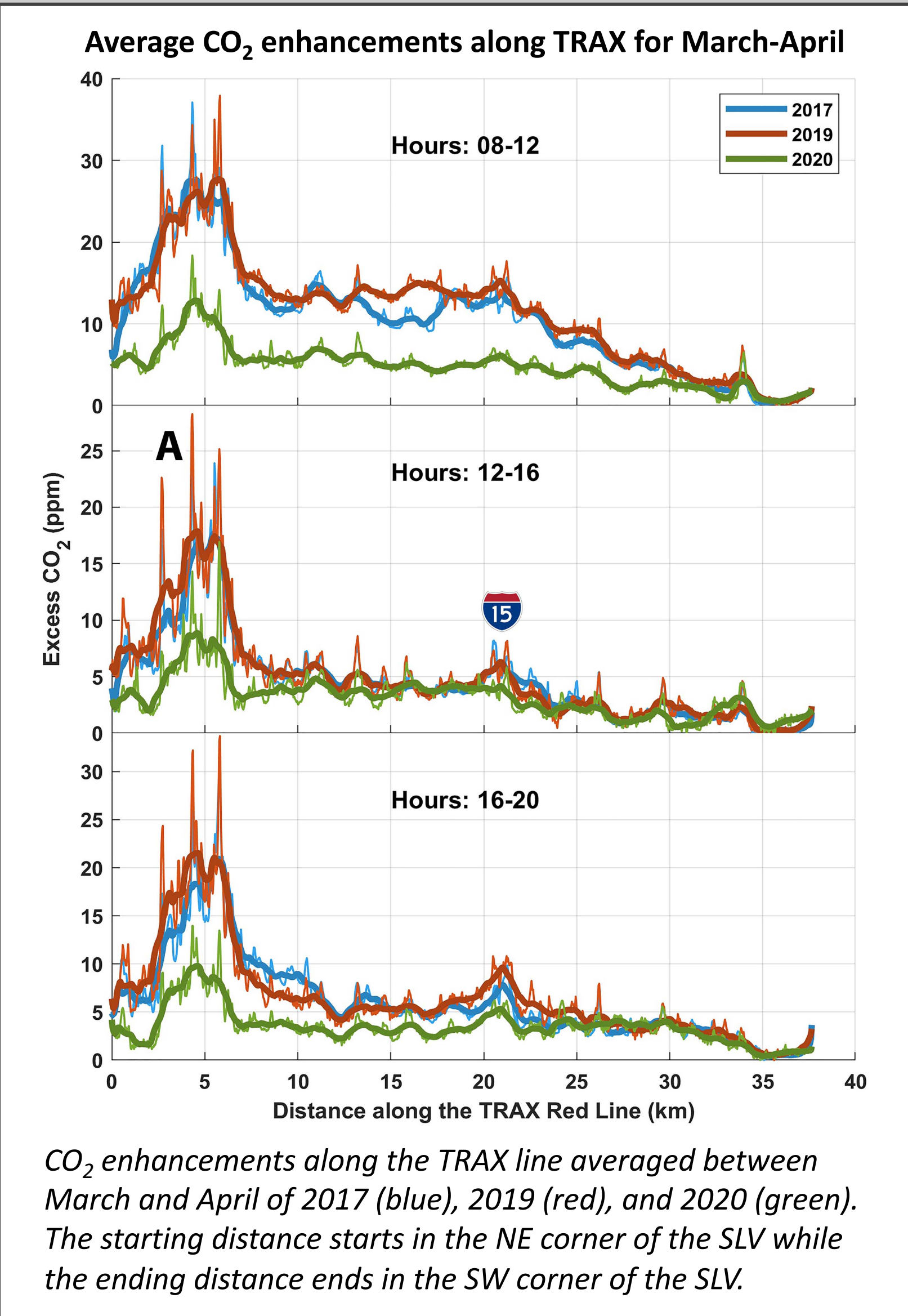
The COVID-19 pandemic resulted in a widespread lockdown during the spring of 2020. The Salt Lake Valley (SLV) observed a large decrease in anthropogenic emissions during the COVID-19 lockdown, which altered the urban atmospheric composition across the valley. During this time, the SLV saw an average decrease of 40% in traffic volume during the lockdown. The COVID-19 lockdown offered a unique opportunity to study how changes in human behavior can impact greenhouse gas (GHG) emissions and air quality.



