Using Source-Resolved Aerial Surveys to Create Measurement-Based Methane Inventories

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Why We Need to Incorporate Measurements in (Methane) Inventories

- Multiple studies, in multiple jurisdictions, using multiple techniques consistently show current oil and gas sector methane inventories are underestimated
 - Airplane source/site-resolved (e.g., Tyner & Johnson, EST 2021; Chen et al., EST 2022)
 - Airplane mass balance (e.g., Johnson et al., EST 2017; Karion et al., EST 2015; Peischl et al., J. Geophys. Res., 2015, 2016; Alvarez et al., Science 2018)
 - Mobile (truck) measurements (e.g., Mackay et al., Sci. Reports, 2021)
 - Inverse modelling of ground station data (e.g., Chan et al., EST 2020; Miller et al., PNAS 2013)
 - Satellite measurements (e.g., Zhang et al., Sci. Adv., 2020)
 - Isotope measurements (e.g., Hmiel et al., Nature, 2020)
- Emissions must be expected to rapidly change!
 - <u>Emission factors</u> and inventories must be *continually updated* if we are to track reductions



Key Challenges: Why We Don't Generally Use Measurements in Inventories

- Inventories must preserve source / site / facility-type resolution
 - Bottom-up resolution is critical for regulatory and mitigation decisions
 - Simple-scaling of bottom-up totals to match some other total measurement misses a key part of the problem
- Unknown / unverified capabilities of available measurement technologies
 - What is the Probability of Detection (POD) of a source under general conditions?
 - What is the quantification uncertainty of a source/site under general conditions?

Protocols to incorporate measurements?

- What about unmeasured sources?
- How do determine required sample sizes with skewed distributions?
- Finite sample effects
- Etc.



Potential for Airborne Measurement Approaches



Scientific Aviation (Conley et al., AMT 2017)



Scientific Aviation (Johnson et al., EST 2017)



Bridger Photonics (Tyner & Johnson, EST 2021)



Kairos Aerospace (Chen et al., EST 2022)



AVIRIS-NG (Cusworth et al., Energy & Climate 2021)



Example Aerial Technology: Bridger Photonics Gas Mapping LiDAR

- Sites have one or more passes
- Flights with detected emissions are revisited in a subsequent day
- Source quantification for inventory development purposes requires interpretation of data from each pass





Source Attribution: Geo-locating Aerial Survey Imagery

 Combining satellite imagery, geolocated aerial photos, plot plans, & ground survey data to attribute







Source Attribution: Match Sources to Plot Plans

- Plot Plans provide a site schematic and equipment list
- Match Sources to Plot Plan





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High Resolution (~1m) Data Enables Attribution to Specific Sources

Key sources:

- a) Tanks
- b) Compressors
- c) Unlit flares

Tyner & Johnson, Environ. Sci. Technol, 2021 (doi: <u>10.1021/acs.est.1c01572</u>)



High Resolution (~1m) Data Enables Attribution to Specific Sources

- Other detected sources in BC:
 - d) Amine boiler unit
 - e) Dehydrator
 - f) Generator
 - g) Cooler
 - h) Etc.



(g)



Tyner & Johnson, Environ. Sci. Technol, 2021 (doi: <u>10.1021/acs.est.1c01572</u>)

Robust, Critical Evaluation of Measurement Technologies



- Fully- and semi-blinded controlled release testing
- B.M. Conrad, D.R. Tyner, M.R. Johnson (2022) Robust Probabilities of Detection and Quantification Uncertainty for Aerial Methane Detection: Examples for Three Airborne Technologies, *Remote Sensing of Environment* (under review: preprint)
- M.R. Johnson, D.R. Tyner, A.J. Szekeres (2021) Blinded evaluation of airborne methane source detection using Bridger Photonics LiDAR, *Remote Sensing of Environment*, 259:112418. (doi: <u>10.1016/j.rse.2021.112418</u>)





1. Fully-Blinded Controlled Release Testing of <u>Sensitivity Limits</u>

- Conducted under cover of parallel survey of oil and gas facilities
 - Airplane has no knowledge they are even being tested





M.R. Johnson, D.R. Tyner, A.J. Szekeres (2021) Blinded evaluation of airborne methane source detection using Bridger Photonics LiDAR, *Remote Sensing of Environment*, 259, 112418. (doi: <u>10.1016/j.rse.2021.112418</u>)



Continuous Probability of Detection (POD) Functions



Probability of detection any source Q for a given wind speed u and altitude h

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2021 Carleton-EERL National Methane Survey

- National-scale effort
 - ~8200 sites across 4 provinces



Research and Innovation Society





environment







- Similar, highly-skewed distributions across all provinces
 - Note these measured sources are ~80% of total methane (shown later)
- 95% of GML measured sources less than 30 kg/h
 - 2/3 of measure methane / ~81% of all methane
 - Not just about "super-emitters"
 - Mid-sized source key and will become more important as mitigation efforts succeed





Measured distributions represent
 ~80% of total methane (shown later)





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- At 13 kg/h sensitivity can see:
 - ~18% of these sources /
 62% of this methane
 - ~50% (0.62*0.8) of all methane





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- At 27 kg/h sensitivity can see:
 - ~7% of these sources / 40% of this methane
 - ~32% (0.4*0.8) of all methane





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- At 200 kg/h sensitivity can see:





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- At 27 kg/h sensitivity can see:
 - ~7% of these sources / 40% of this methane
 - ~32% (0.4*0.8) of all methane
- At 200 kg/h sensitivity can see:
 - <1% of these sources / 5% of this methane
 - ~4% (0.05*0.8) of all methane
- Critical to understand sensitivity limits when Source Em incorporating measurements from different technologies



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2. Semi-Blinded Controlled Release Testing of *Quantification Accuracy*

- Semi-blinded (collaborative) controlled release tests
 - Plane flies laps over controlled release points and quantifies
 - Actual release rates are not shared with plane





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A Measurement-Based Methane Inventory for British Columbia (BC), Canada

- Demonstrate feasibility of measurement-based methane inventories using aerial measurements
- Key enabling pieces:
 - Technology with sufficient sensitivity to capture majority of sources
 - Detailed probability of detection (POD) functions in varying conditions
 - Detailed uncertainty model for technology
 - Bottom-up data for unmeasured sources





A Measurement-Based Methane Inventory for British Columbia (BC), Canada



- Survey includes:
 - 59% of all active facilities
 - 8% of all active wells





Protocol to Create a *"Hybrid"* Bottom-Up *Measurement-*Based Inventory



Johnson et al., (2022) to be submitted



Very powerful approach to quantify, analyze, and *minimize* uncertainty



Measurement-Based Methane Inventory for BC







Stark Differences in Sources Among Provinces



Saskatchewan

British Columbia





Rapid Changes as Sources Evolve and Regulations Take Effect





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Conclusions

- Traditional bottom-up, emission factor based inventories face many challenges
 - Persistent underestimation
 - Rapid evolution of sources and source distributions as regulations take hold
- New aerial technologies are a revolution in possibilities, but:
 - Robust, independently-proven probabilistic sensitivity and uncertainty models are critical
 - Not all technologies are interchangeable and not all are sufficient for creating source- and site-resolved inventories
- Measurement-based methane inventories are possible now using careful application of statistical methods using current technologies
 - Province of BC Canada looking to transition to measurement-based inventories this year!



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