## Leveraging Multiple Continuous Monitoring Sensors for Emissions Alerting on Oil and Gas Facilities

William Daniels<sup>1,2</sup>, Meng Jia<sup>1,2</sup>, Dorit Hammerling<sup>1,2</sup> Shyla Kupis<sup>3</sup>, Nasr Alkadi<sup>3</sup>, Anna Scott<sup>3</sup>

<sup>1</sup>Department of Applied Mathematics and Statistics, <sup>2</sup>Payne Institute for Public Policy, Colorado School of Mines, Golden, CO

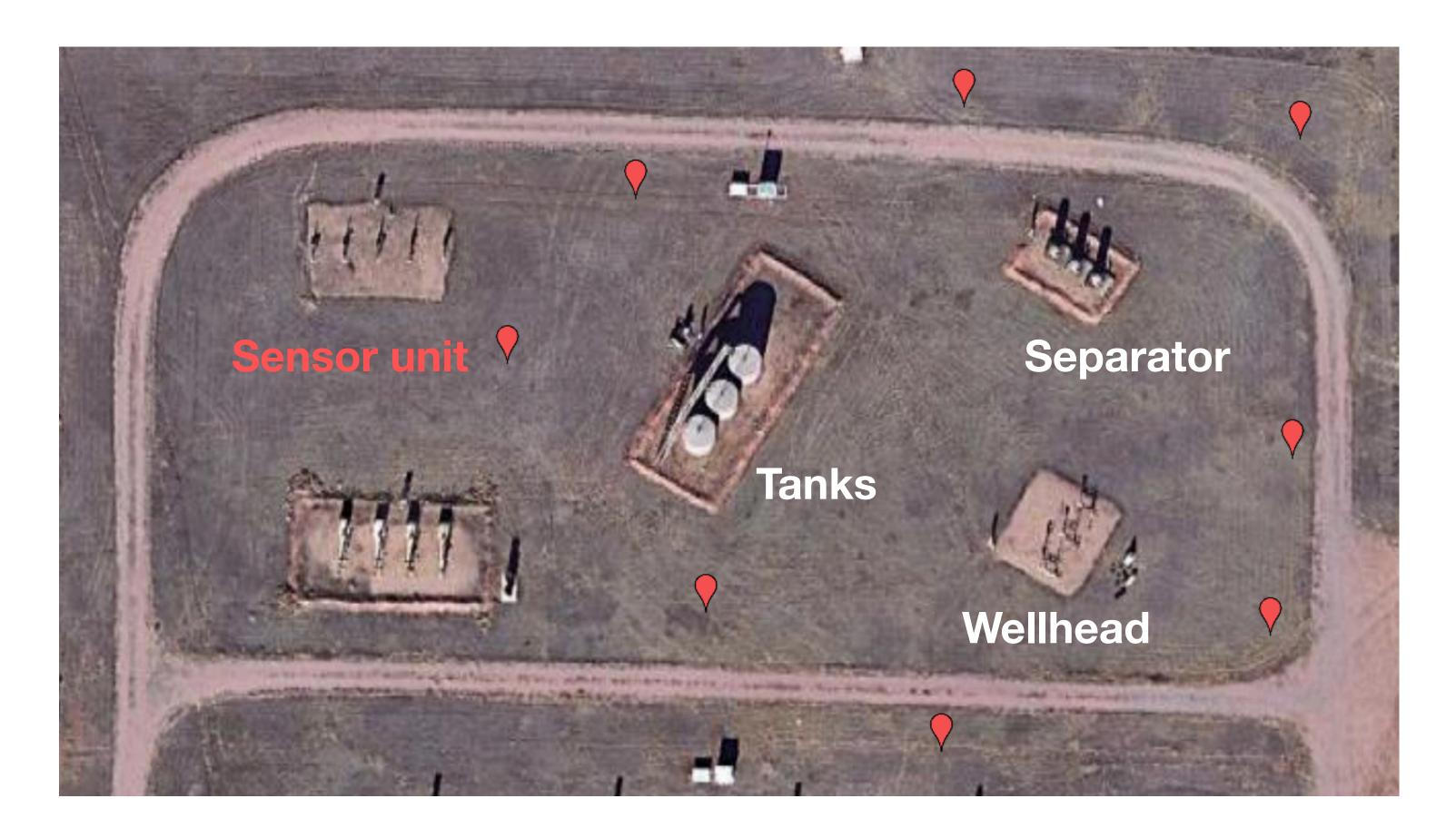


<sup>3</sup>Project Canary, Denver, CO



## The problem

Given a number of continuous monitoring sensors on an oil and gas facility, can we deliver concise alerts when an actionable event occurs?



### Colorado State University's **METEC** Facility

Oil and gas test facility capable of controlled emissions





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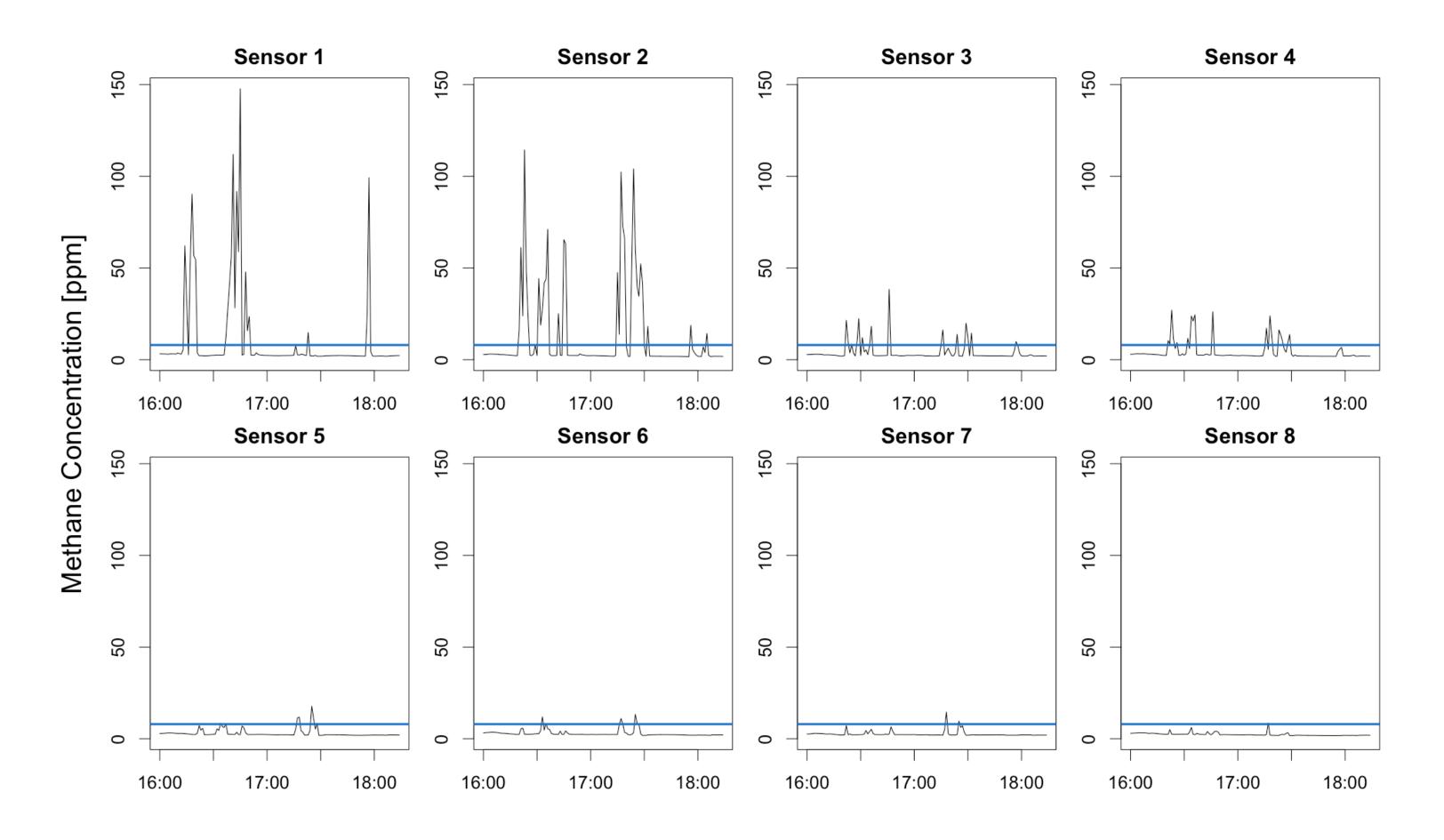
- Continuous monitoring data from Project Canary
  - Sensors report minute-averaged observations -
  - E.g., methane concentration, wind speed, wind direction





### The motivation

Alerting on static thresholds can overwhelm operator and does not utilize information from all units simultaneously



### Alert Log

	А	В	с
1	time	sensor	concentration
2	2/16/21 16:14	1	62.12096162
3	2/16/21 16:15	1	32.81096162
4	2/16/21 16:17	1	51.36796162
5	2/16/21 16:18	1	90.21896162
6	2/16/21 16:19	1	56.69896162
7	2/16/21 16:20	1	54.65696162
8	2/16/21 16:20	2	16.15248963
9	2/16/21 16:21	2	61.10948963
10	2/16/21 16:21	4	10.3509834
11	2/16/21 16:22	2	23.87848963
12	2/16/21 16:22	3	21.48698755
13	2/16/21 16:22	4	8.602983402
14	2/16/21 16:23	2	114.3904896
15	2/16/21 16:23	3	9.827987552
16	2/16/21 16:23	4	27.0479834
17	2/16/21 16:24	2	47.91548963
18	2/16/21 16:24	4	11.3499834
19	2/16/21 16:25	2	23.16148963
20	2/16/21 16:25	3	8.521987552
21	2/16/21 16:26	4	9.316983402
22	2/16/21 16:28	3	9.788987552
23	2/16/21 16:29	3	22.37298755
24	2/16/21 16:31	2	44.24248963
25	2/16/21 16:31	3	12.02098755
26	2/16/21 16:32	2	18.86448963
27	2/16/21 16:32	4	11.6609834
28	2/16/21 16:33	2	27.16248963
29	2/16/21 16:33	6	11.97362448
30	2/16/21 16:34	2	41.88248963
31	2/16/21 16:34	4	23.7939834
32	2/16/21 16:34	5	8.417526971





