



# Onboard Ranging-based Relative Localization and Stability for Lightweight Aerial Swarms

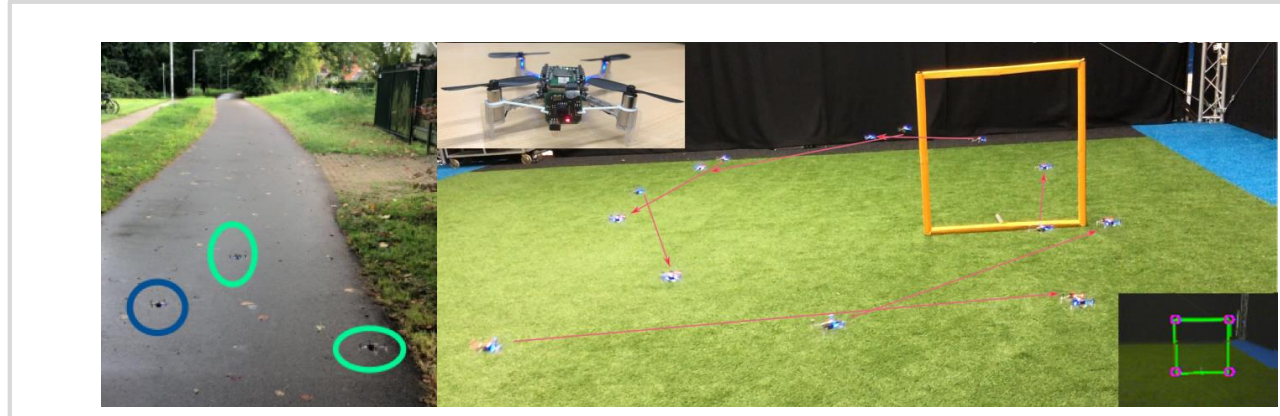
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## 1. Key Contributions

- First fully autonomous onboard relative localization on 13 lightweight aerial vehicles (33 g, 168 MHz MCU, 192 KB memory).
- UWB swarm ranging supports simultaneous communication and ranging without infrastructure.
- Self-regulated convergence: unobservable formation states become observable again through drift and control actions.
- Open-source implementation for off-the-shelf Crazyflie quadrotors.

