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# Localization and Quantification of Trace-gas Fugitive Emissions Using a Portable Optical Spectrometer

E. J. Zhang<sup>1</sup>, C. C. Teng<sup>2</sup>, T. G. van Kessel<sup>1</sup>, L. Klein<sup>1</sup>, R. Muralidhar<sup>1</sup>, C. Xiong<sup>1</sup>, Y. Martin<sup>1</sup>, J. S. Orcutt<sup>1</sup>, M. Khater<sup>1</sup>, L. Schares<sup>1</sup>, T. Barwicz<sup>1</sup>, N. Marchack<sup>1</sup>, S. Kamlapurkar<sup>1</sup>, S. Engelmann<sup>1</sup>, G. Wysocki<sup>2</sup>, N. Sosa<sup>1</sup>, W. M. J. Green<sup>1</sup>

<sup>1</sup>IBM Thomas J. Watson Research Center, Yorktown Heights NY 10598

<sup>2</sup>Dept. Of Electrical Engineering, Princeton University, Princeton NJ 08540

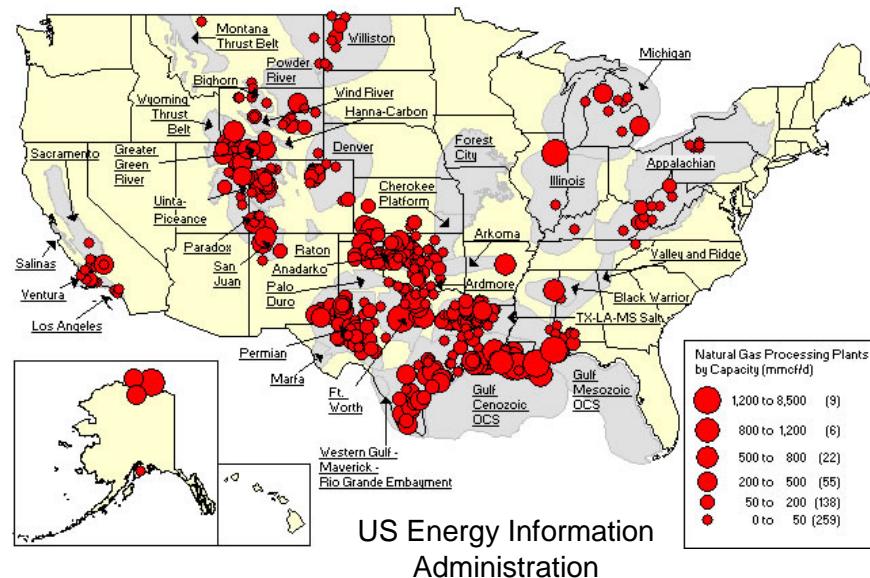
- **Introduction**
  - Motivation:  $\text{CH}_4$  fugitive emissions monitoring
- **Design and characterization of a portable TDLAS sensor**
  - TDLAS sensor construction
  - Sensitivity analysis and chamber response time
- **Field deployment at METEC CSU**
  - Accuracy benchmark vs. MOX VOC sensors
  - AOA localization of  $\text{CH}_4$  fugitive emissions
  - Source magnitude estimation (Gaussian plume / ML algorithms)
- **Toward a next generation integrated photonic chip sensor**
  - Initial results: on-chip evanescent field waveguide TDLAS
- **Concluding remarks**

# Motivation: $\text{CH}_4$ fugitive emissions monitoring



- > 500,000 active oil/gas wells in USA
- $570 \times 10^9 \text{ ft}^3$  of  $\text{CH}_4$  leakage in 2009, (59 % leaks during production phase)
- ~ 30% anthropogenic  $\text{CH}_4$  emissions
- Radiative forcing of  $\text{CH}_4$  is **37x greater than  $\text{CO}_2$**

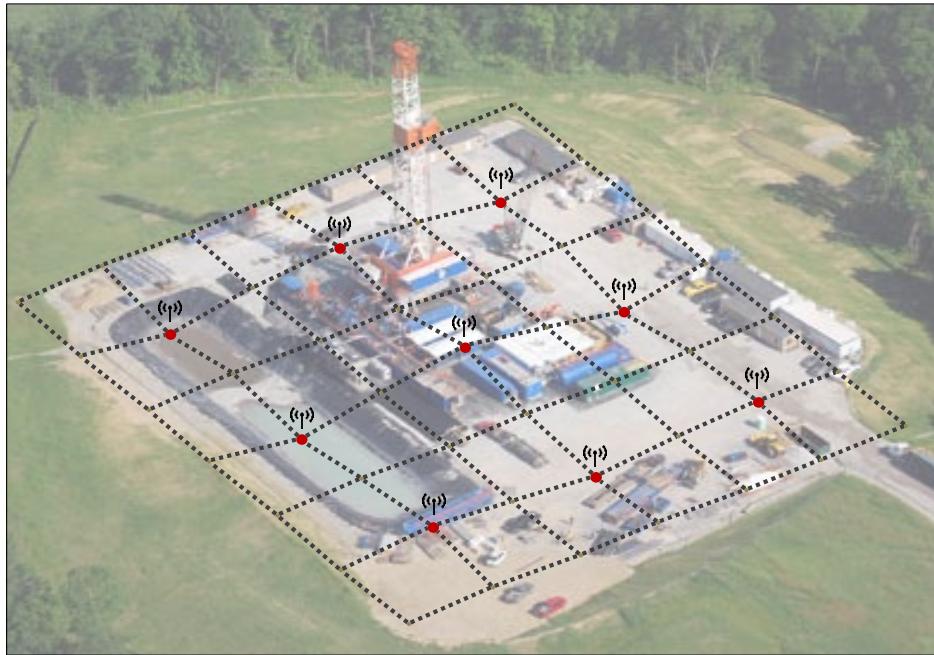
Alvarez et. al., "Greater focus needed on methane leakage from natural gas infrastructure," Proc. Nat. Acad. Sci., 109 (17), pp. 6435-6440, (2012).



- $\text{CH}_4$  leakage rate on oil/gas well pad is **2-10% of total production!**

Cost-effective sensor network for **localization** and precise **quantification** (ppmv-level) of  $\text{CH}_4$  on oil and gas production well-pads

# Use-case for innovative sensor networks



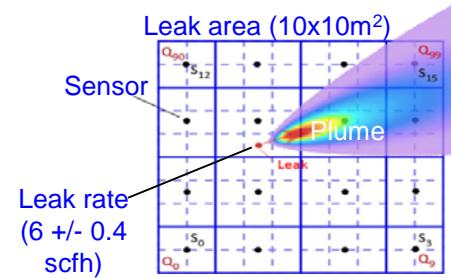
## An Intelligent Multi-Modal Methane Measurement System (AIMS)

Real-time sensor mesh network  
(IBM MMT System)

Aggregate/push to Bluemix  
cloud (MQTT protocol)

### Physical analytics:

- source inference via inversion
- plume dispersion models
- machine-learned model blending



### Technological driver: ARPA-E MONITOR

- Cost-effective sensor network for continuous  $\text{CH}_4$  leak quantification, localization, and repair
- No viable technology today: Alignment of performance with required cost point poses significant challenge

### Opportunity driver: Application of physical analytics/IoT solutions to

- Significantly reduce fugitive  $\text{CH}_4$  emissions across the oil and gas industry
- Improve production efficiency, safety, and compliance with emissions regulations
- Harness the full potential of natural gas as a clean fuel

# Use-case for innovative sensor networks



Metal-oxide (MOX):



