# A data-driven algorithm to optimize the placement of continuous monitoring sensors on oil and gas sites

Meng Jia, Troy Sorensen, Will Daniels, Dorit Hammerling Applied Mathematics and Statistics, Colorado School of Mines

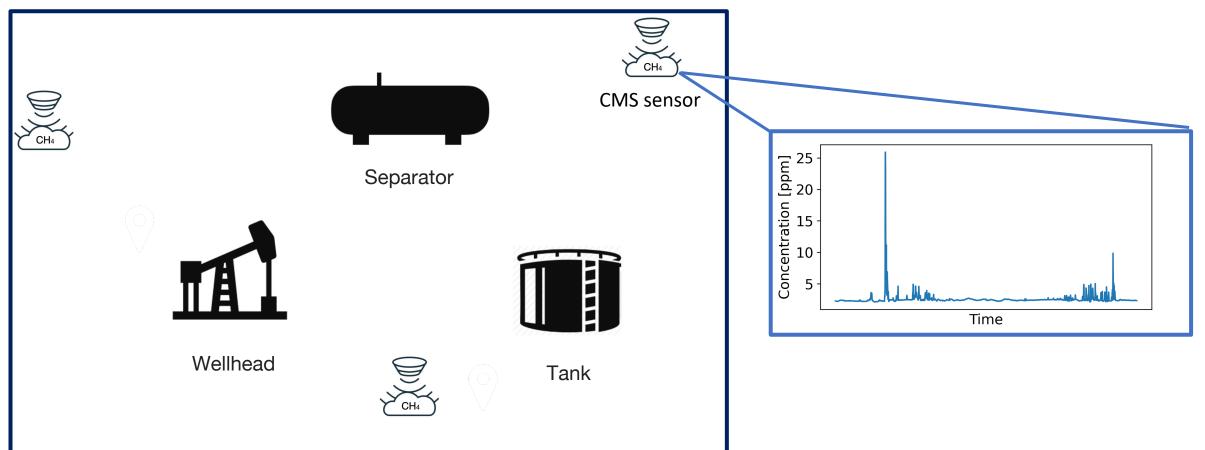
**Colorado School of Mines GRADS 2024** 

April 03, 2024

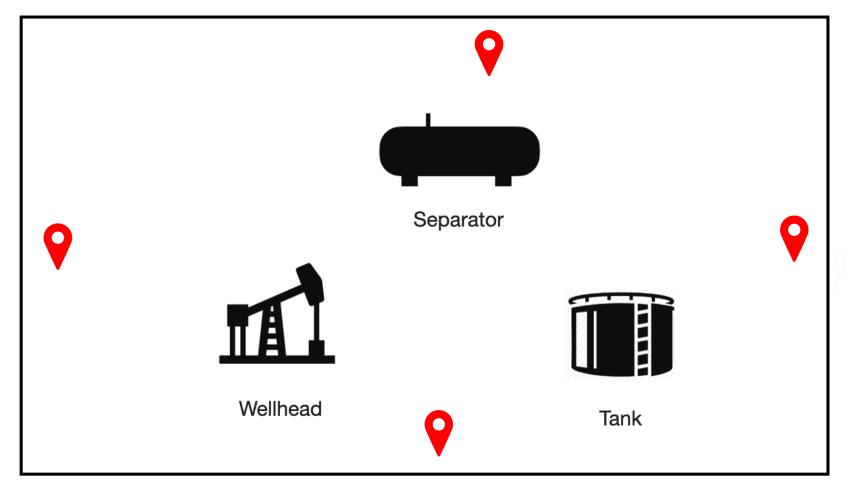
## Background

- Methane, CH4, is the 2<sup>nd</sup> biggest contributor to global climate change after CO2.
- CH4 has higher heat-trapping but shorter lifetime compared to CO2.
- Rapid reduction of CH4 has a quick impact on mitigating global warming.
- Oil & gas sector accounts for ~ 22% of global anthropogenic methane emissions.
- Methane emission monitoring technologies: satellite, aerial, groundbased systems

