

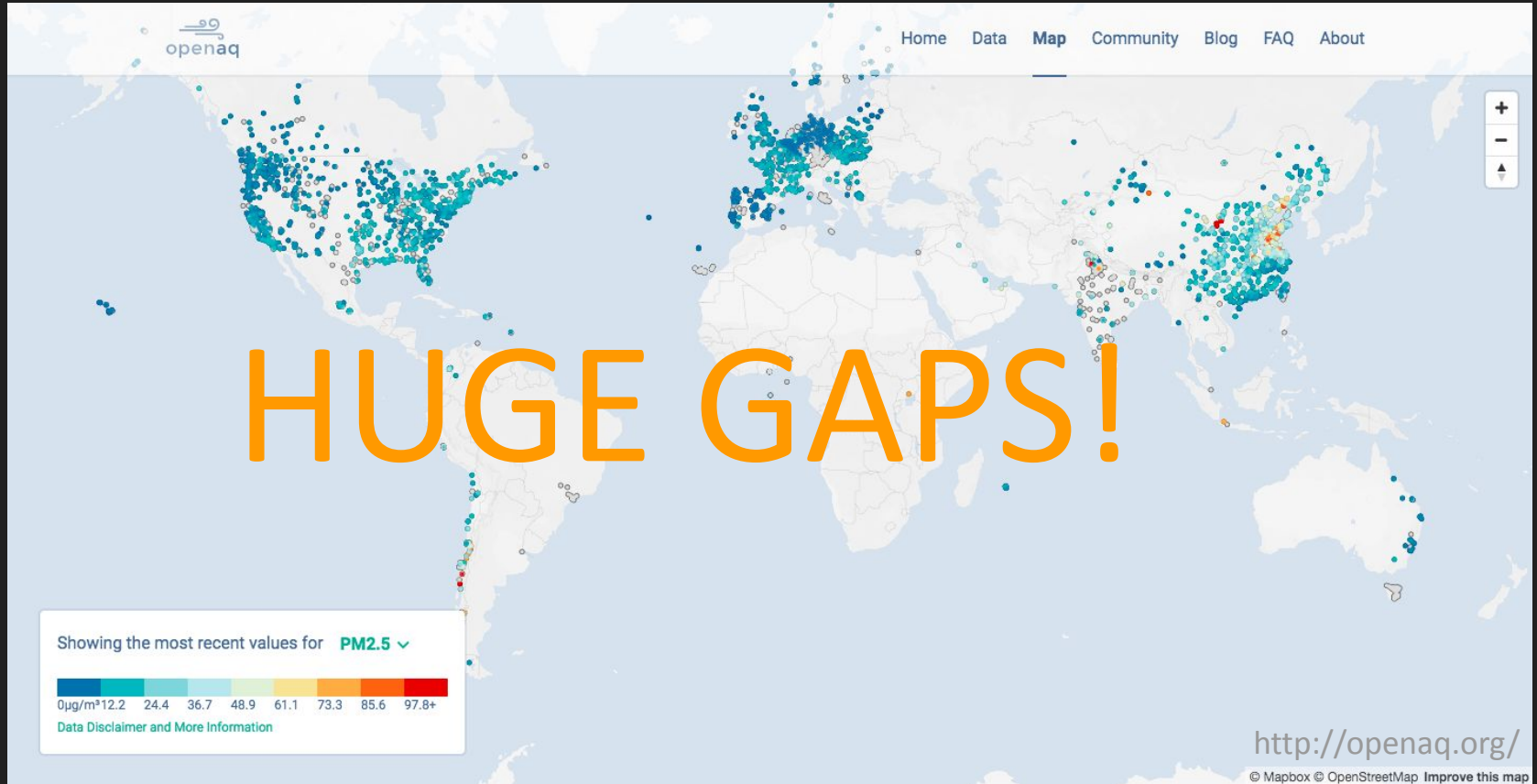
Investigating Use of Low-Cost Sensors to Increase Accuracy and Equity of Real-Time Air Quality Information

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Air Quality Data Reported by Countries



