

2022

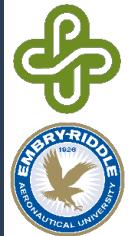
# Consensus-based Communication-aware Formation Control

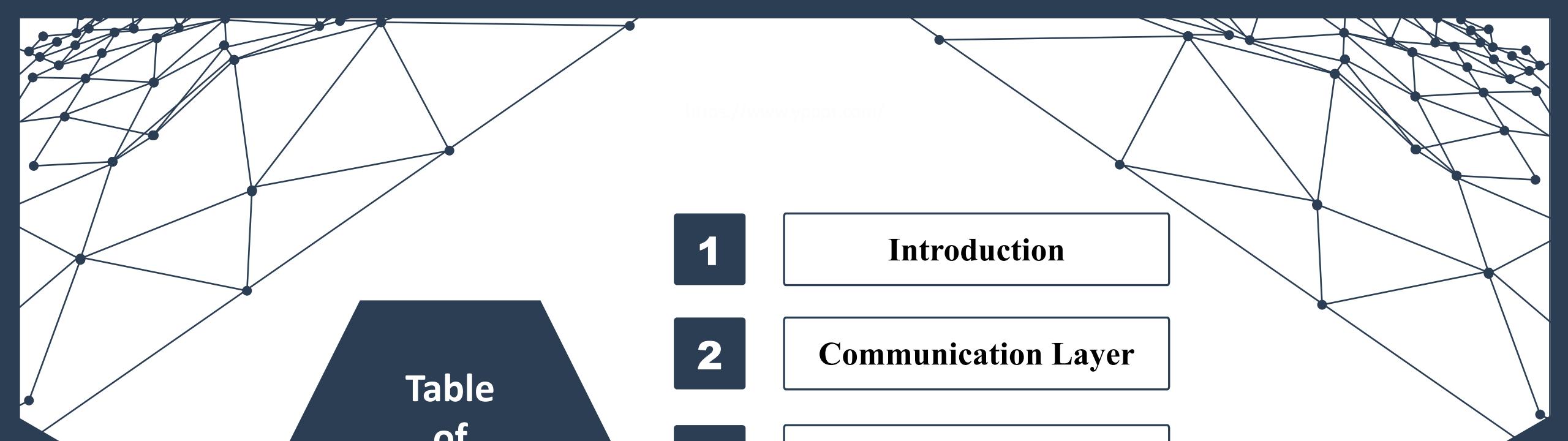
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(ERAU)

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(ERAU)

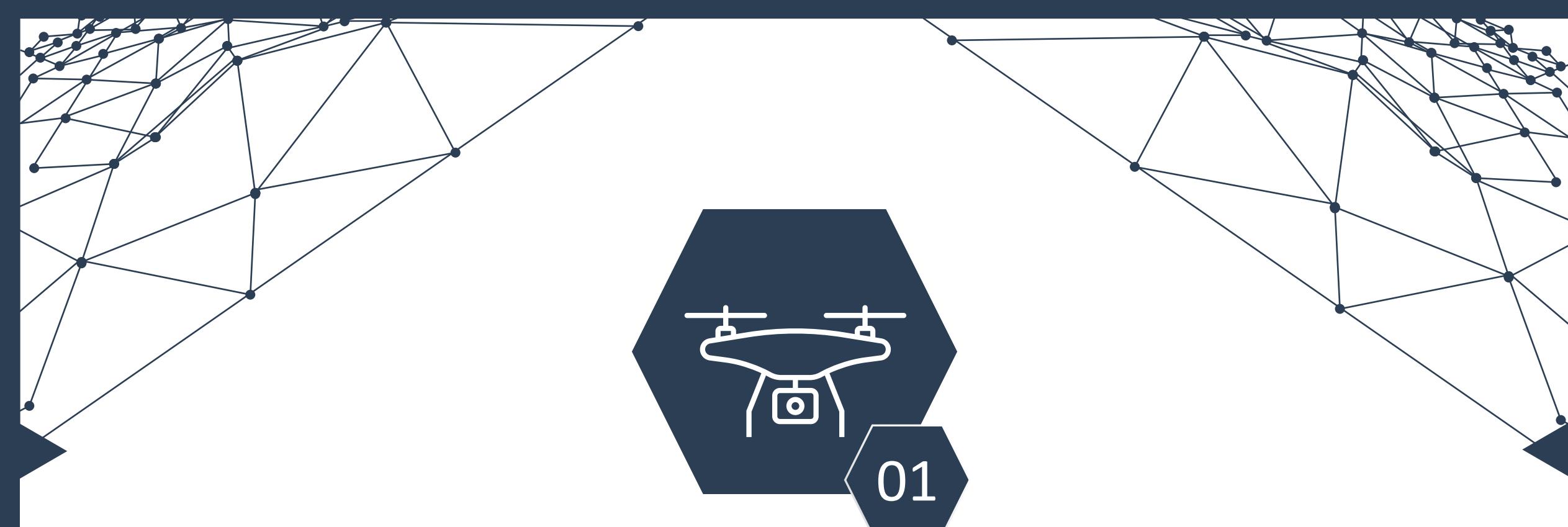




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- 4      Simulation**
- 5      Conclusion**



# Introduction

- Formation Control
- Preliminaries
- Schematic Diagram



# Formation Control

## Classical Formation Control

Agents typically perceive their absolute position relative to the global coordinate system and achieve their desired formation by actively controlling the absolute position.



## Communication-aware Formation Control

Wireless channel has been used in formation control since communications between agents are usually assumed to be ideal within a certain communication range.