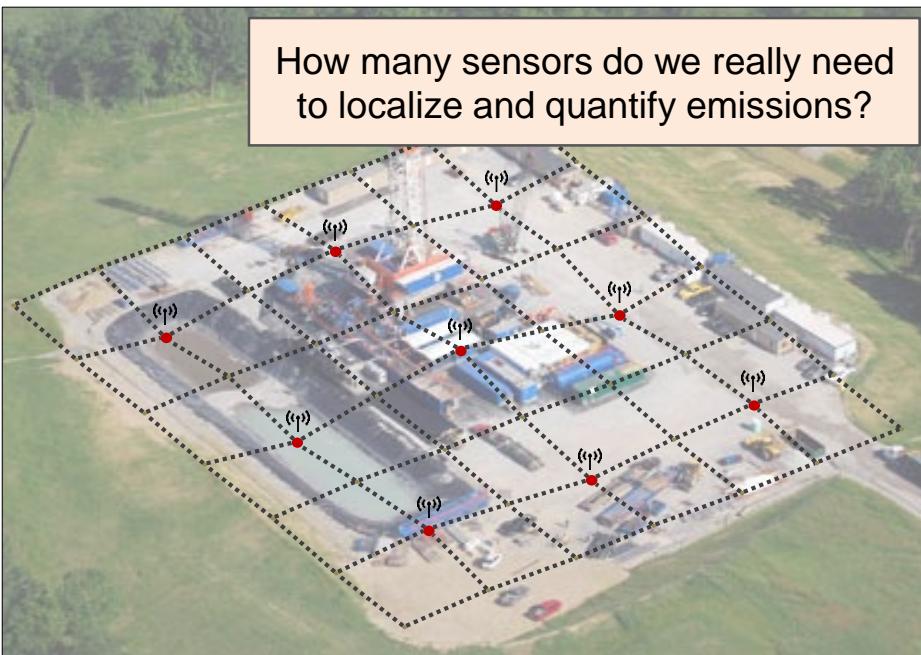
Figaro  
TGS2611

- Chemi-resistive sensor
- $\text{CH}_4$  adsorption (hot  $\text{SnO}_2$ )
- ppmv-level sensitivity
- **~10 USD / sensor**
- Susceptible to other VOCs
- Ruggedized enclosure

Open-path TDLAS:



- Conventional NIR TDLAS
- < 5k USD / sensor
- **Intermediate SWaP-C compromise + benchmark IOS-TDLAS performance**



## An Intelligent Multi-Modal Methane Measurement System (AIMS)

- **Robust sensors** to withstand harsh environments
- Low size, weight, power, and cost (SWaP-C)
- Species **selectivity + sensitivity** (DL < 10 ppmv)
- Lightweight data packaging on each WSN node for wireless connectivity and **real-time analytics**

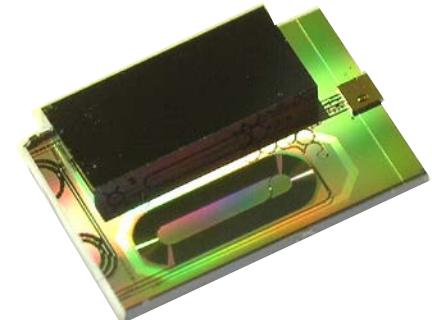
Cavity-ringdown (CRDS):



Picarro G2308

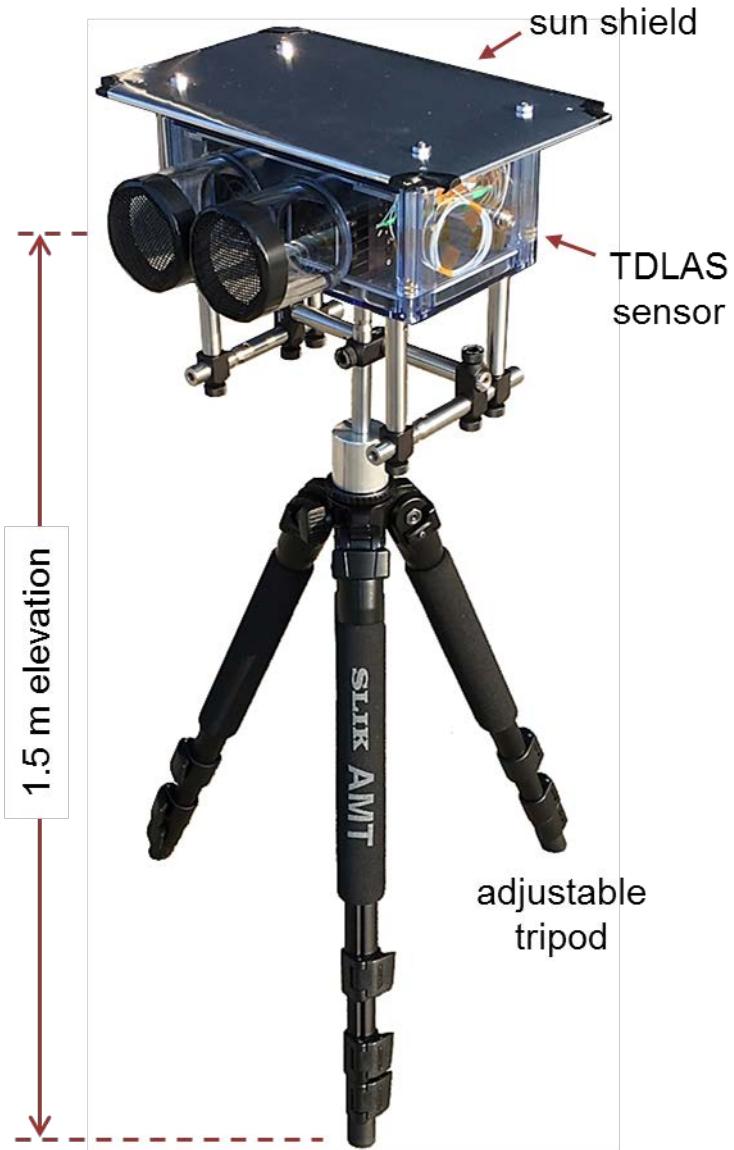
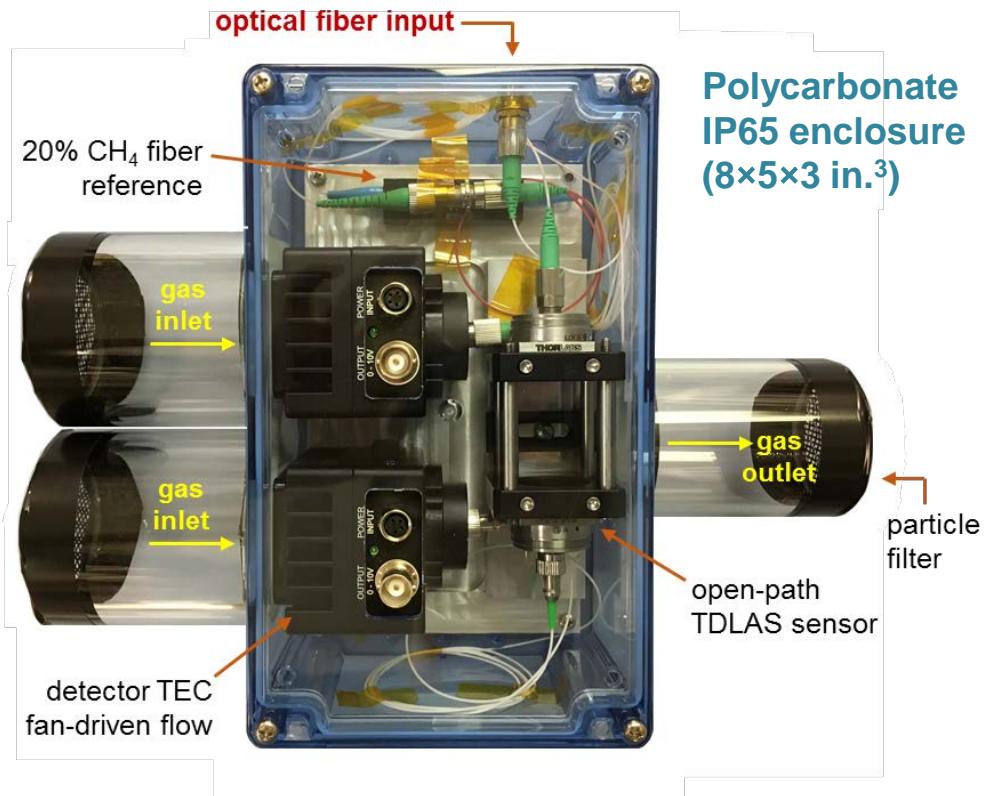
- Cavity-enhanced absorption
- < 10 ppbv sensitivity
- Dynamic range: (200 ppmv)
- 60 lbs, power: ~250 W
- Requires vacuum pump
- **~50k USD / sensor**

Integrated chip sensor:

