

# Area Inspection by Robot Swarms through Exploitation of Information Gain

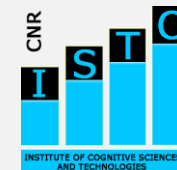
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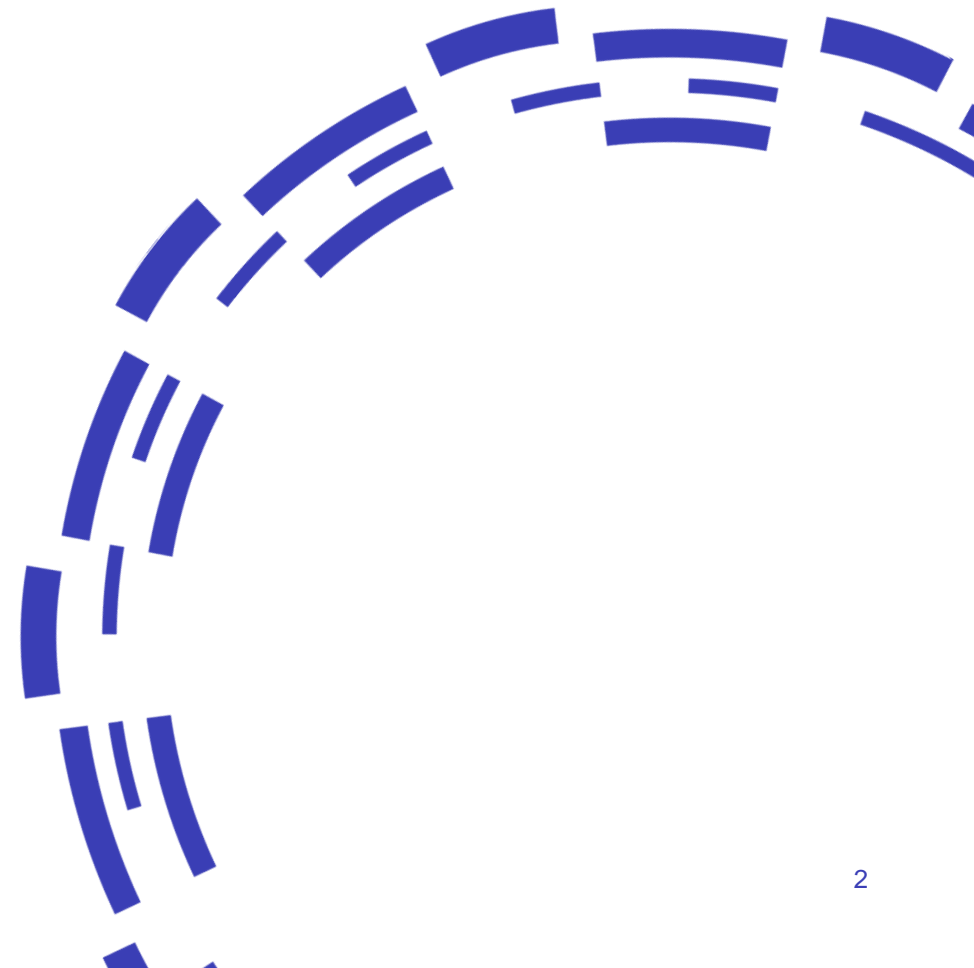


## **Dario Albani**

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Senior researcher at the Technology Innovation Institute in Abu Dhabi. Previously researcher at National Research Council of Italy and Sapienza University of Rome.