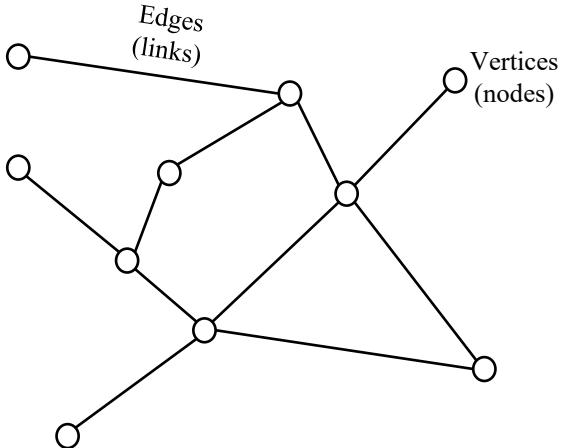


## Consensus-based Communication-aware Formation Control

In our research, we adopted ideas from [1], where author Li constructs a communication-aware formation controller that uses the communication channel quality, which is measured locally by agents to guide agents into a desired formation. Thus, it also optimizes the quality of communication of the formation system. Inspired by [2], We further constrains this formation control to reach a consensus between any pair of connected agents.



# Preliminaries



## Rigid Formation

The formation of groups of mobile agents in which all inter-agent distances remain constant is called **rigid**.

The **relative distance** between agent  $i$  and agent  $j$  is denoted by

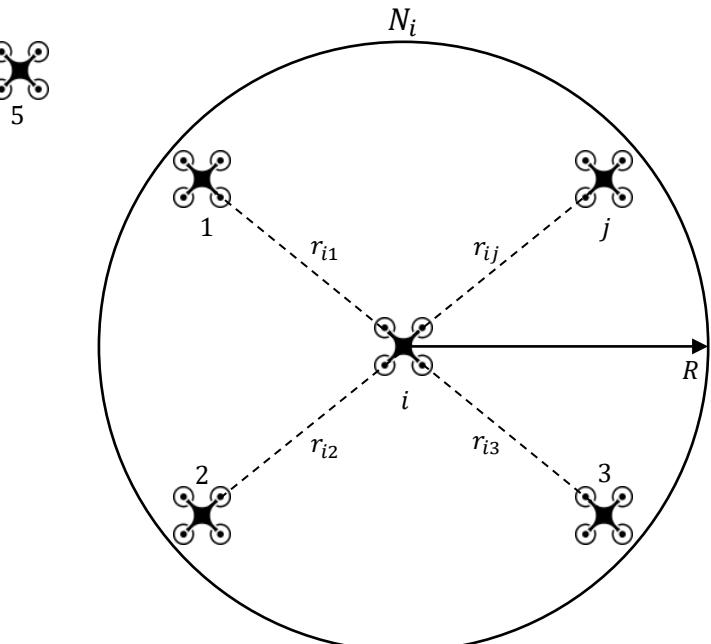
$$r_{ij} = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2} = \|q_i - q_j\|.$$

Let  $R > 0$  denote the **communication range** between two agents. The neighboring set of agent  $i$  can be denoted by