Motivations

- Low-cost sensors (LCS) could fill information gaps between reference monitors
 - Cost per instrument: \$250 vs. \$10,000+
- Concerns about measurement error from LCS
- Persistent air inequality between socioeconomic and demographic groups
 - Discrepancy increasing as overall AQ improves in US (Jbaily et al., 2022)
- deSouza & Kinney (2021): PurpleAir sensors (most common PM_{2.5} LCS) tend to be in more privileged areas
- US EPA recently allocated \$20 million for enhanced community AQ monitoring
 - Priority: environmental justice (EJ) communities

Research Questions

1. How does the **distribution (density and placement)** of low-cost sensors affect real-time air quality (AQ) information, in terms of both **accuracy and equity**?
2. What **mechanisms** drive inaccurate AQ reporting, and how important are these mechanisms relative to each other?
3. How does the **type and amount of sensor measurement error** affect the usefulness of LCS in real-time AQ reporting?

Overview of Simulation Method

- Setting: California, 2016
- Consider daily 1x1km Di et al. PM_{2.5} estimates “truth”
- Assume AQS reference monitors are fixed and have no measurement error
- Select LCS locations (based on one placement strategy), x 50
- Simulate measurement error at “sensor” locations
- Each grid cell “sees” AQ info from the nearest monitor/sensor
- Calculate performance metrics (comparing the “true” AQ to the “shown” AQ), overall and by subgroups, averaged across the 50 trials
- Main analysis: weighted by population density
 - Secondary analysis: unweighted

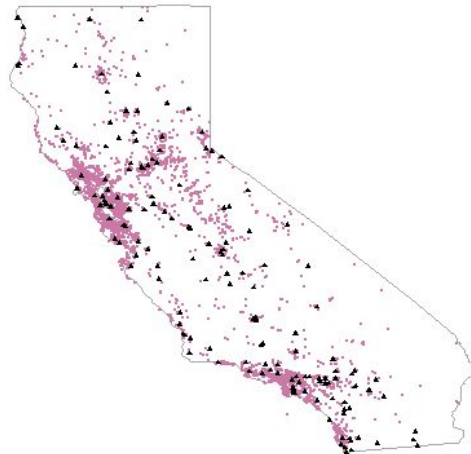
Repeat for different types and amounts of sensor measurement error.

Placement Strategies

Randomly-selected $n = 50, 100, 250, 500,$ and $1,000$:

- Current PurpleAir locations
- Schools
- Weighted by lengths of major roads within a 500m buffer
- Weighted by environmental and socioeconomic marginalization:
 - CalEnviroScreen: Pollution Score and compound CES Score

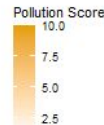
AQS Monitors (triangles) and PurpleAir LCS (dots)



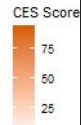
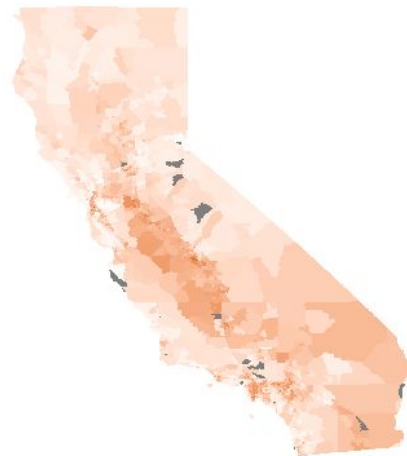
Public Schools and Highways