## 2. Observational Analysis



- Large decreases in CO<sub>2</sub> across much of the SLV in 2020 relative to non-COVID years (2017 & 2019)
- The largest differences in CO<sub>2</sub> were most evident during the early morning (8-12 LST) and across downtown Salt Lake City
- Preliminary results suggest that the COVID-19 had a large impact CO<sub>2</sub> emissions, which was potentially driven by changes in traffic activity
  -  observed some of the largest differences in CO<sub>2</sub> as a percentage



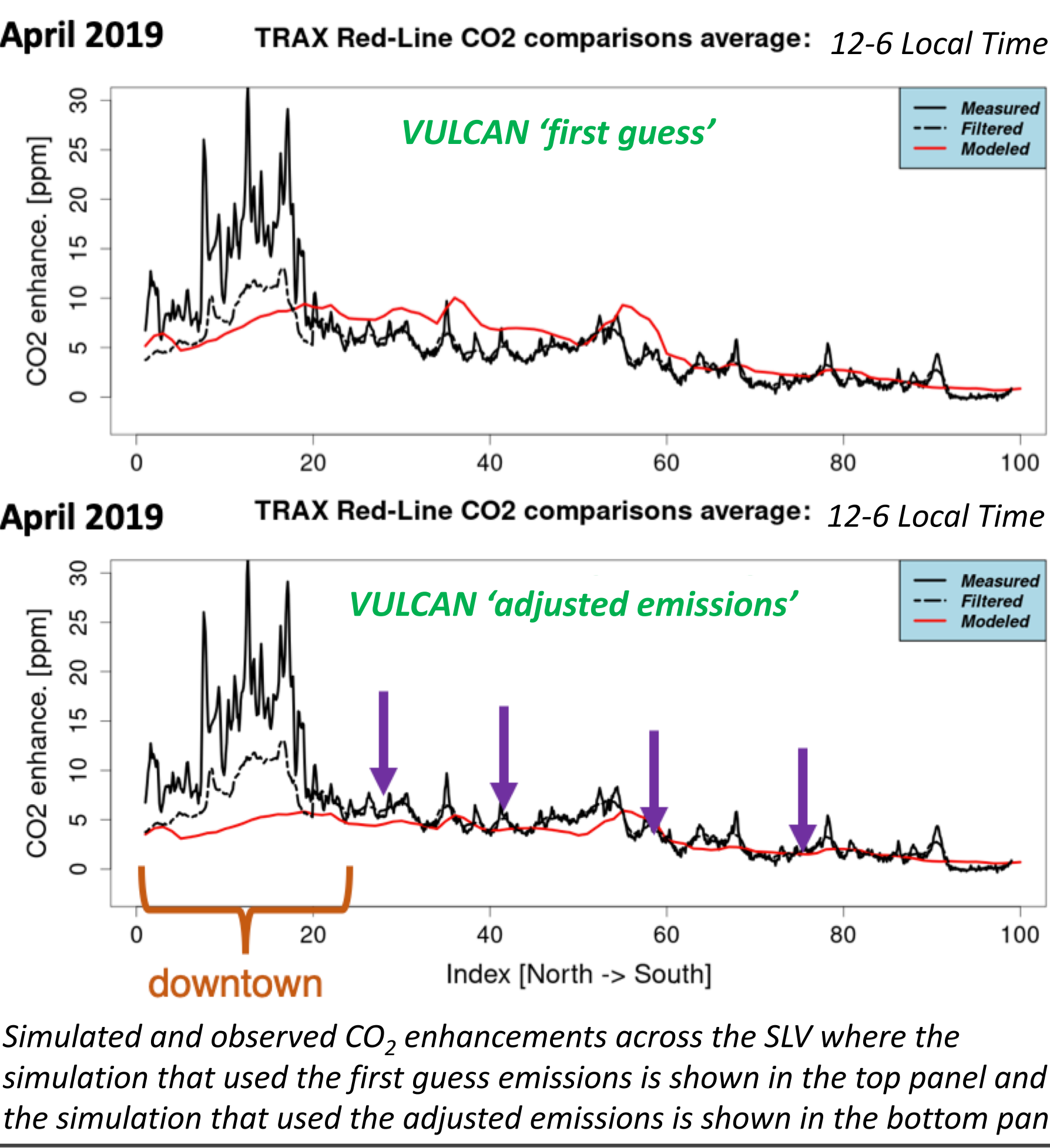
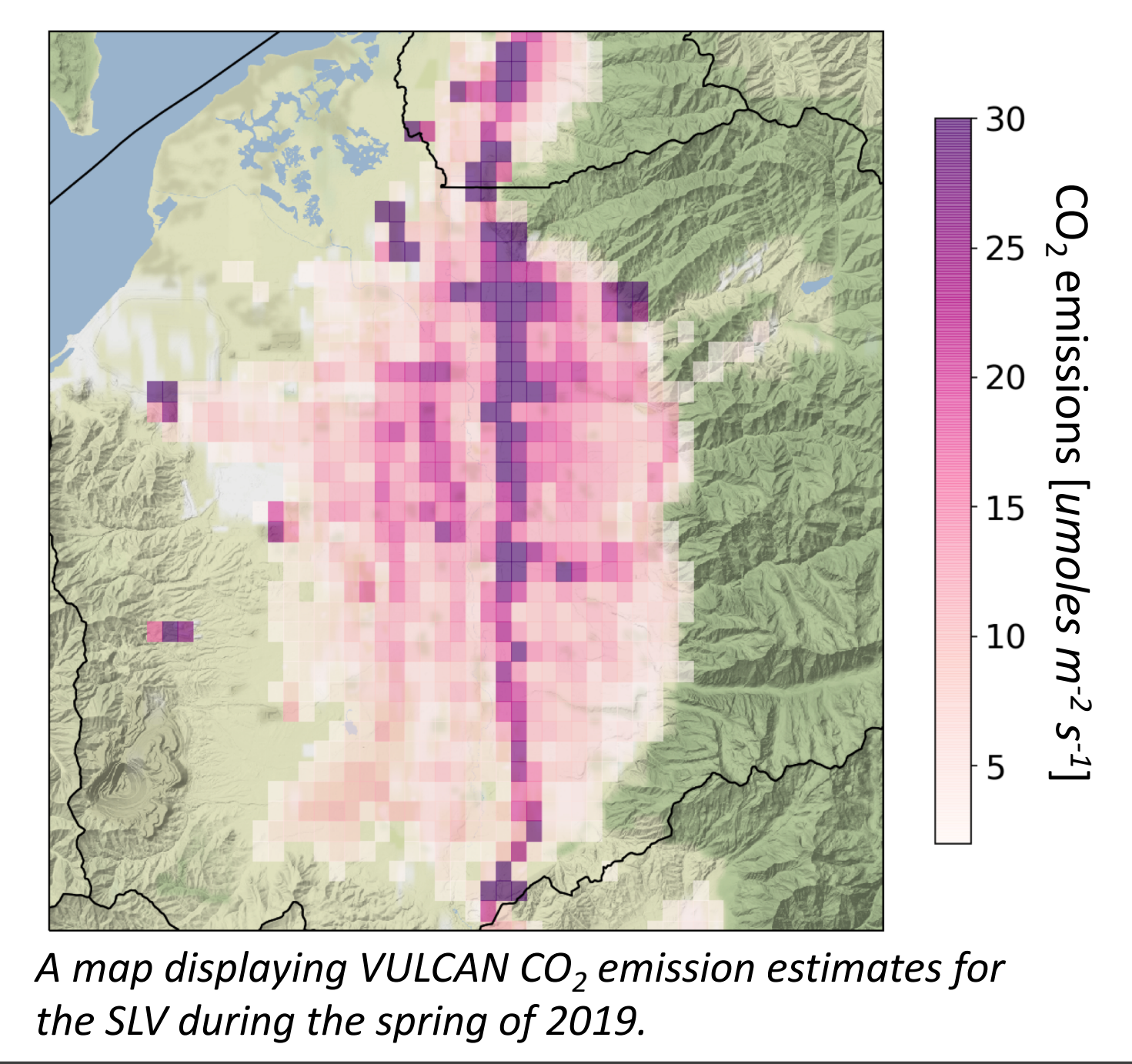
## 3. Results:

An anthropogenic emissions inventory for CO<sub>2</sub> (VULCAN) was combined with an atmospheric transport model (WRF-STILT) to quantify discrepancies between emission inventory estimates during COVID and non-COVID years. Observations and model analyses were used to drive a Bayesian Inverse model to constrain urban emissions:

Adjusted emissions  $\hat{S} = S_p + (HQ)^T (HQH^T + R)^{-1} (Z - Hs_p)$

First guess emissions  $S_p$     Scaling that accounts for model uncertainties, first guess uncertainties, and the correlations of these uncertainties in time and space  $(HQ)^T (HQH^T + R)^{-1}$     Observed CO<sub>2</sub>  $Z$     Modeled predicted CO<sub>2</sub>  $Hs_p$

Equation for the Bayesian Inverse model use to constrain emissions.  $H$  describes the model-predicted spatiotemporal influence for each observation,  $Q$  characterizes the first guess emission uncertainty, and  $R$  describes uncertainties related to the atmospheric model and observations.

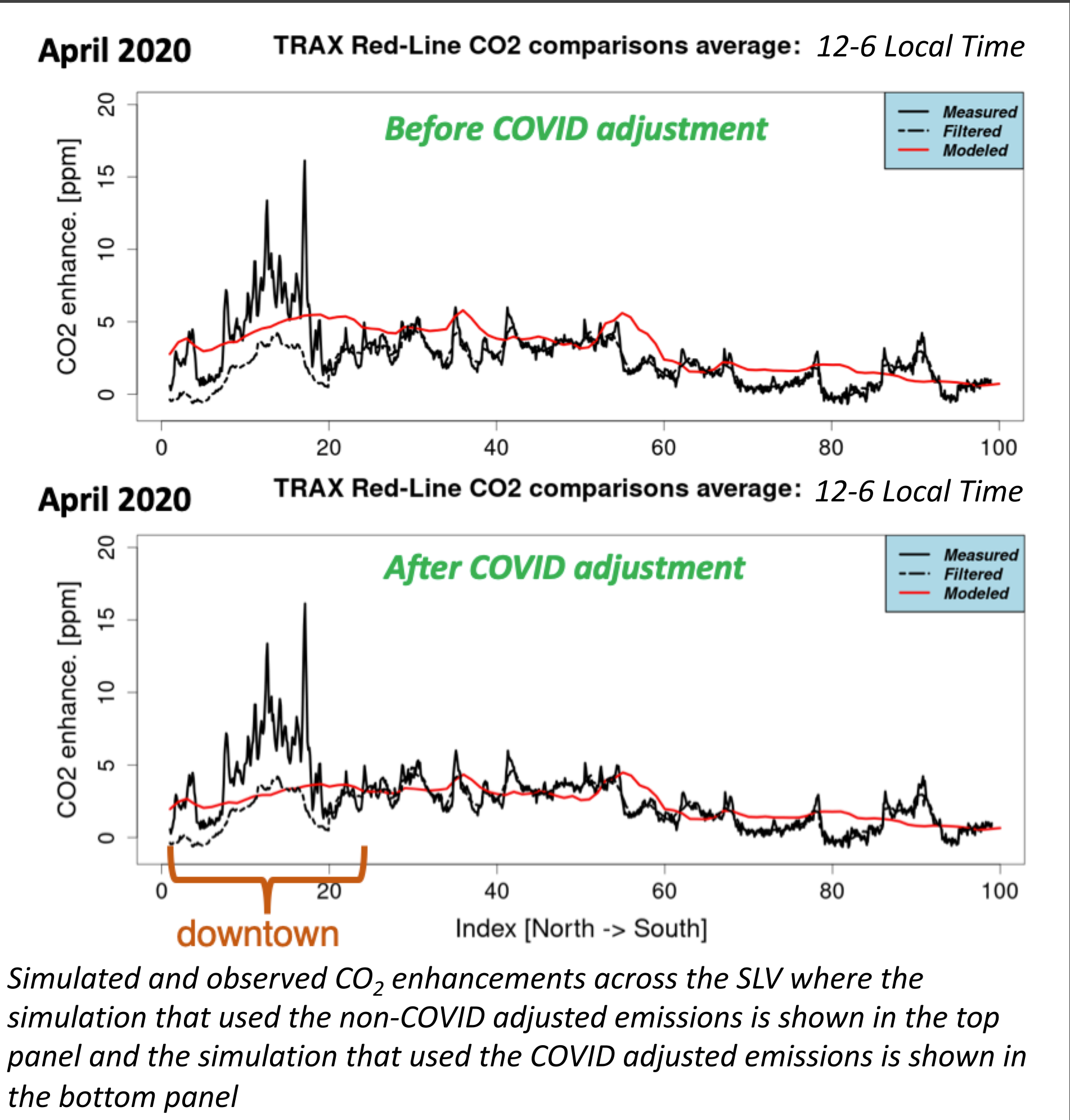


### Step 2: Adjust emissions for COVID-19 lockdown period

- Modeled CO<sub>2</sub> enhancements overestimated, especially over northern half of the SLV when using our re-scaled VULCAN emissions inventory
  - Some overestimations over the southern part of the valley, but less pronounced
- Ran Bayesian Inverse analysis for 2020 to adjust emissions for COVID-19 lockdown
- Inverse analysis aligns modeled CO<sub>2</sub> enhancements more closely with observed concentrations