$$N_i = \{j \in \mathcal{V} \mid r_{ij} \leq R\}.$$

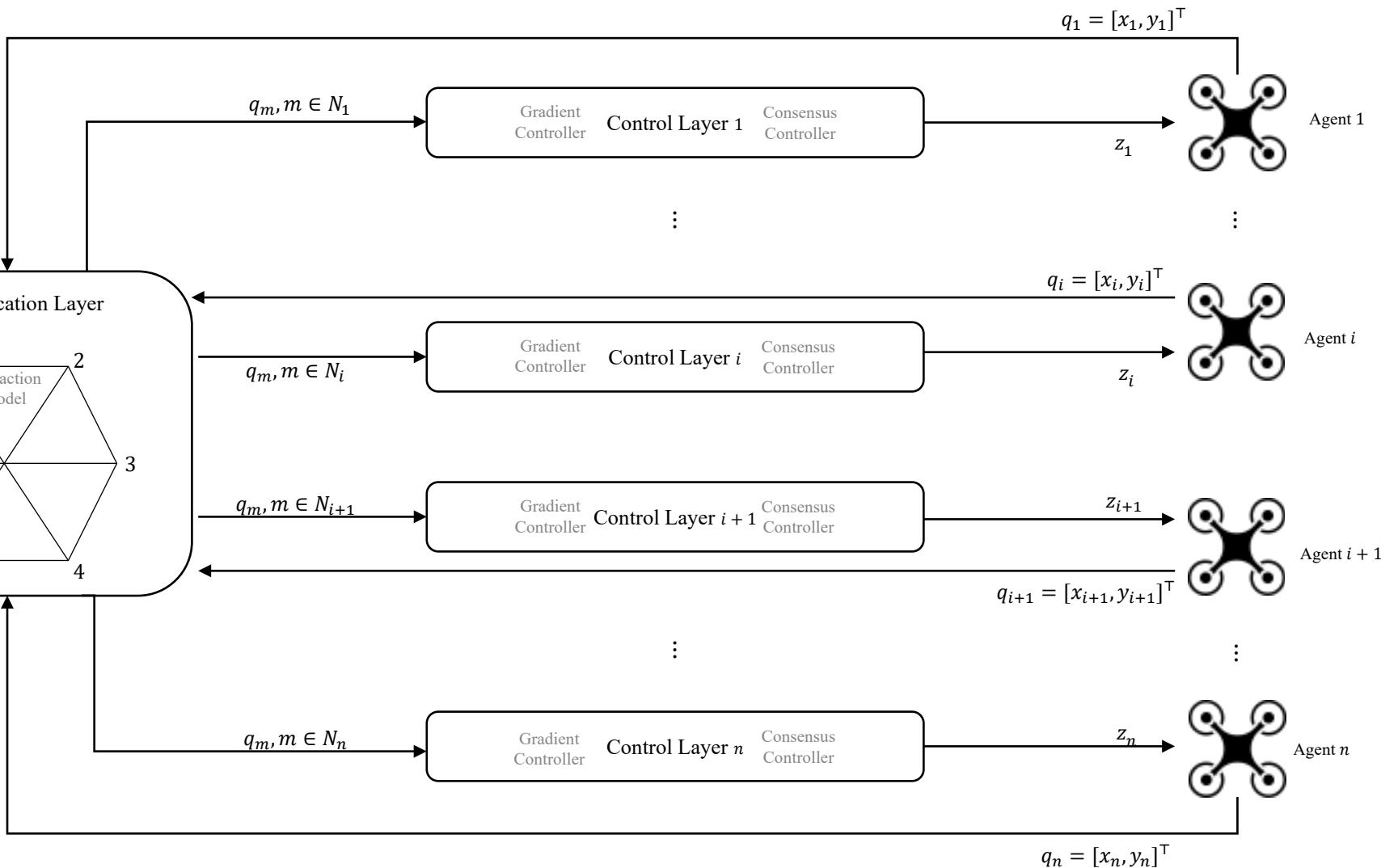
## Graph Theory

A **graph**  $G$  is a pair of  $(\mathcal{V}, \mathcal{E})$  consisting of a set of **vertices**  $\mathcal{V} = \{1, 2, \dots, i, \dots, j, \dots, n\}$  and a set of ordered pairs of the vertices called **edges**  $\mathcal{E} \subseteq \mathcal{V} \times \mathcal{V}$ . I.e.,  $\mathcal{E} = \{(i, j) \mid i, j \in \mathcal{V}, i \neq j\}$ . Here, we assume that  $G$  has no **self-edges** and **undirected**.





# Schematic Diagram

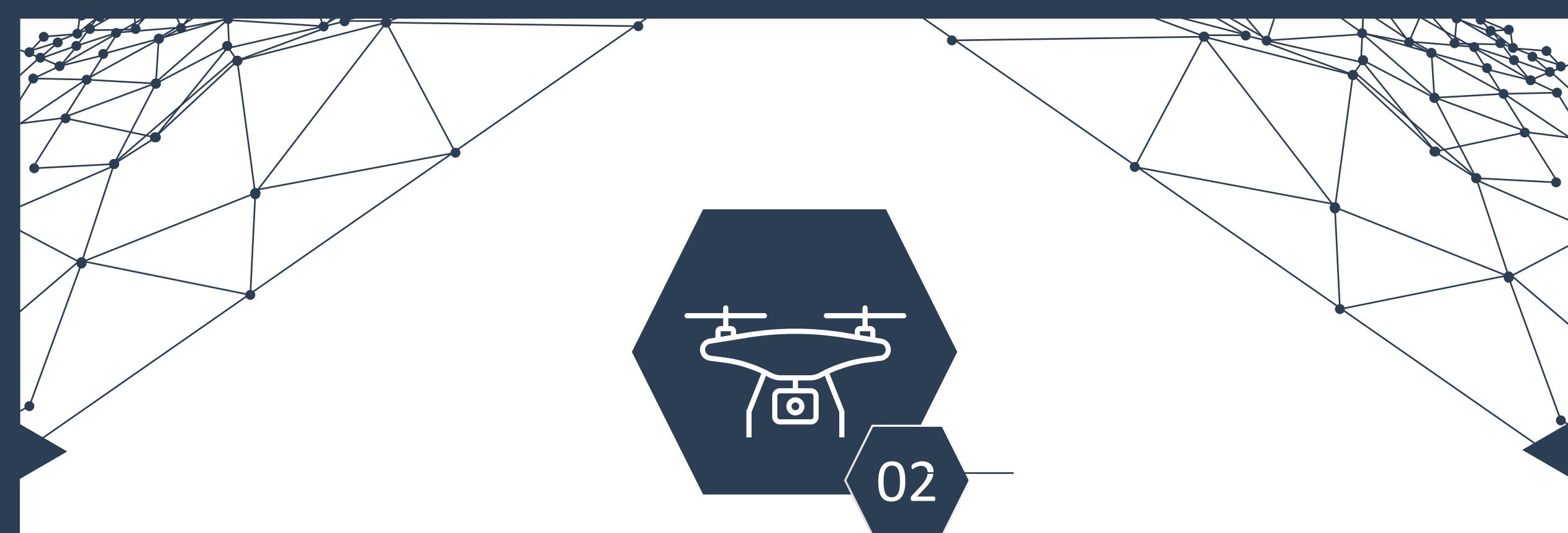


## System Dynamics

The **dynamics** of this multi-agent system is denoted by

$\dot{q}_i = z_i, \quad i = 1, 2, \dots, n,$   
where

$q$ : positions of agents,  
 $z$ : controls of agents.



