# Infotaxis in a Multi-agent Sensor Network

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### BACKGROUND

Source

0

### PROBLEM

How do we find a source when observations are sparse?

#### CHALLENGES:

Observations are too sparse for gradient descent

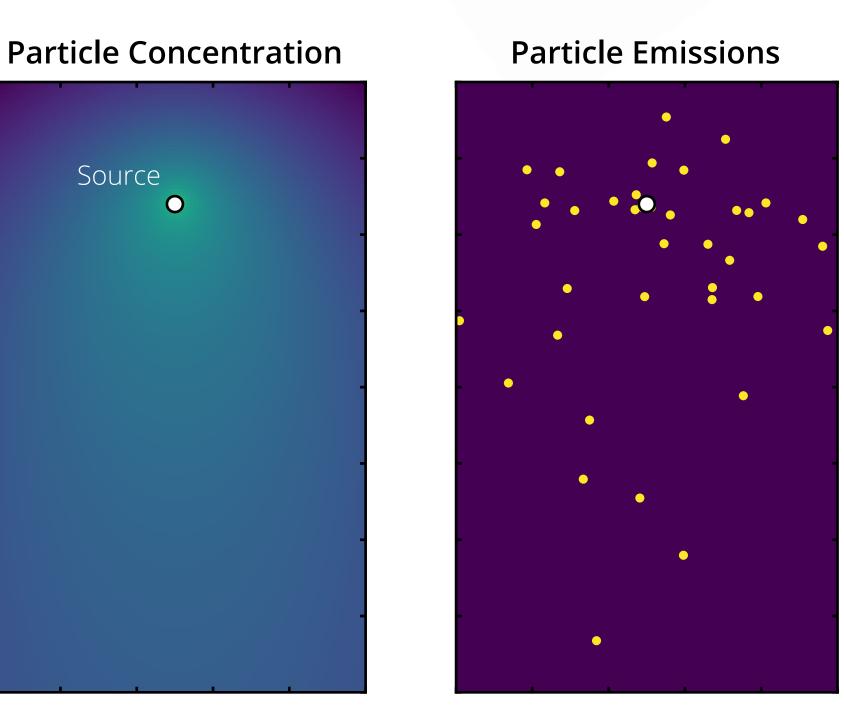
#### SOLUTION:

*Infotaxis:* Follow the information gradient

### SETUP ENVIRONMENT:

Particles emitted from source at constant rate





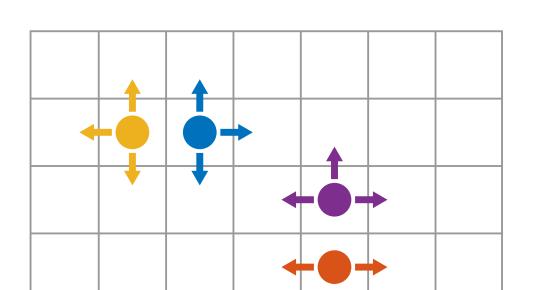
 $\mathbf{r}_{j=4}$ 

High

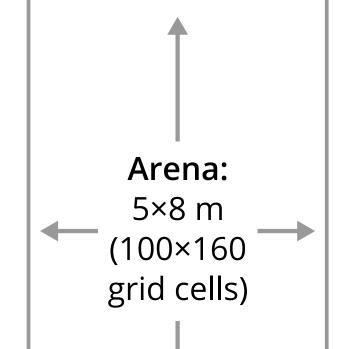
High

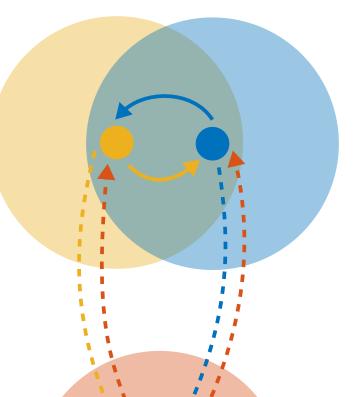
# MULTI-AGENT INFOTAXIS

#### COLLISION AVOIDANCE COMMUNICATION



Continuous communication over fixed range Agents treat others' observations as their own





### EXPERIMENTS

VARY: Communication range

Confined, discretized arena with constant wind velocity

#### AGENTS:

**Goal:** reach source Particle detection ability Move one cell per tick

> **Detection Probability** High Low

### SINGLE-AGENT INFOTAXIS

#### PROBABILITY MAP

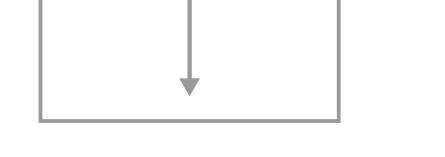
Agents maintain map of the likelihood that the source is in each cell