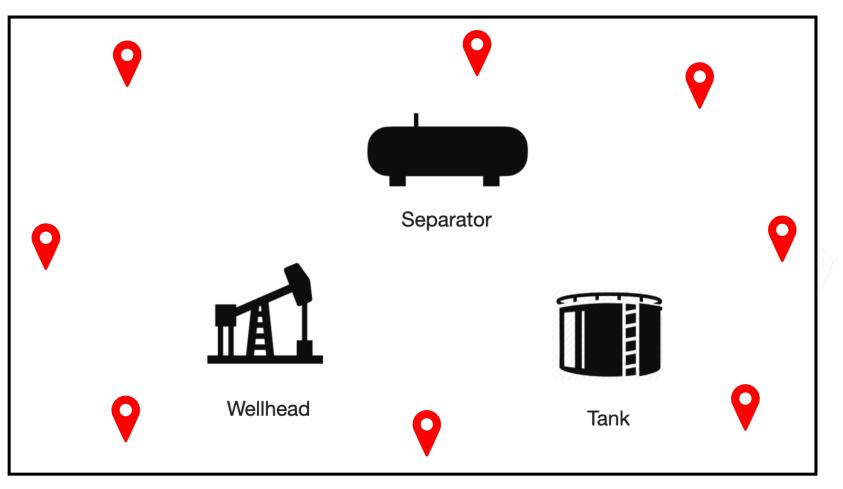
• Continuous monitoring systems (CMS)