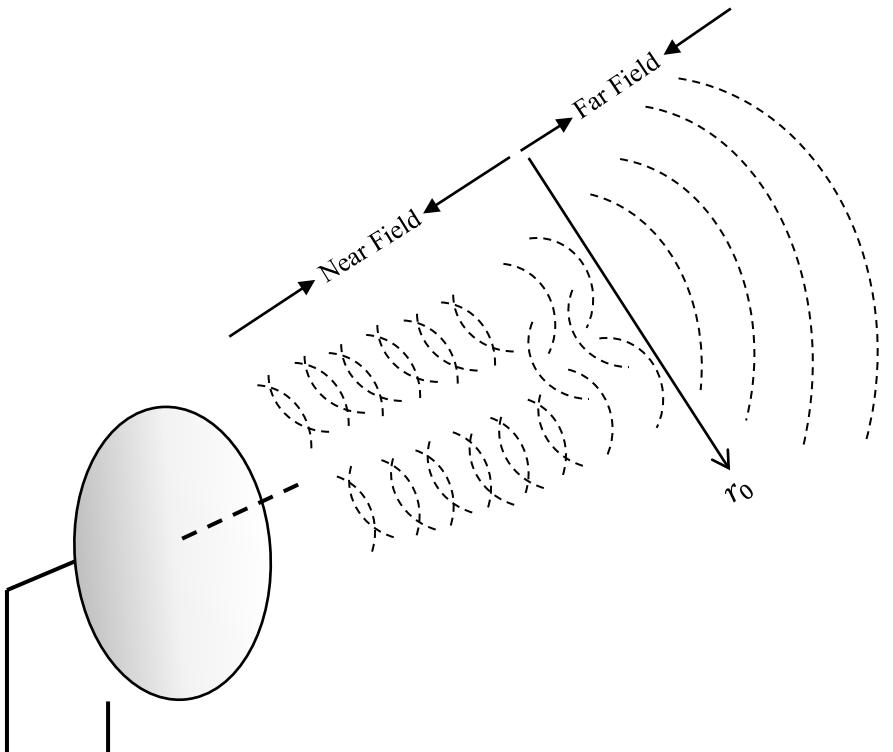
## Communication Layer

- ❑ Antenna Near-field and Far-field
- ❑ Interaction Model



# Antenna Near-field and Far-field

The antenna far field is the area away from the antenna. The boundary between antenna near-field and far-field is vaguely defined by the reference distance  $r_0$ .



## Far-field

The communication channel quality in antenna far-field is denoted by

$$f_{ij} = \exp\left(-\alpha(2^\delta - 1)\left(\frac{r_{ij}}{r_0}\right)^\nu\right),$$

where

$r_0$ : reference distance for antenna near-field,

$r_{ij}$ : Euclidean distance between agent  $i$  and agent  $j$ .

## Near-field

A simple model of antenna near-field communication quality is:

$$n_{ij} = \frac{r_{ij}}{\sqrt{r_{ij}^2 + r_0^2}}$$

where

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# Interaction Model

## Signal Scattering Effect

When a traveling wave encounters a change in the wave impedance, it will reflect, at least partially. If the reflection is not total, it will also partially transmit into the new impedance.

## Path Loss Effect

The reduction in power density (attenuation) of an electromagnetic wave as it propagates through space. As a result, the received signal power level is several orders below the transmitted power level.

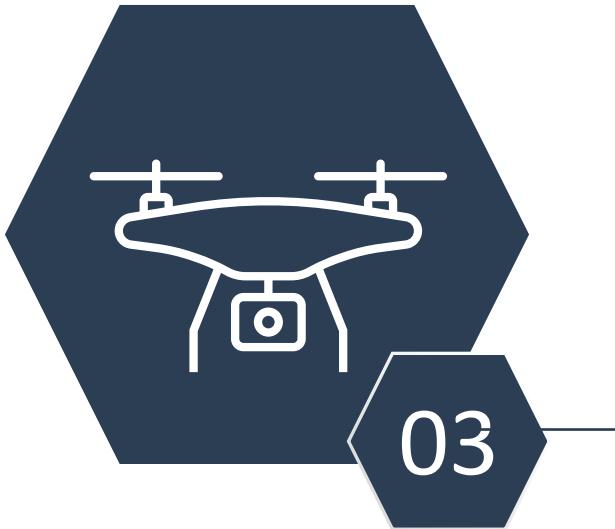
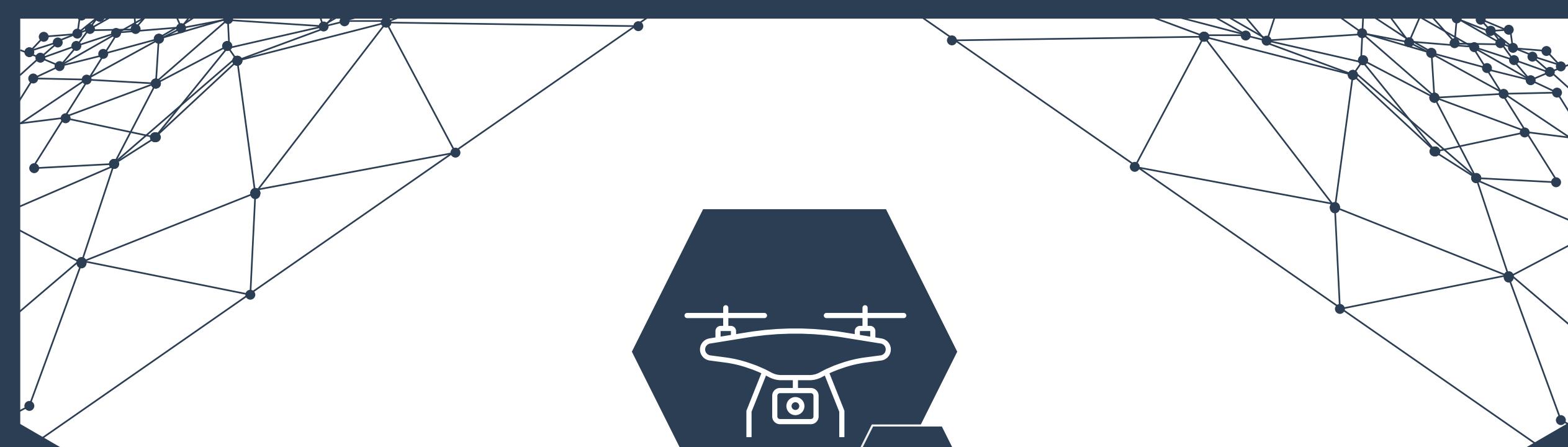
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## Interaction Model

The interaction model is denoted by

$$\phi(r_{ij}) = n_{ij} \cdot f_{ij} = \frac{r_{ij}}{\sqrt{{r_{ij}}^2 + {r_0}^2}} \cdot \exp\left(-\alpha(2^\delta - 1)\left(\frac{r_{ij}}{r_0}\right)^v\right).$$



## Control Layer

- Gradient Controller
- Unicycle Kinematic Model
- Consensus Controller



# Gradient Controller

In order to optimize the communication performance, the interaction model is designed to maximize its communication performance by taking the first-order derivative of interaction model we denote

$$\frac{d\phi}{dr_{ij}} = \varphi(r_{ij}) = \frac{-\beta\nu(r_{ij})^{\nu+2} - \beta\nu r_0^2(r_{ij})^\nu + r_0^{\nu+2}}{\sqrt{(r_{ij}^2 + r_0^2)^3}} \cdot \exp\left(-\beta\left(\frac{r_{ij}}{r_0}\right)^\nu\right),$$

where  $\beta = \alpha(2^\delta - 1)$ .

