

Multi-Robot Exploration

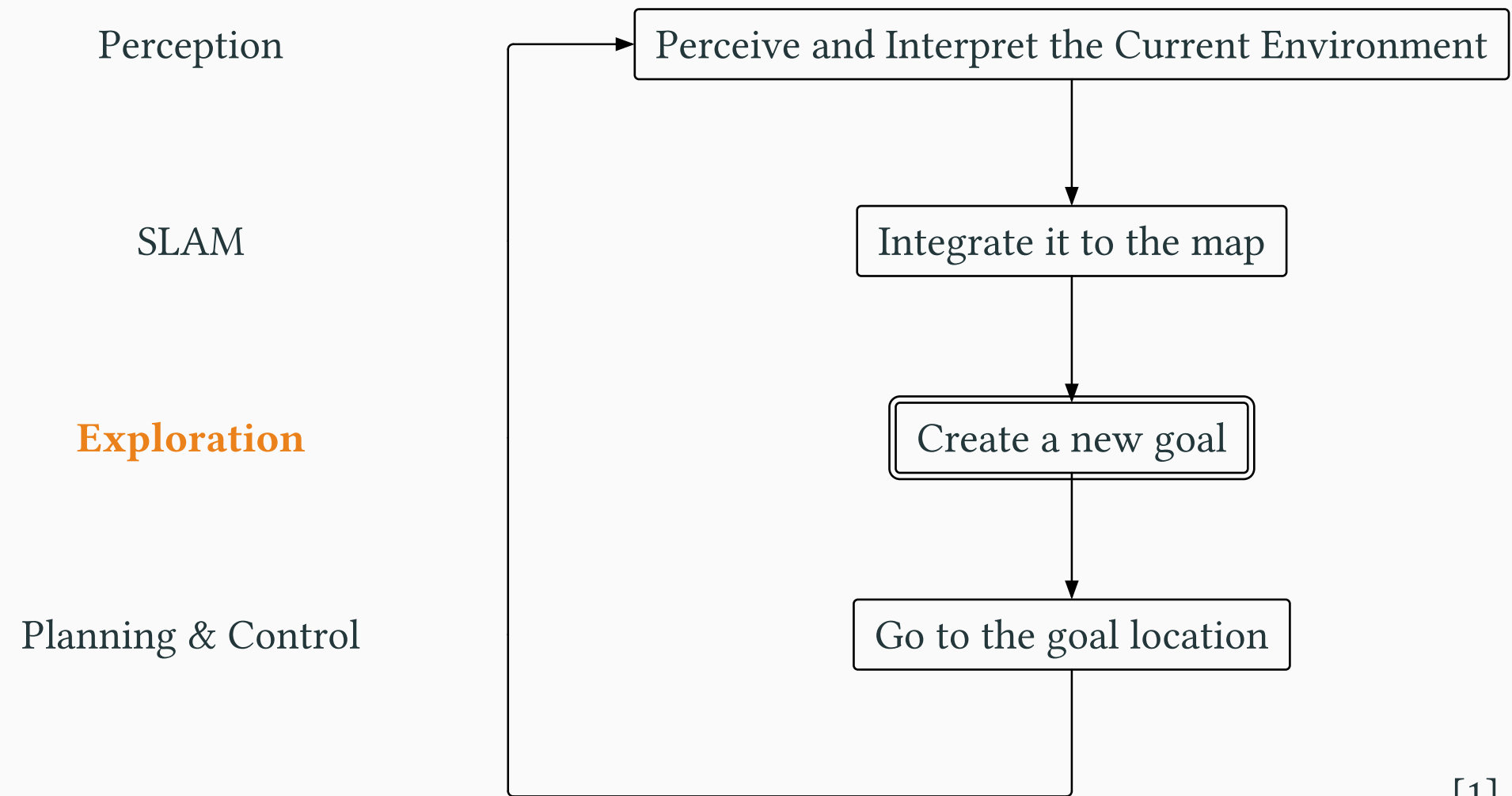
Integrated Seminar: Intelligent Robotics

Jonas Schmidt

November 27, 2025

What is Exploration?

What is Exploration?