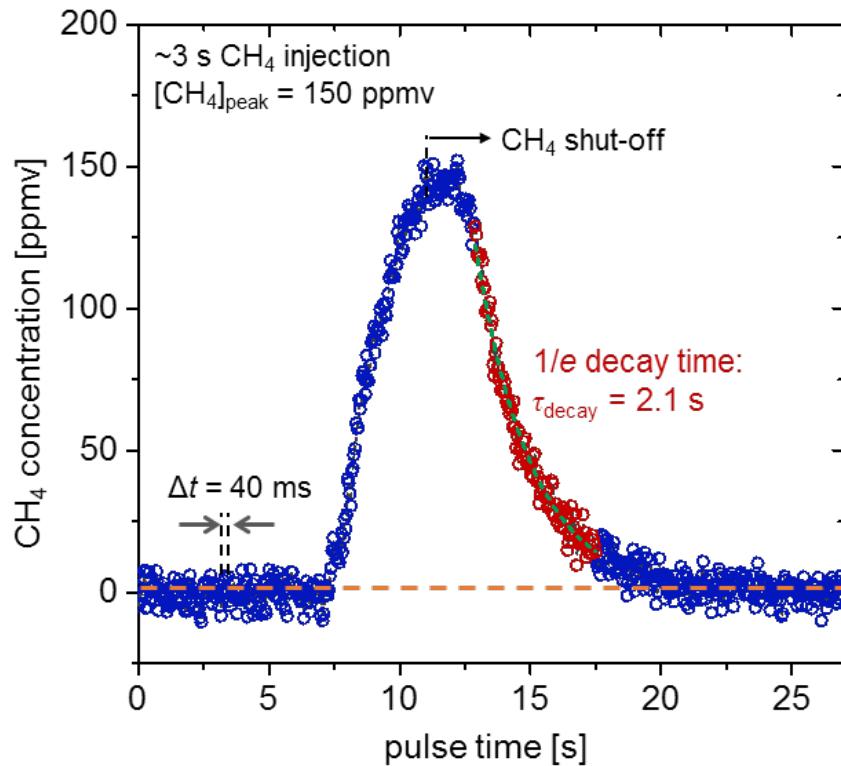
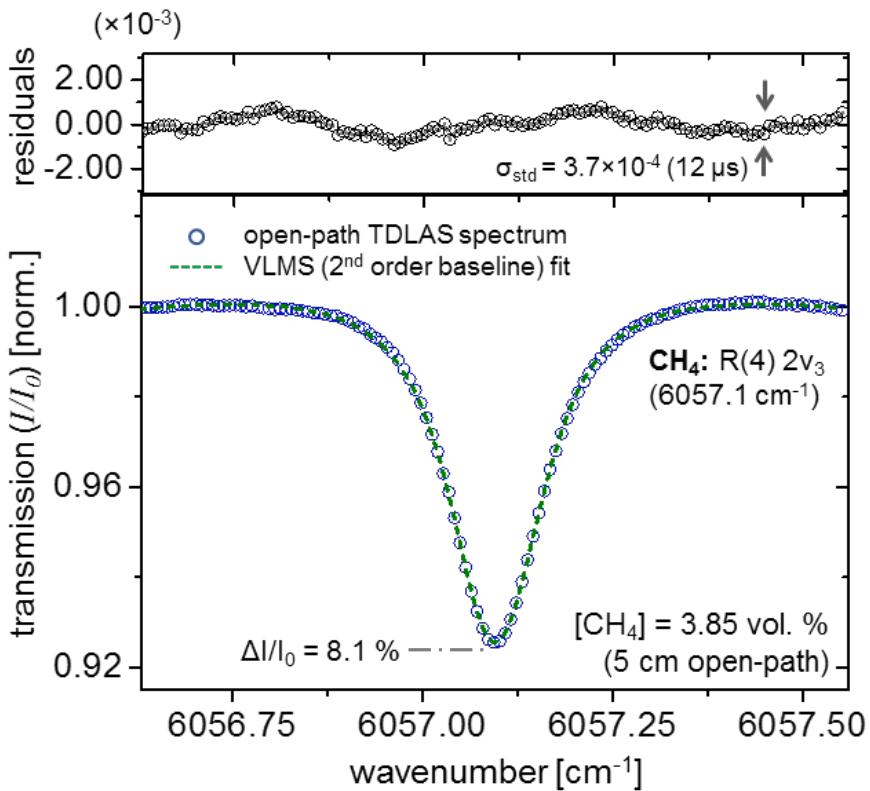


- Integrated optical sensor TDLAS (IOS-TDLAS)
- **Low-volume cost (< 250 USD)**
- Sensitivity:  $6.3 \text{ ppmv}\cdot\text{Hz}^{-1/2}$

# Portable TDLAS sensor construction



- 5 cm open-path fiber-coupled cell (TDLAS)
- Parallel (3 cm) 20 vol. % CH<sub>4</sub> ( $\lambda$  reference)
- Dual InGaAsP photodetector **TEC fans** for **gas exchange** in chamber

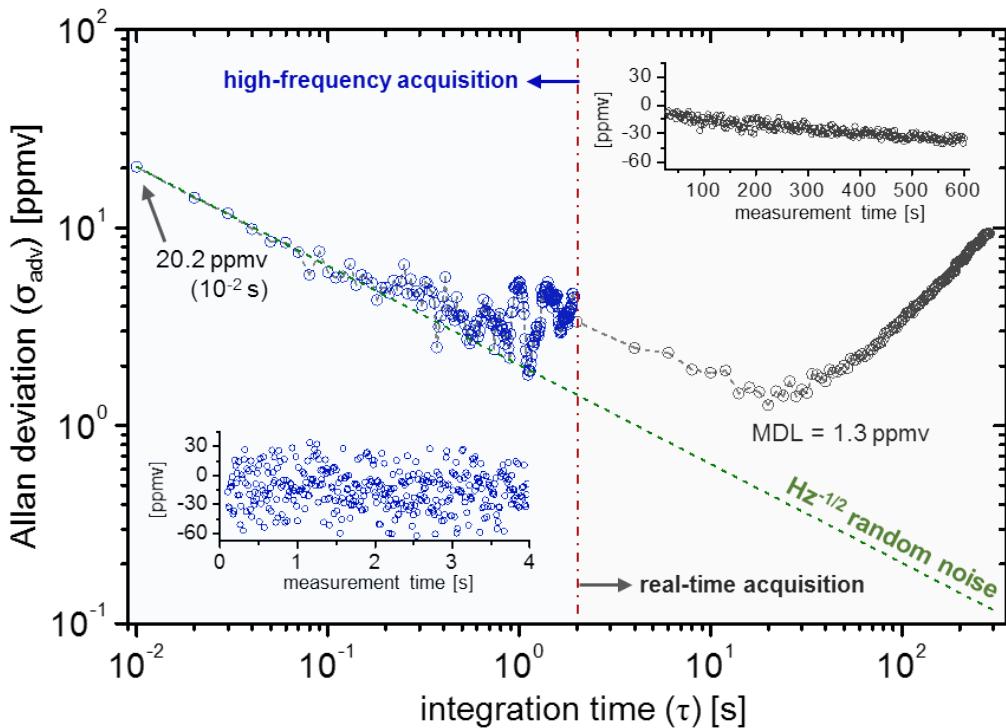


## 2v<sub>3</sub> CH<sub>4</sub> R(4) spectra at 6057.1 cm<sup>-1</sup> (1651 nm)

- Alignment optimized + AR lens for fringe reduction
- Voigt nonlinear regression with 2<sup>nd</sup> order baseline
- 2 ms spectral acquisition (500 Hz laser ramp)
- Residual deviation:  $\alpha_{\text{min}} = 6.5 \times 10^{-7} \text{ cm}^{-1} \cdot \text{Hz}^{-1/2}$

## Chamber response time

- $\sim 3 \text{ s}$  release (1.0 vol. %)  $\rightarrow 150 \text{ ppmv} \text{ CH}_4$
- 40 ms measurement resolution
- 90% to 10%  $\rightarrow 2.1 \text{ s}$  (1/e decay time)
- Typical CH<sub>4</sub> peak:  $\sim 20 \text{ s}$  (field measurements)

