

Source term estimation using unmanned autonomous vehicles

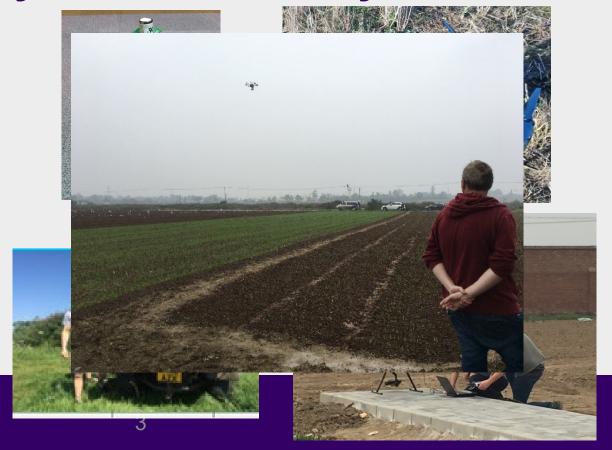
Michael Hutchinson, Dr Cunjia Liu, Prof. Wen-Hua Chen

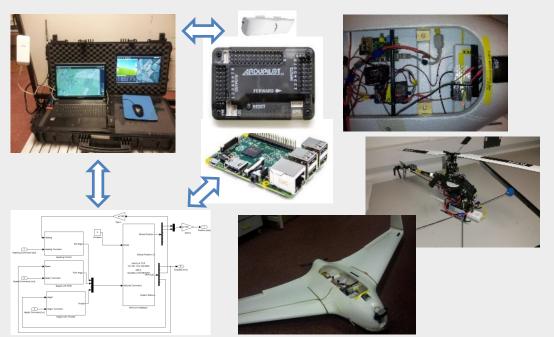












Ladosz, Pawel, Matthew Coombes, Jean Smith, and Michael Hutchinson.

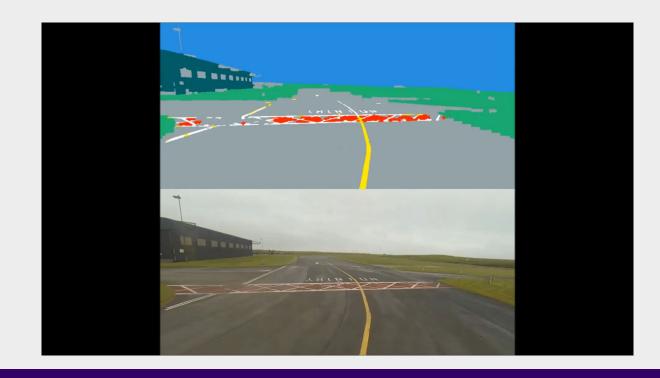
"A Generic ROS Based System for Rapid Development and Testing of Algorithms for Autonomous Ground and Aerial Vehicles."

In Robot Operating System (ROS), pp. 113-153. Springer, Cham, 2019.

Control of small helicopters in wind.



Towards autonomous terminal area operations for UAVs.



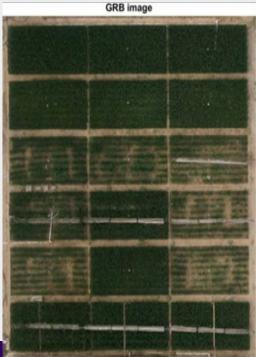
Ground target tracking and computer vision.



Agriculture



Classification map



Background





Loughborough
University

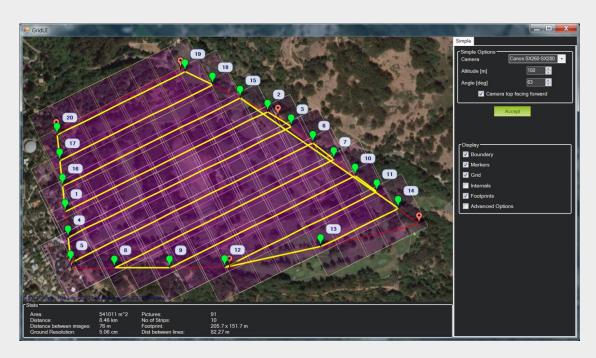


Wind – Sensing and handling

Surveying in wind.

Survey of the wind.







Wind – Sensing and handling

Ultrasonic anemometer mounted on UAV.

