## **Drone swarms for precision agriculture**



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# Why swarms?

- Parallelise operations → higher efficiency
- Collaborative action → higher accuracy
- Redundant systems → higher robustness
- Decentralised algorithms → higher scalability

















# SAGA in a nutshell

Hardware enables: communication among UAVs high-level control and onboard vision

> **Swarm-level control:** collaborative weed mapping decentralised UAV deployment

**Onboard vision enables:**  Iow-altitude weed classification high-altitude density estimation



Hardware



### Aerial Curiosity





2x Real Cortex M4 real-time cores



Raspberry Pi 3 Compute module



8M Pixel CSI camera

**Optical Flow** position sensor



Standard GNSS **GPS** receiver



Dual IMU & magnetometer



25 min flight time

+ UWB: indoor positioning + ZigBee: swarm communication





# **Onboard Vision**





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![](_page_15_Picture_15.jpeg)

![](_page_15_Figure_16.jpeg)

## **Classification with YOLO**

![](_page_16_Figure_1.jpeg)

<sup>(</sup>Redmon et al. 2018)

![](_page_16_Picture_4.jpeg)

### altitude: 3m

![](_page_16_Picture_6.jpeg)

## **Classification with YOLO**

![](_page_17_Figure_1.jpeg)

<sup>(</sup>Redmon et al. 2018)

![](_page_17_Picture_4.jpeg)

### altitude: 3m

![](_page_17_Picture_6.jpeg)

![](_page_18_Picture_0.jpeg)

# Faster RCNN

![](_page_19_Figure_1.jpeg)

![](_page_19_Picture_2.jpeg)

![](_page_20_Picture_0.jpeg)

![](_page_21_Picture_1.jpeg)

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![](_page_21_Figure_3.jpeg)

![](_page_21_Figure_4.jpeg)

![](_page_21_Picture_6.jpeg)

![](_page_22_Picture_1.jpeg)

![](_page_22_Figure_3.jpeg)

![](_page_22_Figure_4.jpeg)

![](_page_22_Picture_5.jpeg)

![](_page_23_Picture_1.jpeg)

![](_page_23_Figure_3.jpeg)

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![](_page_24_Picture_1.jpeg)

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![](_page_25_Picture_1.jpeg)

![](_page_25_Figure_3.jpeg)

![](_page_25_Figure_4.jpeg)

![](_page_25_Picture_5.jpeg)

![](_page_26_Picture_1.jpeg)

![](_page_26_Figure_3.jpeg)

![](_page_26_Figure_4.jpeg)

![](_page_26_Picture_5.jpeg)

# Collaborative Weed Mapping

# Collaborative Weed Mapping

- Full coverage of a cultivated field to inspect for weeds
- Collaboratively map weed presence minimising classification errors
- Deal with UAVs seamlessly entering/leaving an area
- Aim at robustness, efficiency and scalability
- Adapt to environmental heterogeneities
- Avoid collisions with other UAVs
- Proposed solution: reinforced random walks (RRW)

Albani, D., Nardi, D., & Trianni, V. (2017). Field Coverage and Weed Mapping by UAV Swarms Proceedings of the 2017 IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS 2017), Vancouver, Canada, Sept. 2017

### RRW for coverage and mapping Isolated agents perform a correlated random walk m a Random selection among those cells that are closer and not yet visited 3 5 Preferential choice of cells in the motion direction lacksquare(gaussian decay with distance, width $\sigma_A$ ) 3 3 Agents are attracted towards areas of interest (gaussian decay with distance, width $\sigma_B$ ) 3

- Neighbour agents repel each other

## RRW for coverage and mapping Isolated agents perform a correlated random walk

- - Random selection among those cells that are closer and not yet visited
  - Preferential choice of cells in the motion direction lacksquare
- Neighbour agents repel each other (gaussian decay with distance, width  $\sigma_A$ )
- Agents are attracted towards areas of interest (gaussian decay with distance, width  $\sigma_B$ )
- Resultant vector determining
  - Direction of bias
  - Persistence of the correlated random walk

![](_page_30_Figure_9.jpeg)

# RRW for field coverage/mapping

![](_page_31_Figure_1.jpeg)

### Mapping error goes down from 20% to 5%

Coverage

64

3r

![](_page_31_Figure_4.jpeg)

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8	- 2823	2824	2823	2849	2830	2725
16	- 2681	2660	2757	2730	2840	2718
32	- 2591	2546	2509	2500	2734	2737
	- 2519	2522	2521	2507	2484	2723
64						

![](_page_31_Figure_6.jpeg)

 Onboard vision and autonomous control allow for non-uniform coverage

- Onboard vision and autonomous control allow for non-uniform coverage
- High-altitude estimation of weed density

![](_page_34_Picture_7.jpeg)

- Onboard vision and autonomous control allow for non-uniform coverage
- High-altitude estimation of weed density
- Low-altitude collaborative weed mapping

![](_page_35_Figure_8.jpeg)