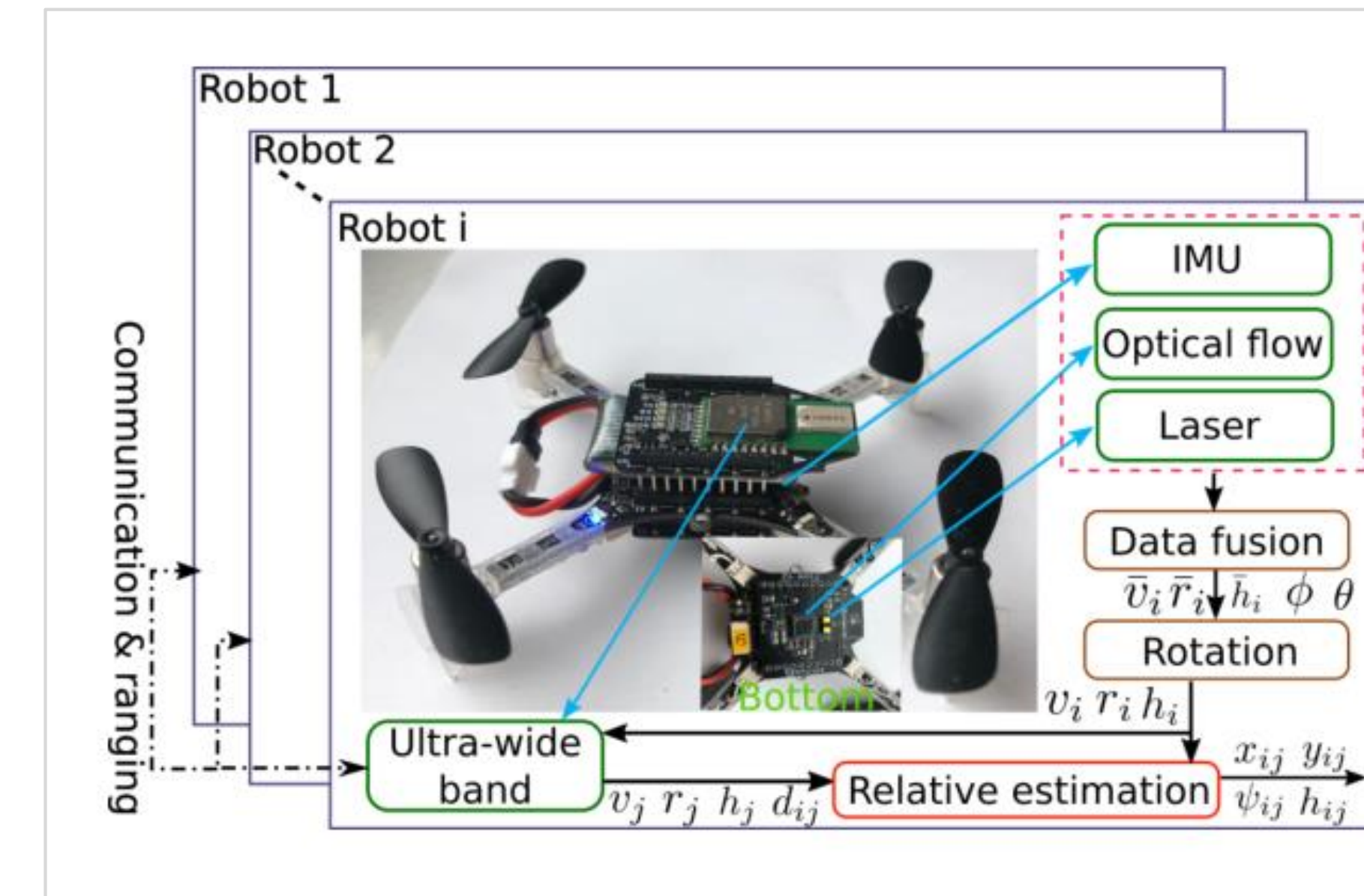
**Tiny + Onboard + Scalable**



Outdoor formation flight & indoor leader follower flight

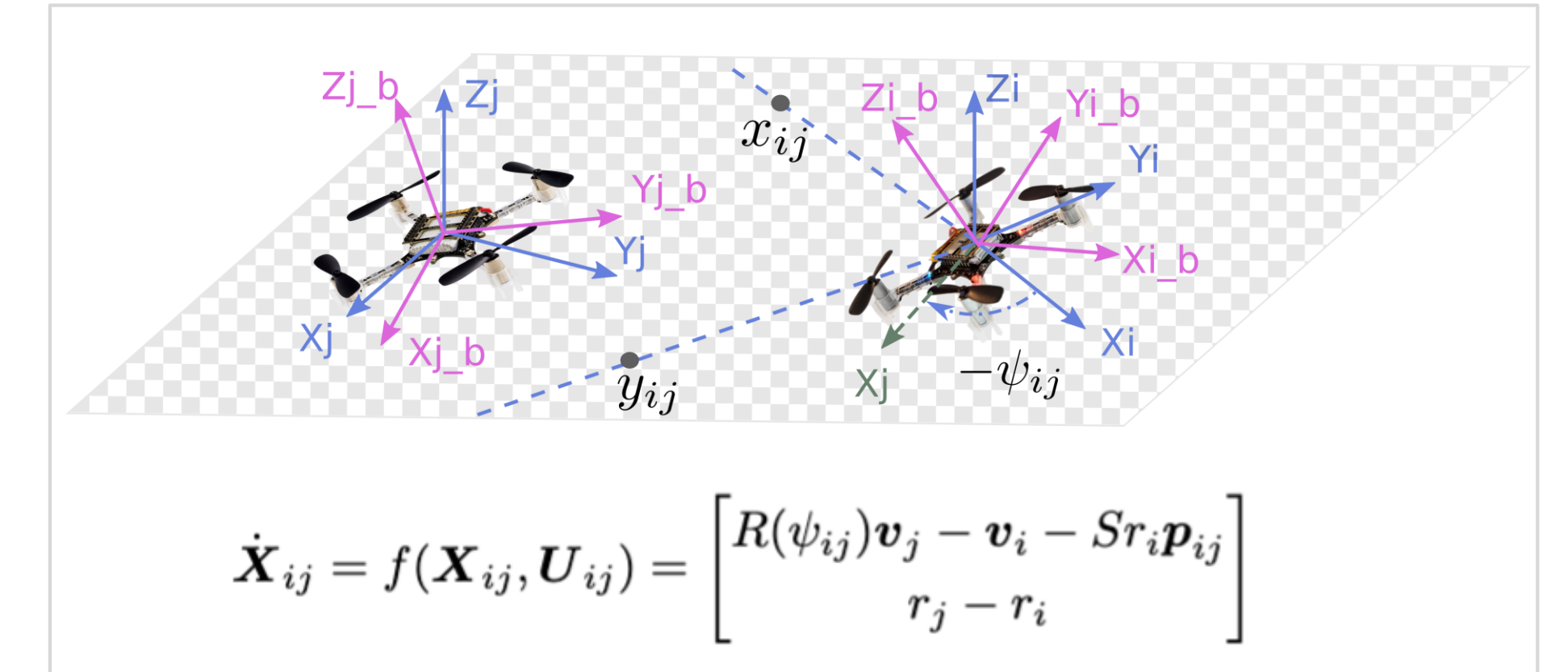
Video & code

## 2. System Overview



### Pipeline

- Sensors: IMU, optical flow, laser height, UWB.
- Shared state: velocity, yaw rate, height.
- Output: ego-centered relative pose  $X_{ij} = [x_{ij}, y_{ij}, \psi_{ij}]^T$ .
- Estimator runs fully onboard on resource-constrained MCU.