[1], [2]

– What is Exploration?

mobile robot exploration is an iterative process in which the robot integrates its observations into its world model, interprets the world model to determine which parts of the environment are unknown, and plans a path to visit and observe these unknown areas

Formal Definition

“Given a team of m robots $R = \{r_1, r_2, \dots, r_m\}$,
deployed in a continuous and bounded environment Env,
the goal for the robots is to select a sequence of poses $Q = \langle Q_0, Q_1, \dots, Q_n \rangle$
so that at each pose they perceive features in the environment
through a sensor mounted on-board.”

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Motivation

— Motivation

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- examples: Delivery, Find an Rescue,
- having a map is only the case in veryfew settings
- even if a map exists, the environment might change

Motivation

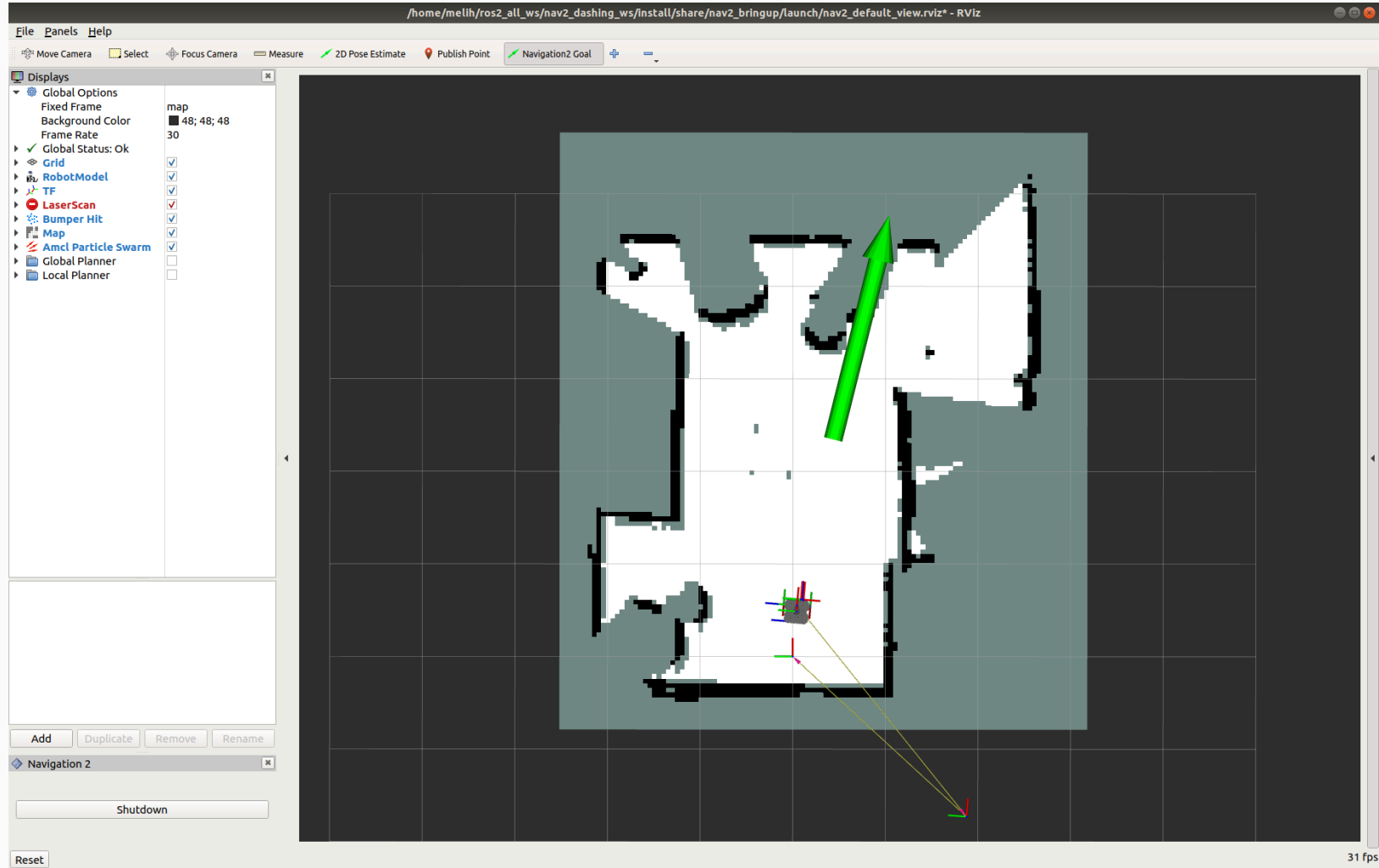
How does Exploration integrate with SLAM and Mapping?

