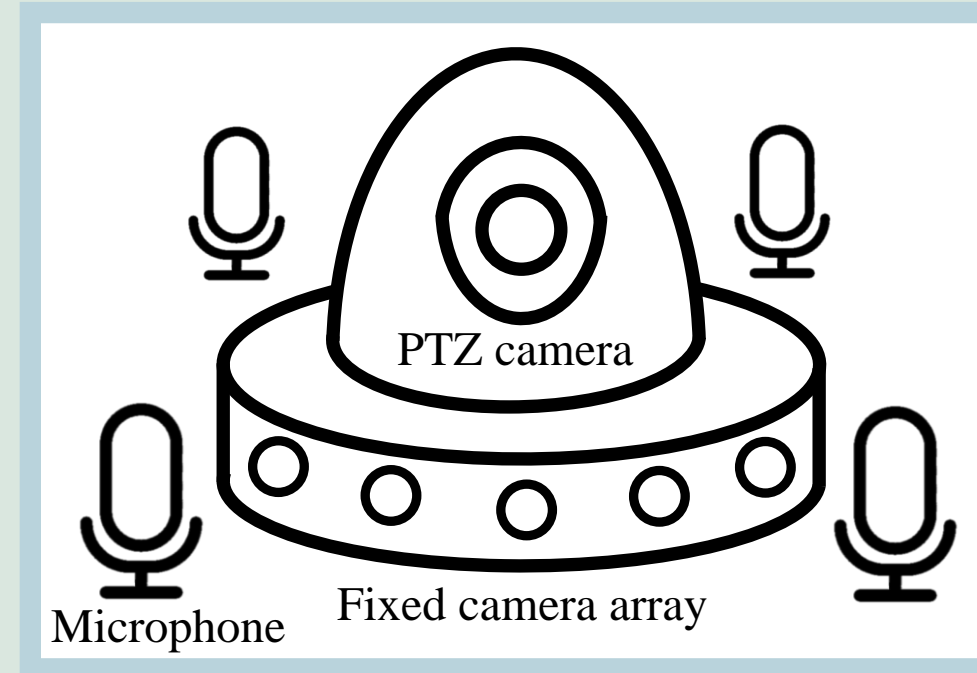


## Contribution

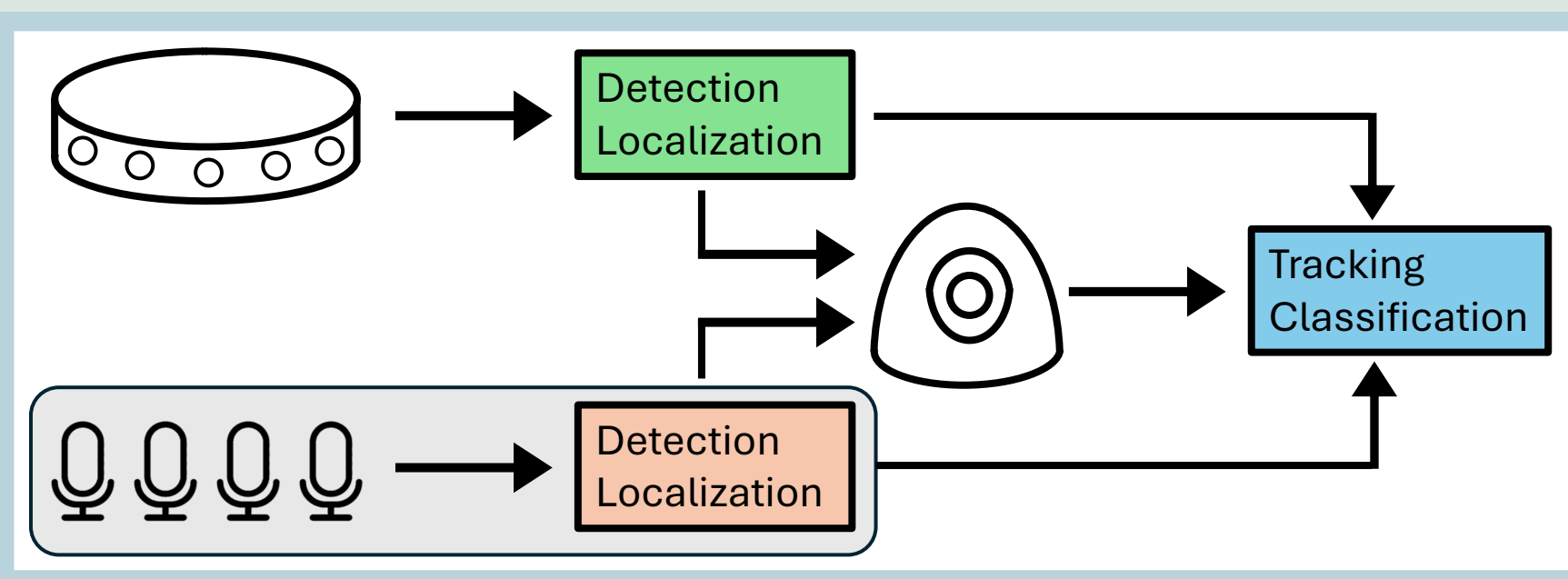
- A robust multi-channel algorithm for locating drones relative to an ad-hoc microphone array.
- A new dataset with multi-channel audio recordings, with ground truth position of drones.
- Evaluation both against ground truth and against localizations by human observers.