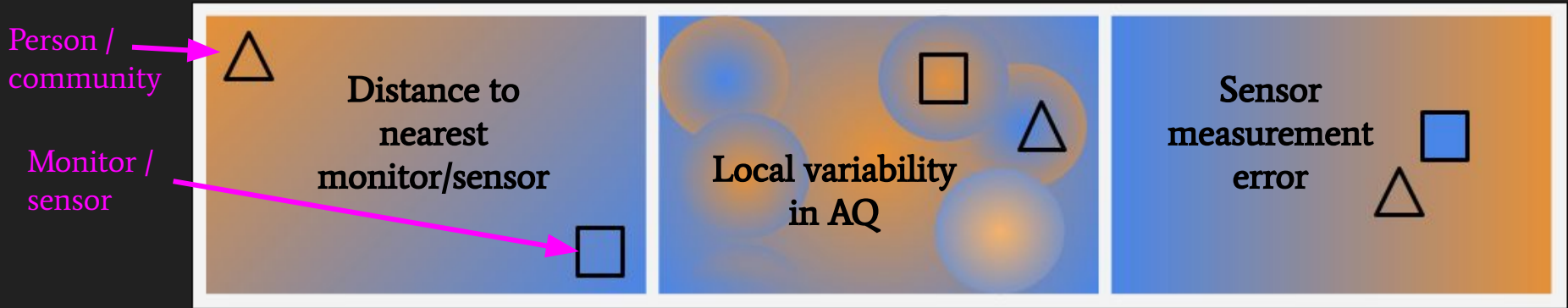
Pollution Score



CES Score



Investigating Mechanisms Driving Errors in Reported AQ

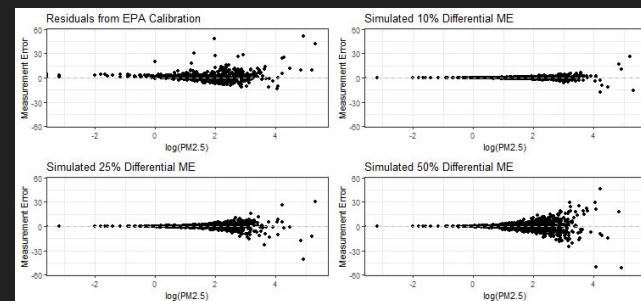


Calculations:

- Distance to nearest monitor/sensor
 - Overall and among misclassified observations
 - “Classes” from EPA AQI
- Sensitivity analysis: simulations without adding sensor measurement error

LCS Measurement Error Simulation

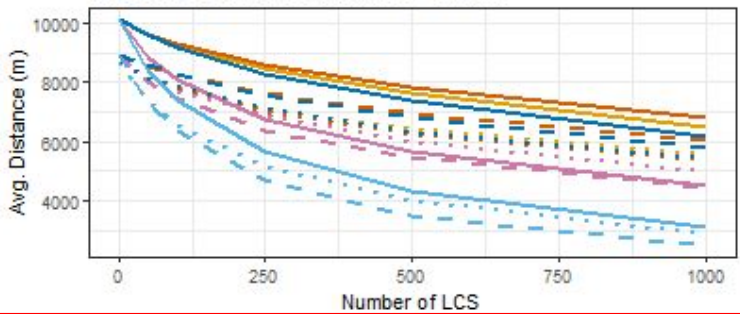
- Williams, et al. (2019):
 - Regulatory monitoring: $\pm 10\%$ of the average AQ levels in an area
 - Mapping spatial gradients and monitoring microenvironments: $\pm 25\%$
 - $\pm 50\%$ accuracy still useful for tracking large sources of air pollution
- EPA developed linear correction for PurpleAir in US – Barkjohn, et al. (2021)
- Our simulations:
 - No LCS measurement error
 - Non-differential measurement error, 10% and 25% of avg.
 - Differential measurement error, 10% and 25% of “truth”
 - Sampled empirical residuals from EPA correction
 - Obtained from collocated monitor-sensor pairs



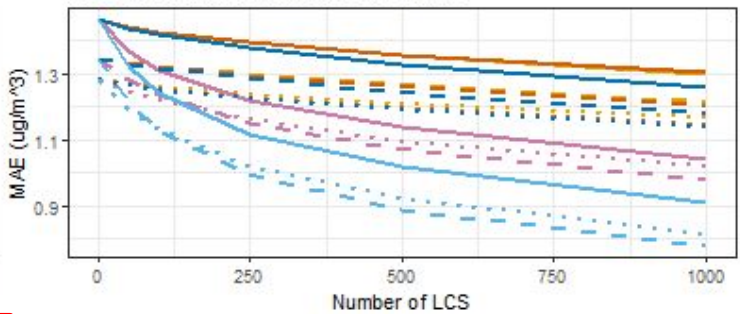
Now for some results...