## The plan

### Proposed solution: Semi-real time event detection and localization utilizing:

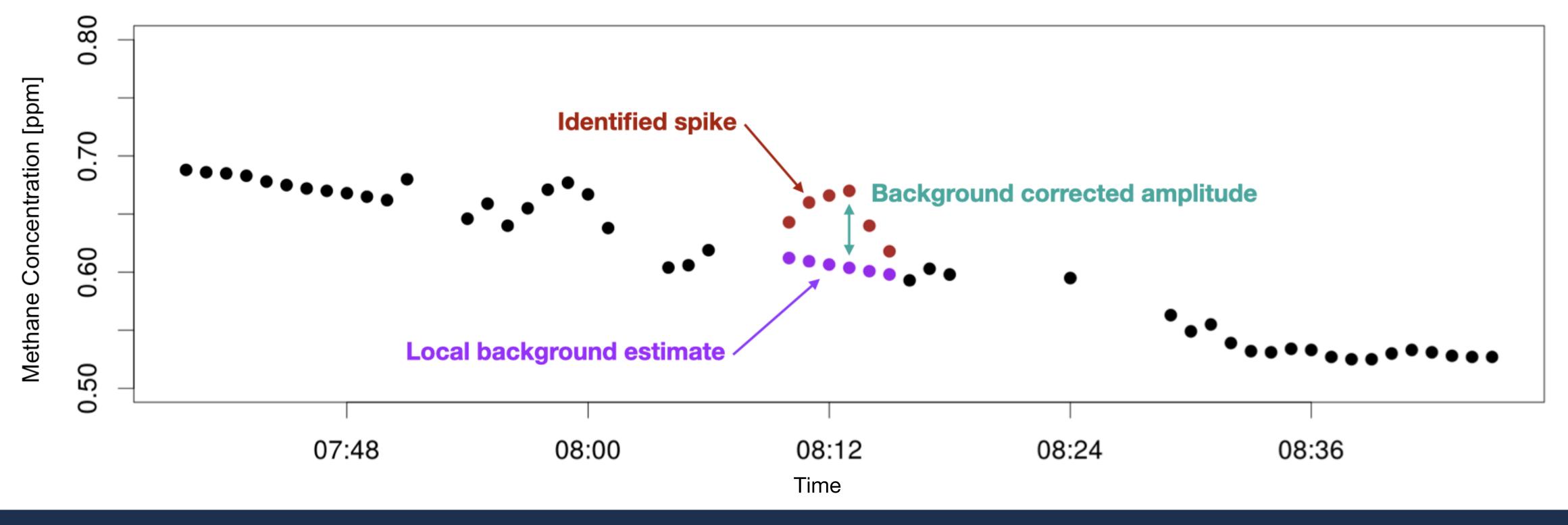
- 1. Site geometry, including knowledge about all potential sources
- 2. Information from all available methane and wind sensors

### Method:

- 1. Remove background from sensor observations
- 2. Simulate concentrations at sensor locations from all potential emission sources
- 3. Pattern match simulated concentrations and observations via custom metric to identify most likely source for each sensor
- 4. Use wind data and site geometry to combine information across sensors



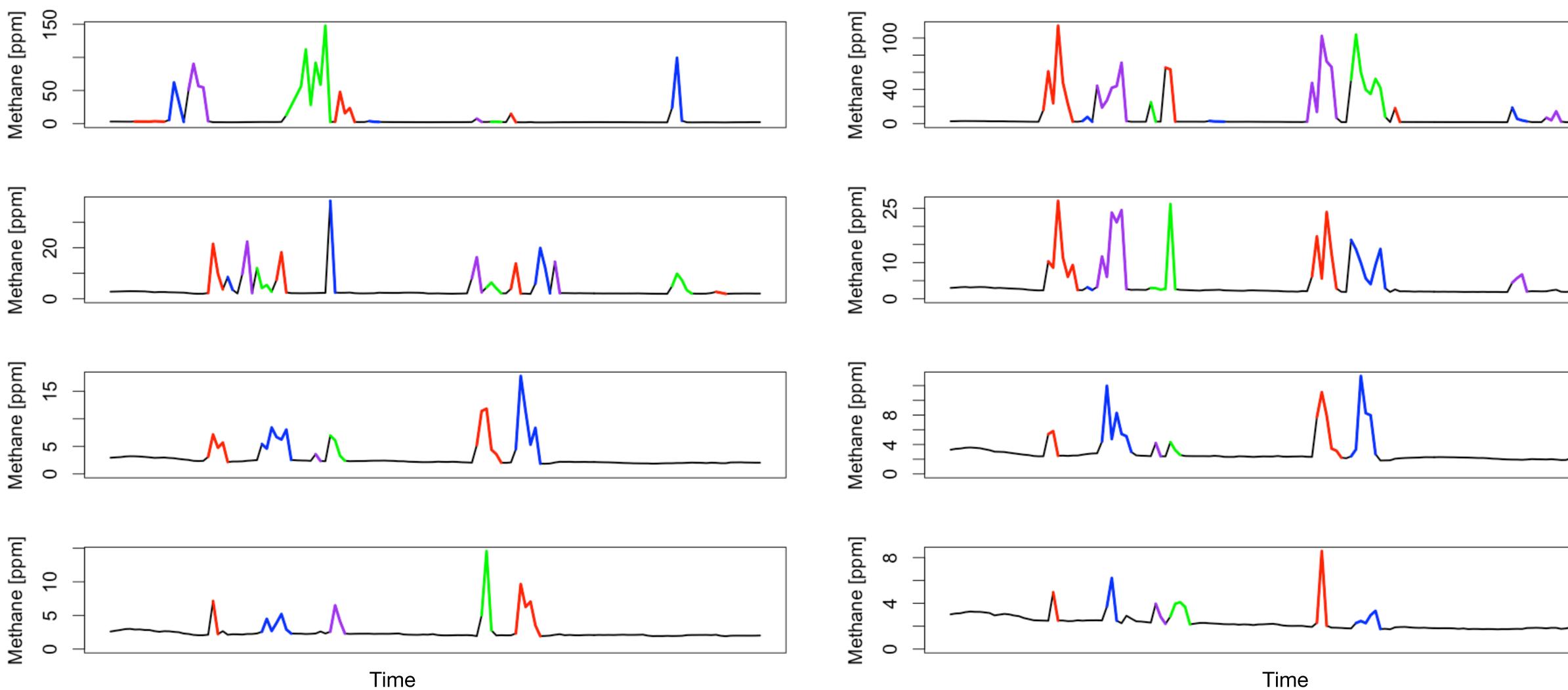
- Detect spikes via custom gradient-based method
- Estimate background via non-parametric regression fit to local "non-spike" observations