Multi-Robot Exploration

— Motivation

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Coverage

Map every point within a region of interest

Target Localization

Explore a region to find a target of interest

[2]

— Tasks in Exploration

Coverage: Map every point within a region of interest

- Mapping every point = Coverage by the robot sensor(s)
- Common approach:
 1. Decomposition of the environment into cells
 2. Optimization of the cells' visit order
 3. Generation of a trajectory to visit each cell
- Limiting the number of locations to consider:
 - Term: Frontier
 - Locations between known and unknown portions of the environment

Coverage

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Target Localization

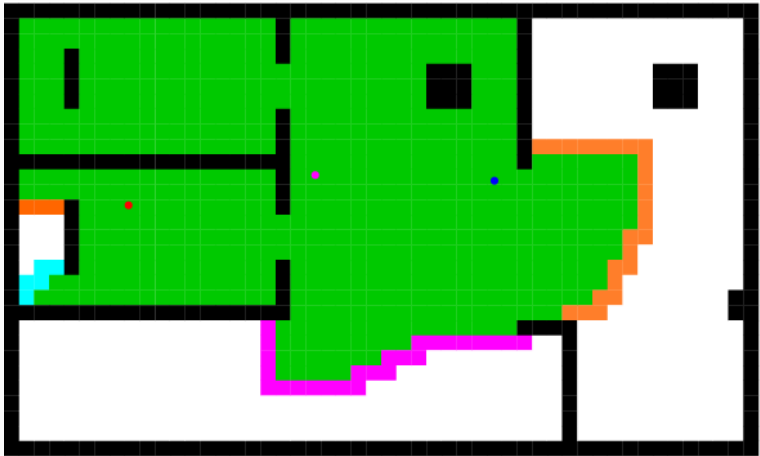
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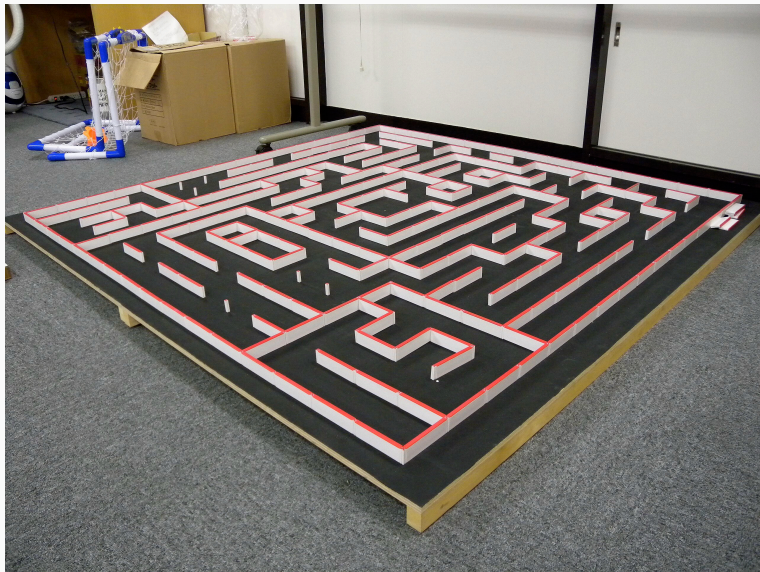
— Tasks in Exploration

Target Localization: Explore a region to find a target of interest

- limit the number of locations to consider
- integrates semantic & topology knowledge



[3]



[4]

Prerequisites for Exploration:

- Sensor Data Fusion
- Probabilistic Mapping
- Planning Methods

[1], [5], [6], [7], [2]

Historical Overview

- End of 1980s, Beginning of 1990s

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Advances over the Years

2D \Rightarrow 3D

Without a priori knowledge \Rightarrow Integrating a priori knowledge

Constant Environment \Rightarrow Dynamic Environment

Indoor Environment \Rightarrow Outdoor Environment

Mapping Occupancy \Rightarrow Mapping other Parameters

Optimal Robot Conditions \Rightarrow Integrating Robots Health

[2]