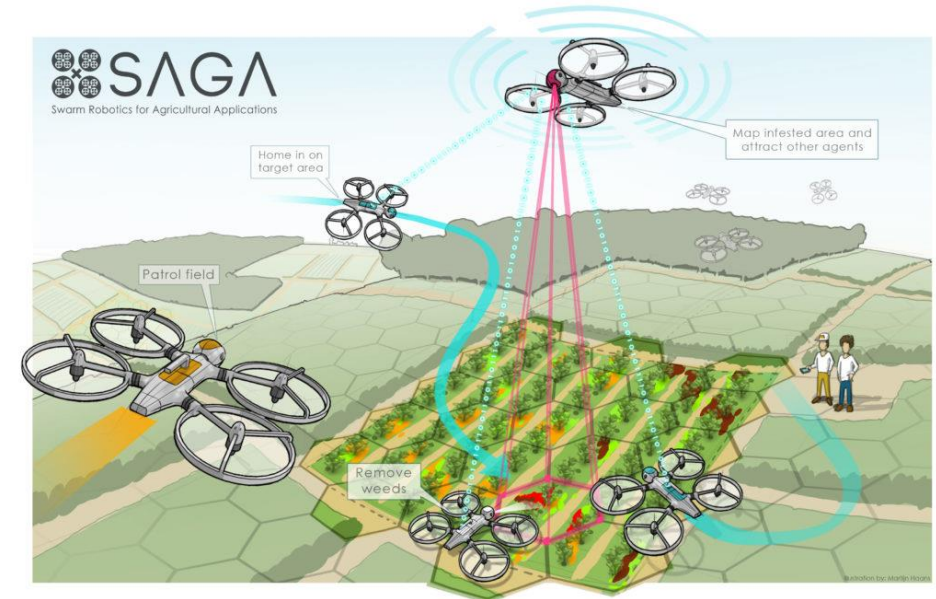
My main research interests span over swarm robotics, self-organizing systems as well as multi-robot systems.



# Study Case

## Precision Agriculture:

- An application domain presenting expansive fields and needs for collaboration
- Robotics technologies promise a remarkable impact
- Far from exploiting the full potential of autonomous robots and multi-robot collaboration



## The approach can fit other study cases as:

- Search and rescue
- Monitoring and mapping applications

# In a Nutshell

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- A **decentralized** approach for area coverage and mapping
- Grid world assumption
- Probabilistic area selection based on information **entropy** and **information gain**
- Direct consideration of neighboring agents through **broadcast communication**
- Relies only on **local** knowledge
- No **free parameters** and no pre-operational tuning needed

# Information Gain ...

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In its simplest form the information gain writes:

$$IG_n(c) = \underbrace{H_n(c)} - \underbrace{H_n(c|o_c(n))}$$

Difference between:

- the residual uncertainty that a robot has about a cell
- the conditional entropy given a robot observation

## ... and Entropy

$$H_n(c) = - \sum p_n(c) \log(p_n(c))$$

the entropy computed over the current knowledge of the robots about the cell

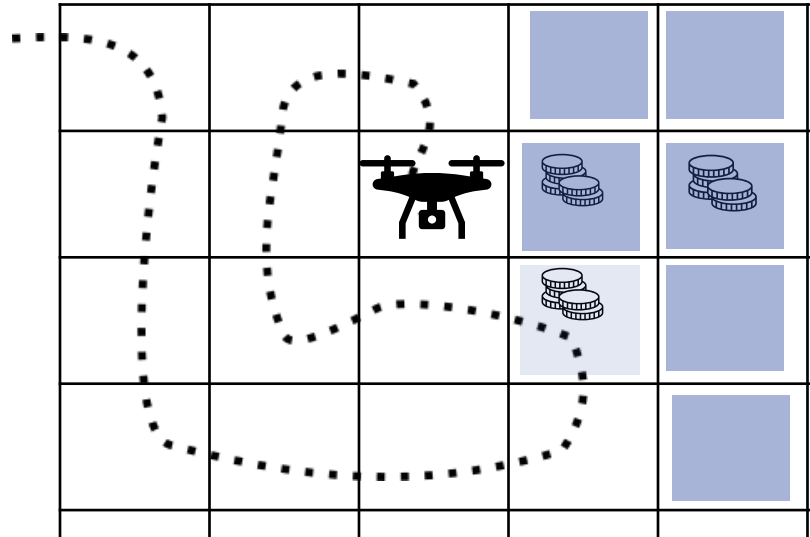
$$H_n(c_{i,j} | \tilde{o}_{i,j}) = - \sum_o p_n(\tilde{o}_c) \sum_c [p_n(c | \tilde{o}_c) \log(p_n(c_c | \tilde{o}_c))]$$

the conditional entropy expressed according to the robot observation

# Information Theory Enriched Random Walk - 1

We use the IG to assign probabilities to the cells:

- Undiscovered regions have a greater associated IG
- Already discovered regions with unmapped POIs have a greater associated IG



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That is:

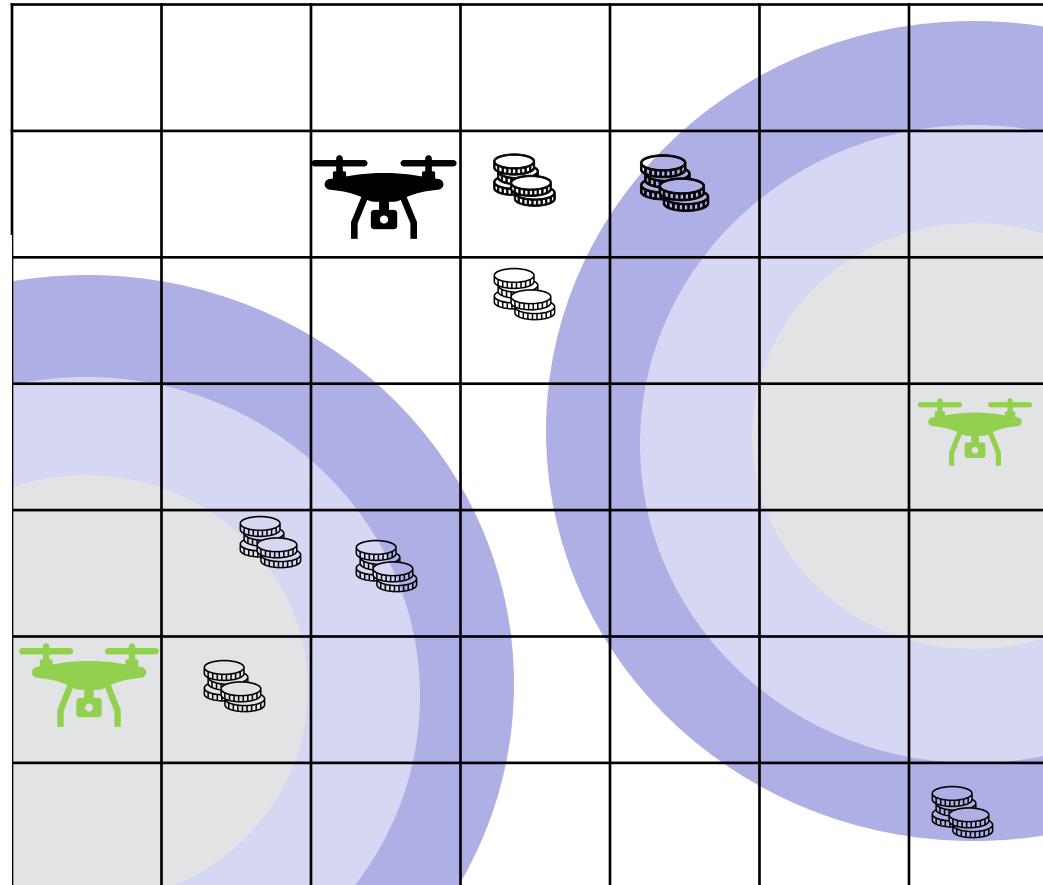
$$P_n(c) = \frac{IG_n(c)}{\sum_{\tilde{c} \in C} IG_n(\tilde{c})}$$



# Information Theory Enriched Random Walk - 2

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- Regions closer to other robots have a smaller associated IG



## Information Theory Enriched Random Walk - 2

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- Regions closer to other robots have a smaller associated IG