Relative kinematic model

**16 Hz Ranging/localization frequency**

**13-robot swarm in real world**

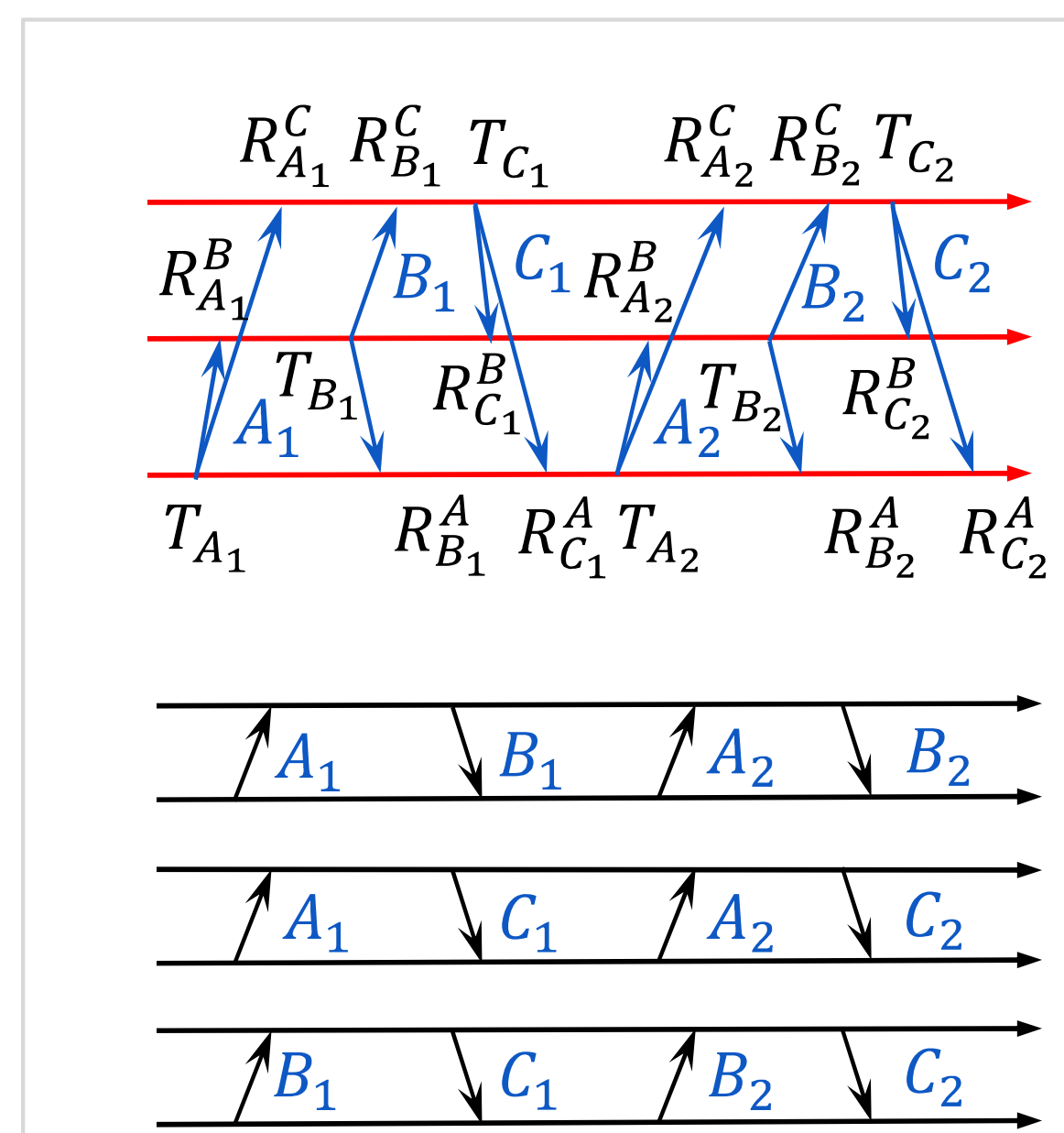
**<0.2 m Position error**

**33 g Drone mass**

**192 KB MCU memory**

## 3. Method: UWB Swarm Ranging + Tiny EKF

### Ranging protocol



An illustration of the swarm ranging protocol

STEP	Ranging Table			
Rearrange the Table	$R_p = R_{A_1}$	$T_r =$	$R_f =$	$R_e =$
A Transmits $A_2$	$T_p = T_{A_1}$	$R_r = R_{Y_1}$	$T_f = T_{A_2}$	$R_e =$
A Receives $Y_2$	$R_p = R_{A_1}$	$T_r = T_{Y_1}$	$R_f = R_{A_2}$	$R_e = R_{Y_2}$
Compute Distance	$R_p = R_{A_1}$	$T_r = T_{Y_1}$	$R_f = R_{A_2}$	$R_e = R_{Y_2}$
Rearrange the Table and A Transmits $A_3$	$R_p = R_{A_2}$	$T_r =$	$R_f =$	$R_e =$