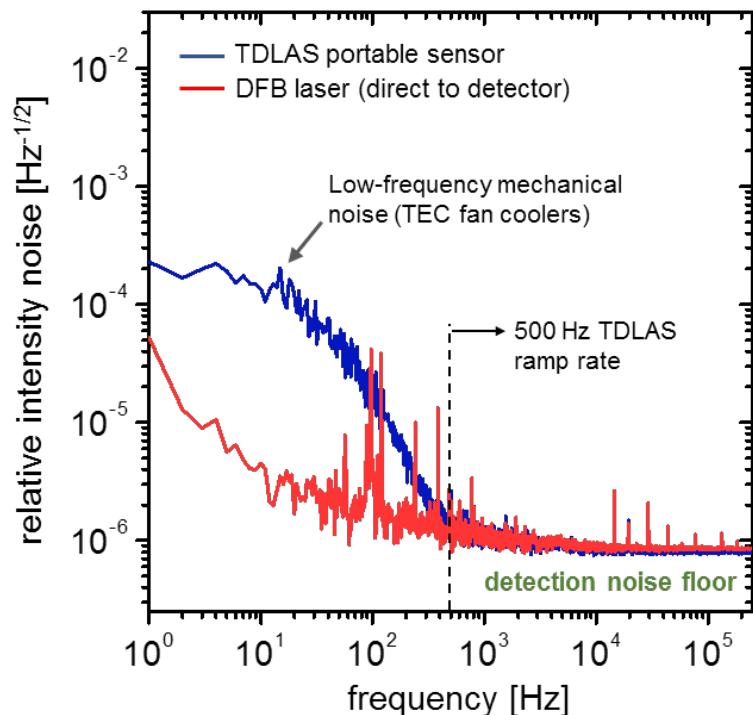


## TDLAS sensor Allan-deviation and RIN analysis:

- High-frequency (2 ms) and real-time (2 s) operation
- Low-frequency noise (< 500 Hz) due to fan-cooled TECs for InGaAsP photodetectors
- Laser  $f_{ramp} = 500$  Hz to avoid low-frequency noise
- NEP =  $7.2 \times 10^{-10} \text{ W} \cdot \text{Hz}^{-1/2}$  (100 kHz) → sensor operates at 2.2× detection noise floor

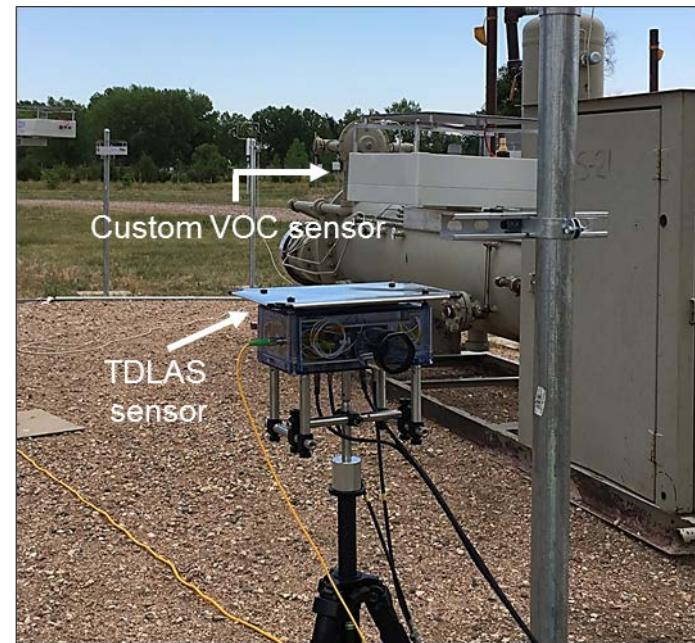
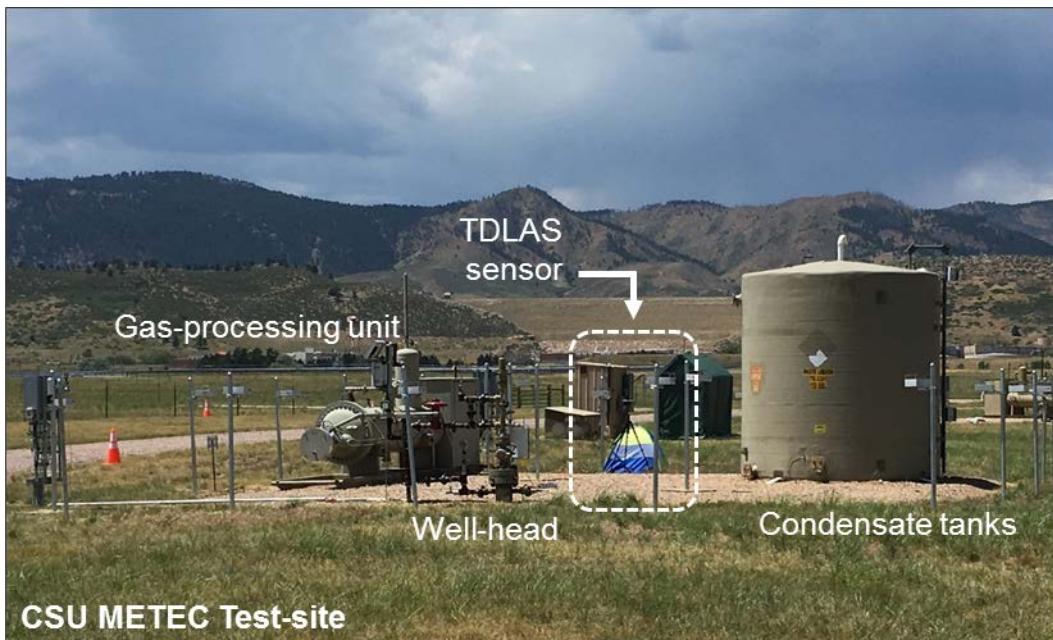
**Detection sensitivity:**  
 $\sigma_{adv} = 2.0 \text{ ppmv} \cdot \text{Hz}^{-1/2}$

**Min. fractional absorption:**  
 $(\alpha L)_{min} = 4.5 \times 10^{-6} \text{ Hz}^{-1/2}$



## Methane Emissions Technology Evaluation Center (METEC) field deployment:

- 5-day deployment (July 17-21, 2017): 16.6 hours CH<sub>4</sub> data
- **4.4 hours** control, 12.2 hours blind (**1.9 hours** CH<sub>4</sub> data)
- Blind measurements: TDLAS sensor not always downwind
- TDLAS sensor co-located with a customized VOC MOx sensor for accuracy benchmark