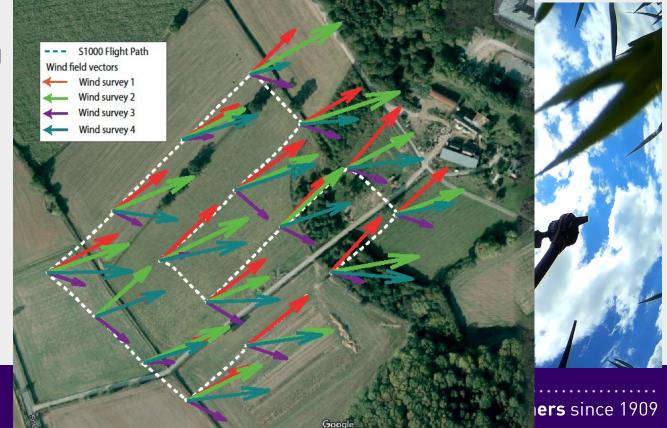
Stability test.

Effects on wind sensing?



Wind – Sensing and handling

Experiments measuring the wind field.





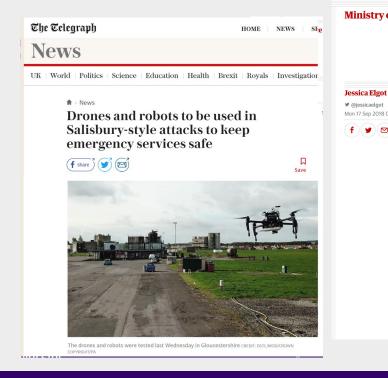
SOURCE TERM ESTIMATION USING A UAV

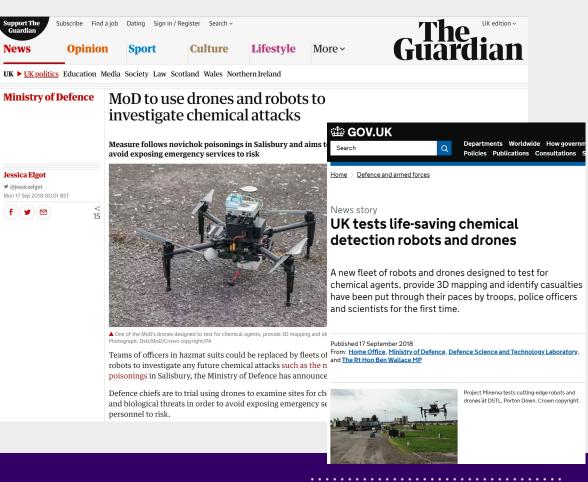


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News

Mon 17 Sep 2018 00.01 BST

INTRODUCTION AND MOTIVATION



Introduction

- Emergency HAZMAT event.
- Where is the HAZMAT coming from?
- Where has it spread?
- The <u>Source term</u>
 - Source location (x, y, z)
 - Emission rate (Q [grams/s])



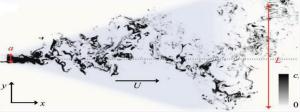




Background

 Finding an emitting source is not as easy as climbing concentration gradients!





- Sensor observations are noisy and sporadic, with more non-detections of hazard concentration then detections. – This is made worse with short sampling periods!
- A robust method is required to handle such poor sensing characteristics.

Estimation And Planning

Simple algorithm overview

ALGORITHM DETAILS



Conceptual solution - Estimation

- Bayes probability theory is used to update probability densities of the source term θ in response to new sensor data d.
- Source term Parameters include Meteorological uncertainty:

$$P(\theta|d,I) = \frac{P(\theta|d)P(d|\theta)}{P(d)}$$

$$\theta = [X_0 Y_0 A_0 U_0 \phi_0 D_0 \tau_0]^\mathsf{T}$$

Algorithm inputs (prior information):

X,Y location

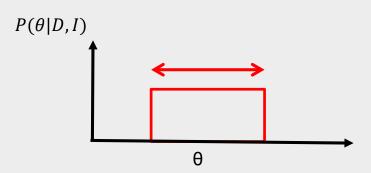
Emission rate

Wind speed

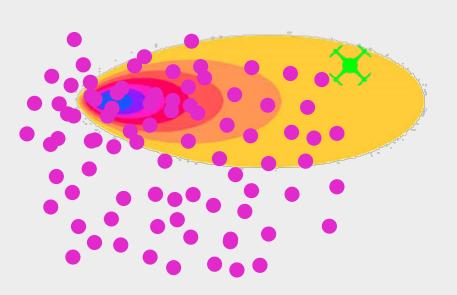
Wind direction

Diffusivity

Lifetime



Conceptual solution - Estimation



 Bayesian estimation is performed using a sequential Monte Carlo algorithm.