— Historical Overview

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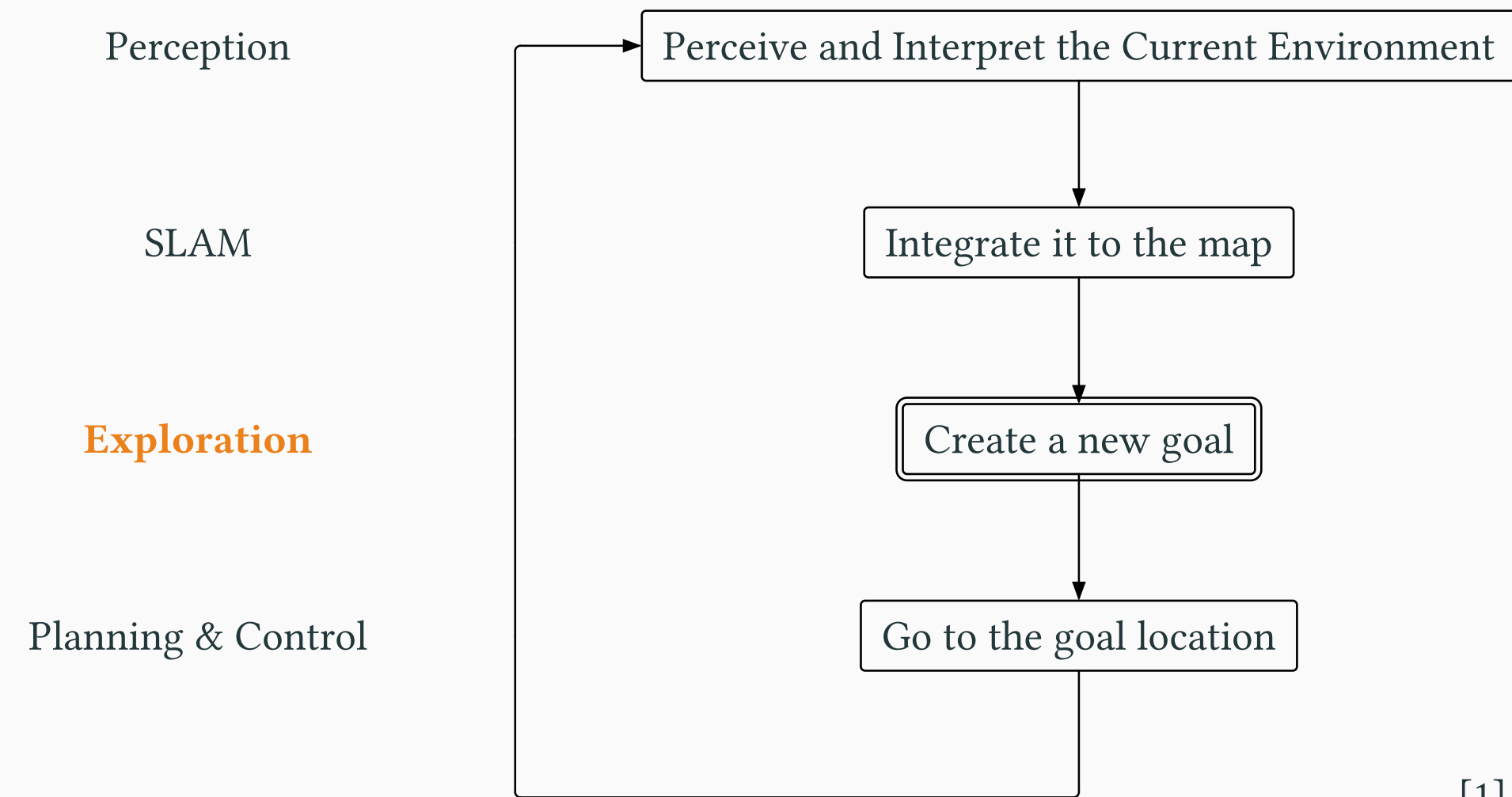
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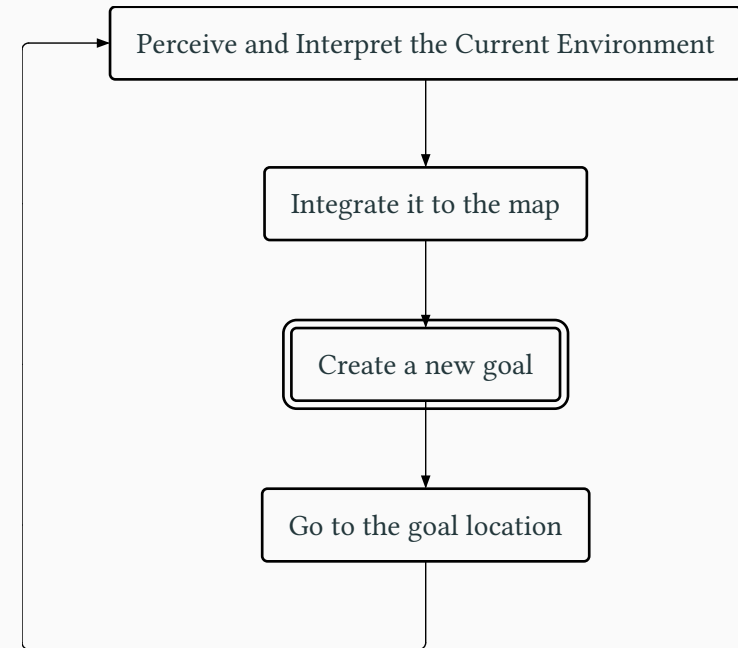