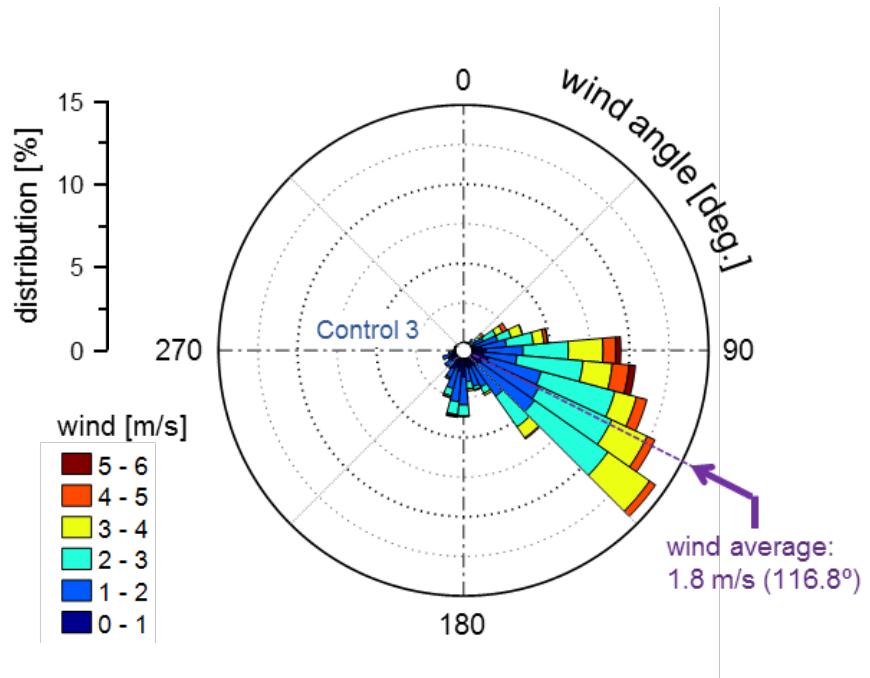
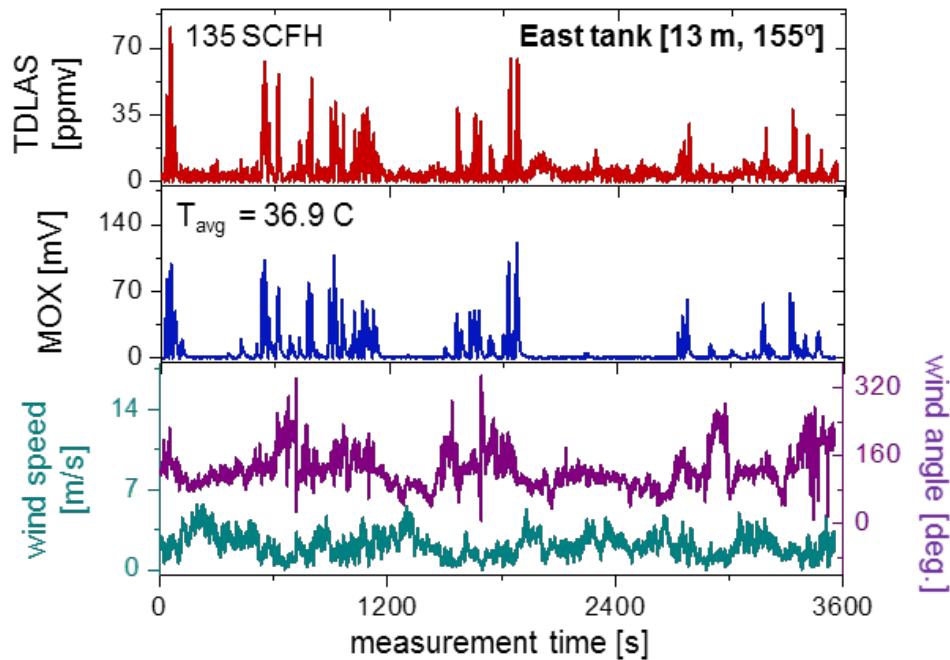


## Experimental Configuration:

- Control: 68 SCFH – 135 SCFH
- Blind: 0 SCFH – 40 SCFH
- Concurrent anemometer measurement

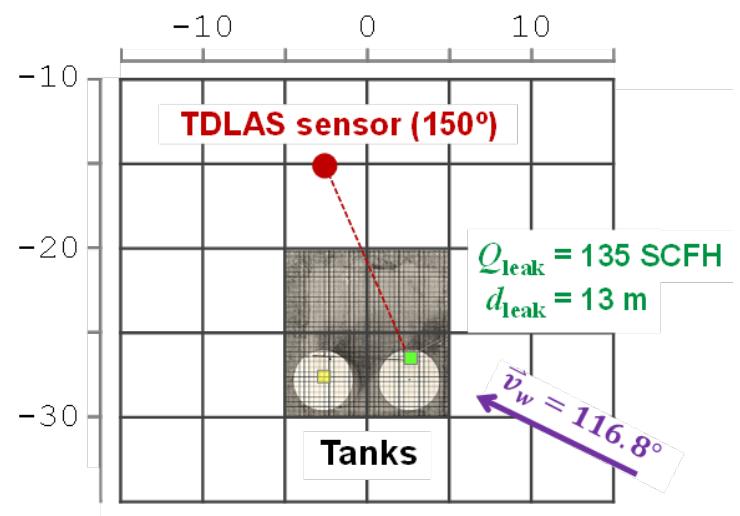
**Goal: single-sensor source AOA and magnitude estimation**

# Control experiment: tank $\text{CH}_4$ release



## Concurrent TDLAS / MOX / anemometer data:

- Downwind placement of TDLAS sensor ( $150^\circ$  LoS)
- 1 hour measurement, 2 s time-resolution (TDLAS/MOX)
- East tank thief hatch (13 m, 135 SCFH) control  $\text{CH}_4$  leak
- Good visual correspondence between TDLAS/MOX units
- Real-time acquired temperature (MOX unit thermistor)

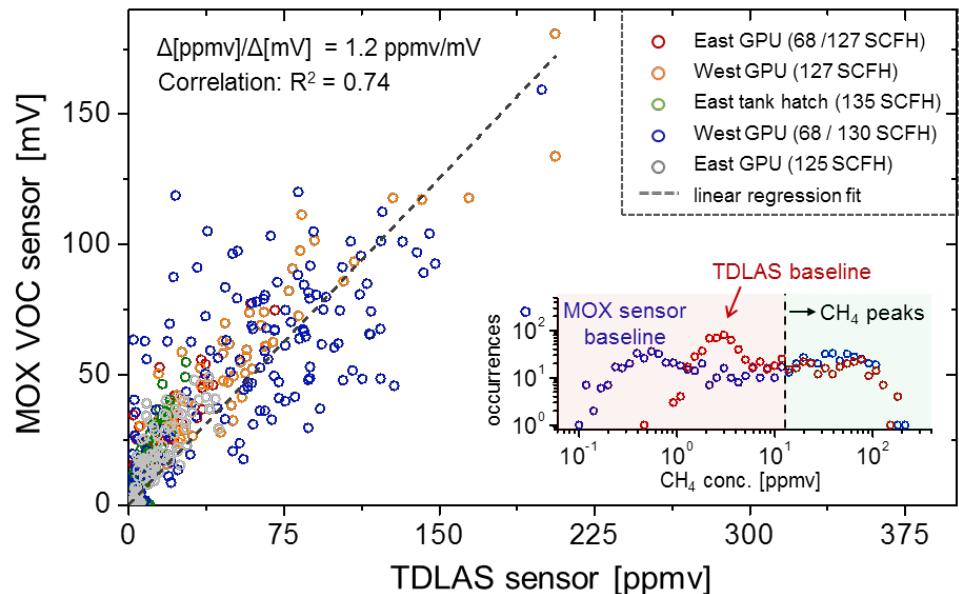
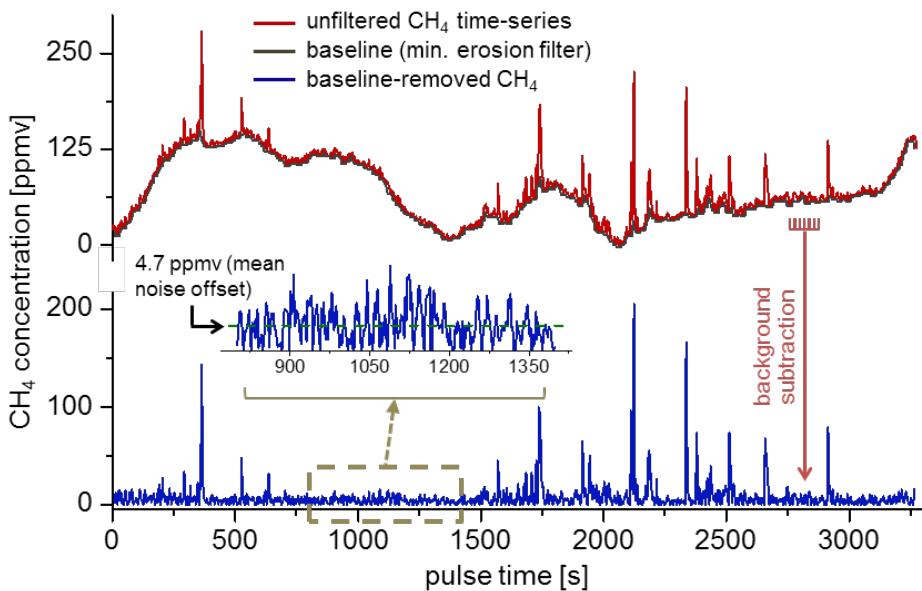