- Random samples (Pink dots) with low weights are resampled around highly weighted ones.
- One of the great challenges from experiments is in sensor modelling.

Modelling

Plume model

$$\mathcal{M}(\mathbf{p}_{k}, \Theta_{k}) = \frac{q_{s}}{4\pi\zeta_{s1}||\mathbf{p}_{k} - \mathbf{p}_{s}||} \exp\left[\frac{-||\mathbf{p}_{k} - \mathbf{p}_{s}||}{\lambda}\right] \times \exp\left[\frac{-(x_{k} - x_{s})u_{s}\cos\phi_{s}}{2\zeta_{s1}}\right] \exp\left[\frac{-(y_{k} - y_{s})u_{s}\sin\phi_{s}}{2\zeta_{s1}}\right].$$

Sensor model split

Detections
Non detections

$$p(z_k|\Theta_k) = \begin{cases} p(\overline{z}_k|\Theta_k) & \text{if } z_k > z_{thr} \\ p(\underline{z}_k|\Theta_k) & \text{if } z_k \le z_{thr} \end{cases}$$

Probability of detection

$$p(\underline{z}_k|\Theta_k) = \left(P_b \times \frac{1}{2} \left[1 + p(\underline{z}_k) \left(\frac{z_{thr} - \mu_b}{\sigma_b \sqrt{2}\sigma_k} \right) \right]_{\overline{\mathcal{I}}_{\pi}} + eR_{n} \left[+ \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{2} \right) \frac{1}{2} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{2} \right) \right]_{\sigma_k} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{2} \right) \right]_{\sigma_k} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{2} \right) \right]_{\sigma_k} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{\sigma_k \sqrt{2}} \right) \right]_{\sigma_k} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{\sigma_k \sqrt{2}} \right) \right]_{\sigma_k} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{\sigma_k \sqrt{2}} \right) \right]_{\sigma_k} + erf \left(\frac{\overline{z}_k - \mathcal{M}(\mathbf{p}_k, \Theta_k)}{\sigma_k \sqrt{2}} \right) \right)$$

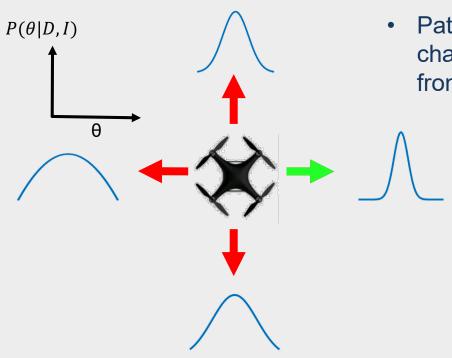
Background noise

Intermittency or missed detection

Below sensor threshold



Conceptual solution - Planning



 Path planning is based on the expected change in what we know about the source from a manoeuvre.

- Consider new locations to take a measurement from
- At each position:
 - Given what we know
 - Estimate what we expect to see
 - Approximate what I expect to learn (KL divergence)
- Move to most informative choice