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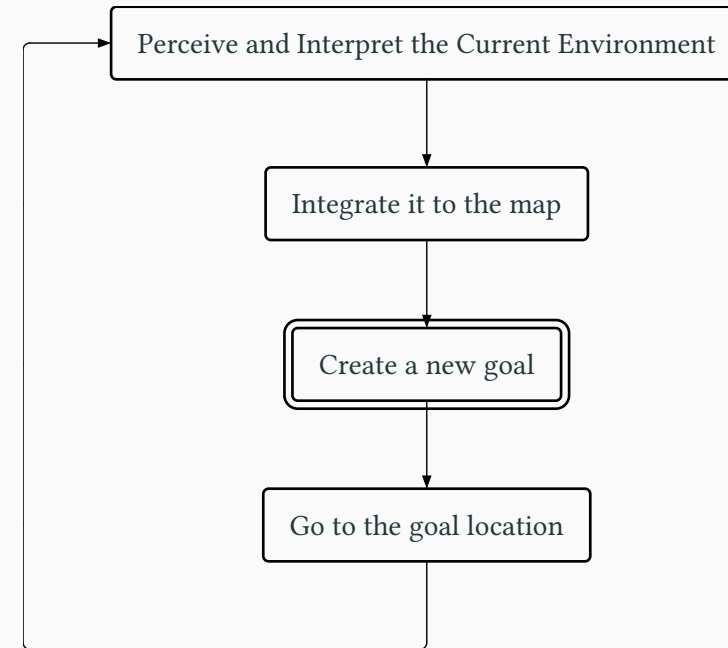
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Robot 1