## Real-time $CH_4$ retrieval:

- 500 Hz line-scan rate, VLMS (2 s resolution)
- Account for temperature dependent:
  - i. transition line-strength  $S_{\eta \rightarrow \eta'}(T)$
  - ii. air-broadening coefficient  $\delta\nu_L(p, T)$
- Post-analysis baseline erosion:

$$[CH_4](t_i) \rightarrow [CH_4](t_i) - \min_{t_i \in B_i} [CH_4](t_i)$$

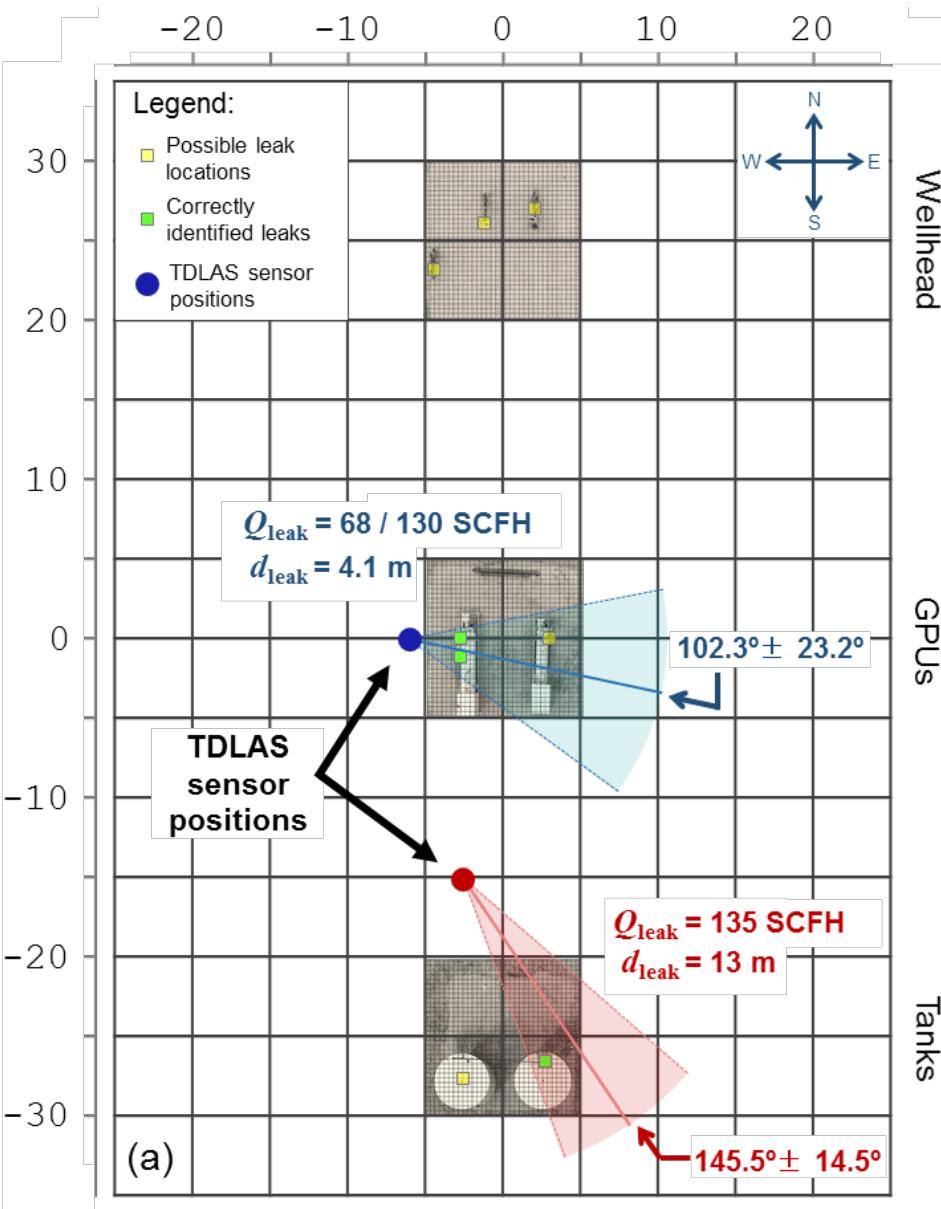
$$B_i = \left\{ \lfloor t_i / T_B \rfloor \cdot T_B \leq t < (\lfloor t_i / T_B \rfloor + 1) \cdot T_B \right\}$$



## TDLAS / MOX sensor comparison:

- 5 Control experiments (4.4 hours), spans leak rates 68 – 135 SCFH
- TDLAS sensor placed downwind from leak
- Good  $R^2$  correlation (0.74); non-ideal orientation
- Erosion noise floor peak: ~ 3 ppmv (consistent with Allan-deviation sensitivity)
- **TDLAS/MOX sensor agrees for  $[CH_4] > 12$  ppmv**

# Determining emission source location

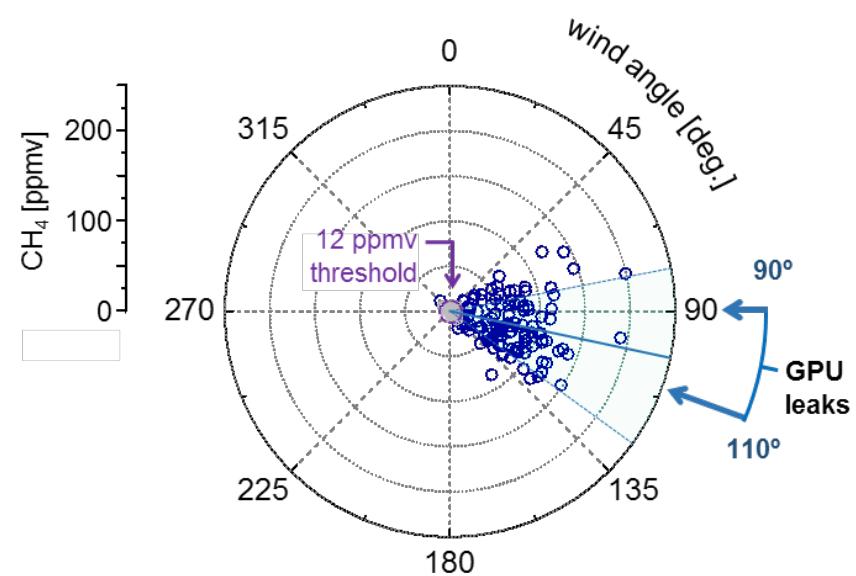


## Leak angle-of-arrival (AOA) localization:

- Bin data in 20 s intervals; 12 ppmv threshold
- Correlate wind angles with  $\text{CH}_4$  concentrations

$$\langle \alpha_{AOA} \rangle = \frac{\sum_i [CH_4](t_i) \cdot \varphi_i}{\sum_i [CH_4](t_i)}, \quad \varphi_i = \tan^{-1} \left( \frac{v_x}{v_y} \right)$$

- Single-sensor can only determine AOA, need  $\geq 2$  sensors for true localization



Experiment	Leak duration	Leak component	Flow rate (SCFH)	Leak location [distance, angle]	Average wind-velocity ( $v_w$ )	Leak AOA ( $\alpha_{AOA}$ ) $\pm \delta\alpha_{AOA}$
Control 1	3554 s	East GPU (Pad 3)	68/127	9.0 m (90°)	1.96 m/s (137.2°)	102.3° $\pm$ 22.0°
Control 2	1757 s	West GPU (Pad 3)	127	4.1 m (90°, 110°)	2.60 m/s (125.6°)	109.9° $\pm$ 24.1°
Control 3	3553 s	East tank (Pad 3)	135	13.0 m (155°)	1.77 m/s (116.8°)	145.5° $\pm$ 14.5°
Control 4	3552 s	West GPU (Pad 3)	68/130	4.1 m (90°, 110°)	1.58 m/s (97.0°)	102.3° $\pm$ 23.2°
Control 5	3459 s	East GPU (Pad 3)	125	9.0 m (90°)	2.80 m/s (61.6°)	82.6° $\pm$ 13.2°
Blind 1	3442 s	Tank (Pad 1)	36.1	3.4 m (60°)	0.58 m/s (68.9°)	76.3° $\pm$ 53.6°
Blind 2	3470 s	Wellhead (Pad 2)	4.4	6.9 m (125°)	1.41 m/s (124.8°)	119.3° $\pm$ 25.2°