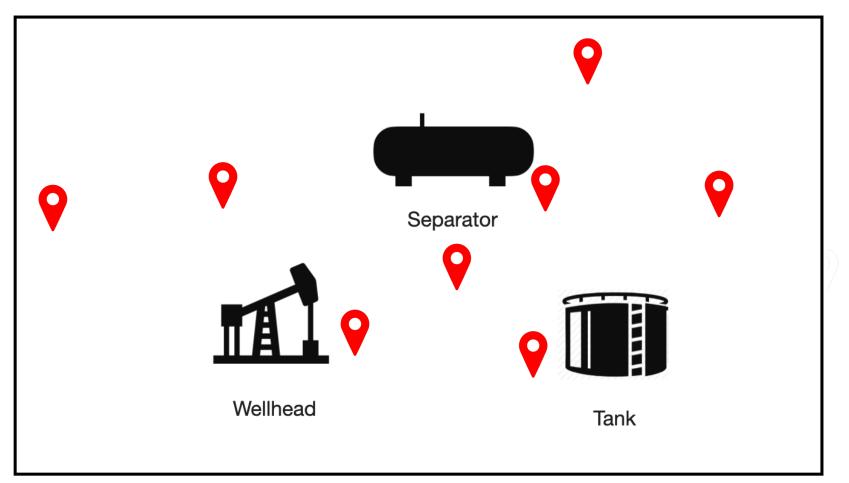
• CMS sensor placement



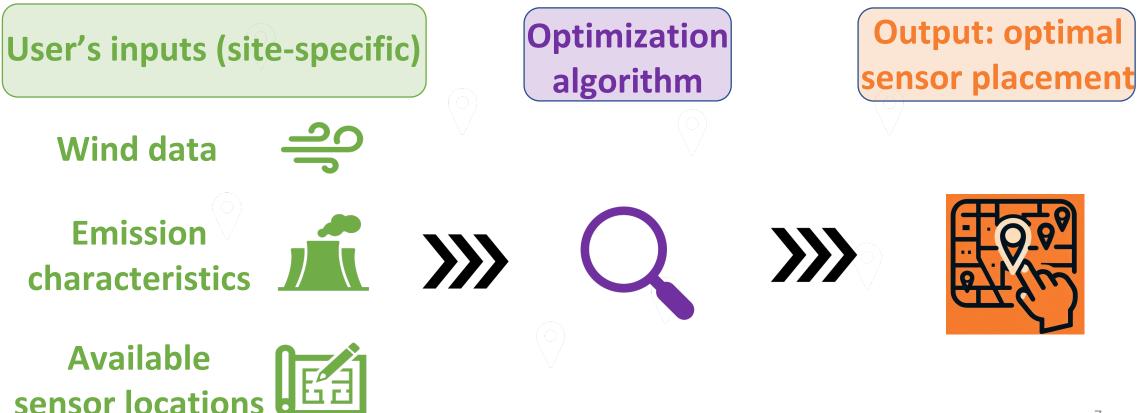
• CMS sensor placement



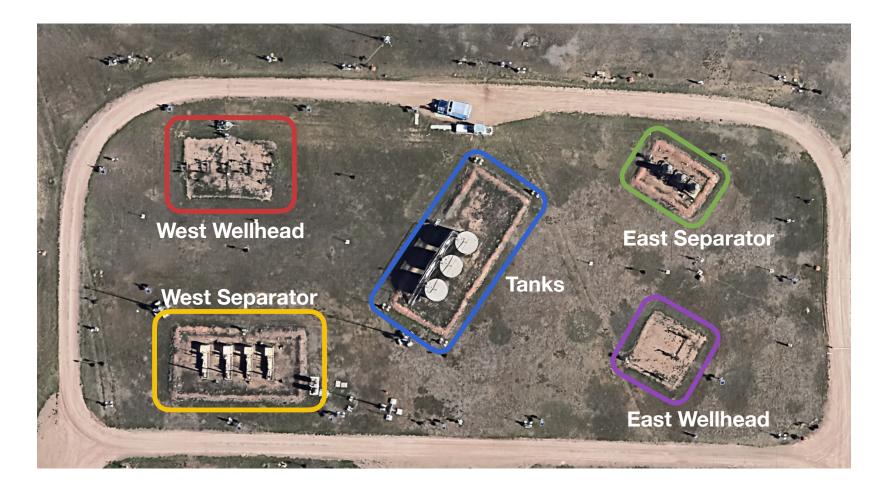
• CMS sensor placement



 A data-driven algorithm to optimize sensor placement for best emission detection



### **Experiment Data**



METEC facility, 5 potential emission sources

# Algorithm



Generate emission scenarios



Set possible sensor locations