We find that interaction model has the best communication performance  $\phi^*$  at  $r_{ij}^*$ .

A gradient controller can be designed for agents converge in the formation with the maximized communication performance of function  $\phi(r_{ij})$ .

## Gradient Control Model

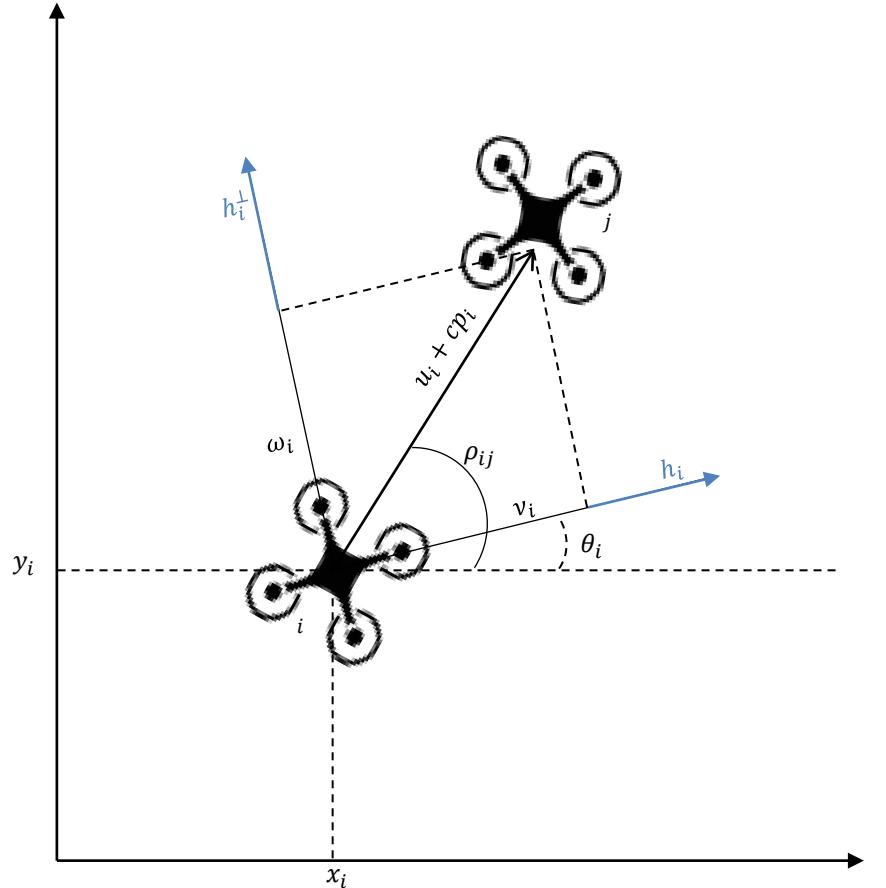
The gradient control model of agent  $i$  is denoted by

$$\mathcal{G}_i = \sum_{j \in N_i} [\nabla_{q_i} \phi(r_{ij})] = \sum_{j \in N_i} [\varphi(r_{ij}) \cdot e_{ij}]$$

where  $e_{ij} = (q_i - q_j)/\sqrt{r_{ij}}$ .



# Unicycle Kinematic Model



## Unicycle Kinematic Model

The unicycle kinematic model of agent  $i$  is denoted by

$$\dot{x}_i = v_i \cos(\theta_i)$$

$$\dot{y}_i = v_i \sin(\theta_i)$$

$$\dot{\theta}_i = \omega_i,$$

$h_i$ : Heading vector, defined as  $\begin{bmatrix} \cos(\theta_i) \\ \sin(\theta_i) \end{bmatrix}$

$h_i^\perp$ : Perpendicular heading vector, defined as  $\begin{bmatrix} -\sin(\theta_i) \\ \cos(\theta_i) \end{bmatrix}$

$\theta_i$ : Heading angle

$v_i$ : Linear velocity vector

$\omega_i$ : Angular velocity vector

$\rho_{ij}$ : Line of sight, defined as  $\arctan2(q_i - q_j)$

$u + cpi$ : consensus control vector

## Dubins Constraints

Due to its physical capabilities, the airspeed and heading angle of the UAV are limited. These physical limits can be represented by the constraints

$$\begin{aligned} v_{min} &\leq v_i \leq v_{max}, \\ |\omega_i| &\leq \omega_{max}, \end{aligned}$$