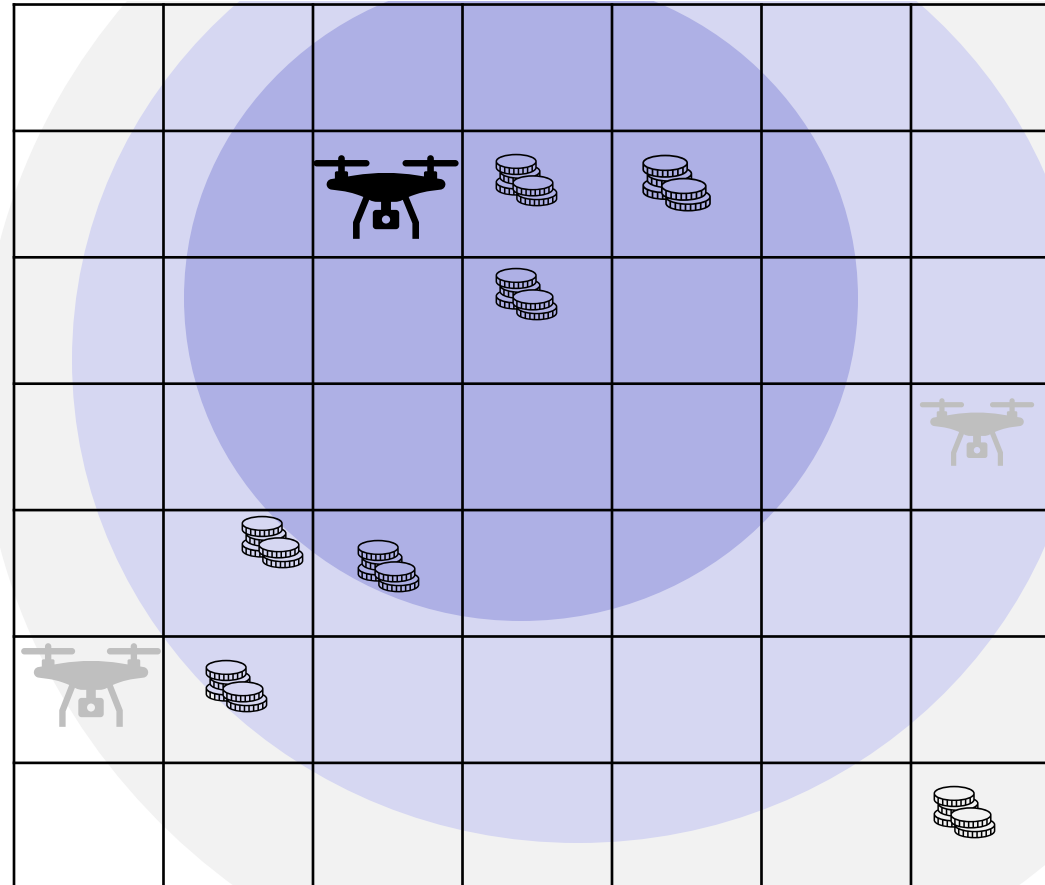
That is:

$$P_n(c_{i,j}) = \frac{IG_n(c)}{\sum_{\tilde{c} \in C} IG_n(\tilde{c})} \prod_{\tilde{n} \neq n} \left[ 1 - \frac{IG_{\tilde{n}(c)}}{\sum_{\tilde{c} \in C} IG_{\tilde{n}}(\tilde{c})} \right]$$

# Information Theory Enriched Random Walk - 2

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- Regions too far from the robot have a smaller associated IG