Will Daniels - wdaniels@mines.edu



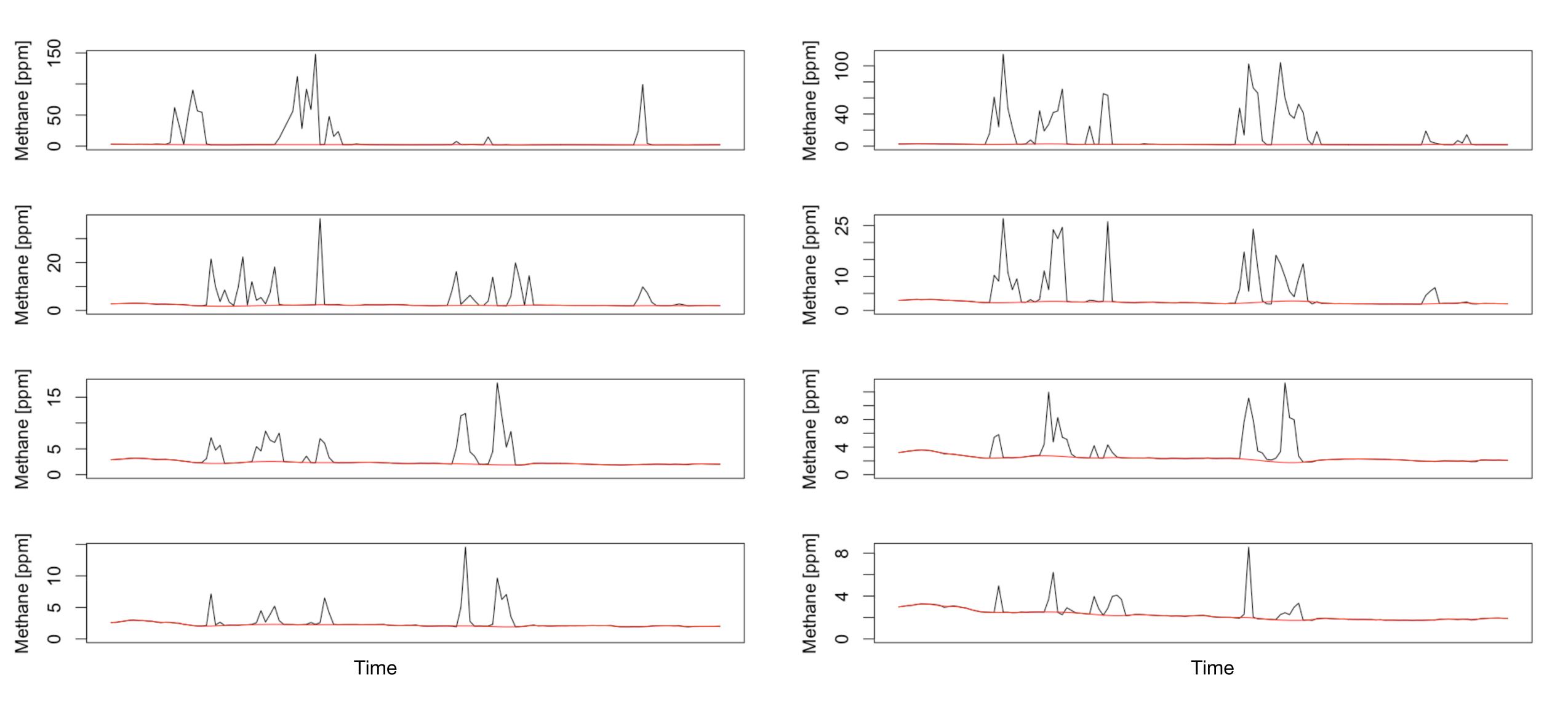
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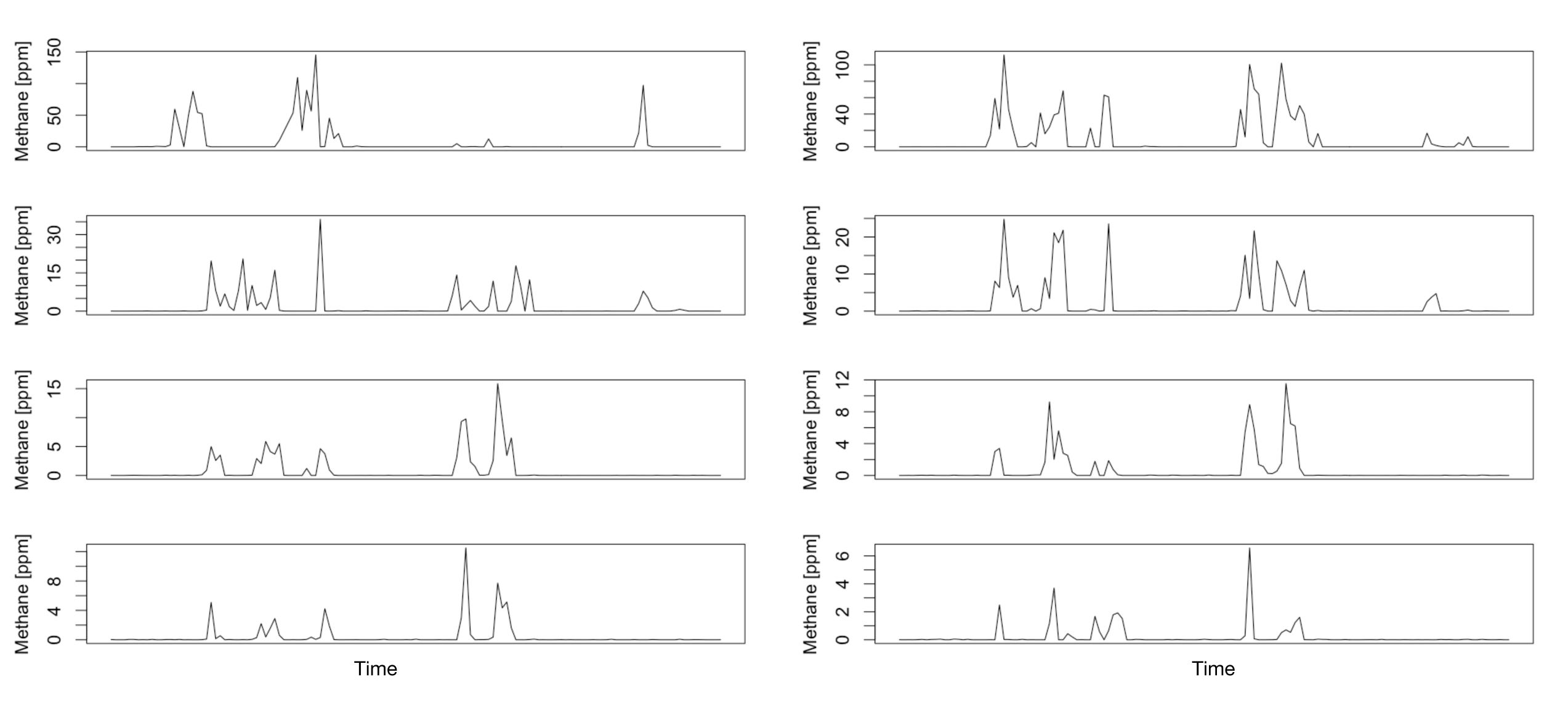


Colors distinguish between different spikes









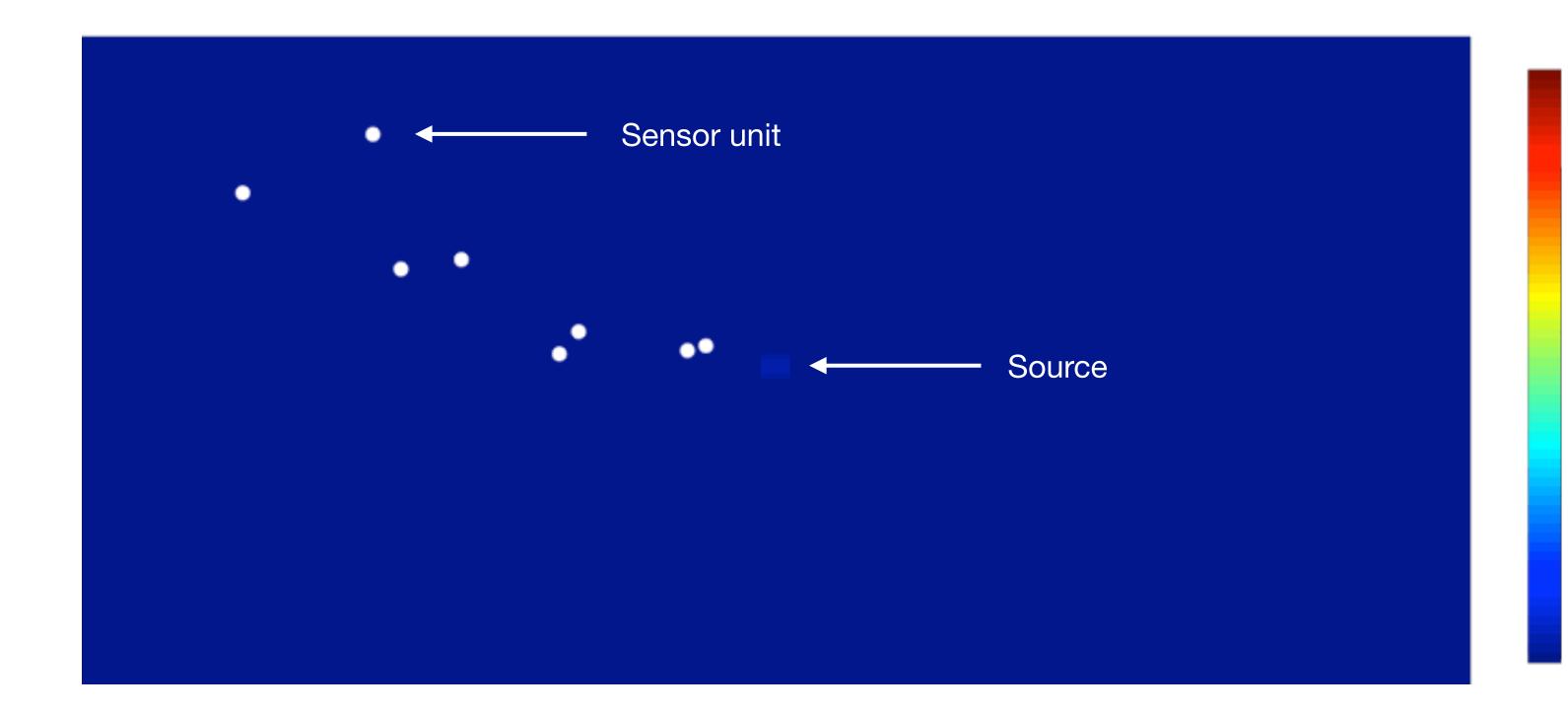


Forward model: Gaussian puff with different horizontal and vertical variances  $\bullet$ 

$$C(x, y, z, t) = \frac{Q_t}{(2\pi)^{\frac{3}{2}} \sigma_y^2 \sigma_z} \exp\left(-\frac{(x - ut)^2 + y^2}{2\sigma_y^2}\right) \left[\exp\left(-\frac{(z - H)^2}{2\sigma_z^2}\right) + \exp\left(-\frac{(z + H)^2}{2\sigma_z^2}\right)\right]$$

- Where:
  - C(x, y, z, t) is the predicted concentration at location (x, y, z) and time t
  - $Q_t$  is the amount of methane released at time t
  - u is the wind speed at time t
  - H is the height of the source \_