A Transmits $A_3$	$T_p = T_{A_2}$	$R_r = R_{Y_2}$	$T_f = T_{A_3}$	$R_e =$
A Receives $Y_3$	$R_p = R_{A_2}$	$T_r = T_{Y_2}$	$R_f = R_{A_3}$	$R_e = R_{Y_3}$
Compute Distance	$R_p = R_{A_2}$	$T_r = T_{Y_2}$	$R_f = R_{A_3}$	$R_e = R_{Y_3}$
Rearrange the Table and A Transmits $A_4$	$R_p = R_{A_3}$	$T_r =$	$R_f =$	$R_e =$
A Transmits $A_4$	$T_p = T_{A_3}$	$R_r = R_{Y_3}$	$T_f = T_{A_4}$	$R_e =$

Swarm ranging steps

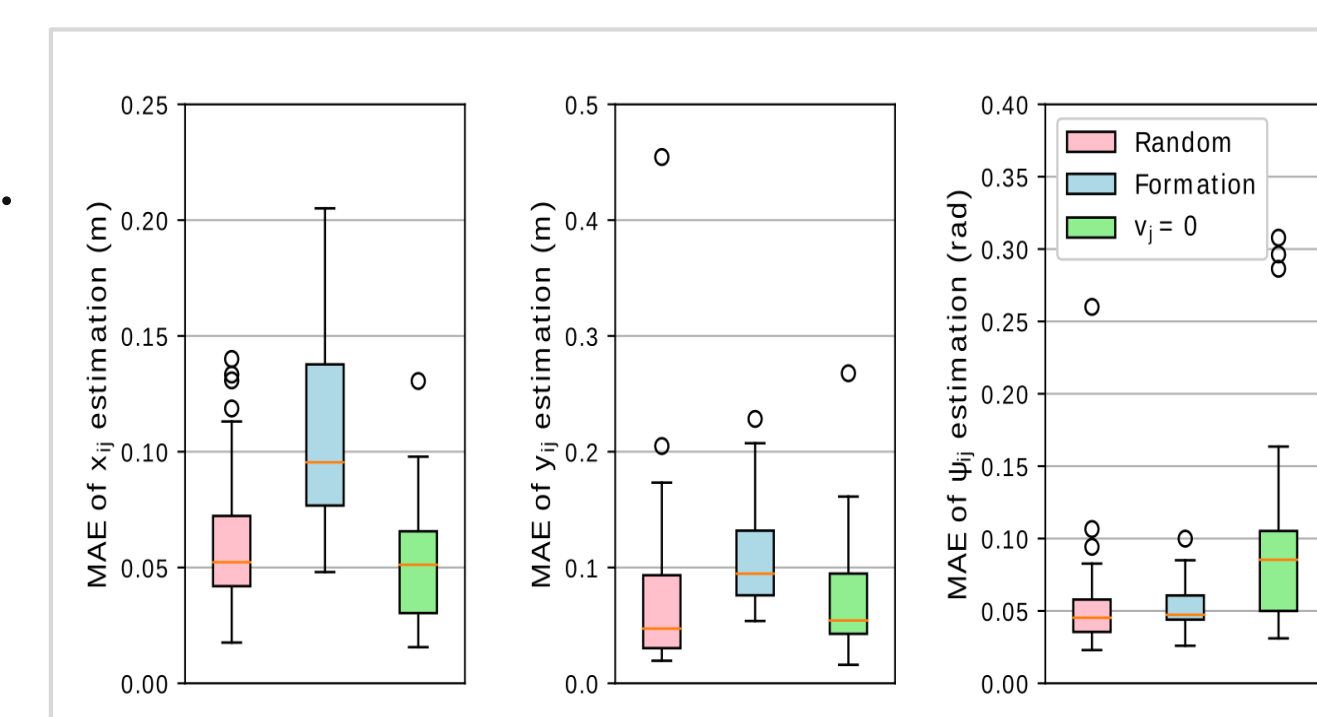
- One broadcast ranging message, not pairwise token passing.
- Each packet carries sender ID, sequence number, transmit timestamp, neighbor reception timestamps, velocity, yaw rate and height.
- Six timestamps recover ToF; a per-neighbor ranging table keeps the state compact.