Indoor experiments with a ground robot

EXPERIMENTS

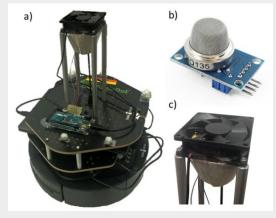


Indoor experiment



Smoke source









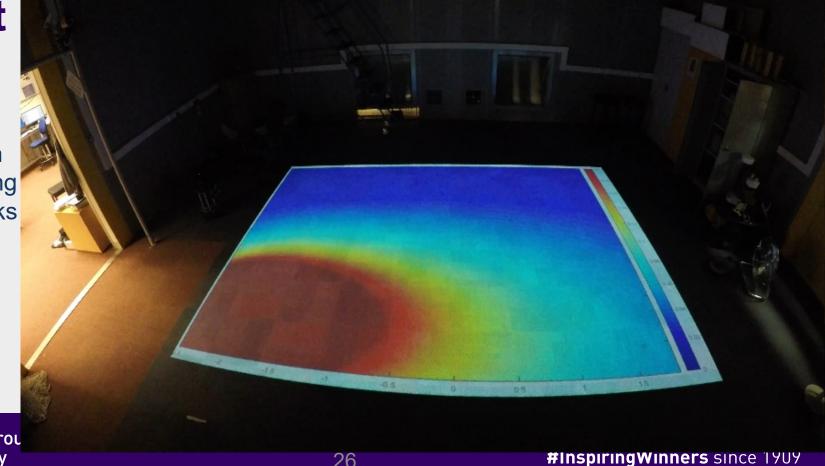
Indoor experiment





Result

Example run with 2 burning incense sticks



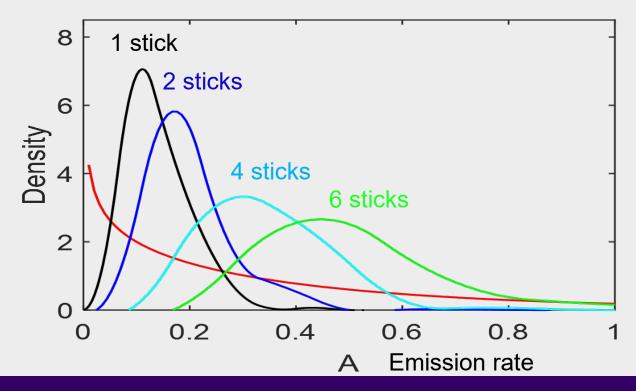
Result

Example run with 6 burning incense sticks.

Search is much more effective, towards the source.



Result – Emission rate estimates





Summary – Indoor experiments with a ground robot

- First STE experiments performed using an autonomous robot with an information based planning algorithm.
- The system was very robust and accurate.
- Next step was to extend and test the system in an outdoor environment using a UAV.

Experiments with a UAV

EXPERIMENTS - SYSTEM





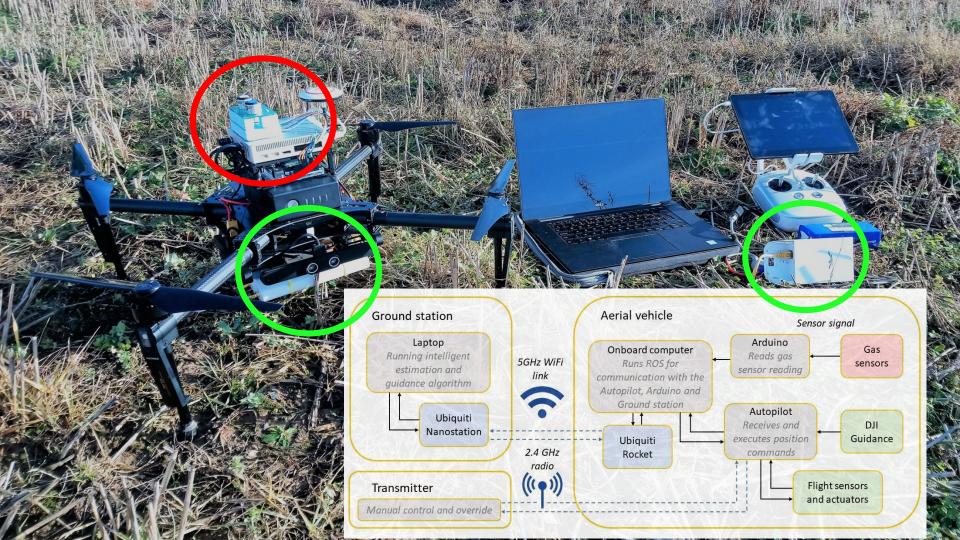


System overview

	PID-A12	PID-AH2
Power (max)	85mW	85mW
Weight	<8g	<8g
Min conc (isobutylene)	50ppb	1ppb
Conc range	6000ppm	50ppm
Material detected	Anything ionised by a 10.6eV UV lamp (see next slide)	