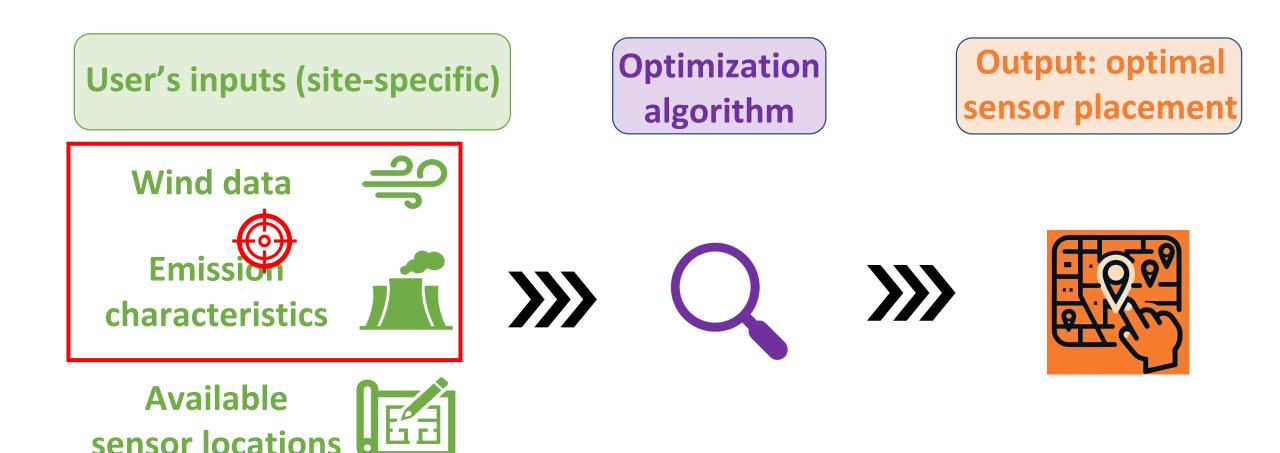


Simulate concentrations & Check detection



Optimize sensor placement

### Step 1 Generate Emission Scenarios

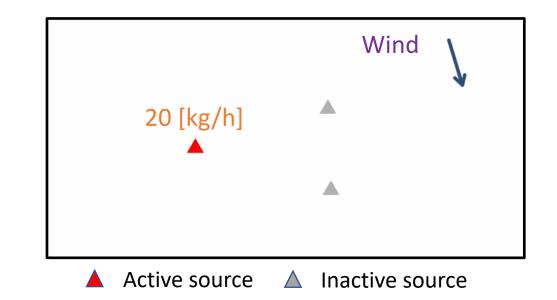


### Step 1 Generate Emission Scenarios

#### A combination of

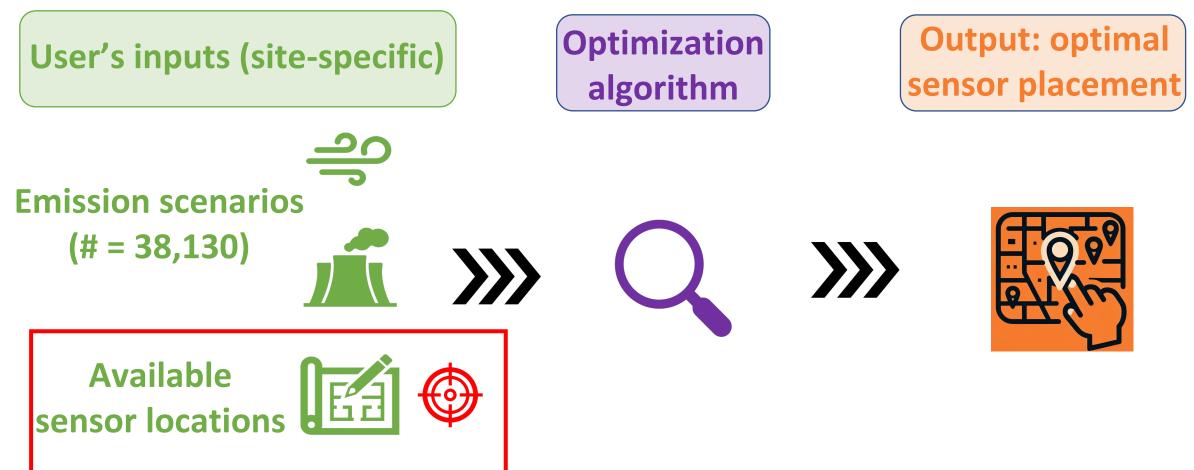
- wind speed time series
- wind direction time series
- emission source location
- emission rate

defines an emission scenario.