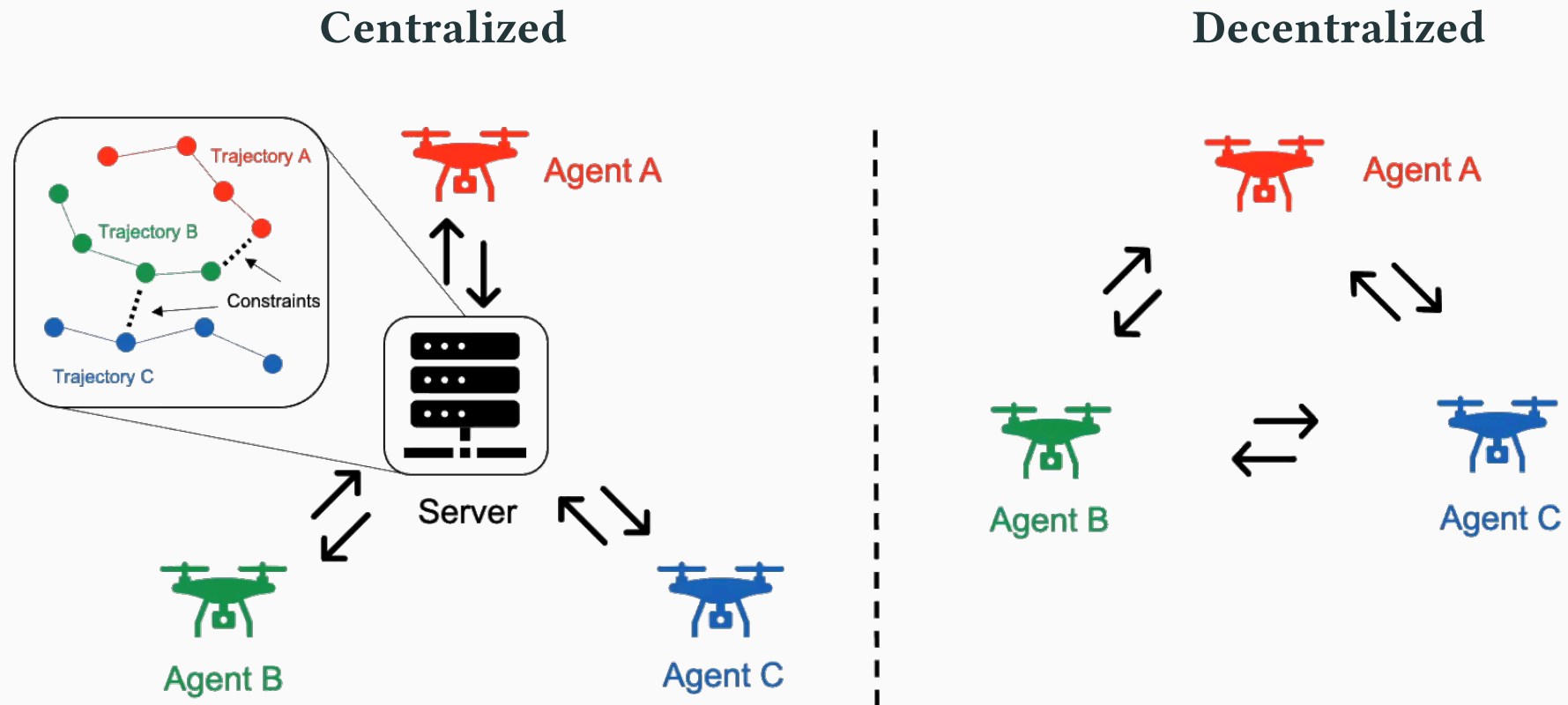
Robot 2



[8], [2]

— Starting with one robot. What changes if another robot helps with Exploration?

- Sensor Data Fusion
- **Shared or Distributed Map:**
 - Map Merging
 - Decentralized SLAM
- Multi-Robot Planning
 - neglecting intersecting paths and let the local controller avoid collisions
 - *Conflict-based search*



Collaborative SLAM

Centralized

- Agents communicate via server
- server:
 - stores trajectories
 - establishes constraints
 - finds a globally consistent solution
 - server is the bottleneck

Decentralized

- Agents communicate peer-to-peer
- computation can be distributed
- more scalable
- challenge is to work with the local view of each agent
- requires a more complex approach to global optimization and finding loop closures

Starting with one robot. What changes if another robot helps with Exploration?

Exploration Techniques

Examples:

1. **Robots work independently of each other; Visited Areas are marked.**
2. Frontier-based exploration