### Relative localization

Type of data	Data of message
Onboard Sensing Data	Message Identification
	Flight Velocity $v_i$   Yaw Rate $r_i$   Flight Height $h_i$
	Last transmission Timestamp $T_i$
Ranging Data for Neighbor 1	Sequence Number   Reception Timestamp $R_1^i$
...	...
Ranging Data for Neighbor N	Sequence Number   Reception Timestamp $R_N^i$

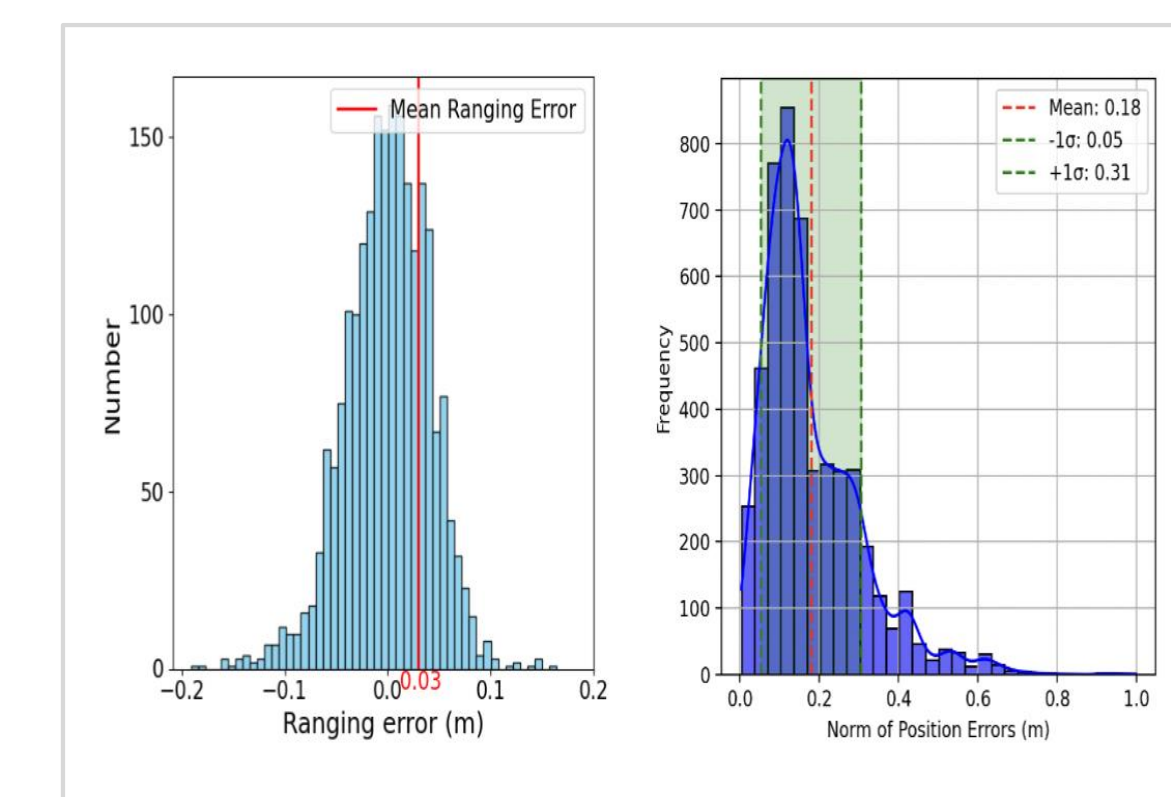
The proposed ranging message that supports relative localization.

- EKF is selected for low compute overhead.
- Prediction: relative kinematics driven by both robots' velocities and yaw rates.
- Update: UWB range  $z = \sqrt{(x^2 + y^2 + \Delta h^2)}$ .
- Stochastic initialization removes the need for manual initial relative poses.