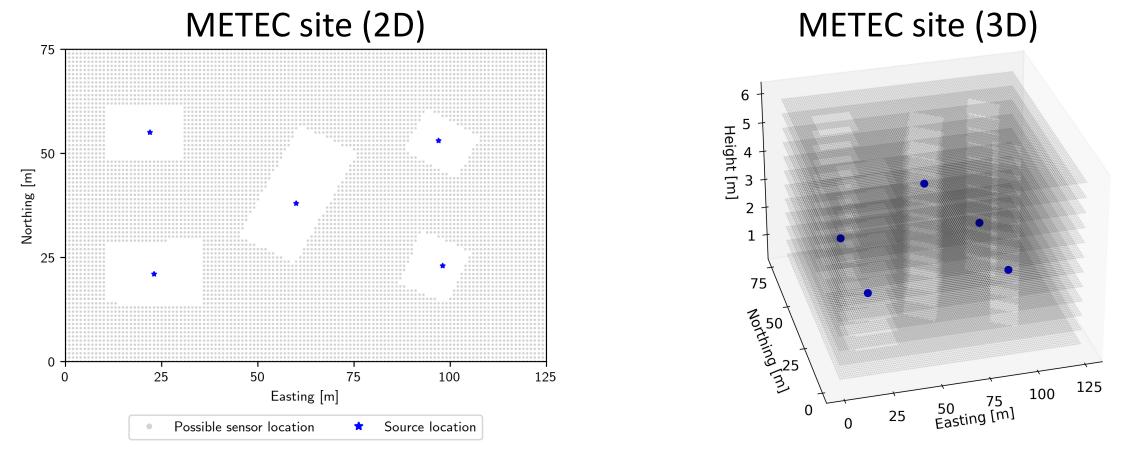


 Estimate probability distributions of wind & emission to sample → 38,130 emission scenarios