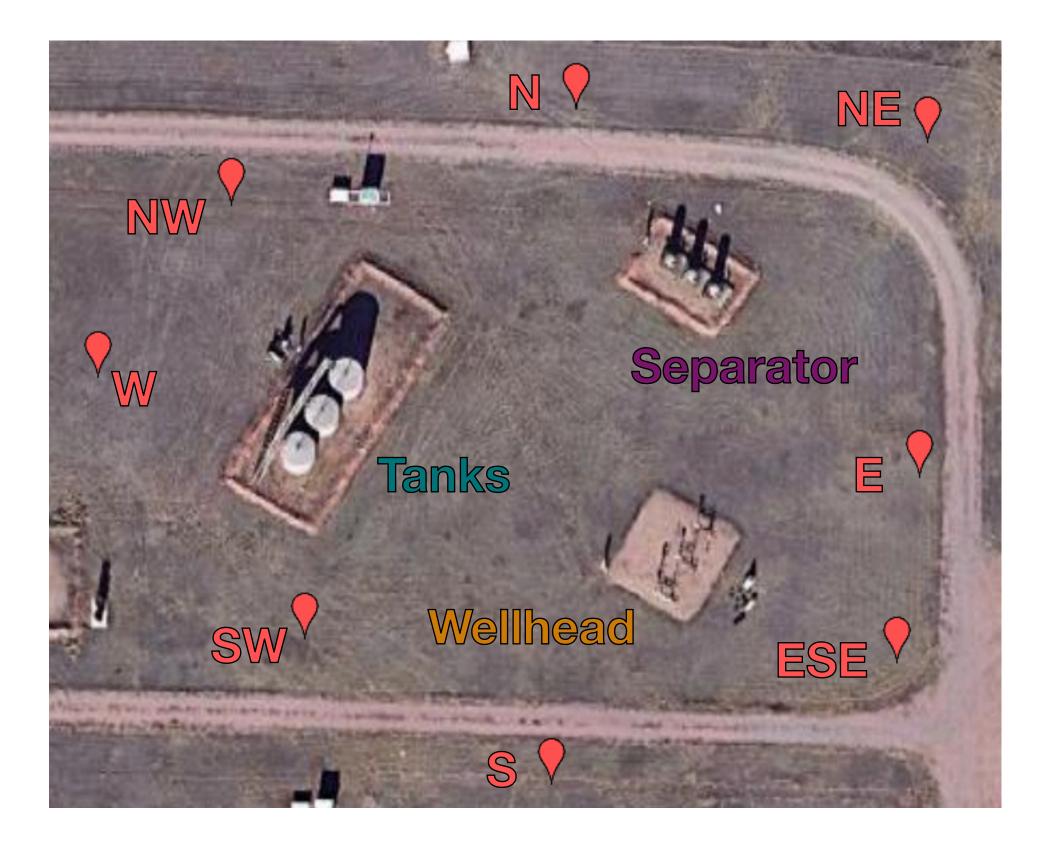
Relative Methane Concentration

Will Daniels - wdaniels@mines.edu



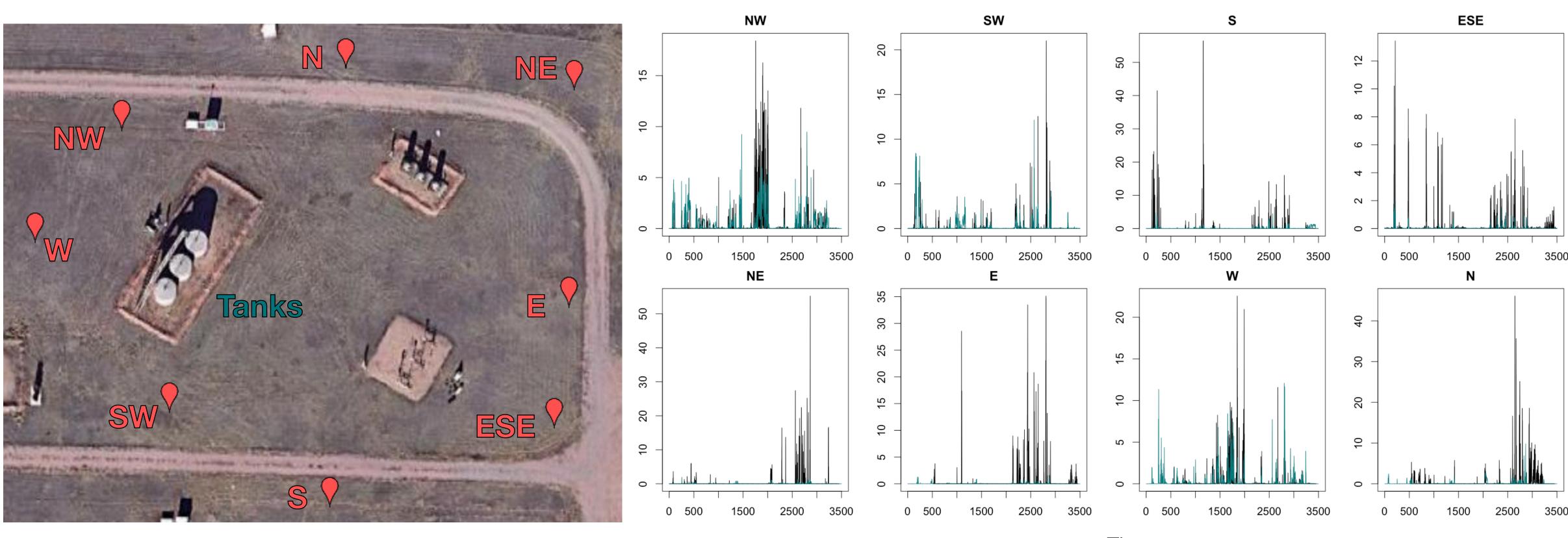
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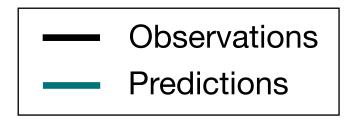
Compute simulation predictions from all possible sources





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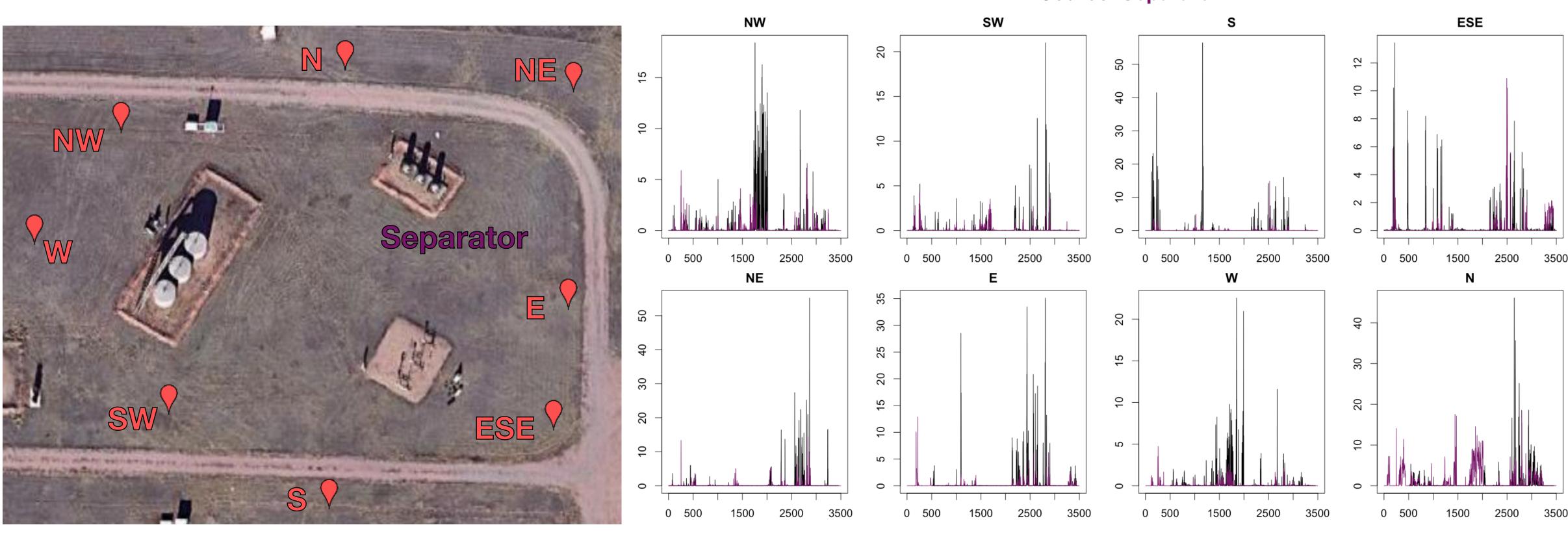