- Onboard vision and autonomous control allow for non-uniform coverage
- High-altitude estimation of weed density
- Low-altitude collaborative weed mapping
- Attention should be focused only to those areas that contain weed patches

![](_page_36_Figure_10.jpeg)

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- Onboard vision and autonomous control allow for **non-uniform coverage**
- High-altitude estimation of weed density
- Low-altitude collaborative weed mapping
- Attention should be focused only to those areas that contain weed patches
- The problem translates to utility-dependent UAV deployment

![](_page_37_Picture_12.jpeg)

### Collective decision

- There are less UAVs than the • UAVs are in excess with respect optimal number for a single area to the optimal number
- Collaboration improves mapping Area utilities do not vary efficiency considerably

Identify a deployment strategy that can be easily tuned

# Decentralised UAV Deployment

Task allocation VS

> Determine optimal number of UAVs given the mapping dynamics

![](_page_38_Picture_8.jpeg)

![](_page_39_Figure_0.jpeg)

$$\delta = 3E-6, \xi = 4E-8 \quad \rightarrow \quad n^* = 5$$

# UAV Deployment Strategy

- UAVs explore and estimate the utility of areas during high-altitude/low-resolution inspection
- UAVs form a wireless communication network and **recruit** other UAVs to areas of high utility
- UAVs are inhibited from monitoring a certain area when
  - other areas of high utility need attention (cross-inhibition among UAVs deployed to different areas)
  - there are too many teammates (self-inhibition among UAVs deployed to the same area)

UAVs prioritise low-altitude/high-resolution inspection for high-utility areas

uncommitted:

- high-altitude inspection
- estimate area utility

deployment:

- spontaneous (utility-driven)
- interactive (recruitment)

- interactive (inhibition)
- spontaneous (mapping completed)

abandonment:

С

deployed to an area: low-altitude mapping recruit/inhibit teammates

![](_page_41_Picture_11.jpeg)

![](_page_41_Picture_12.jpeg)

### Decentralised Deployment Model spontaneous inhibition abandonment M $\dot{x}_i = \gamma_i x_u - \alpha_i x_i + \rho_i x_u x_i - \sum_{i=1}^{m} x_j \beta_{ji} x_i, \qquad x_u = 1 - \sum_{i=1}^{m} x_i$ j=1recruitment 1.0 ODEs 0.8 Parameterisation choice Multi-agent 0.6 $\propto h u_i$ x<sub>i</sub>,u<sub>i</sub> $\propto h u_i$ 0.4 0.2 $(\gamma_i, ho_i)$ 0.0 100 150 200 250 50 300

![](_page_42_Figure_1.jpeg)

$$\begin{aligned} \gamma_i \propto k u_i & \rho_i \\ \alpha_i = 0 \text{ unless } u_i \approx 0 & \beta_{ij,i \neq j} \\ \beta_{ii} = f( e^{-i \lambda_i} ) \end{aligned}$$

Reina, A., Marshall, J. A. R., Trianni, V., & Bose, T. (2017). Model of the best-of-N nest-site selection process in honeybees. Physical Review E, 95(5), 052411–15.

Time [s]

## Decentralised Deployment Model spontaneous inhibition abandonment $\dot{x}_i = \gamma_i x_u - \alpha_i x_i + \rho_i x_u x_i - \sum_{i=1}^{M} x_j \beta_{ji} x_i, \qquad x_u = 1 - \sum_{i=1}^{M} x_i$ j=1

![](_page_43_Figure_1.jpeg)

### Parameterisation choice

$$\begin{aligned} \gamma_i \propto k u_i & \rho_i \\ \alpha_i = 0 \text{ unless } u_i \approx 0 & \beta_{ij,i \neq j} \\ \beta_{ii} = f( e^{-i \lambda_i} ) \end{aligned}$$

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recruitment

 $\propto h u_i$ 

 $\propto h u_i$ 

 $(\gamma_i, 
ho_i)$ 

We study the ratio r=h/kbetween interactive and spontaneous transitions

![](_page_44_Figure_0.jpeg)

# $n_i$

![](_page_45_Figure_0.jpeg)

- Decentralised deployment and re-deployment provides
  - ability to focus only on areas of high interest
  - ability to enforce utility-responsive strategies
- The proposed strategy can be tuned by a single parameter • Utility-proportional deployment (r = 0)• Winner-takes-all deployment  $(r \ge 1, N \le n^{\star})$

- Utility-responsive deployment  $(r \ge 1, N > n^{\star})$ lacksquare
- Strategy tested with UAV simulations
  - Spatial distribution of agents over areas influences deployment
  - Communication range must be sufficiently high

# Results Achieved

# Summing up

- Collaborative field monitoring and mapping provides
  - parallel operation (efficiency) and collaboration (accuracy)
  - robustness and scalability: group size can vary in real time
- Decentralised deployment and re-deployment provides
  - ability to focus only on areas of high interest
  - ability to enforce utility-responsive strategies

## Thanks for your attention