Outdoor experiments with a UAV

EXPERIMENTS - SETUP



Experiment setup

Acetone source, release roughly 1.5grams/s



Search area, starting position, source position and wind direction

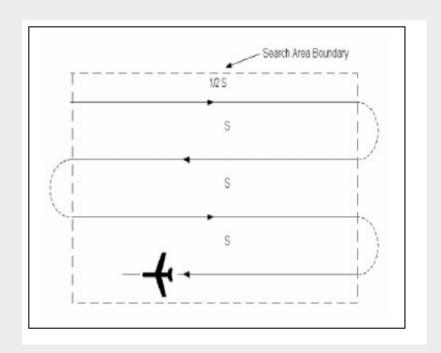


Outdoor experiments with a UAV

EXPERIMENTS – UNIFORM FLIGHT PATTERN

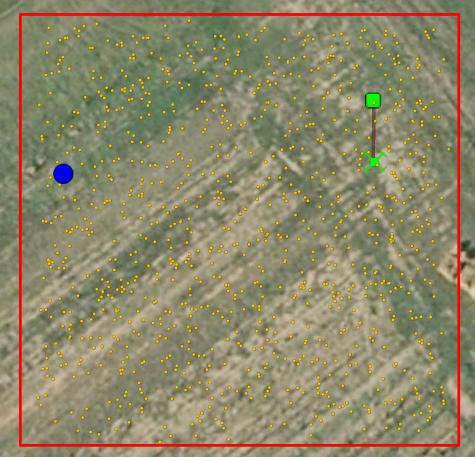


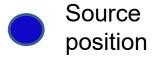
Experiments – Uniform flight pattern





Example 1







UAV position

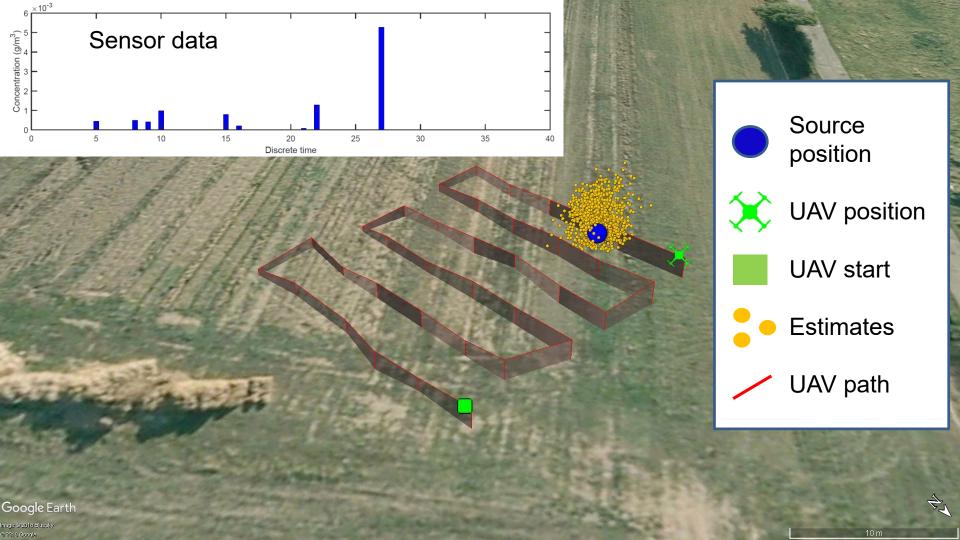


UAV start

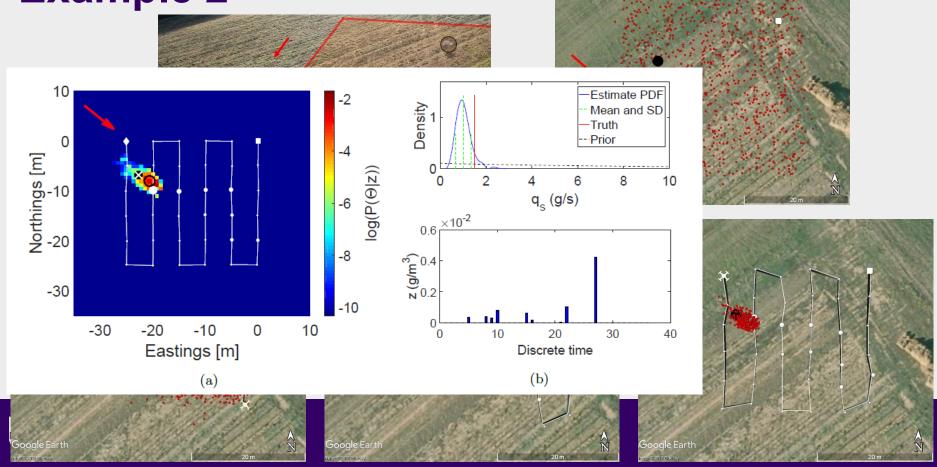


Estimates





Example 2



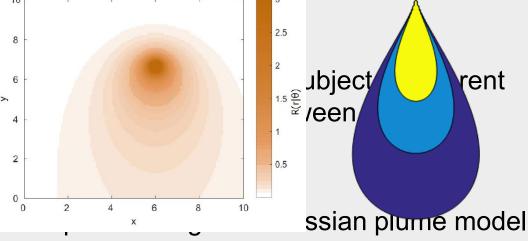
Results

We performed 27 STE experiments using a uniform sweep flight

pattern

• The performance of the flying altitudes, wind sp a measurements.