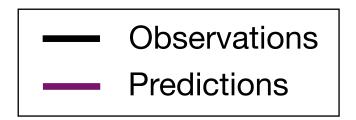
**Source: Tanks** 

Time



Compute simulation predictions from all possible sources



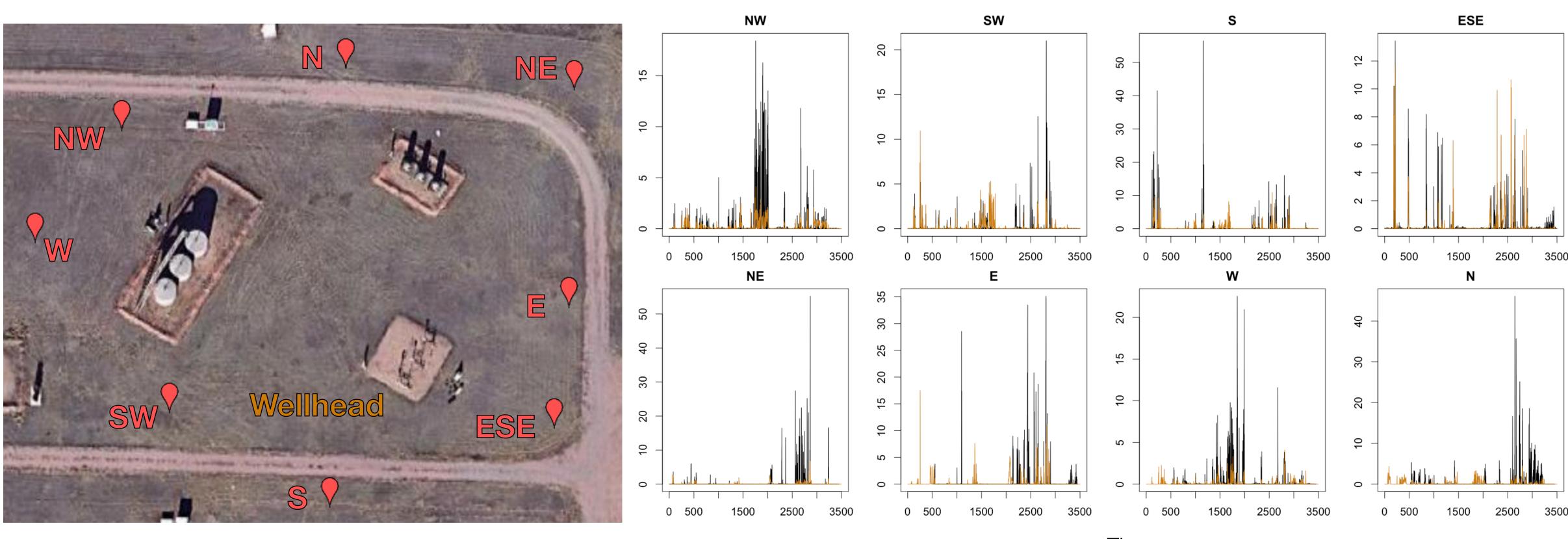


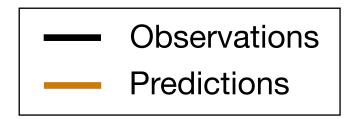
**Source: Separator** 

Time



Compute simulation predictions from all possible sources





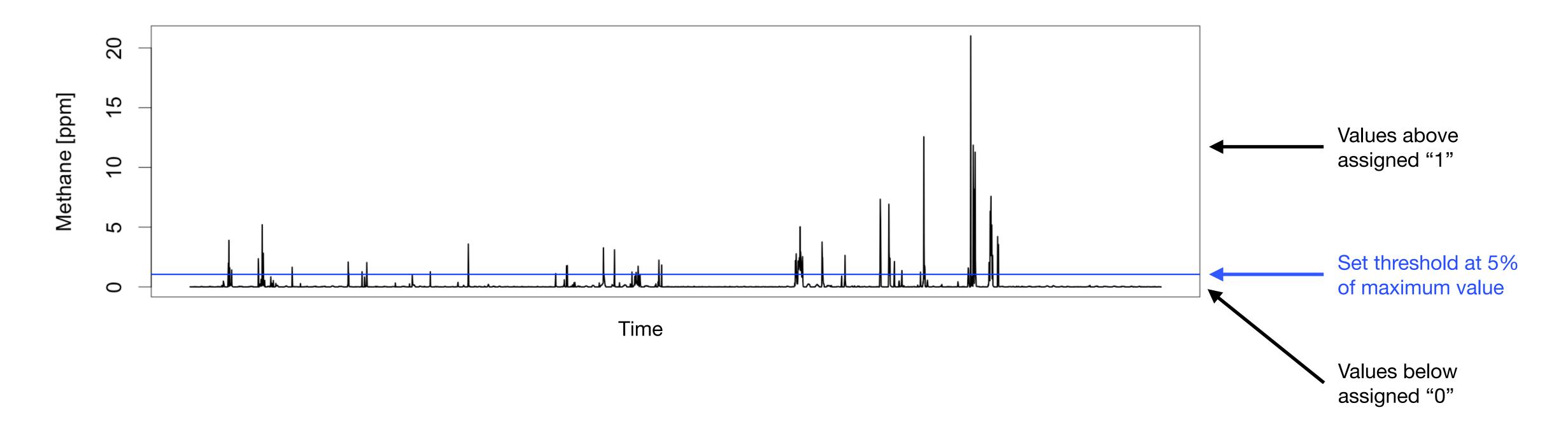
### **Source: Wellhead**

Time



## Step 3: Pattern match

- alignment
- Convert observations and predictions into a binary representation: high or low  $\bullet$

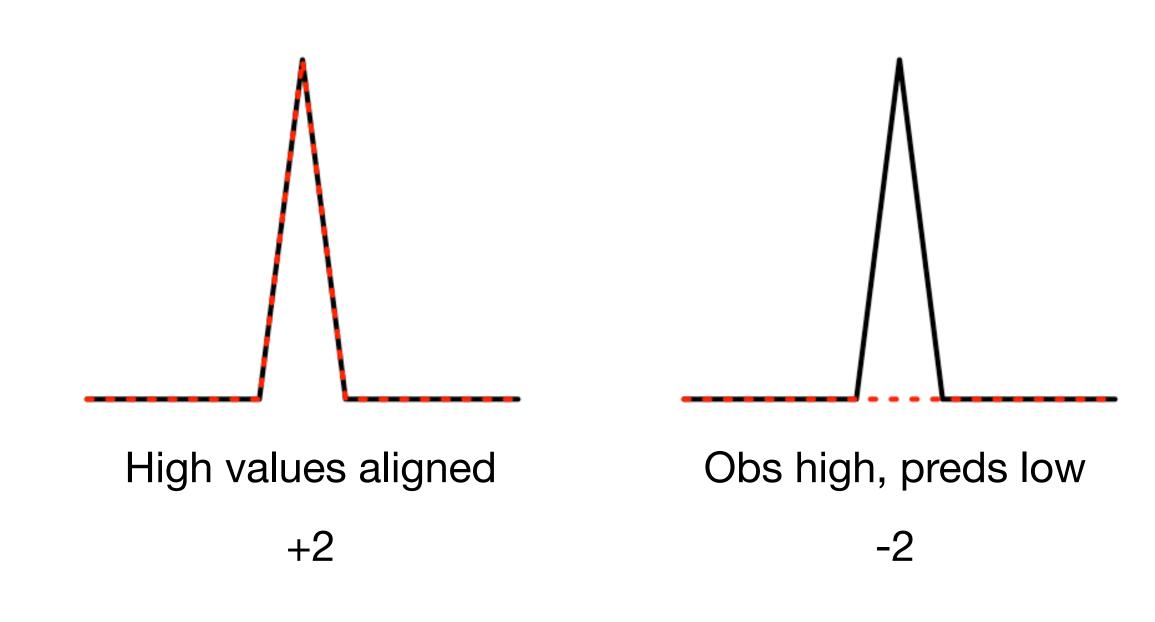


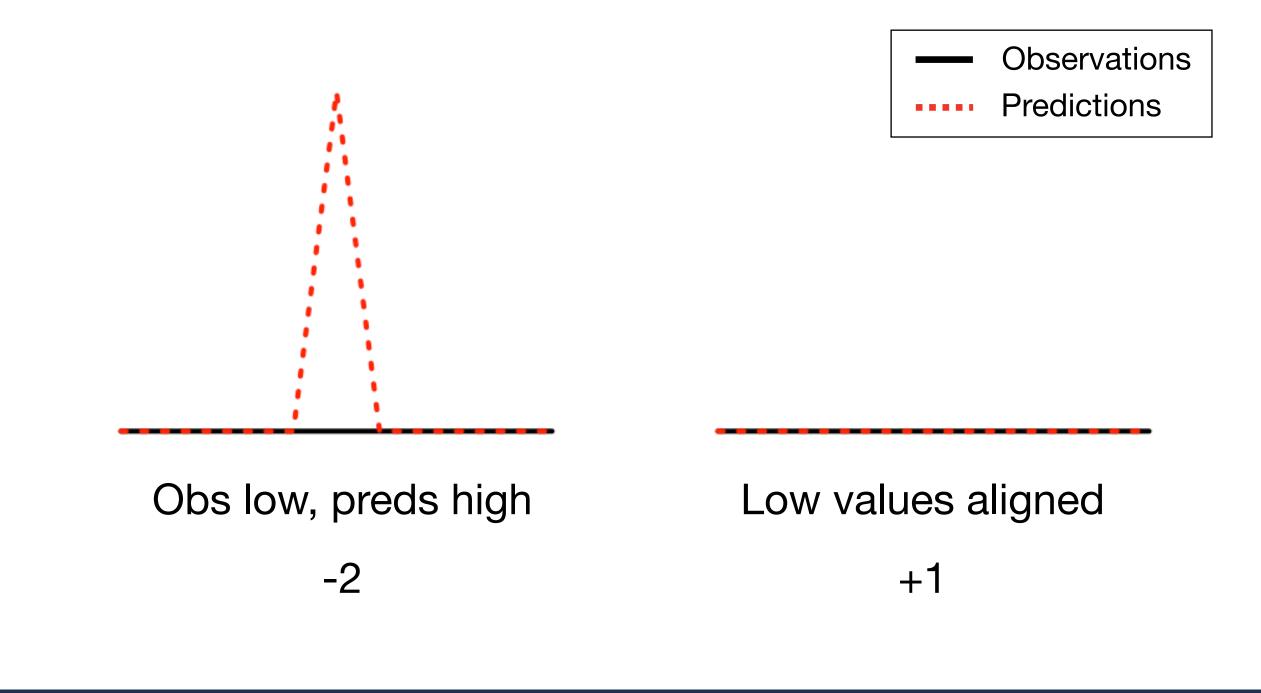
### • True emission rate unknown in practice: focus on spike alignment, not on amplitude



## Step 3: Pattern match

- True emission rate unknown in practice: focus on spike alignment, not on amplitude alignment
- Convert observations and predictions into a binary representation: high or low
- For each simulation, compute "points" in the following manner

