# Characterizing Sampling and Detection Properties of AMLD



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### MOTIVATION

- Natural gas (NG) leaks are a potential safety hazard, economic loss, and climate change contributor because NG is mainly composed of the greenhouse gas methane ( $CH_4$ ).
- Advanced mobile leak detection (AMLD) using highly sensitive CH<sub>4</sub> analyzers has developed in recent years for surveying local NG distribution systems in urban areas.
- However, recommendations for driving survey protocols and the effects of driving effort on quantifying and detecting CH<sub>4</sub> point source emissions remain open questions.

- Data included
- We conducted three analyses

#### 0.75 o 0.50 Birmingham Boston Chicago - City A Dallas Elizabethtowr Jacksonville Los Angeles New Jersey Overall Pittsburgh Staten Island Syracuse

Emission Rate (log10 L/min)

#### **ANALYSIS 2: DETECTION AND VERFICATION PROBABILITY**

Statistical Model:  $logit(p_{ij})$  $= \alpha_i + \beta_0 + \beta_1 x_{ij}$ 

- $\alpha_i$ : city specific RE
- $\alpha_i \sim N(0, \sigma_\alpha^2)$
- $x_{ii}$ : estimated emission rate for city *i*, methane source *j*
- Model the probability of detecting a leak on a single drive-by as a function of leak size using logistic regression. Estimate the probability of verifying a leak (2+ detections) as a function of sampling effort.

- Larger leaks are easier to detect and are more likely to be verified with increasing sampling effort relative to small leaks. • Different levels of sampling effort may be needed to achieve similar leak capture proportions across cities.

### METHODS

We analyzed mobile survey data from 15 U.S. cities to inform driving survey protocols and performance.

 Observed peaks: locations with a rise and fall in CH<sub>4</sub> concentrations, suggesting detection of a leak. Verified peaks: locations with 2+ observed peaks. Location detection and drive-by (sampling) history.

• Analysis 1: Error in Size Estimates

• Analysis 2: Detection and Verification Probability Monte Carlo Synthesis

## ANALYSIS 1: ERROR IN SIZE ESTIMATES

- Use a resampling technique to estimate the percent error in estimated leak emission rate as a function of the number of times a leak is detected.
- More detections of a leak improve estimates of leak emission rates. Positive bias in emissions estimates occurs due to occasional large overestimates of leak size and the thresholds in place for flagging the detection of a leak.







### MONTE CARLO SYNTHESIS



R Core Team (2018). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria. URL https://www.R-project.org/. von Fischer, Joseph C., et al. "Rapid, vehicle-based identification of location and magnitude of urban natural gas pipeline leaks." Environmental Science & Technology

Weller ZD, Yang DK, von Fischer JC (2019) An open source algorithm to detect natura gas leaks from mobile methane survey data. PLoS ONE 14(2): e0212287.

- Combine analyses 1 and 2 to conduct a Monte Carlo simulation to demonstrate the use of AMLD for leak detection and quantification in a simulated city with 200 leaks.
- We recommend 5-8 drives as an attainable surveillance effort that will capture a majority (>90%) of leaks and provide reasonable (-10% to +15%) emissions quantification.
- Adaptive AMLD surveys could increase or decrease survey effort based on the empirical detection probability.