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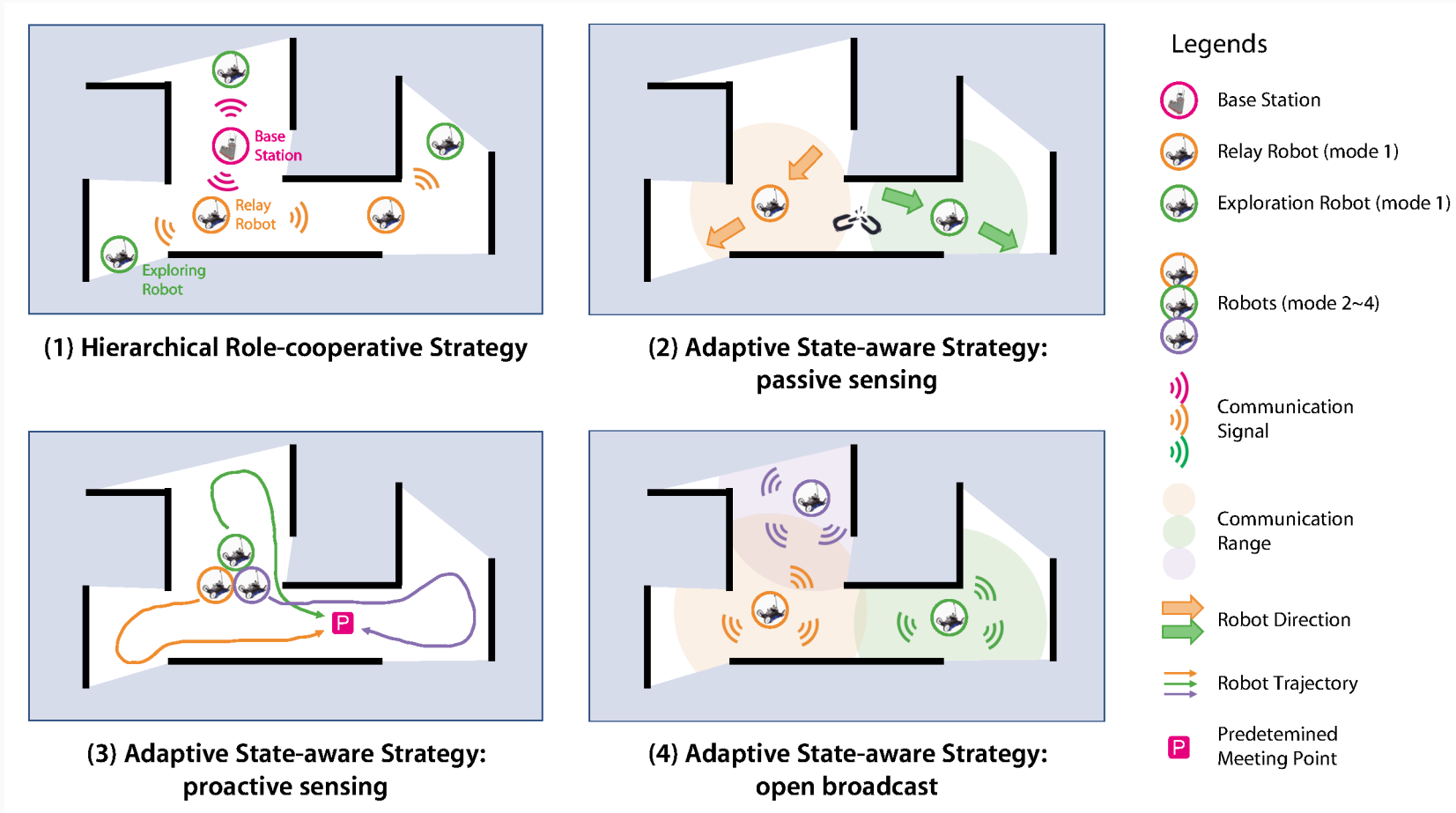
Exemplory Capabilities

- Actuators (one might be able to interact with the environment)
- Sensors
- Velocity limits
- Communication Capabilities
- Operating in multiple Domains (e.g. [11])

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Approach: Distributing Tasks



[10]

— Adding Robots with Different Capabilities

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Application Example

<https://vimeo.com/893930171>

Exploring Unstructured Environments using Minimal Sensing on Cooperative Nano-Drones

**Pedro Arias-Perez¹, Alvika Gautam², Miguel Fernandez-Cortizas¹,
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Scenario

- Task: Covering
- Variable Swarm Sizes
- Variable Obstacle Densities
- Nano Drones

[12]


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vision & aerial
robotics


CONSEJO SUPERIOR DE INVESTIGACIONES CIENTÍFICAS


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Approach

“ExploreBug, a novel hybrid frontier range bug algorithm”

“an intra-swarm collision avoidance system is integrated”

[12]

1. What is Exploration?

- > Generation of a pose target suitable for exploring the environment.

2. What do we need to consider for multi-robot Exploration

- > How do we share current knowledge about the environment and each generate a new pose target?
- > Robot Capabilities, Tasks, Multiple Domains

3. Example: Exploration with Nano-Drones

Sources



[https://drive.proton.me/
urls/0W2WAQHM58#J7
LByzUsANy7](https://drive.proton.me/urls/0W2WAQHM58#J7LByzUsANy7)

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Sources

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