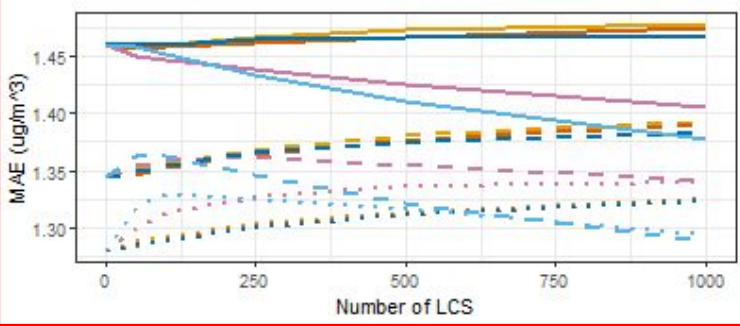
Distance to Nearest Monitor or Sensor



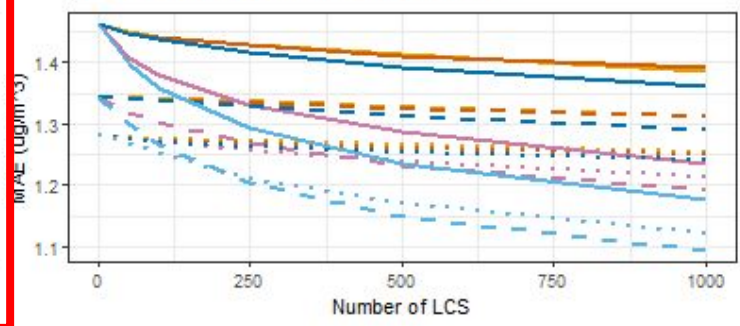
No Sensor Measurement Error: MAE



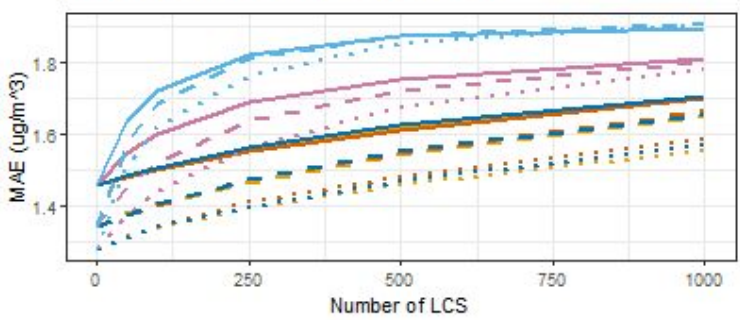
25% Non-differential Measurement Error: MAE



