# **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!**
	- Emission factors 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** (Johnson et al., EST 2017)



**Scientific Aviation** (Conley et al., AMT 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**

6

**Combining satellite imagery, geo**located 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: [10.1021/acs.est.1c01572\)](https://doi.org/doi:%2010.1021/acs.est.1c01572))

# **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.





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#### **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](https://doi.org/10.31223/X5S05F))
- 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: [10.1016/j.rse.2021.112418\)](https://doi.org/10.1016/j.rse.2021.112418)





# **1. Fully-Blinded Controlled Release Testing of** *Sensitivity Limits*

- 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: [10.1016/j.rse.2021.112418\)](https://www.sciencedirect.com/science/article/pii/S003442572100136X?via%3Dihub)



#### **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









**NSERC** 

CRSNG







- 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





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



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16

- **Neasured distributions represent** ~80% of total methane *(shown later)*
- At 13 kg/h sensitivity can see:
	- ~18% of these sources / 62% of this methane
	- $\sim$ 50% (0.62 $*$ 0.8) of all methane





- **Neasured distributions represent** ~80% of total methane *(shown later)*
- At 13 kg/h sensitivity can see:
	- $\sim$ 18% of these sources / 62% of this methane
	- $\sim$ 50% (0.62 $*$ 0.8) of all methane
- At 27 kg/h sensitivity can see:



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- **Neasured distributions represent** ~80% of total methane *(shown later)*
- At 13 kg/h sensitivity can see:
	- ~18% of these sources / 62% of this methane
	- $\sim$ 50% (0.62 $*$ 0.8) of all methane
- At 27 kg/h sensitivity can see:
	- ~7% of these sources / 40% of this methane
	- $\sim$ 32% (0.4\*0.8) of all methane





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- At 13 kg/h sensitivity can see:
	- ~18% of these sources / 62% of this methane
	- $\sim$ 50% (0.62\*0.8) of all methane
- At 27 kg/h sensitivity can see:
	- ~7% of these sources / 40% of this methane
	- $\sim$ 32% (0.4\*0.8) of all methane
- At 200 kg/h sensitivity can see:





- **Neasured distributions represent** ~80% of total methane *(shown later)*
- At 13 kg/h sensitivity can see:
	- ~18% of these sources / 62% of this methane
	- $~$  ~50% (0.62 $*$ 0.8) of all methane
- At 27 kg/h sensitivity can see:
	- ~7% of these sources / 40% of this methane
	- $\approx$  32% (0.4\*0.8) of all methane
- At 200 kg/h sensitivity can see:
	- <1% of these sources / 5% of this methane
	- $\sim$ 4% (0.05 $*$ 0.8) of all methane
- *Critical to understand sensitivity limits when incorporating measurements from different technologies*



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21

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

22

- **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** **a) Measured Sources b) Unmeasured Sources** *Joint PDF Aerial survey data at flight pass-level Additional emission (wind speed & altitude) factor and site-level count data Randomized (e.g., Pneumatics) POD functions High-sensitivity Pass-level source rates measurement data*  $POD(Q, u, h)$ (Miss) *(e.g., Prior OGI*   $\ddot{ }$  $h = ?$ *study) Average rate Total Rate (each source) Probability of successful detection for unmeasured sources*  $\sum_{sources}$   $\longrightarrow$   $\frac{1}{x}$  $\ddot{ }$ *Conrad et al. (2022) [in review] Total missed (by site) Pull B<sub>MC</sub>* Draws *Site-level emission factor Pop.*  $\boldsymbol{N}$  $\lambda$ . ×  $\boldsymbol{\mathcal{X}}$ *Measured Inventory For each MC draw* sources ( *bootstrap samples; times) Sam.*  $\overline{n}$  $N \begin{array}{ccc} & & \end{array}$ *Pop. Estimated inventory*  **c) Total Inventory** *for unmeasured sources*  $B_{MC}$  × BSS estimates of emissions *inventory for measured sources* + **Legend** Bridger GML characteristics and assorted data Monte Carlo analysis of quantification uncertainty and detection sensitivity Population scaling, including bootstrap analysis of sample size effects Estimated partial inventory; measured and unmeasured sources Estimated total inventory

*Johnson et al., (2022) to be submitted*



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

#### **2021 Measurement-Based Methane Inventory for BC**







#### **Stark Differences in Sources Among Provinces**







#### **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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Environnement et Changement climatique Canada



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