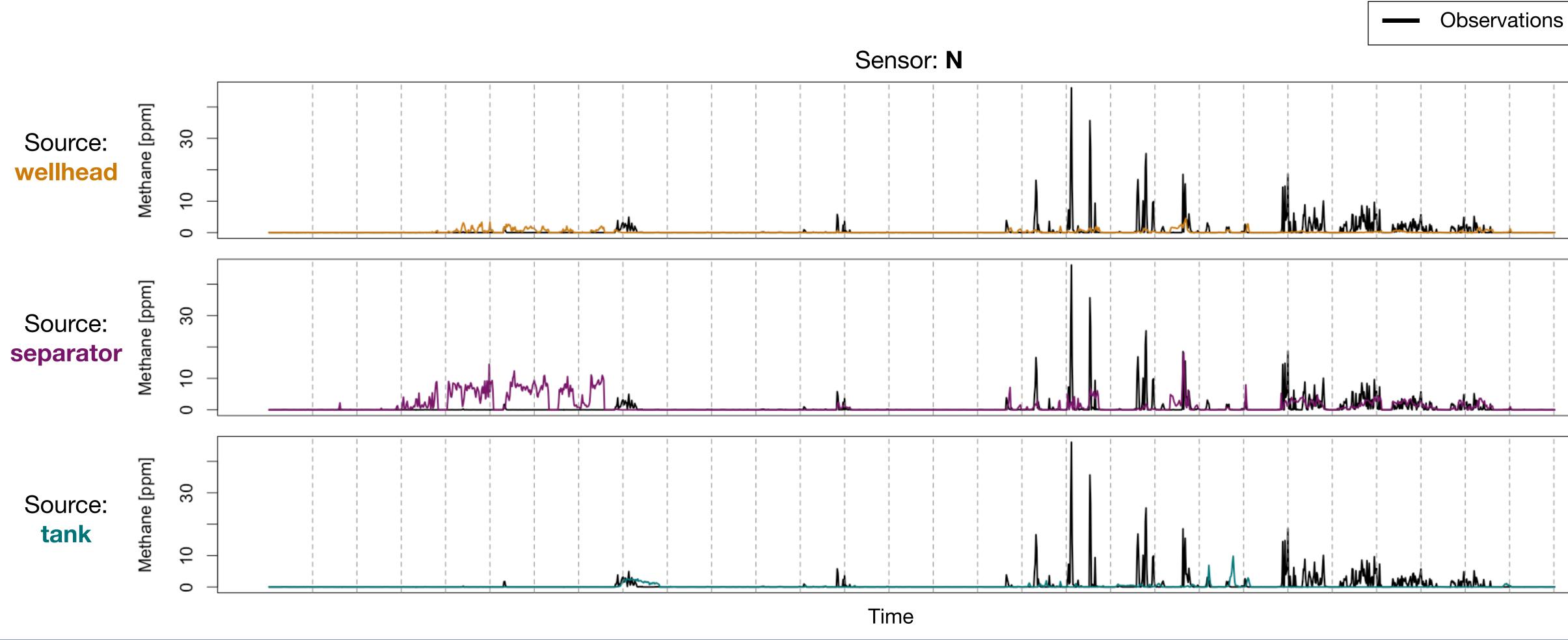






## Step 3: Pattern match

 $\bullet$ sources



### Perform pattern matching algorithm on small time chunks to account for time varying

Will Daniels - wdaniels@mines.edu



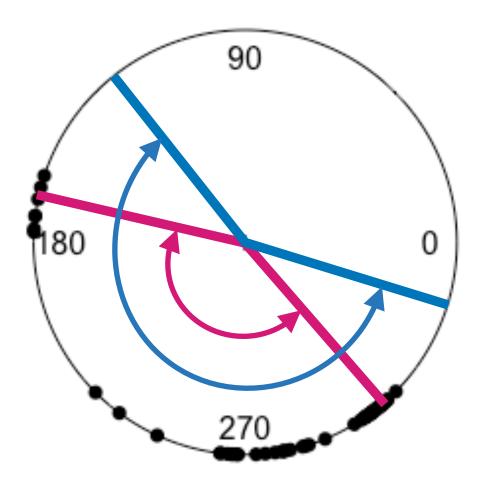
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	1	1
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## **Step 4: Combine sensors**

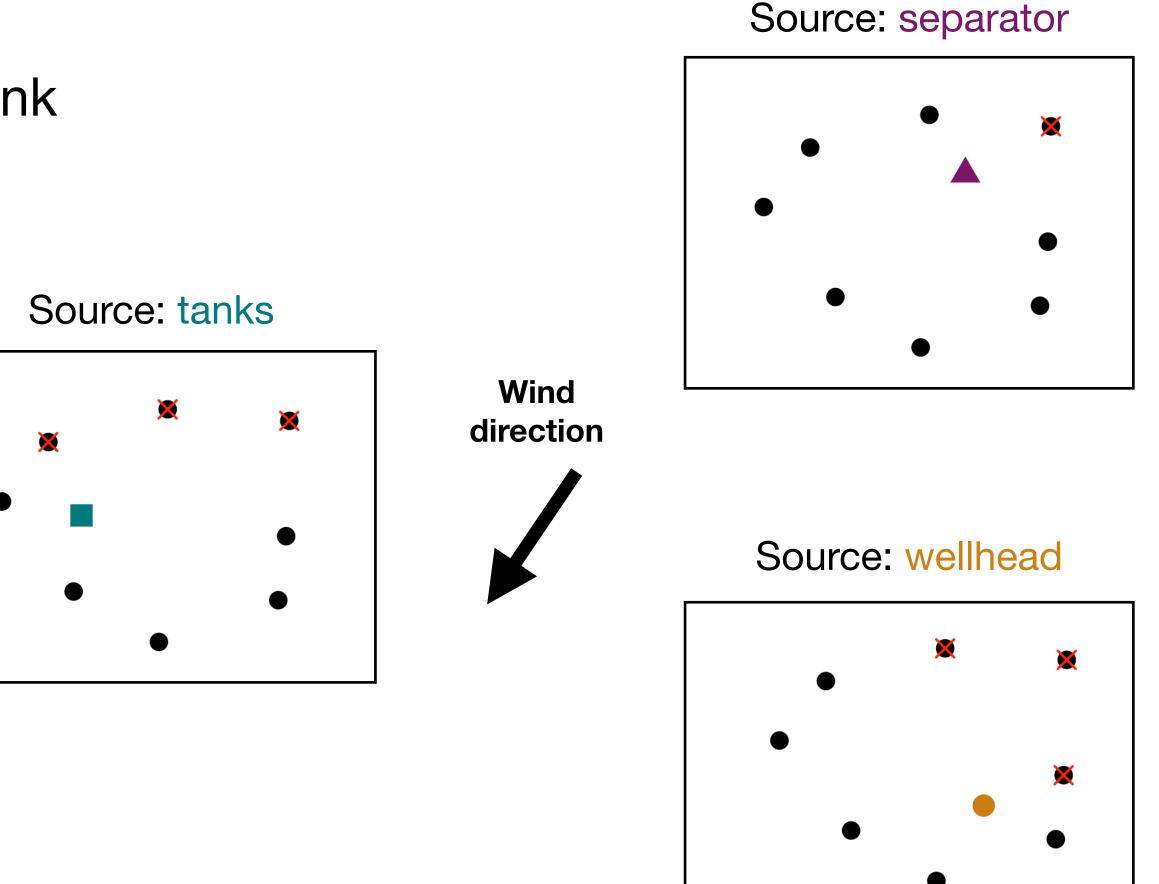
metric across downwind sensors

**Example:** Consider a single 60 minute time chunk



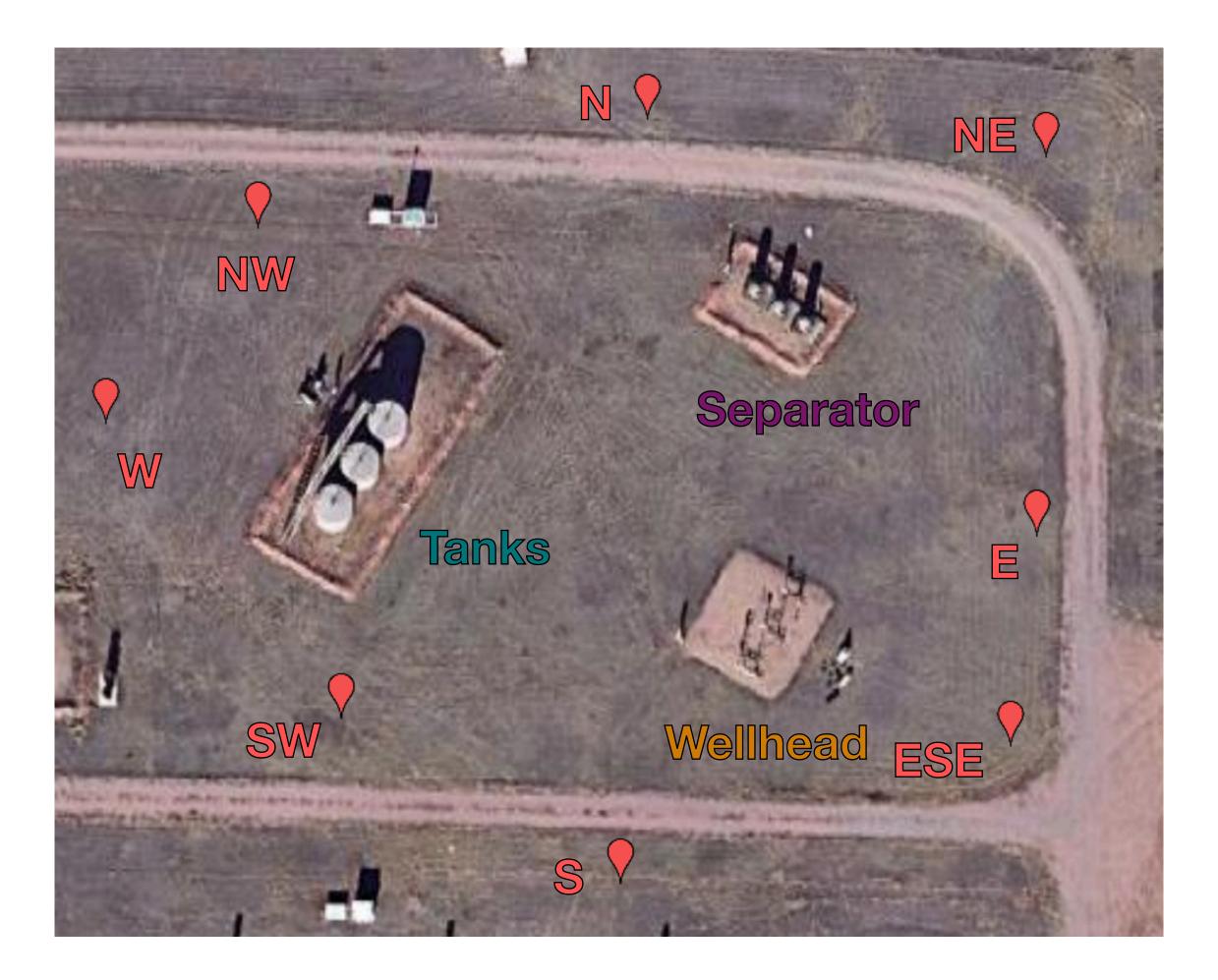
- Wind direction of each observation
- 10th and 90th percentiles
- Extended downwind range