# Consensus Controller

## Consensus Control Model

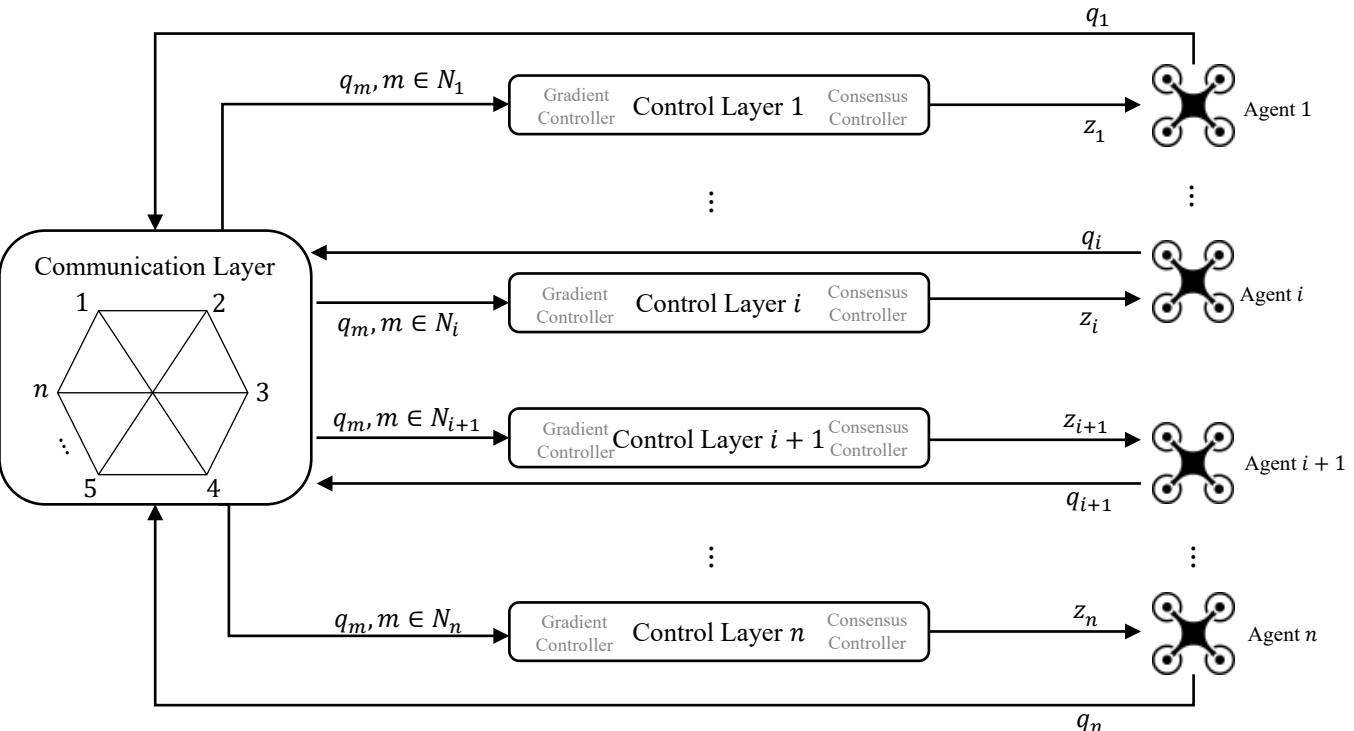
The projections of consensus control vector  $u + cp_i$  along the heading direction  $h_i$  and its perpendicular vector  $h_i^\perp$  are then calculated and used as the linear and angular velocity vectors, respectively. Specifically, the linear and angular velocity controls are given by

$$v_i = h_i^T(u + cp_i) \cos(\rho_{ij} - \theta_i)$$

$$\omega_i = h_i^{\perp T}(u + cp_i) \sin(\rho_{ij} - \theta_i).$$

And the consensus motion of agents  $i$  can be collectively expressed as

$$\mathcal{C}_i = h_i h_i^T(u + cp_i) \cos(\rho_{ij} - \theta_i).$$



## Final Formation Controller

$$z_i = \mathcal{G}_i + \mathcal{C}_i$$

$$= \sum_{j \in N_i} [\varphi(r_{ij}) \cdot e_{ij}] + \sum_{j \in N_i} [h_i h_i^T(u + cp_i) \cos(\rho_{ij} - \theta_i)]$$

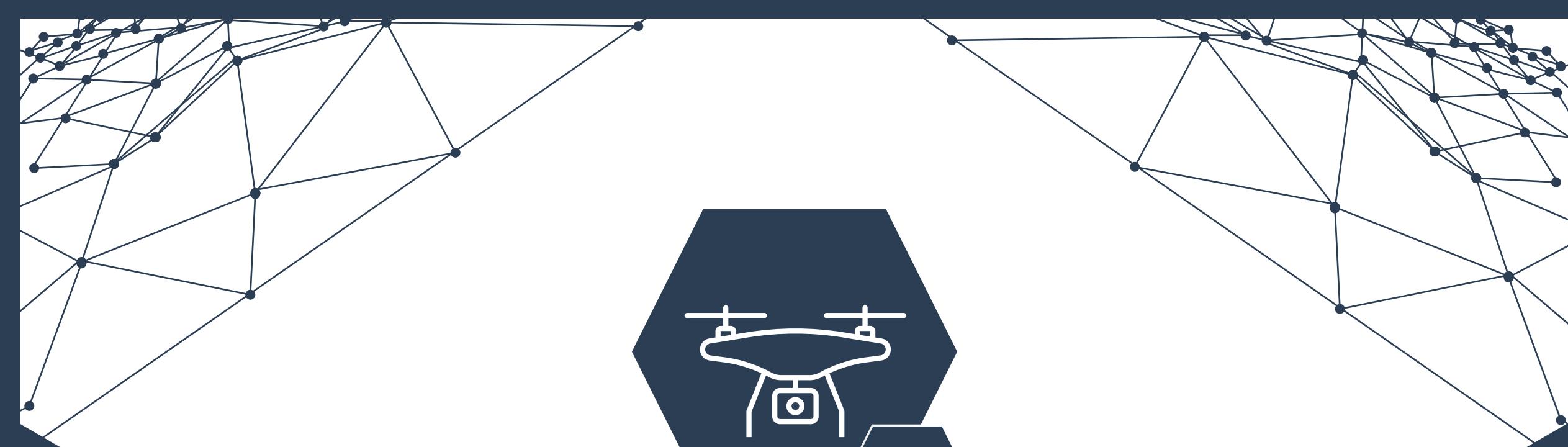
## Dynamics

The dynamics of this multi-agent system is denoted by

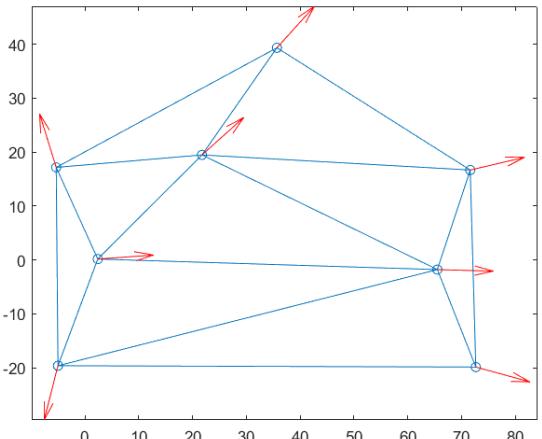
$$\dot{q}_i = z_i, \quad i = 1, 2, \dots, n,$$