• The STE accuracy was and the Isotropic plume model.

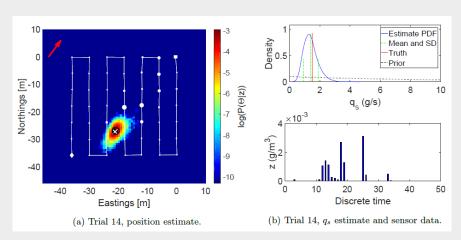




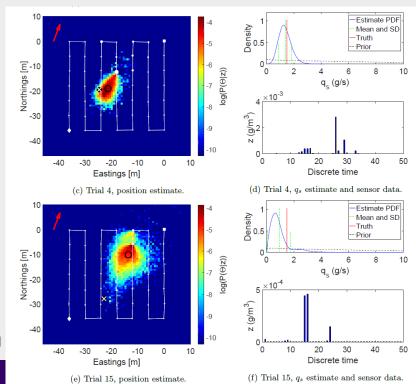
Results – Different altitudes

Altitude = 1.75m





Altitude = 1.4m

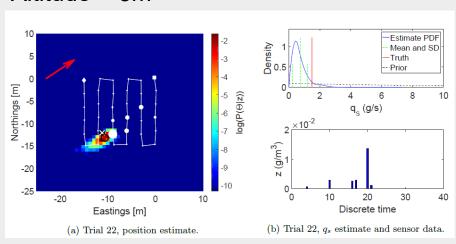




Results – Different step sizes during flight

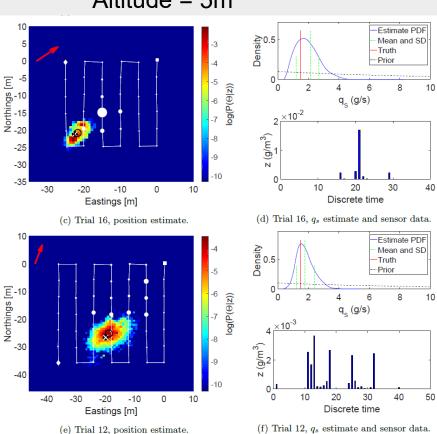
Altitude = 5m





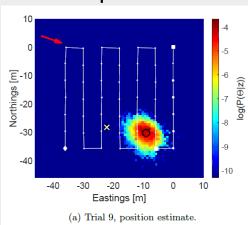
Altitude = 6m

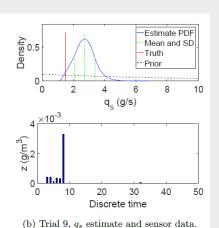




Results – Different wind speeds Wind speed = 5m/s

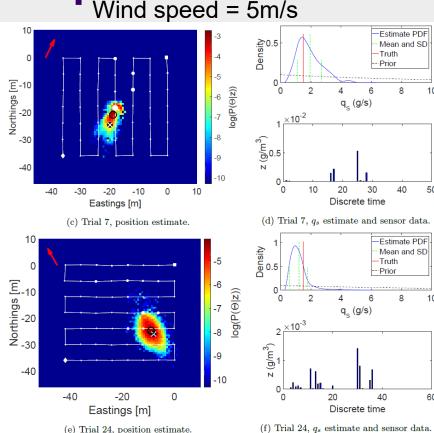
Wind speed = 3m/s





Wind speed = 7m/s





Results – Gaussian plume vs Isotropic plume model

Table 2: RMSE in the position and emission rate estimates using the GP and IP dispersion models.