## Background

- Detecting drones is important in many settings. Audio and video are both passive detection technologies, which can be advantageous.
- Our system works with a combination of spatially distributed microphones, an array of fixed cameras and a high resolution PTZ camera.
- Here, we focus on detection and coarse localization using audio.
- The localization can then be refined using the PTZ camera.

## Setup



This work has been supported by Sweden's Innovation Agency VINNOVA and the Swedish Armed Forces within the programme "Civil-Militärt Innovationsprogram", project nr 2024-03191, the strategic research project ELLIT and the Wallenberg AI, Autonomous Systems and Software Program (WASP), funded by the Knut and Alice Wallenberg (KAW) Foundation.

## Method

- We have  $i = 1 \dots 12$  known microphone positions  $\mathbf{r}_i$  and received sound signals  $\mathbf{s}_i^{(k)}(t)$  coming from  $k$  unknown sound source positions  $\mathbf{x}^{(k)}$ .
- To find the DOA of the sound source positions, compare TDOAs estimated from sound signals  $\mathbf{s}_i^{(k)}$ 

$$\tilde{R}_{\mathbf{s}_\kappa \mathbf{s}_i}^{(k)}(T) = \mathcal{F}^{-1} \left( \frac{\mathcal{S}_\kappa^{(k)}(f) \bar{\mathcal{S}}_i^{(k)}(f)}{|\mathcal{S}_\kappa^{(k)}(f) \bar{\mathcal{S}}_i^{(k)}(f)| + \varepsilon} \right)$$

$$\tilde{\tau}_{\kappa i}^{(k)} = \arg \max_T \tilde{R}_{\mathbf{s}_\kappa \mathbf{s}_i}^{(k)}(T)$$
to DOA candidates for different angles
$$\tau_{\kappa i}(\theta, \phi) = \frac{[\cos(\theta) \cos(\phi), \sin(\theta) \cos(\phi), \sin(\phi)](\mathbf{r}_i - \mathbf{r}_\kappa)}{c}$$
- Two models: 1) only azimuth angle, 2) azimuth and elevation angle.
- Done for every sound segment, leading to a track of the drone movement.

### Algorithm 1 Find Direction of Arrival (DOA)

**Input:** Microphones  $\mathbf{r}_1, \dots, \mathbf{r}_N$ , sounds  $\mathbf{s}_1, \dots, \mathbf{s}_N$

**Output:** DOA estimates  $\hat{\theta}^{(k)}$  (or  $\hat{\theta}^{(k)}$  and  $\hat{\phi}^{(k)}$ )

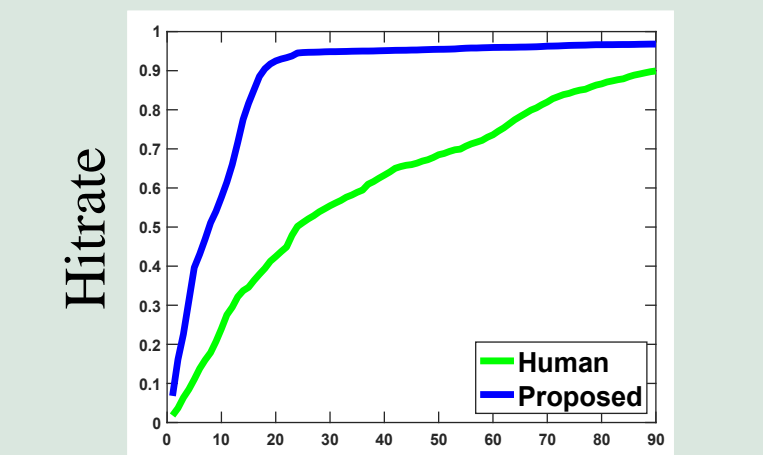
- 1: Divide sound signals  $\mathbf{s}_i$  in 0.1 s long segments  $\mathbf{s}_i^{(k)}$
- 2: **for** each  $k$  **do**
- 3:   Compute the GCC-PHAT  $\tilde{R}_{\mathbf{s}_\kappa \mathbf{s}_i}^{(k)}(T)$
- 4:   Find measured TDOAs  $\tilde{\tau}_{\kappa i}^{(k)}$
- 5: **end for**
- 6: Update  $\tilde{\tau}_{\kappa i}^{(k)}$  to be a moving median of  $\tilde{\tau}_{\kappa i}^{(k)}$
- 7: Grid space in azimuth  $\theta$  for Model 1 (or  $\theta$  and elevation  $\phi$  for Model 2)
- 8: **for** each angle  $\theta$  (or angle combination  $(\theta, \phi)$ ) **do**
- 9:   Compute angle dependent TDOA  $\tau_{\kappa i}(\theta)$  (or  $\tau_{\kappa i}(\theta, \phi)$ )
- 10: **end for**
- 11: **for** each  $k$  **do**
- 12:   Estimate DOA by comparing  $\tilde{\tau}_{\kappa i}^{(k)}$  and  $\tau_{\kappa i}(\theta)$  (or  $\tau_{\kappa i}(\theta, \phi)$ )
- 13: **end for**