### Step 2 Set Possible Sensor Locations

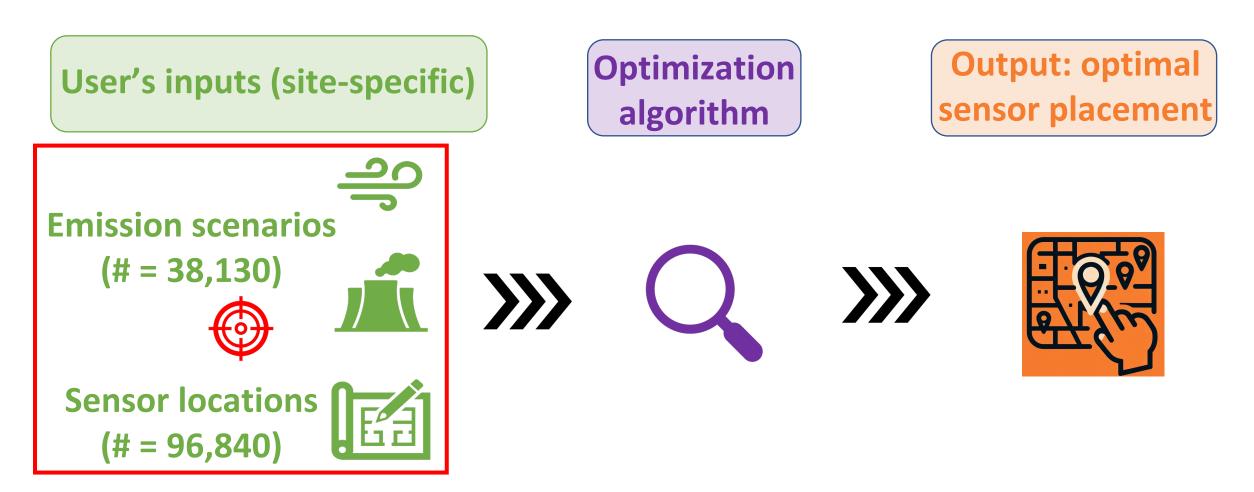


#### **Step 2 Possible Sensor Locations**

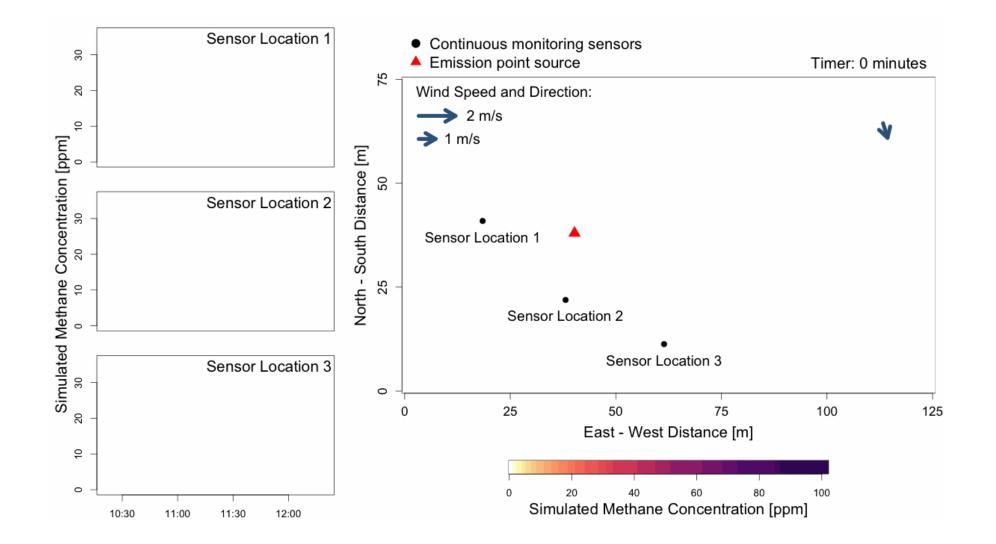


resolution = 1 m for Northing & Easting; = 0.5 m for vertical # possible locations = 96,840

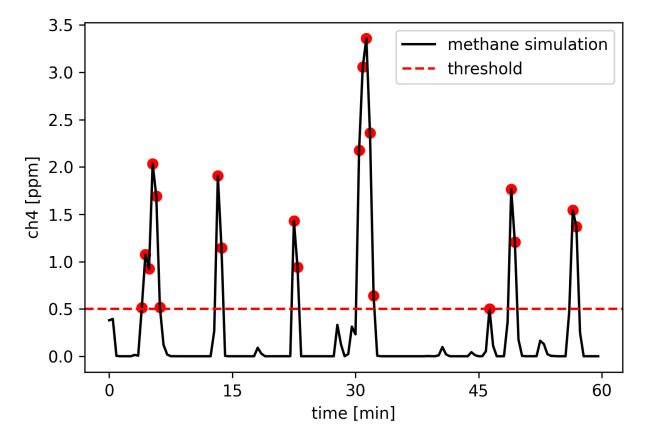
### **Step 3 Concentration Simulation & Detection**



#### Step 3.1 Gaussian puff simulation

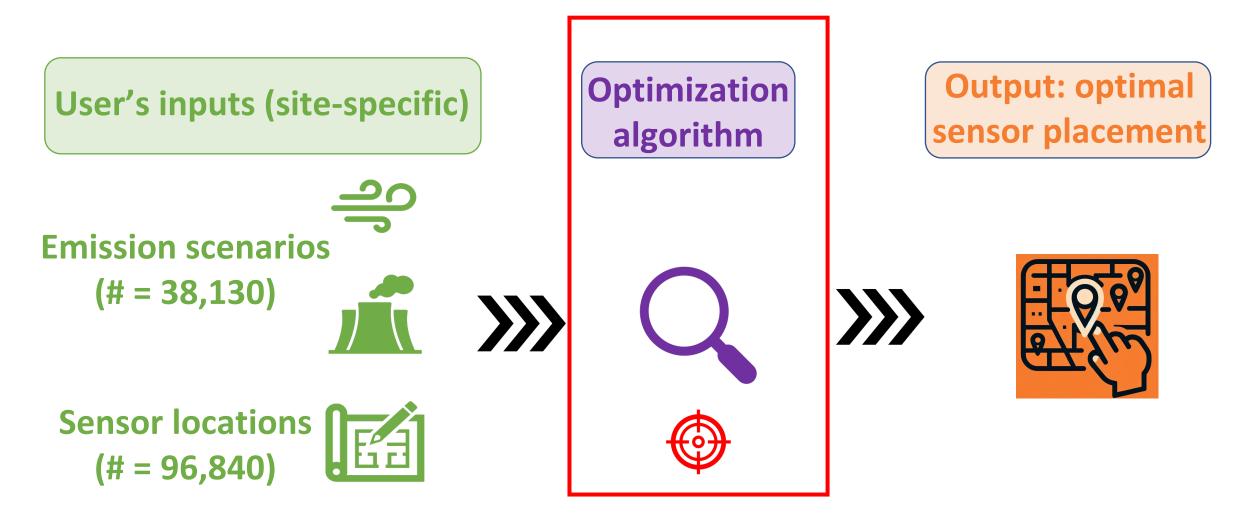


#### **Step 3.2 Detection**



Example of simulated concentrations and detection for Emission Scenario *j* at Sensor Location *i* 