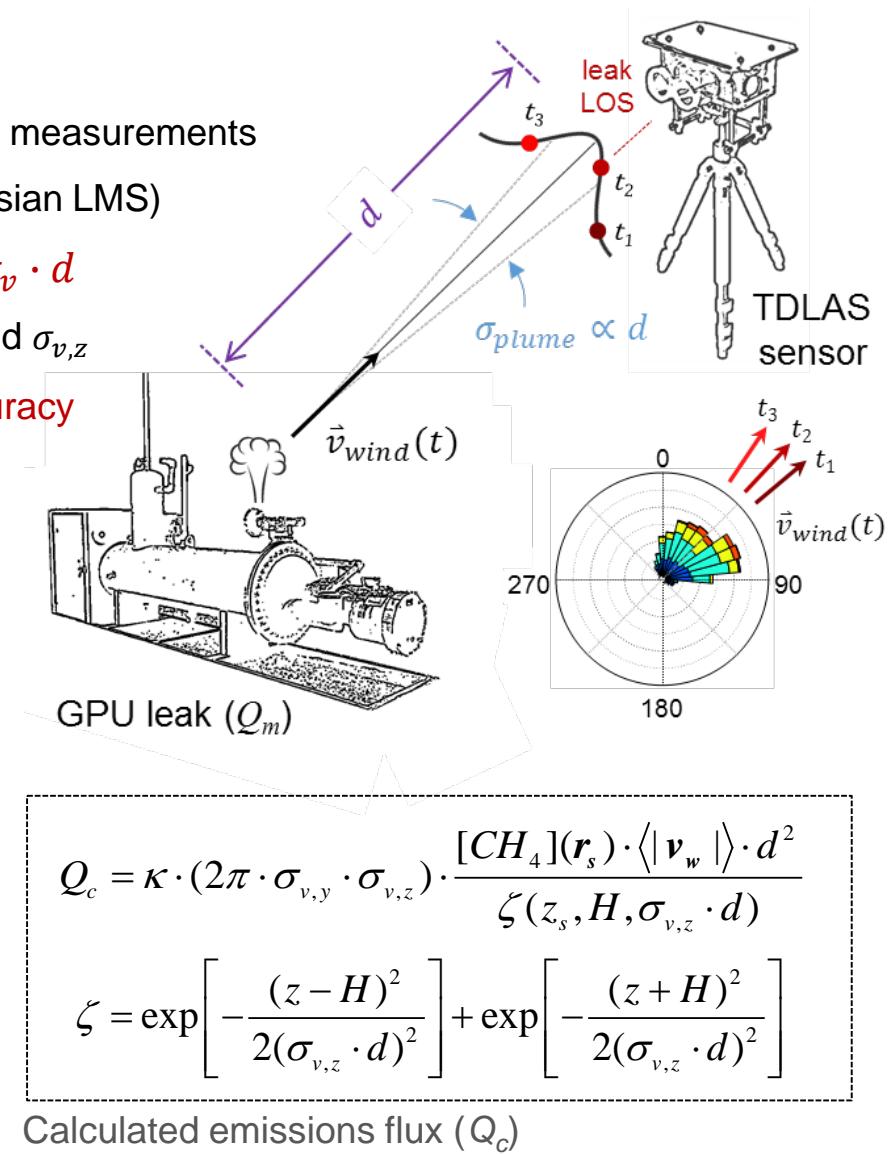
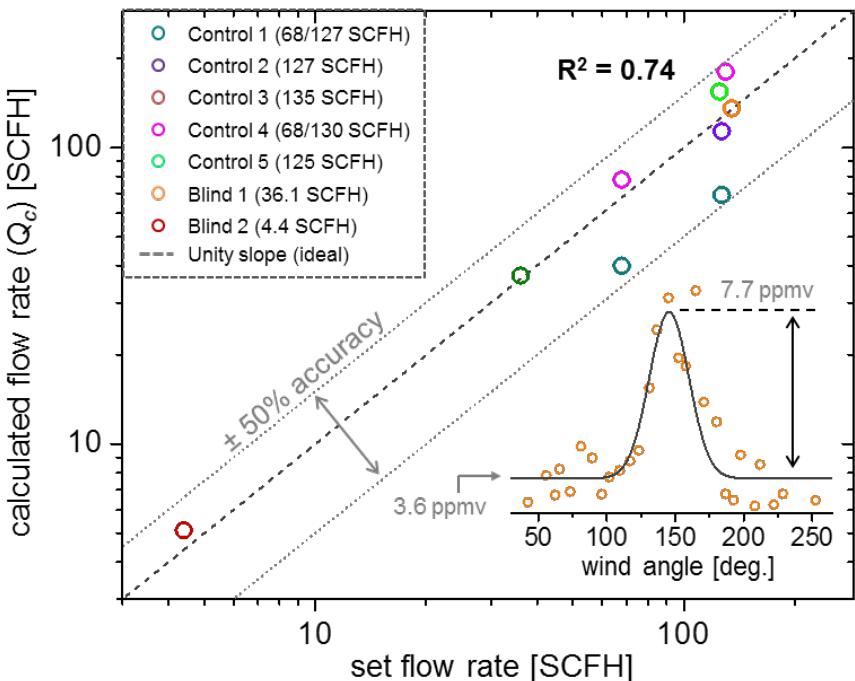
## Calculated source angle-of-arrival (AOA) for single-sensor “localization”:

- Total 6.3 hours control + blind CH<sub>4</sub> release (85 % CH<sub>4</sub>, 10.2 % C<sub>2</sub>H<sub>6</sub>, 0.7 % C<sub>3</sub>H<sub>8</sub>)
- AOA consistent with known source-detector line-of-sight (LoS), and downwind of CH<sub>4</sub> leak
- Single-sensor cannot distinguish between two leaks along a single LoS → choose placement wisely
- Can we estimate the source magnitude from a single sensor?
- Extract the shape of the plume for modeling (multiple sensors)

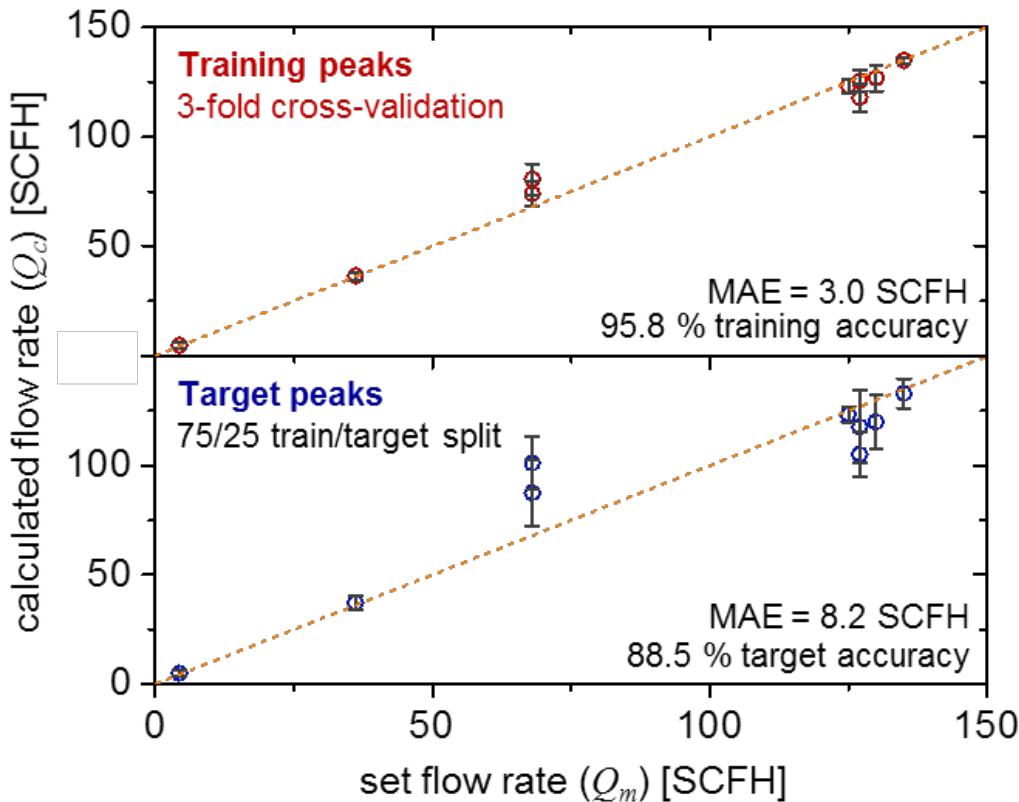
Need plume profile information  
→ use wind variability