### **The plan:** For each time chunk and for each source, omit data from upwind sensors and average





### Results



### Experimental setup

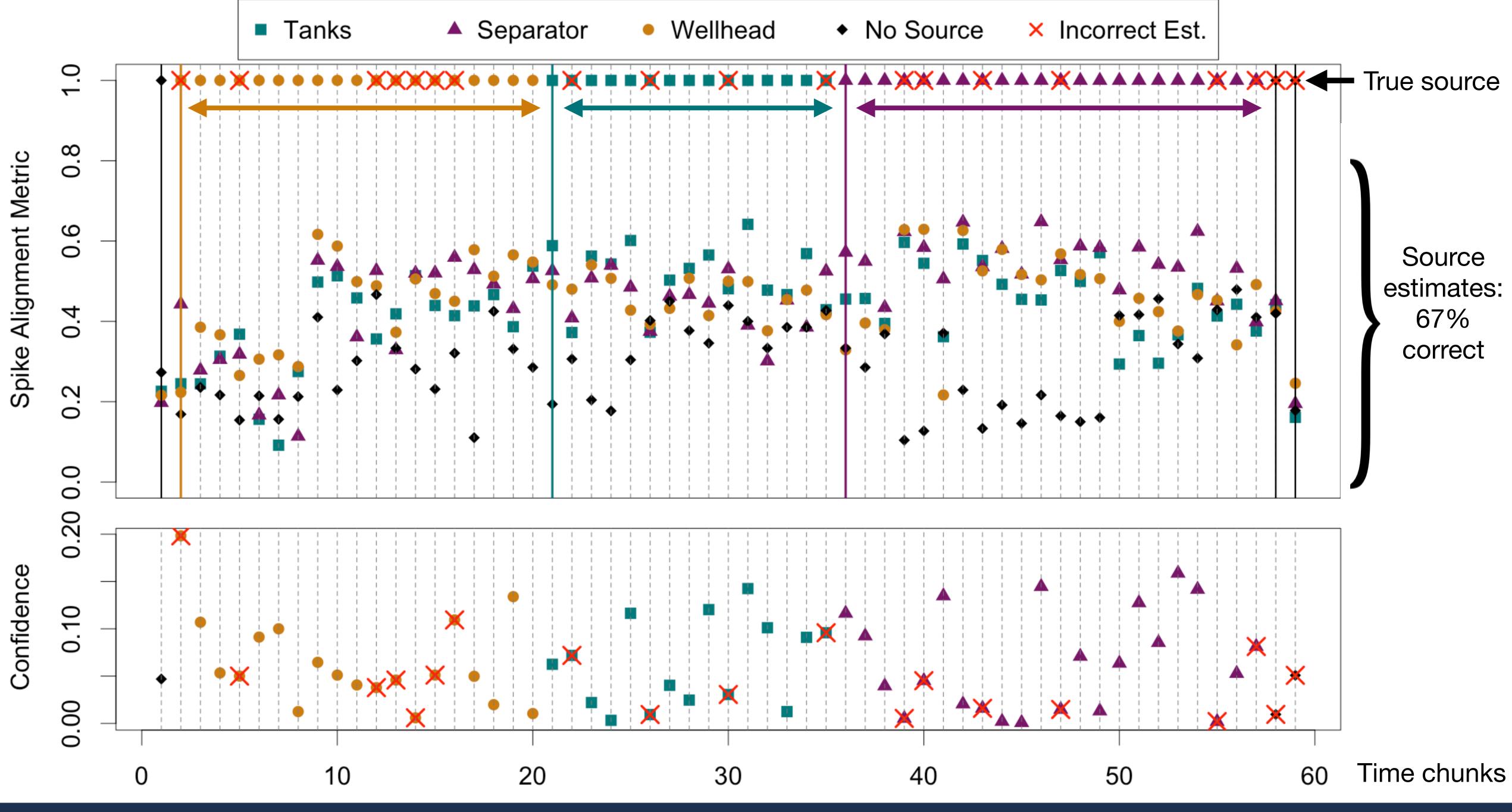
- 3 potential sources: tanks, separator, wellhead
- 8 sensors
- 58 hours of data (observations taken every minute)  $\bullet$

### **Emission profile**

- Controlled experiment
- Only one source emits at a time
- True source and emission rate changes over time





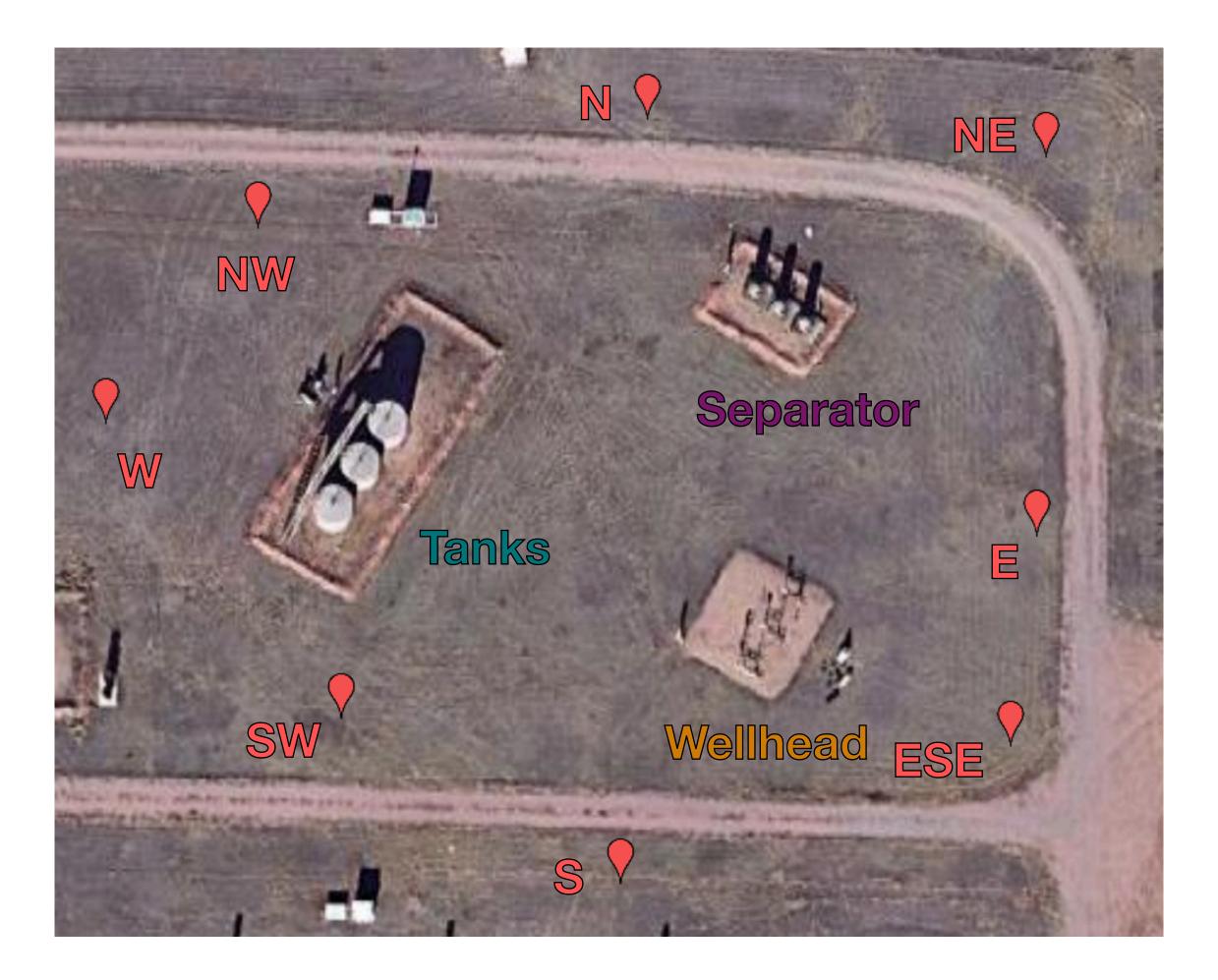


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Will Daniels - wdaniels@mines.edu