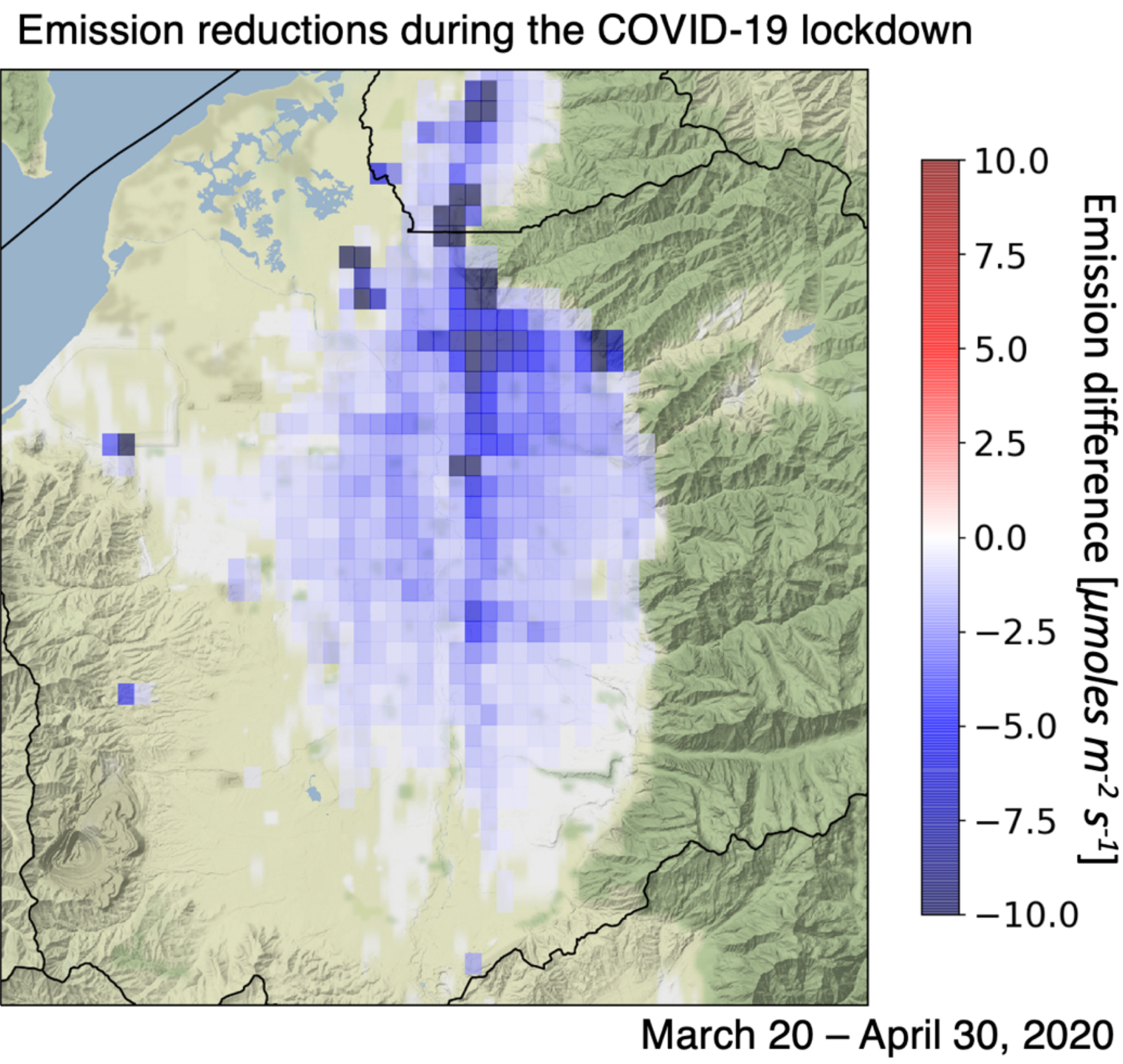
### Step 1: Correct inherent inventory errors for non-COVID year

- Objective of 2019 inverse analysis was to remove errors within our first guess emission inventory
  - Uncertainties at the city-scale can be large (> 30%)
- CO<sub>2</sub> emissions were overestimated by ~30% across the SLV during non-COVID years
- This analysis was used to re-scale VULCAN emissions before running an inverse model analysis for the COVID-19 lockdown in 2020



## 4. Conclusions

- Bayesian Inverse analysis for COVID-19 lockdown (March 20-April 30<sup>th</sup>) resulted in a 18% decrease in CO<sub>2</sub> emissions across the SLV
  - Emission adjustments confined to major roadways and towards the northern part of the SLV
- Emission reductions in agreement with back-of-the-envelope calculations where:
  - 48% of the emissions in the SLV are from traffic
  - ~40% reduction in traffic volume
  - = Expected decrease of 19% for CO<sub>2</sub> emissions



## 5. What's next?

- Large adjustments applied to point sources but are these adjustments real? For example, electricity power generation was likely flat during COVID-19
- Use CO<sub>2</sub> adjustments to correct emission estimates of co-emitted atmospheric pollutants (PM<sub>2.5</sub>, CO, NO<sub>x</sub>, VOCs)
- Can we use this framework to quantify the air quality benefits associated with reduced traffic emissions, e. g., the electrification of SLV's vehicle fleet!

## Acknowledgements

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