## Parametrized Gaussian plume model:

- Plume reconstruction from  $\vec{v}_w$  variation in short-term measurements
- Identify nominal LoS  $\text{CH}_4$  peak concentration (Gaussian LMS)
- $d \ll D_L$  (Lagrangian integral dist.  $\sim 10^2$  m)  $\rightarrow \sigma = \sigma_v \cdot d$
- Optimization (i.e. “calibration”) of dispersions  $\sigma_{v,y}$  and  $\sigma_{v,z}$
- Training data shows  $Q_c$  agreement within 50 % accuracy



# Source estimation using ML models



- Random-forest model accuracy:  $\pm 8.2$  SCFH (target MAE)  $\rightarrow 88.5\% Q_c$  accuracy
- Top features ( $d$ ,  $[CH_4]$ ,  $\vec{v}_w$ ) consistent with GP model:

Flow rate:

$$Q_c \propto [CH_4](r_s) \cdot \langle |\vec{v}_w| \rangle \cdot d^2$$

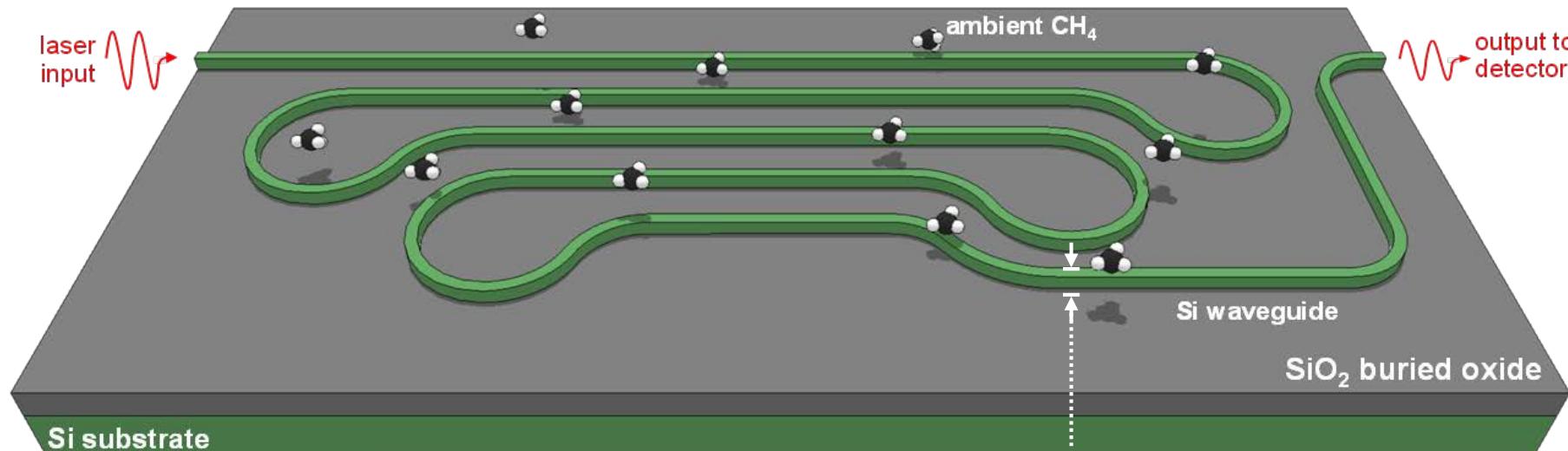
## RF-based supervised learning:

- Random-forest (RF) regressor model, 75/25 training-to-test split ratio
- Supervised training (20 s  $CH_4$  data input bins), 3-fold CV w/ randomized HP optimization

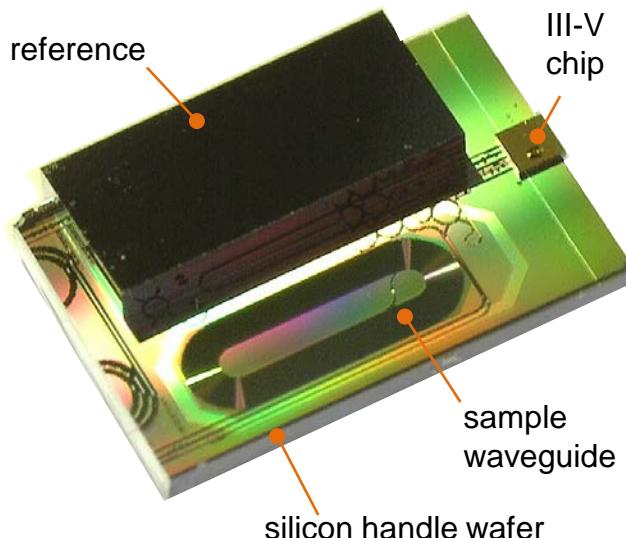
Feature	Relative Significance [%]
Source distance	$56.0 \pm 6.6$
$CH_4$ amplitude	$20.9 \pm 11.1$
Wind speed (max.)	$10.9 \pm 10.2$
Wind angle (rel.)	$5.0 \pm 1.1$
Wind speed (avg.)	$3.9 \pm 5.8$
Wind angle (abs.)	$1.8 \pm 0.8$
Wind speed (dev.)	$1.6 \pm 0.6$

# Silicon photonic chip sensor design

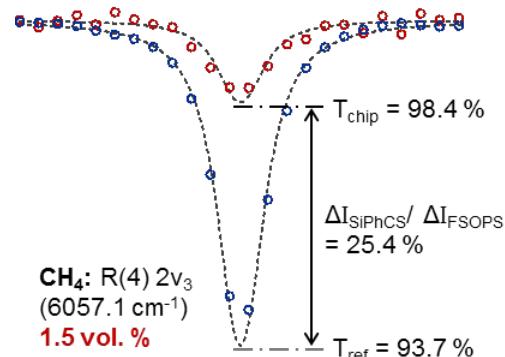
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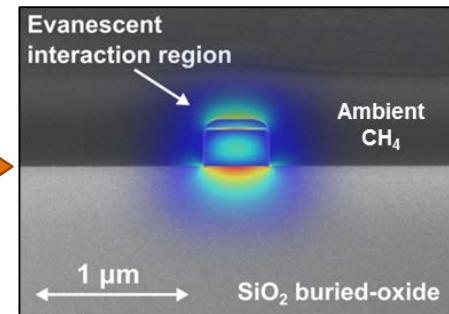
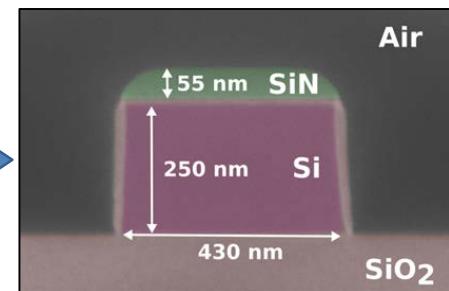
## Integrated photonic chip sensor



**Detection sensitivity:**  
 $\sigma_{\text{adv}} = 6.3 \text{ ppmv} \cdot \text{Hz}^{-1/2}$   
**Noise-equivalent absorp:**  
 $(\alpha L)_{\text{min}} = 3.3 \times 10^{-5} \text{ Hz}^{-1/2}$



x-section  
mode simulation



## Fugitive emissions monitoring of CH<sub>4</sub>:

- CH<sub>4</sub> as a clean fuel → reduce GHG loading via leak monitoring
- Requirements: spatial + temporal resolution (real-time, large-area SN)

## Demonstration of a field-deployable portable TDLAS sensor:

- Fiber coupled open-path (5 cm) absorption sensor ( $\alpha_{\min} = 6.5 \times 10^{-7} \text{ cm}^{-1} \cdot \text{Hz}^{-1/2}$ )
- Benchmark performance for next generation integrated photonic chip sensors
- 5-day field deployment at METEC facility (5 Control, 2 Blind, 6.3 hours data)
- Demonstrate correspondence vs. custom MOX sensors (TDLAS → specificity)

## Physical analytics for source localization/quantification:

- AOA calculation via CH<sub>4</sub> weighted mean wind-angle
- Single-sensor source estimation via Gaussian plume + ML models
- Generalizable to alternative sensing modalities (or multiple sensors in WSN)

**Next generation: integrated photonic chip sensor nodes for significant SWaP-C benefits → facilitate large-scale deployment of real-time WSNs**

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Yorktown Heights, NY, USA