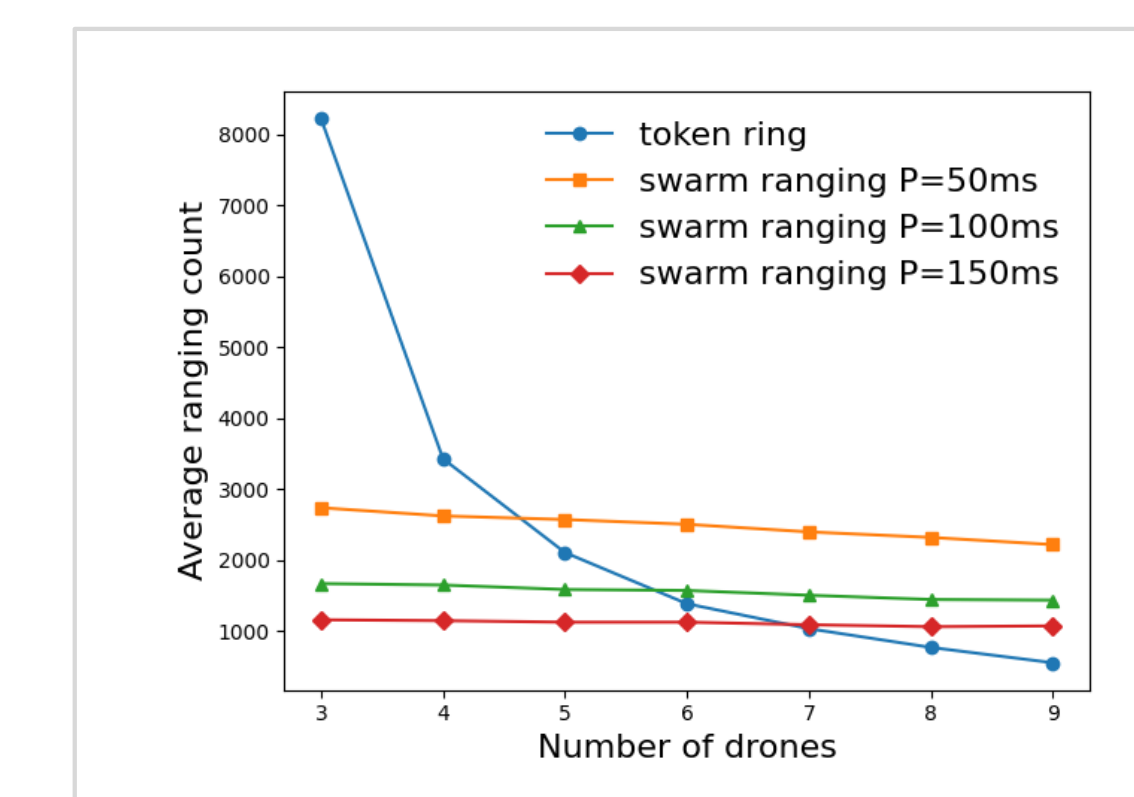
Error distribution of relative localization in different unobservable situations.

## 4. Real-world Experimental Results



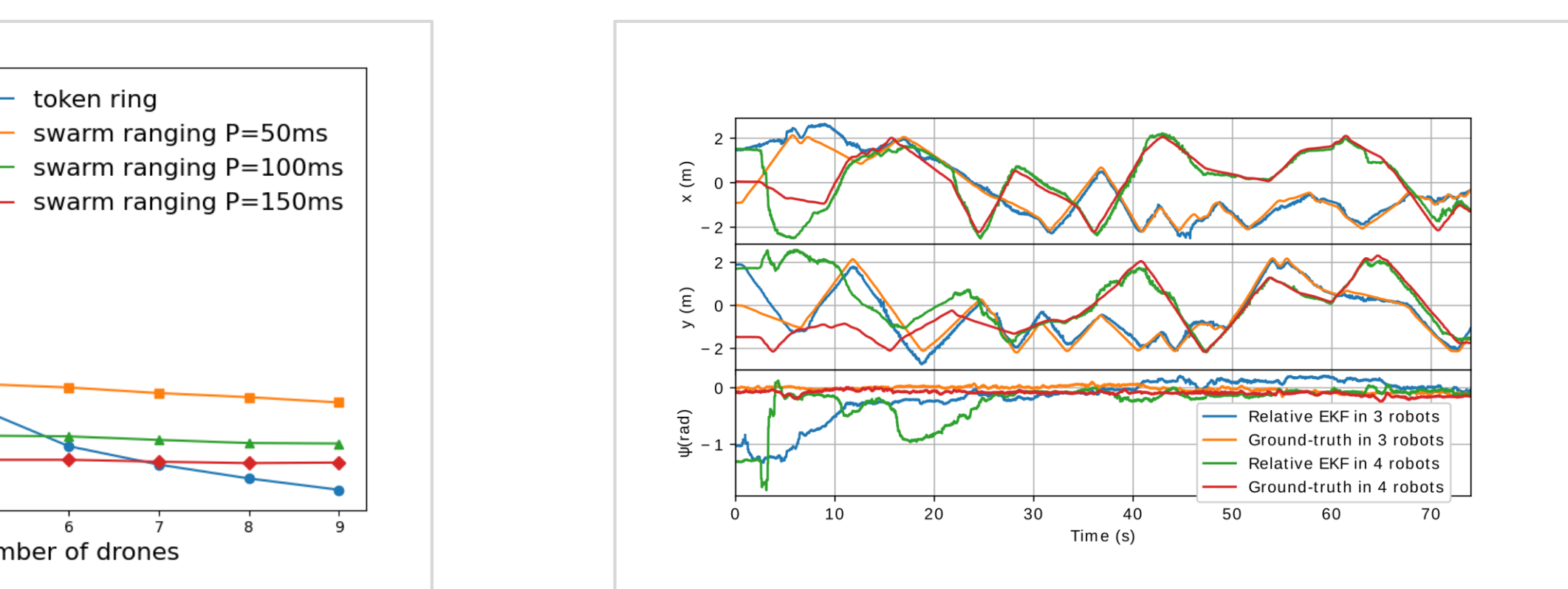
UWB measurements and position norm error distribution