Observation history (trace):

$$\Gamma_t = \{(r_a, t_a)\}$$

Number of agents

#### INVESTIGATE:



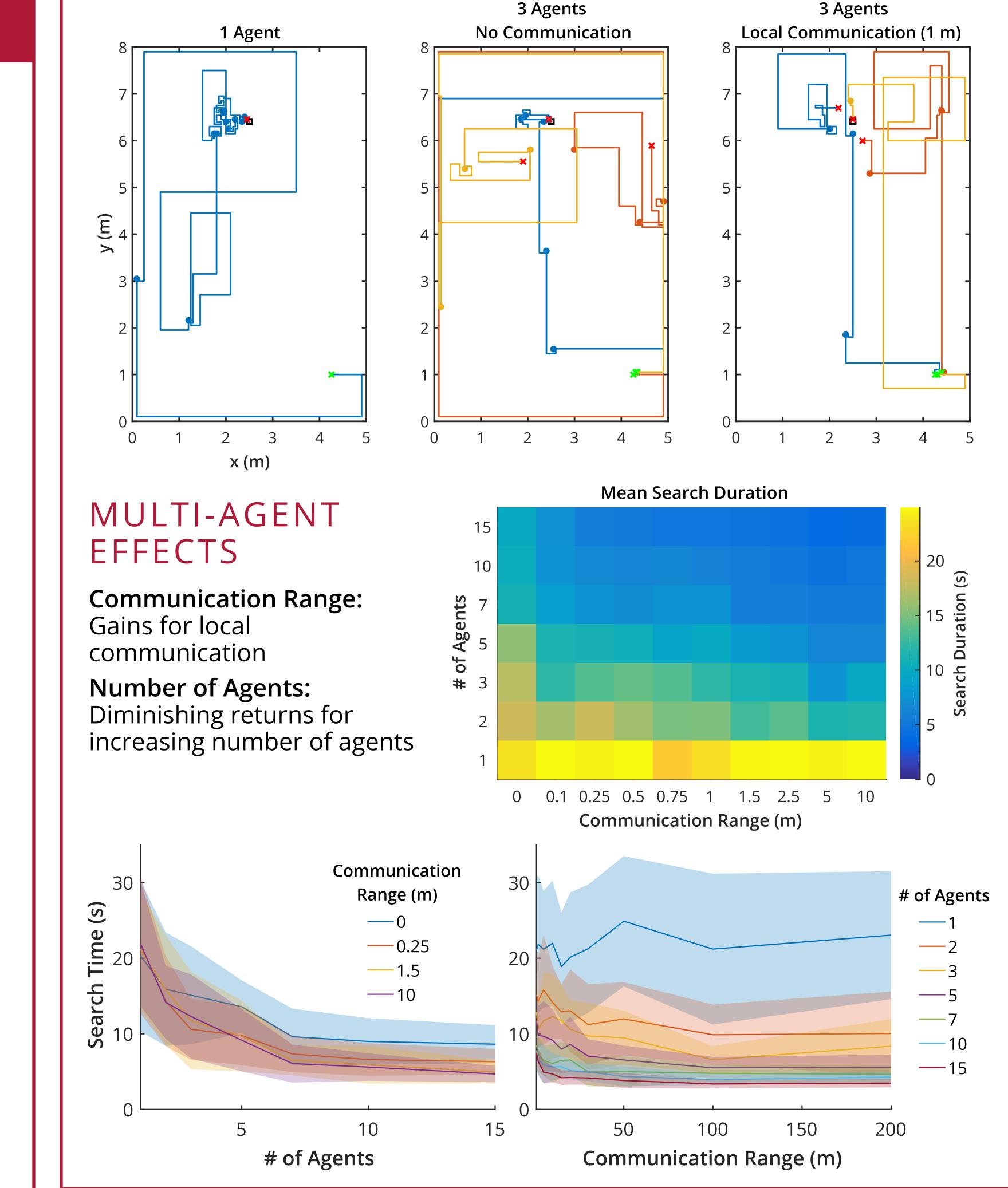
Time for first agent to reach source

# RESULTS

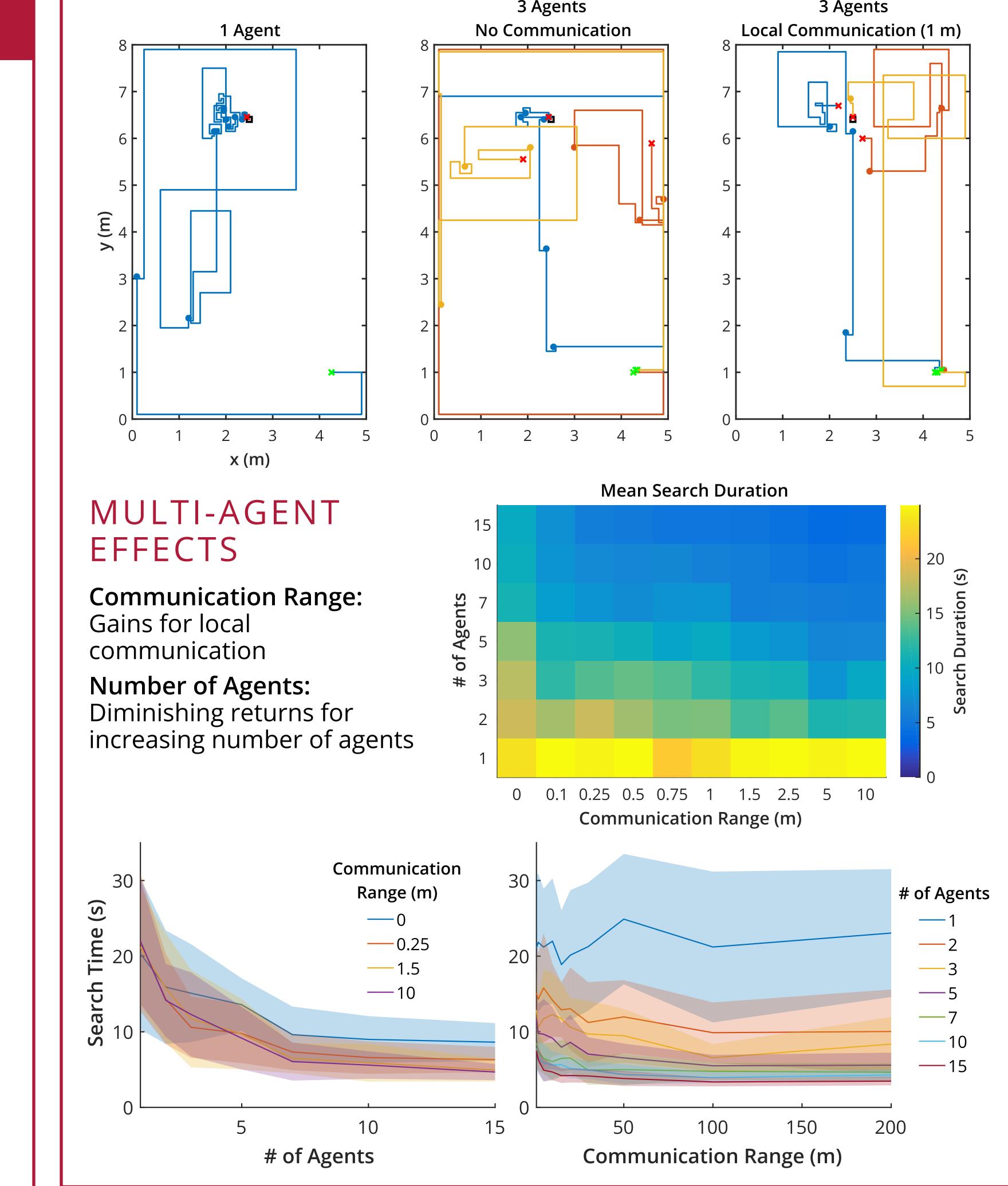
### AGENT TRAJECTORIES

Sweeping search patterns as observed in moths and dogs Early Behavior: Exploration dominates; agents spread out Late Behavior: Exploitation dominates; agents cluster

- Source
- × Start
- End
  - Particle Detections



3 Agents



Likelihood that source is at  $\mathbf{r}_0$ :

 $R(\mathbf{r}|\mathbf{r}_0) = \text{estimated particle detection rate at } \mathbf{r} \text{ if source is } \mathbf{r}_0$ 

$$L(\Gamma_t | \mathbf{r}_0) = \exp\left(\sum_a \sum_{t' \in T_a} R(\mathbf{r}_a(t') | \mathbf{r}_0) dt')\right) \prod_{b=1}^{T_b} R(\mathbf{r}(t_b), t_b | \mathbf{r}_0)$$

Prior

Probability map over full arena:

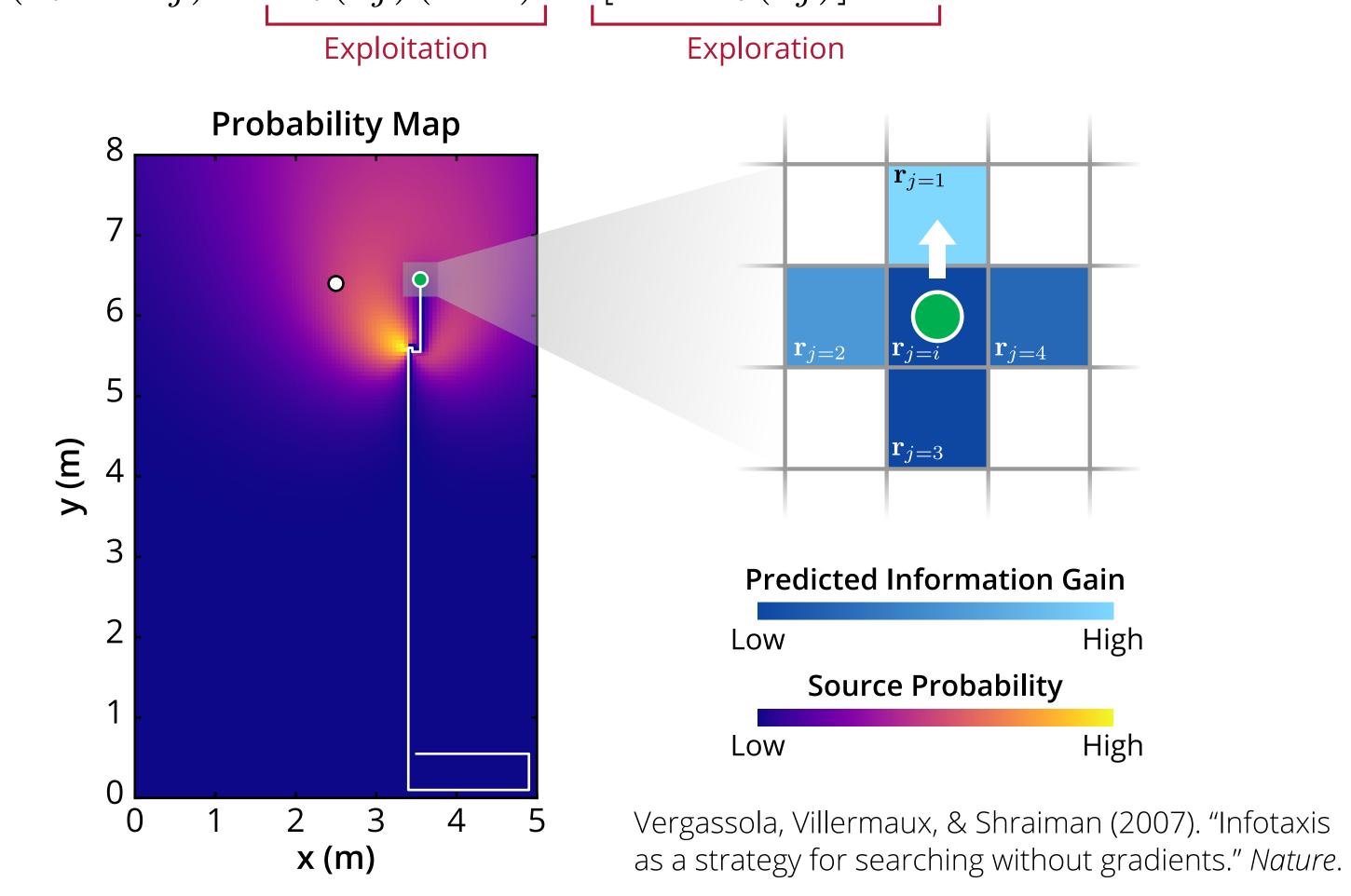
 $P_t(\mathbf{r}_0|\Gamma_t) = \frac{L(\Gamma_t|\mathbf{r}_0)}{\sum_a L(\Gamma_t|\mathbf{r}_q)d\mathbf{r}_q}$ 

#### MOVEMENT DECISION

Agents move to cell that maximizes entropy

Expected information gain:

 $\Delta S =$  Information from expected detections at  $\mathbf{r}_i$ (Poisson distribution)  $H = \text{Entropy of } P_t(\mathbf{r}_i)$  $\Delta H(\mathbf{r}_i \to \mathbf{r}_j) = P_t(\mathbf{r}_j)(-H) + [1 - P_t(\mathbf{r}_j)] \Delta S$ 



### FUTURE WORK

Multi-agent behavior: Quantify group dynamics, test robustness **Reduce assumptions:** Lossless communication, perfect localization **Localization:** Noisy, non-discrete, filtered **Communication:** Asynchronous, lossy **Computational complexity:** Continuous probability map (mixture model)