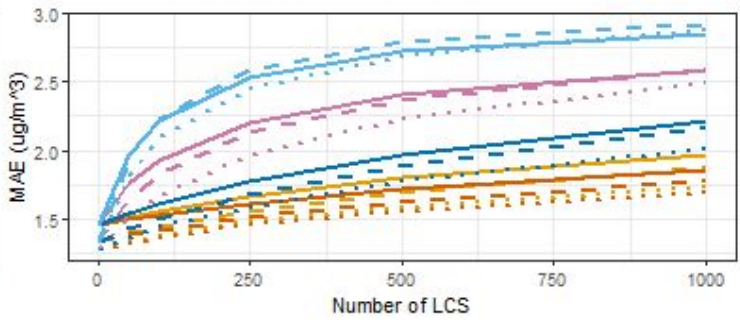
10% Differential Measurement Error: MAE



25% Differential Measurement Error: MAE

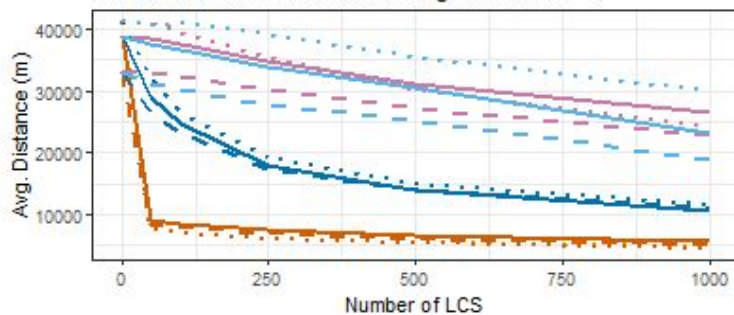


EPA Calibration Residual Draws: MAE

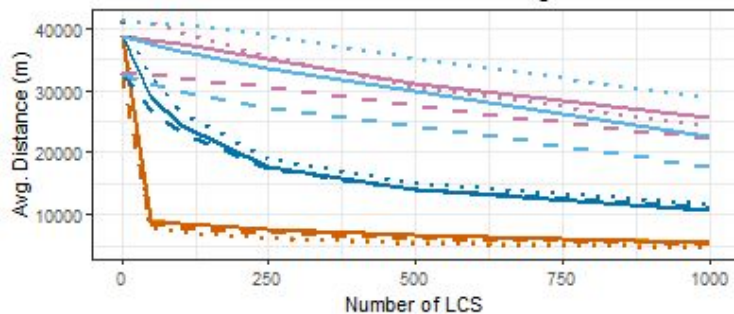




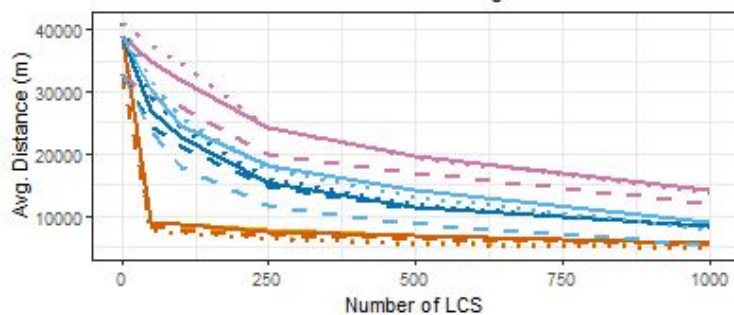
No Sensor ME: Distance Among Misclass. > 1



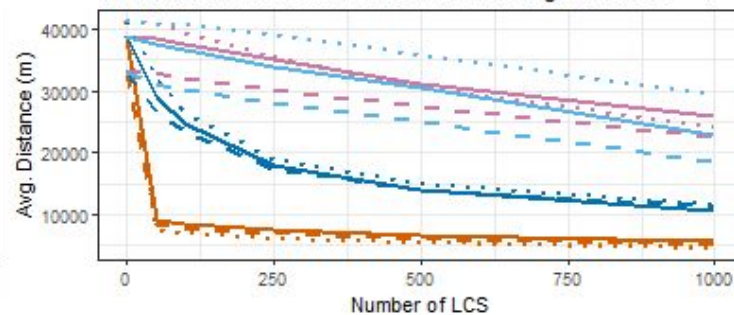
25% Non-differential ME: Distance Among Misclass. > 1