where

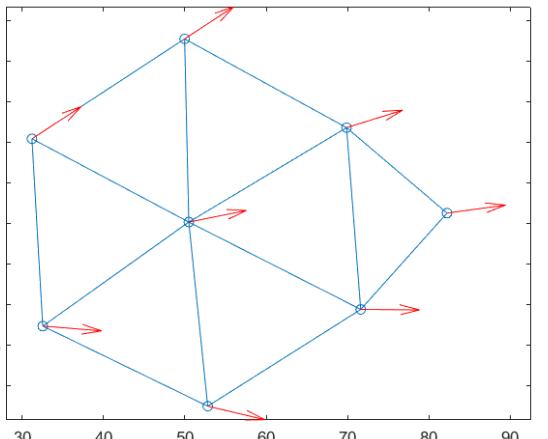
$q$ : position input of agents,  
 $z$ : control input of agents.



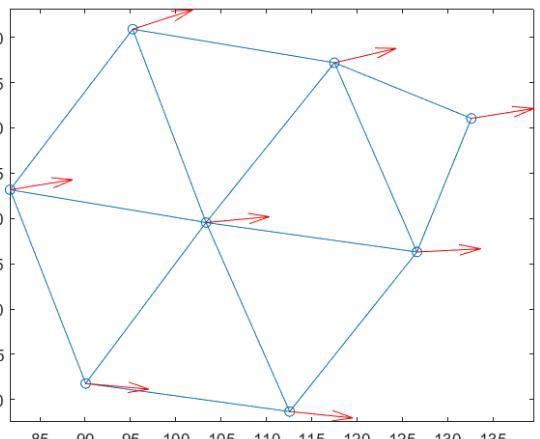
## Simulations



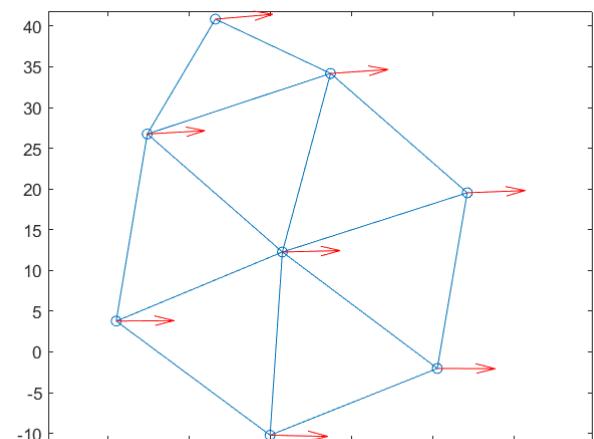
$t = 0s$



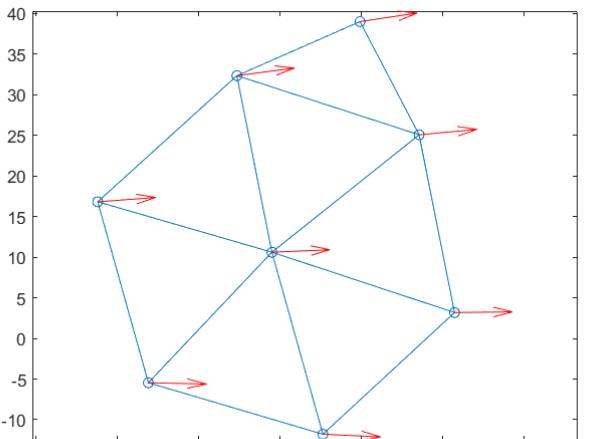
$t = 10s$



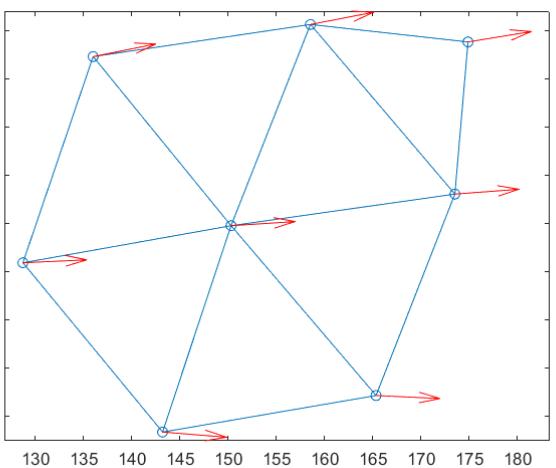
$t = 30s$