**0.03 m mean UWB ranging error**



Comparison with ranging based on token ring

**~80% packet reception with >20 drones**



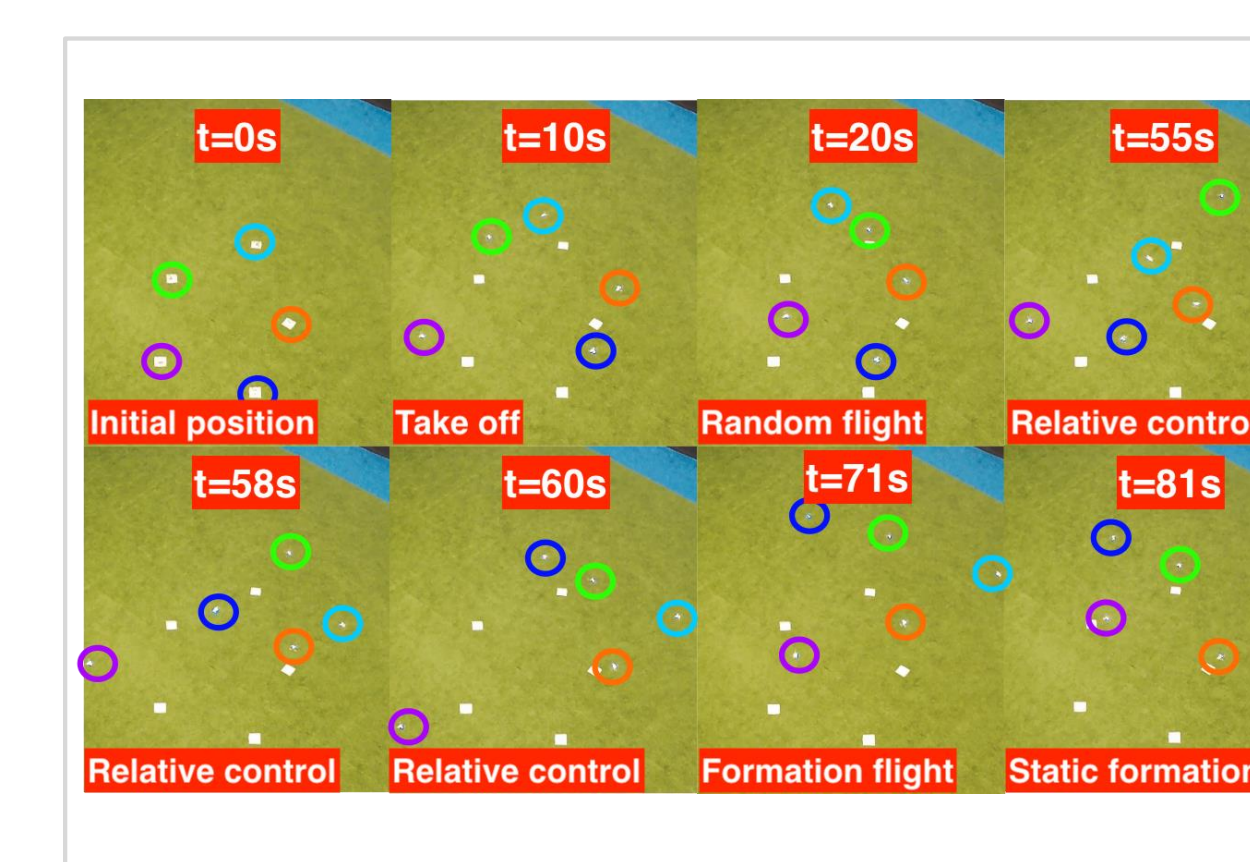
3-dimensional localization error

**0.662 ms processing delay, 26 drones**

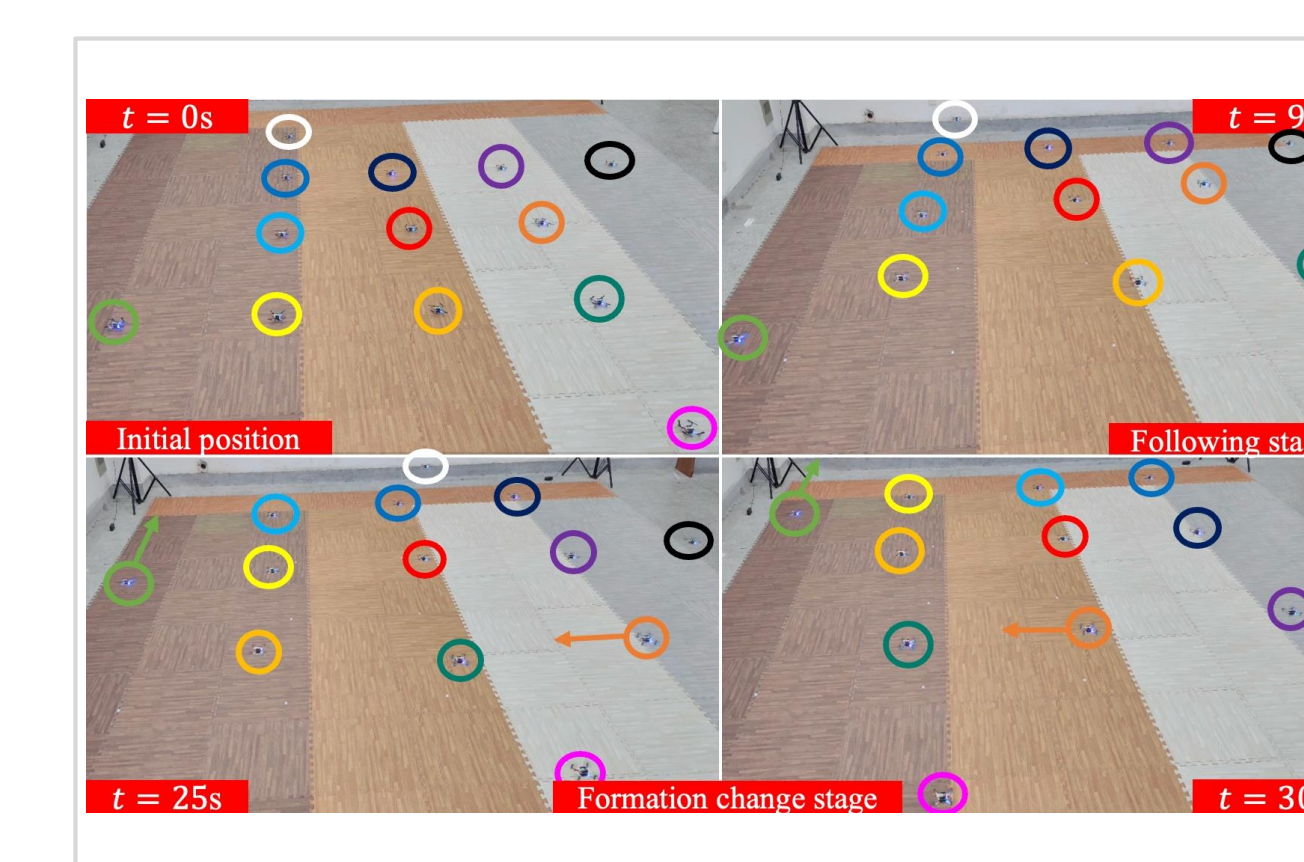
**~0.13 m relative localization**

Method	Dependence				Accuracy and Number
	IMU	Velocity	Height	North	
2019, [21]	•	•	•	•	40Hz, ~0.2m, 3
2020, [20]	•	•	•	•	16Hz, ~0.22m, 3
2020, [22]	•	•	•	•	10Hz, ~1.4m, 2
2021, [23]	•	•	•	•	10Hz, ~0.6m, 2
Ours	•	•	•	•	16Hz, ~0.13m, 13

Comparisons with the state-of-the-art UWB relative localization methods



Top view of the formation flight of 5 robots



Top view of the formation flight of 13 robots.

- Outperforms token-ring ranging as swarm size grows.
- Initial states converge in about 20 s in real flights.
- 5-drone and 13-drone formation flights validate closed-loop use of onboard localization.

Theoretical result: formation flight can be locally unobservable, but estimation drift triggers control corrections that restore observability.

Results use OptiTrack only for ground-truth validation; localization and control run onboard.