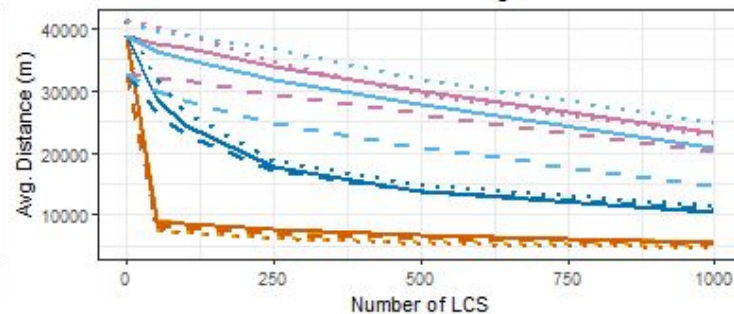
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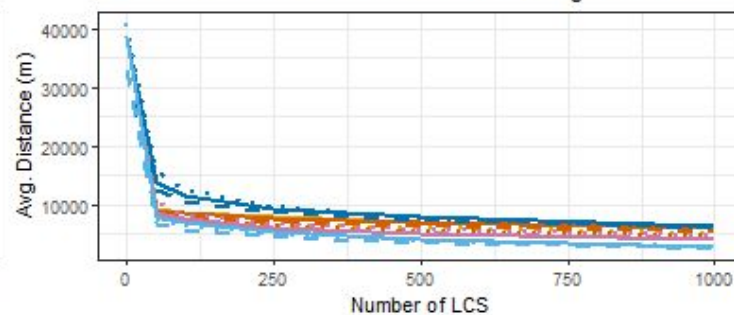
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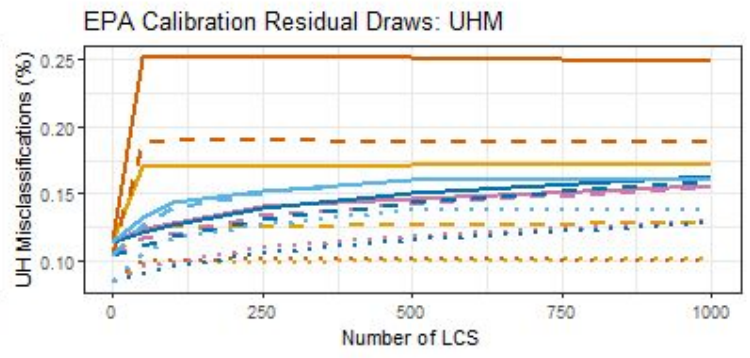
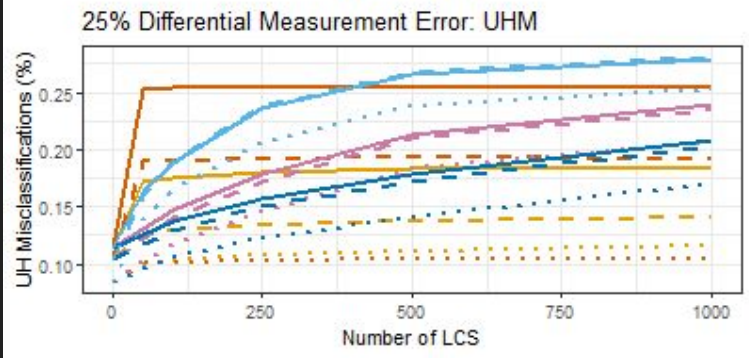
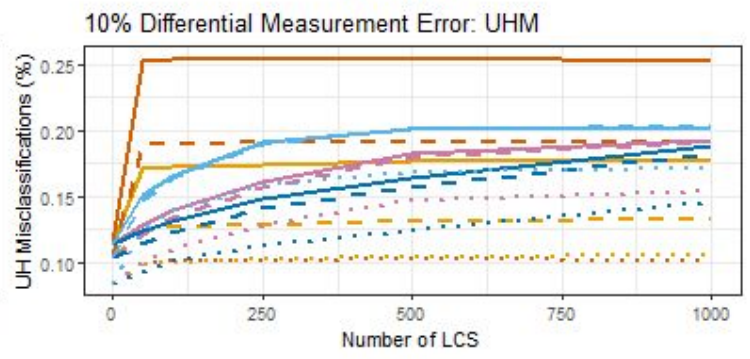
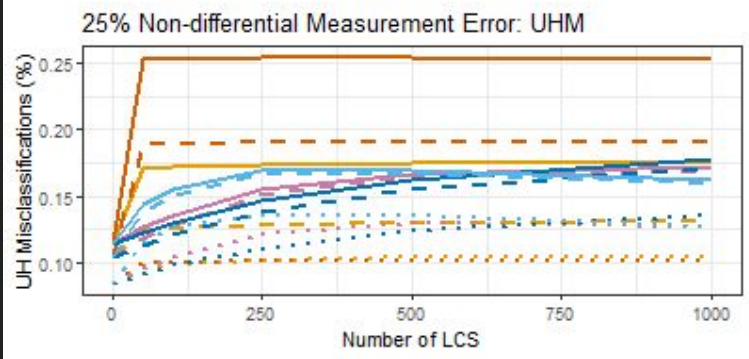
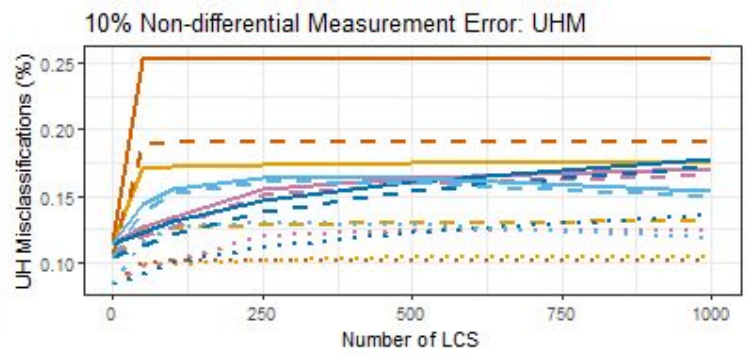
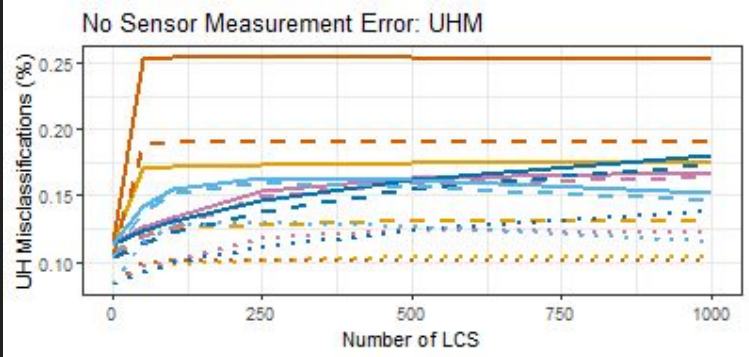
10% Differential ME: Distance Among Misclass. > 1