### Results



### What about a more realistic sensor arrangement?

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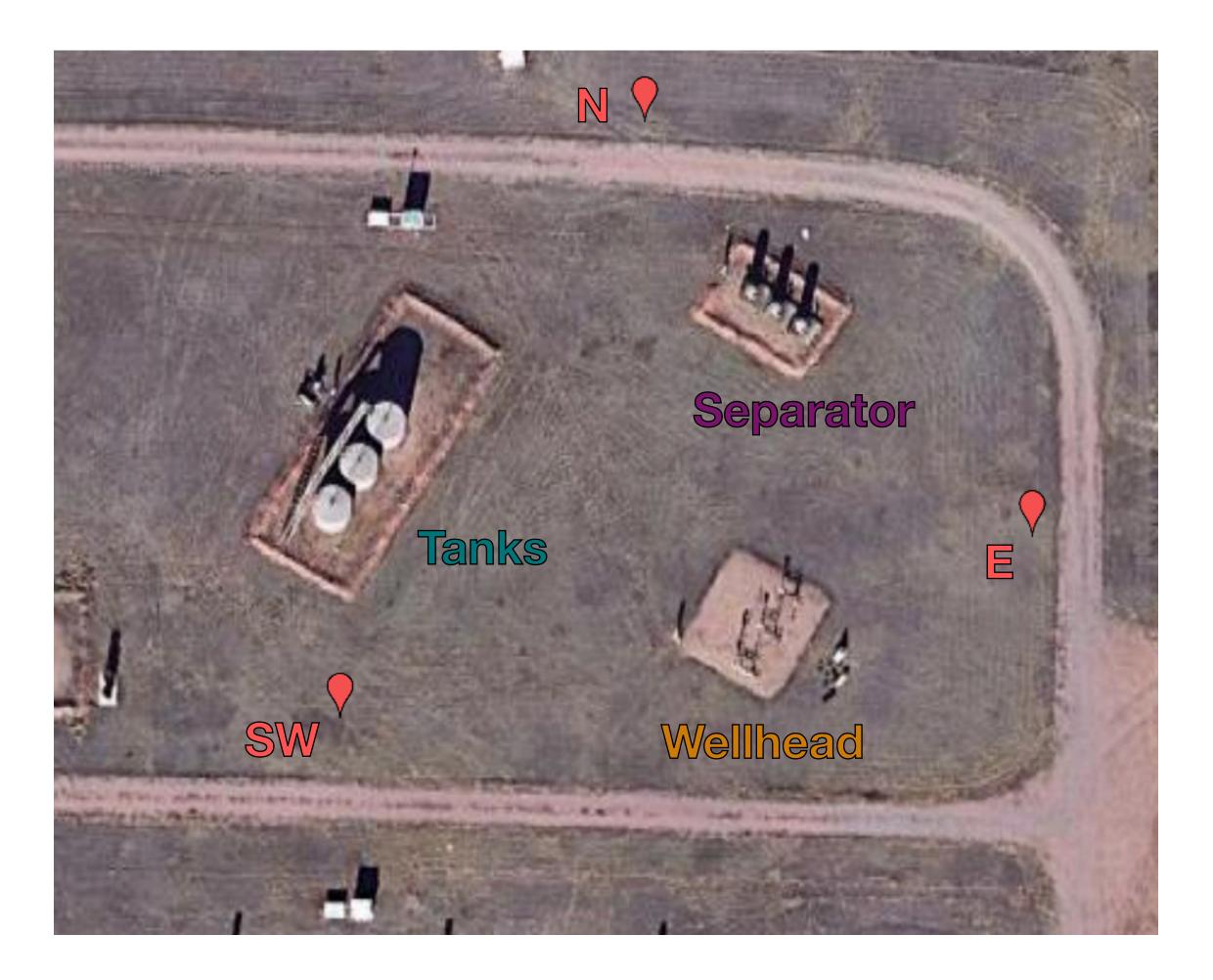
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### Results



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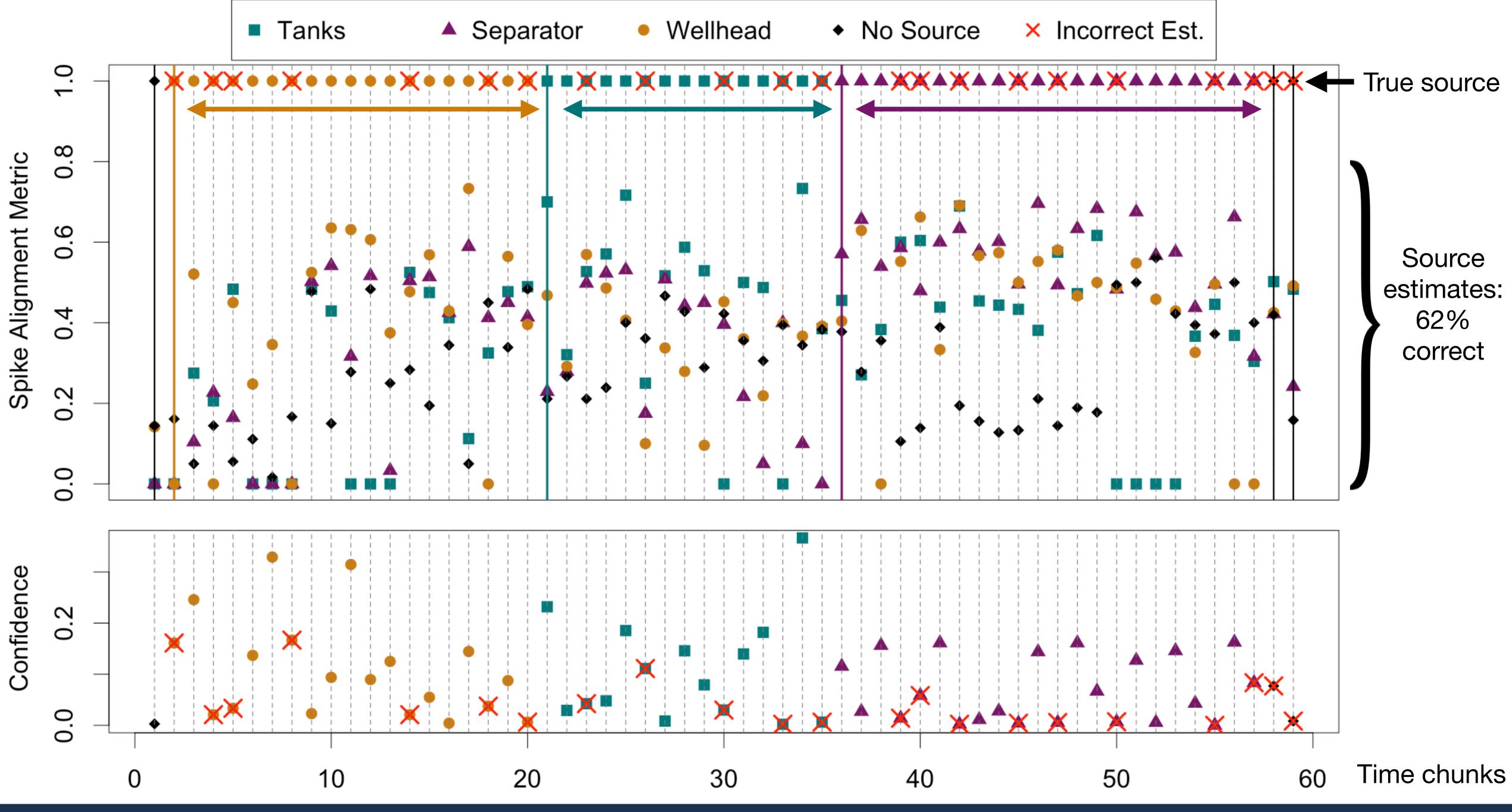
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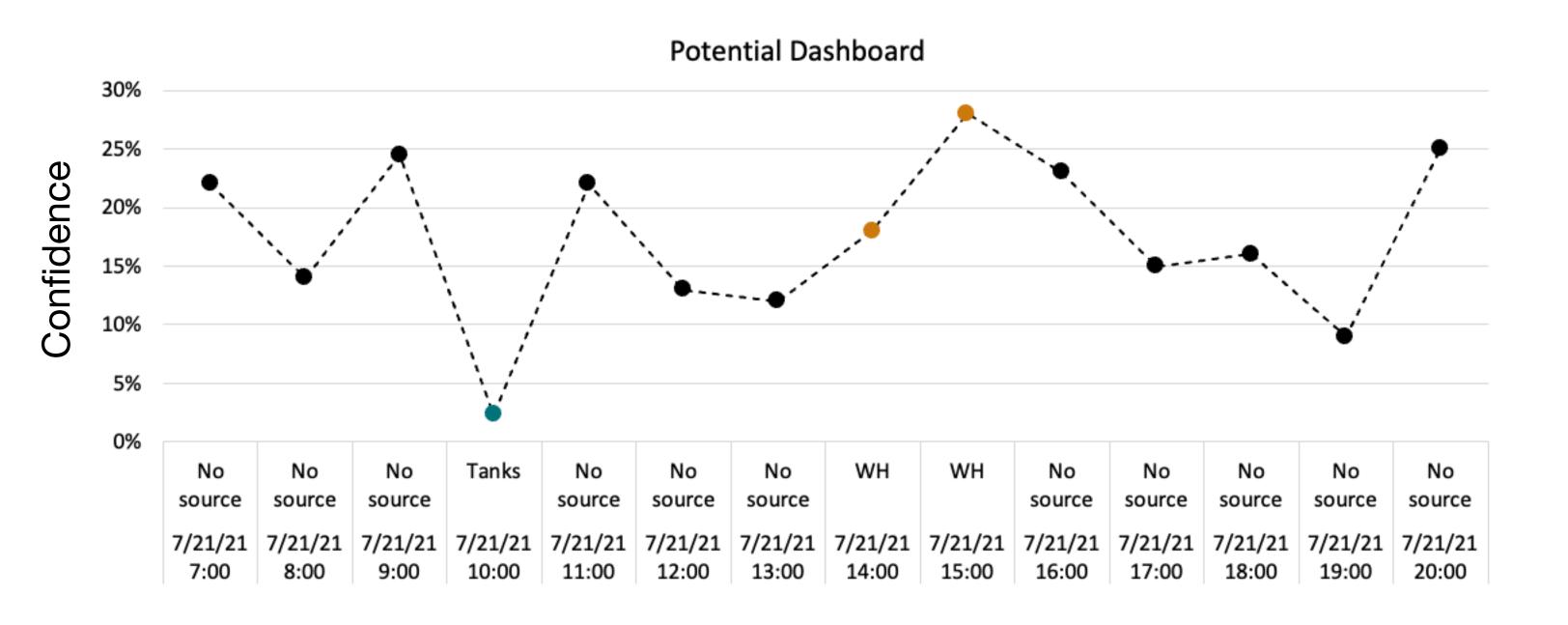


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- 1. Not doing a full inversion, but using a forward model for each potential source to inform localization
- 2. Using wind direction when combining sensors maximizes contribution of meaningful signal
- 3. Framework performs well in practical scenario
- 4. Framework does not depend on true emission rate







# wdaniels@mines.edu