## Results

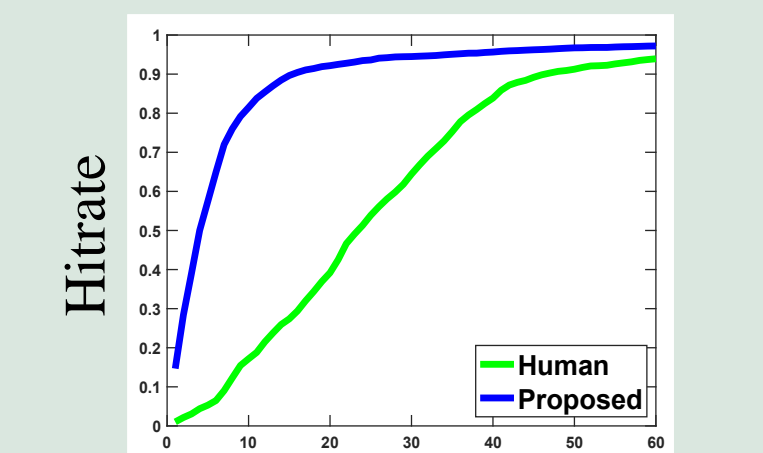
The red rectangles show estimated angles from our system



Comparison between our system and a human listener

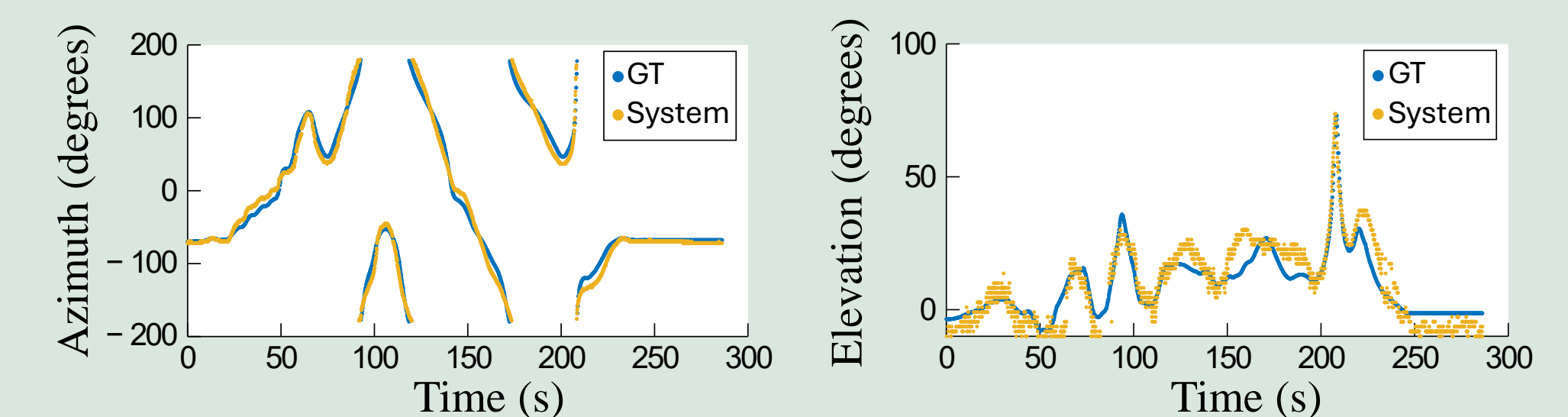


Azimuth error (degrees)



Elevation error (degrees)

Comparison between our system and ground truth on one test sequence



## Conclusion

- A drone detection and tracking system based on multimodal sensor input.
- Azimuth and elevation angles can be estimated reasonably well.
- Works up to several hundred meters distance.
- A new dataset with sound and GT positions released.
- Future work: integrate coarse localization from sound with the camera array and then use the PTZ camera for fine localization.



Paper  
Code  
Dataset