EPA Calibration Residuals: Distance Among Misclass. > 1



Weighted by Population Density



Overall Conclusions

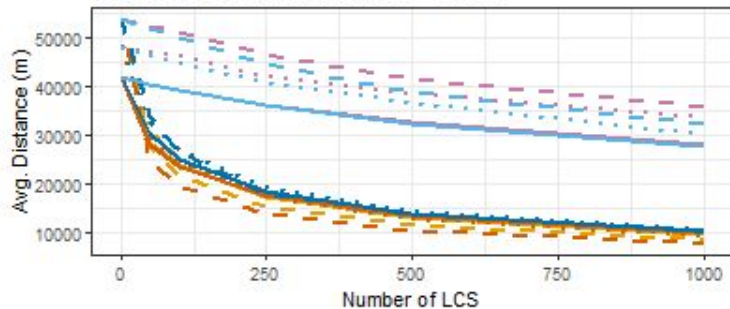
- The value of using LCS for real-time AQ reporting depends strongly on type and amount of sensor measurement error (ME)
 - 25% non-differential ME appears to be workable for some placement strategies, but not others
 - LCS corrections may need to be more localized / advanced than a national linear correction
- With low-to-moderate amounts of ME (depends on type): placing LCS at schools results in the greatest decrease in MAE, for all demographics considered, as the number of LCS increases
- Placing LCS in EJ hotspots may help the immediate community, but can cause issues when integrated into wider AQ reporting platforms
- Balancing policy priorities will be tricky (e.g. urban vs. rural)

Limitations / Future Directions

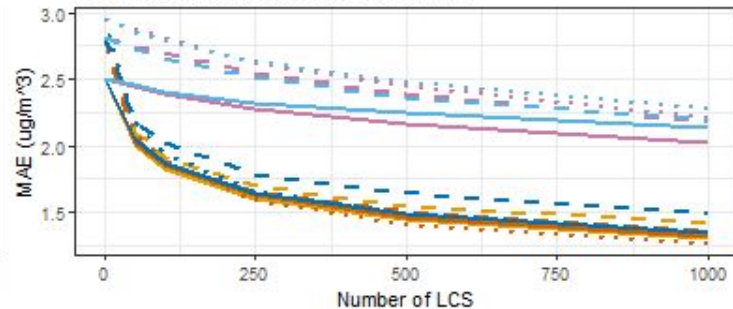
- Real-time AQ info is only a first step for public health
 - Behavioral insights...
- Looking beyond location of residence
 - Other sensors and activity patterns...
- Results from California in 2016 not necessarily generalizable
 - Value of adding LCS depends on coverage of existing reference monitors
- More advanced placement strategies, more targeted EJ placements
- Spatiotemporal resolution of truth surface...
 - Other air pollutants may exhibit even more local variability
- More advanced / regional calibrations or simulations for measurement error