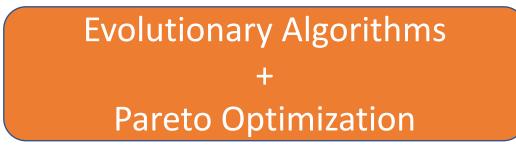
### Step 4 Optimize Sensor Placement

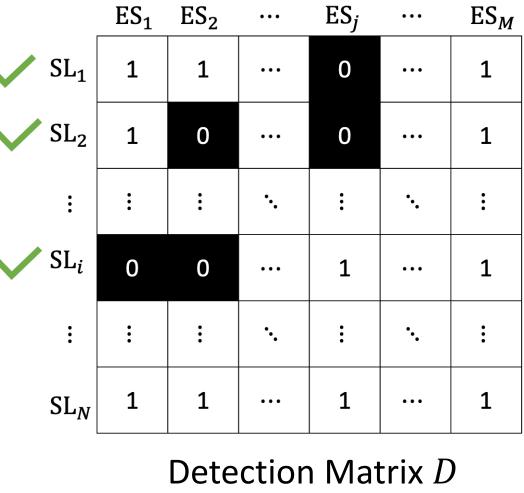


### **Step 4 Optimization**

Rows of *D*: Sensor Locations (SL) Cols of *D*: Emission Scenarios (ES)  $D_{ij} = 0$ , if SL<sub>i</sub> can detect ES<sub>j</sub>;