	Position RMSE [m]		Emission RMSE $[g/s]$	
Data subset	GP model	IP model	GP model	IP model
All data	4.75	3.35	0.75	0.65
Step size $= 6$ m	5.37	4.08	0.77	0.63
Step size $= 5$ m	3.38	2.69	0.61	0.73
Step size = 3m	3.55	1.72	0.70	0.66
Height ≥1.5m	5.22	3.33	0.95	0.66
Height <1.5m	3.88	2.45	0.67	0.69
Wind speed ≤3m/s	6.45	4.65	0.85	0.62
Wind speed >3m/s	3.86	2.63	0.69	0.66

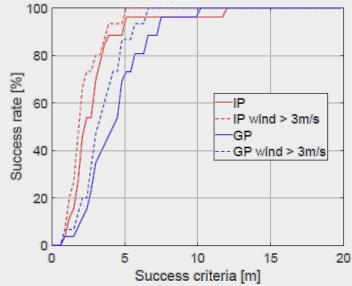


Figure 13: Success rates of the system.



Summary – Uniform sweep STE

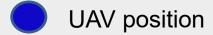
- Our first successful STE experiments performed in outdoor conditions using a UAV.
- The system performed best flying at lower altitudes in higher wind speeds (further inside the plume in a more stable atmosphere).
- Next step was to test the information based planning algorithm.