# Information Theory Enriched Random Walk - 2

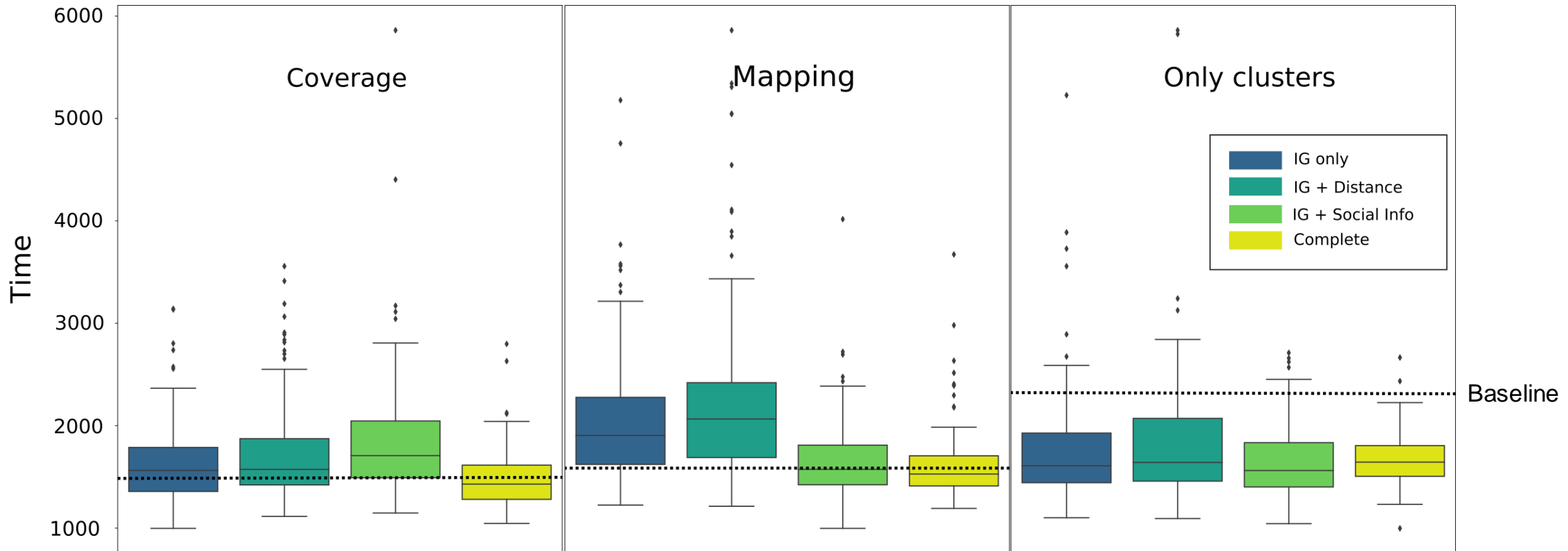
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- Regions too far from the robot have a smaller associated IG

That is:

$$P_n(c) = \frac{d_n(c)^{-1}IG_n(c)}{\sum_{\tilde{c} \in C} d_n(\tilde{c})^{-1}IG_n(\tilde{c})} \prod_{\tilde{n} \neq n} \left[ 1 - \frac{d_{\tilde{n}}(c)^{-1}IG_{\tilde{n}}(c)}{\sum_{\tilde{c} \in C} d_{\tilde{n}}(\tilde{c})^{-1}IG_{\tilde{n}}(\tilde{c})} \right]$$

# Preliminary Results



Results compared against the RRW presented in Field coverage and weed mapping by UAV swarms  
D Albani, D Nardi, V Trianni - 2017 IEEE/RSJ/ROCS

# Pros and Cons

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## No Parameters

PRO

- No free parameters to be set
- No pre-operational tuning needed
- Allows or fast deployment

## Swarm Properties

PRO

- Completely Decentralized
- Robust to failures
- Scalable
- Relies on local knowledge

## Probabilistic Selection

CON

- Might lead to wrong decisions
- Wrong decisions increase operational time

## Communication and Performances

CON

- Performances should be improved
- Lack of specific information sharing protocols
- Is it really scalable?

## Future Work

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Being this a preliminary work there is room for improvement

- Belief propagation (via Gaussian Processes) to:
  - Decrease the communication overhead
  - Decrease the computational load
  - Increase the scalability
- Introduction of more accurate sensor models for perception
- Real world testing

# Area Inspection by Robot Swarms through Exploitation of Information Gain



- A decentralized approach for area coverage and mapping
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