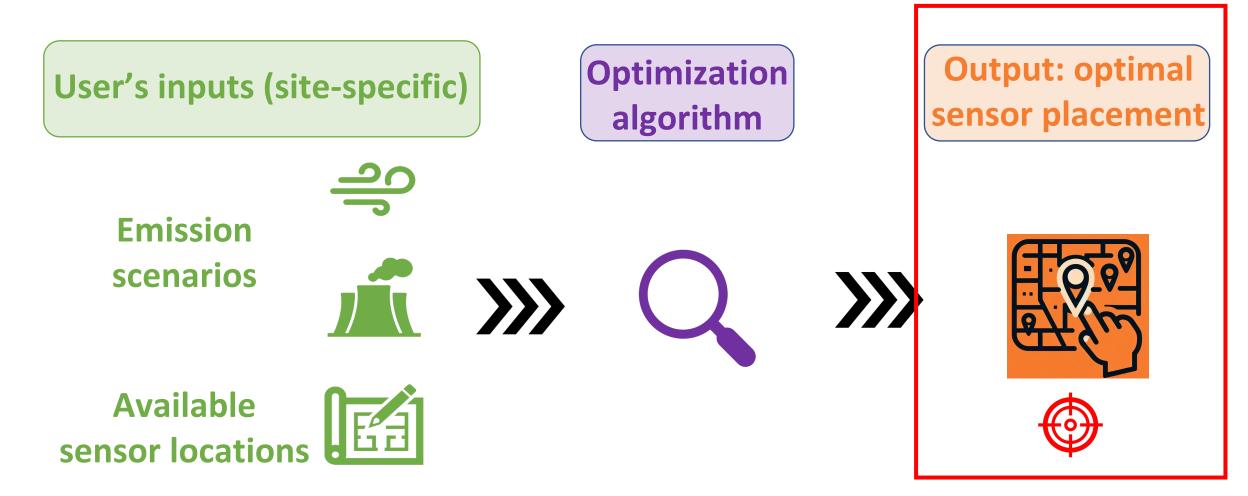
 $D_{ij} = 1$ , otherwise



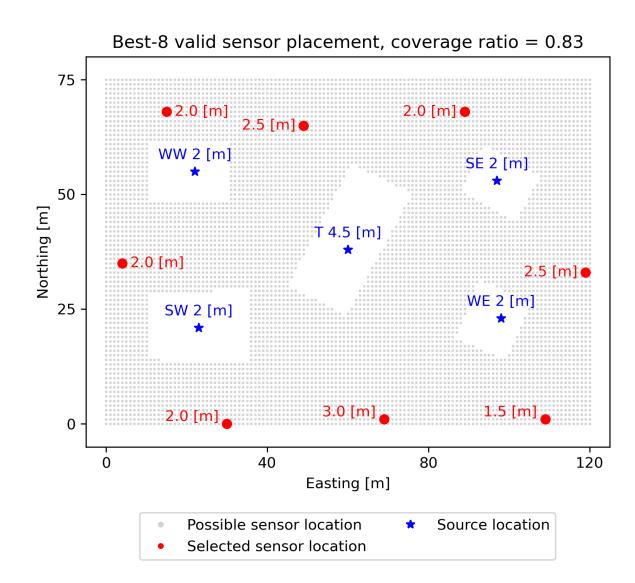


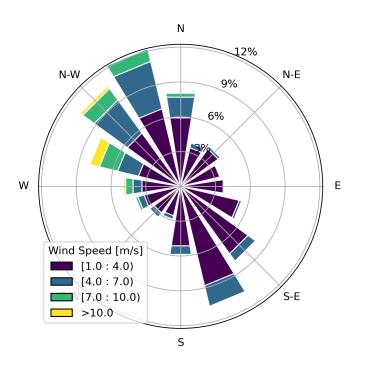
N = 96,840; *M* = 38,130

### Results

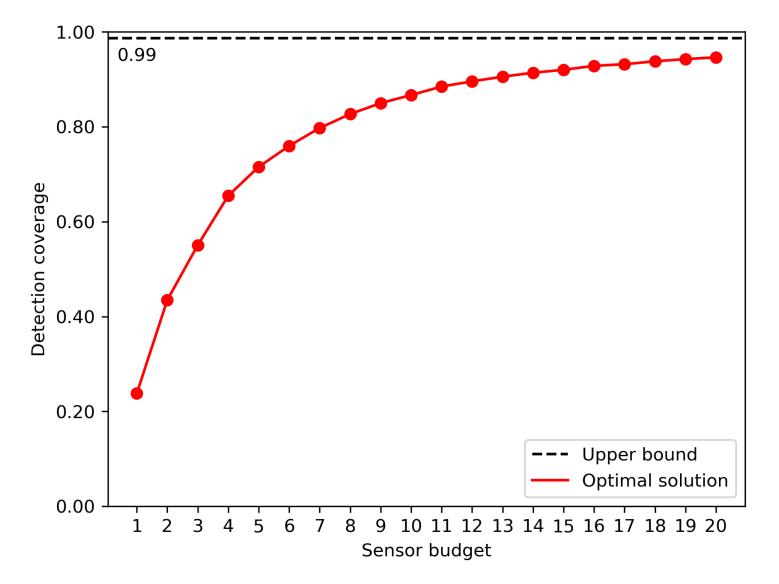


#### **Results:** Best 8-sensor placement





#### Results: Budget vs. coverage



### Summary

- We propose a data-driven algorithm for optimizing sensor placement on oil and gas sites.
- Our algorithm's high modularity allows users to incorporate their own methods and models.
- The algorithm can be expanded for wider applications such as emission localization and quantification by changing the objective functions.

### **Relevant Presentations in GRADS**

- Dynamic spatiotemporal thresholds for the Gaussian Puff atmospheric dispersion model using dynamic spatiotemporal thresholds Ryker Fish
- Exploring Optimal Continuous Monitoring Sensor Configurations on a Prototypical Midstream Oil and Gas Site – Troy Sorensen
- Estimating methane emission duration with continuous monitoring systems – William Daniels
- Sampling Frequency Strategies for Methane Emissions from Oil & Gas Olga Khaliukova
- Estimating Theoretical Error Distributions for Overflight Methane Measurements – Cal Richards-Dinger

Thank you for attending! Questions?







for more details!