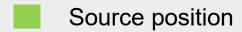
Outdoor experiments with a UAV

EXPERIMENTS – INFORMATIVE SEARCH ALGORITHM

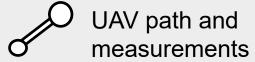


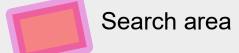
Video

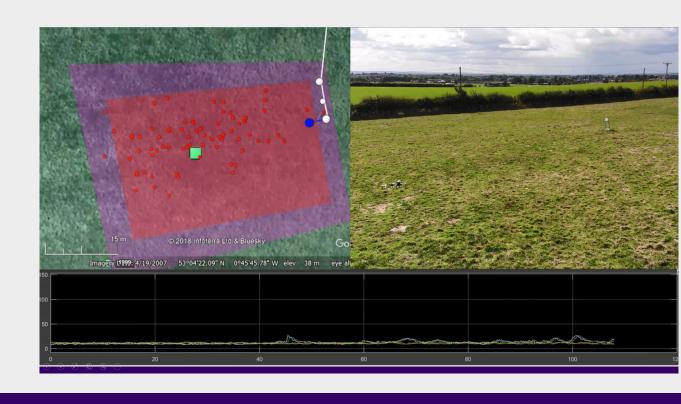




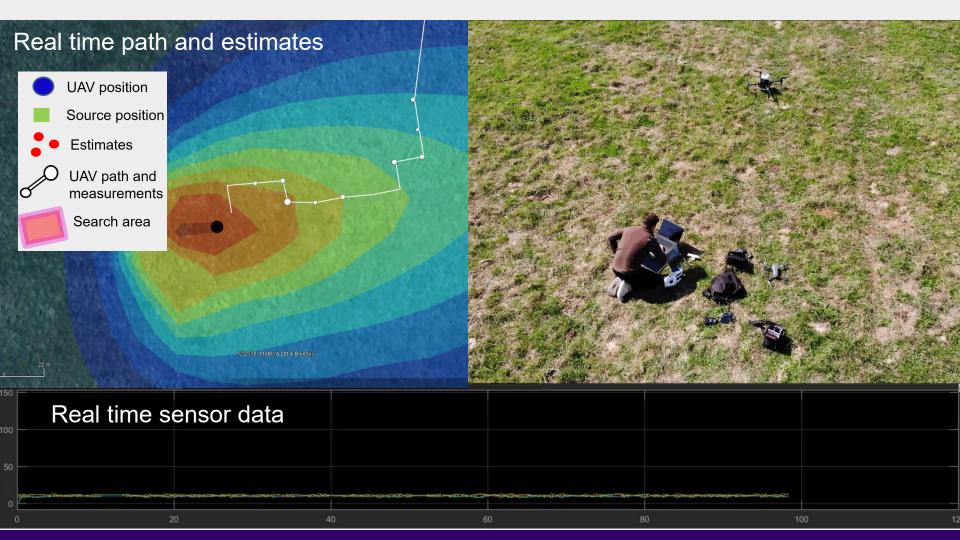


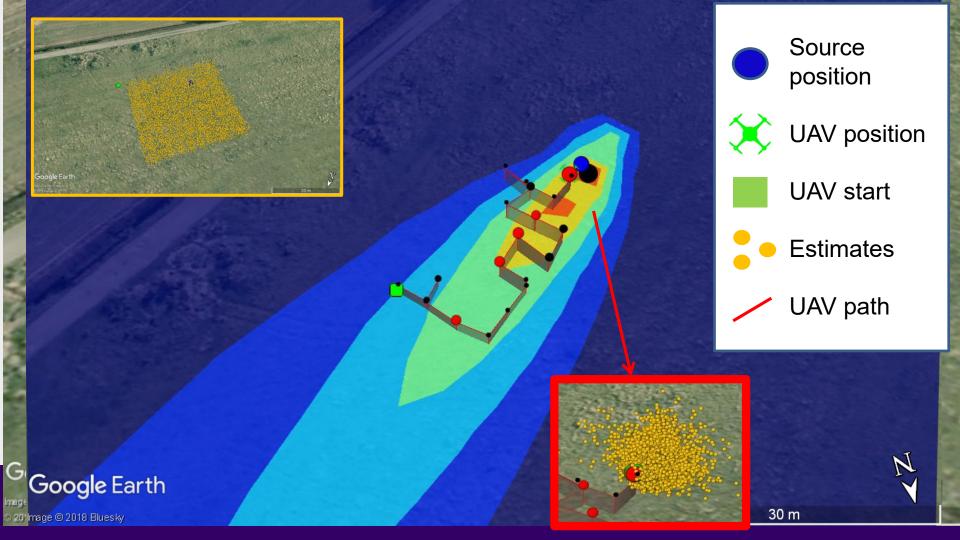




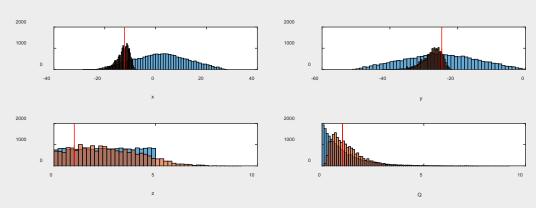


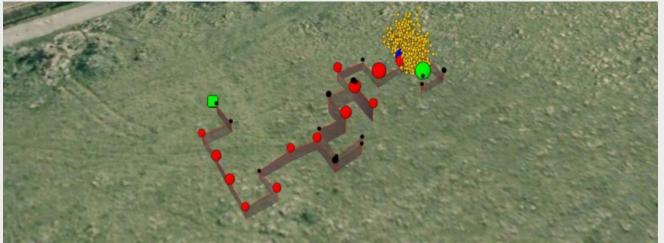










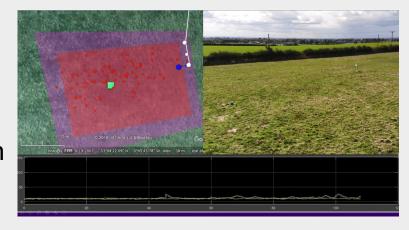




Informative Search Results

 Similarly to the sweep, accuracy is relative to the step size between measurements.

 Estimates the source term in less than half the time of the sweep (on average).



Struggles in changing wind conditions due to model assumptions.

Summary

- Source term estimation using mobile sensors has been taken from purely theoretical work to real world experiments.
- We have shown that it is a viable method of response to HAZMAT incidents.

 Next steps will continue to extend the system to work in different environments and improve its performance.

Next steps

- Extend to handle changing wind conditions.
- Handle cluttered or urban environments (plan informative paths around obstacles)
- Plan more informative paths considering measurements in the future and whilst moving.
- Multiple vehicle coordination for STE and large area reconnaissance and survey for HAZMAT.
- Multiple source STE using UAV.
- Mapping using multiple UAVs
 - Include physics and account for non-detections