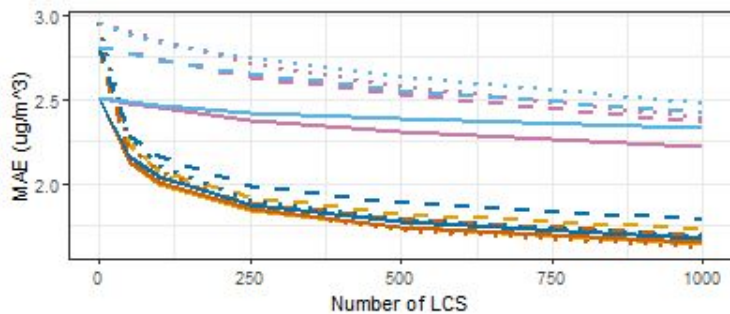
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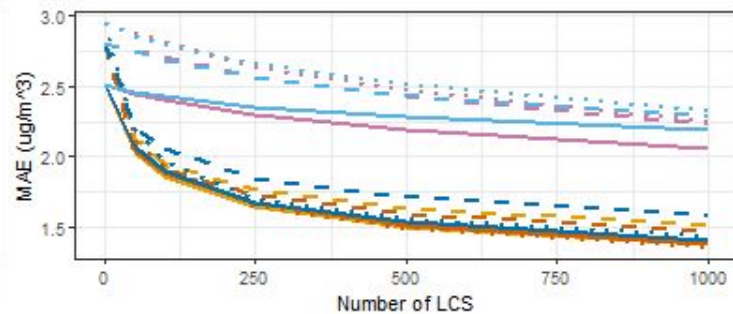
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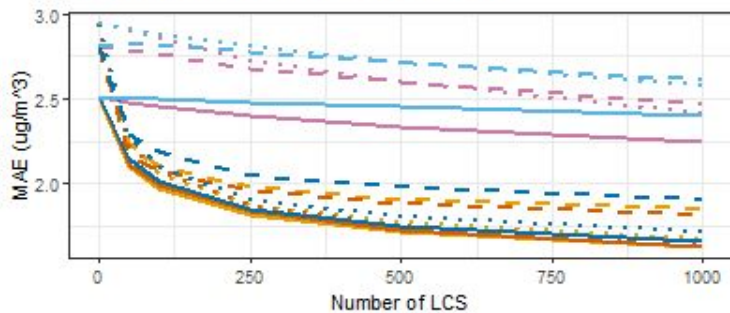
25% Non-differential Measurement Error: MAE



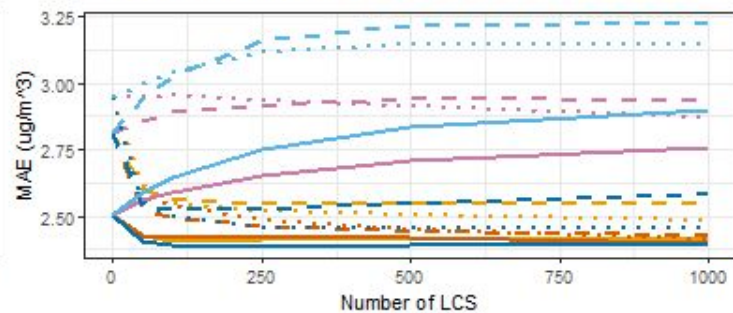
10% Differential Measurement Error: MAE



25% Differential Measurement Error: MAE



EPA Calibration Residual Draws: MAE



Unweighted by Population Density