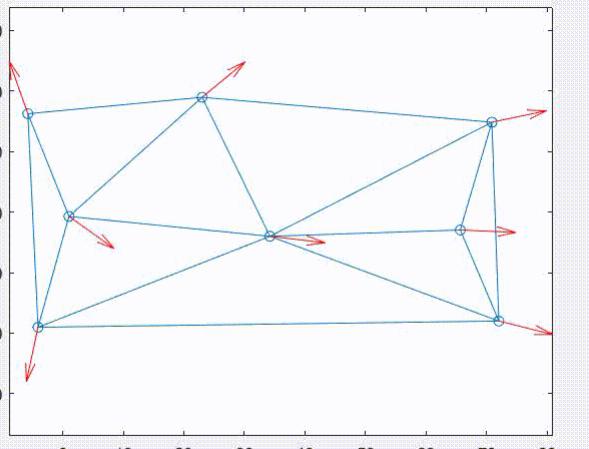
$t = 180s$



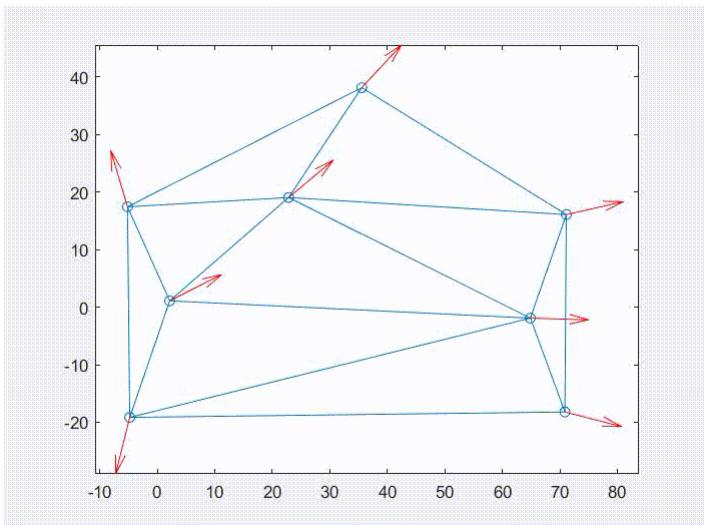
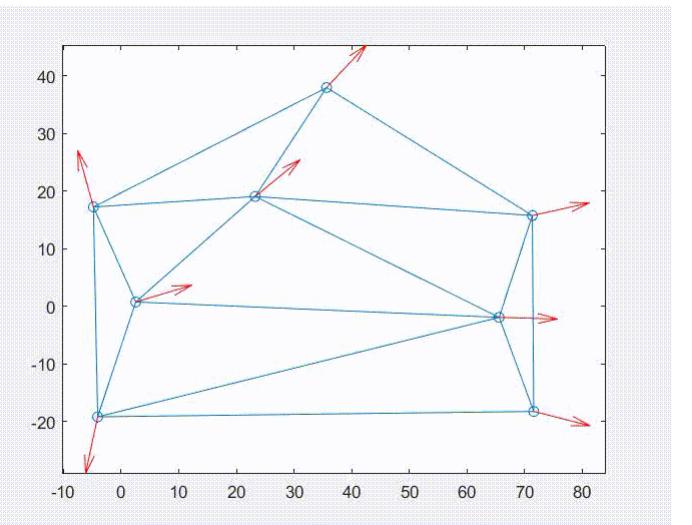
$t = 90s$

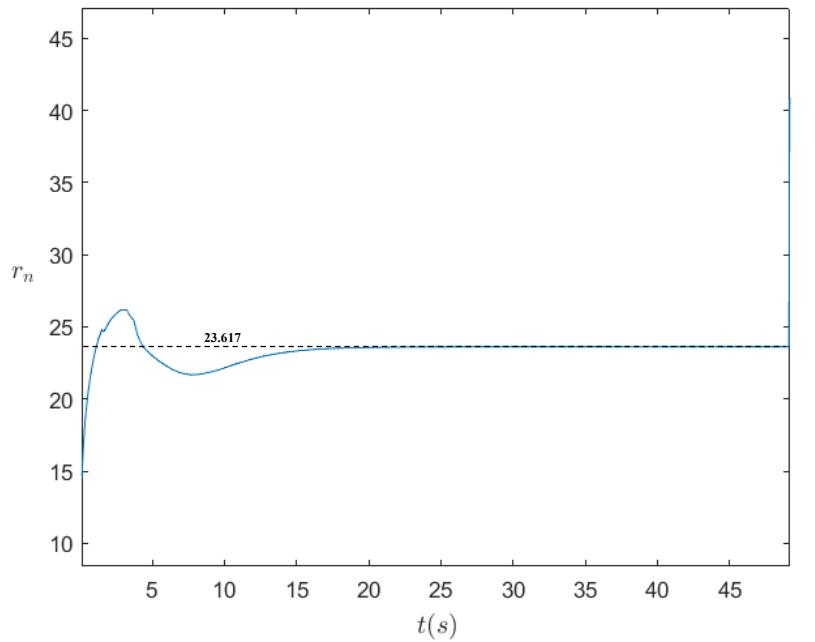


$t = 60s$



Before proposed

After Proposed  
Traveling in NE directionAfter Proposed  
Traveling in in E direction

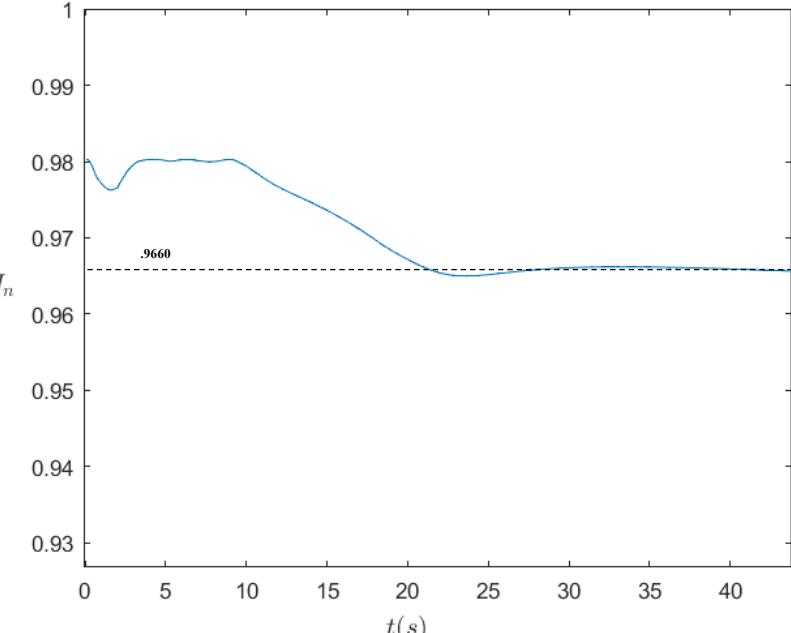
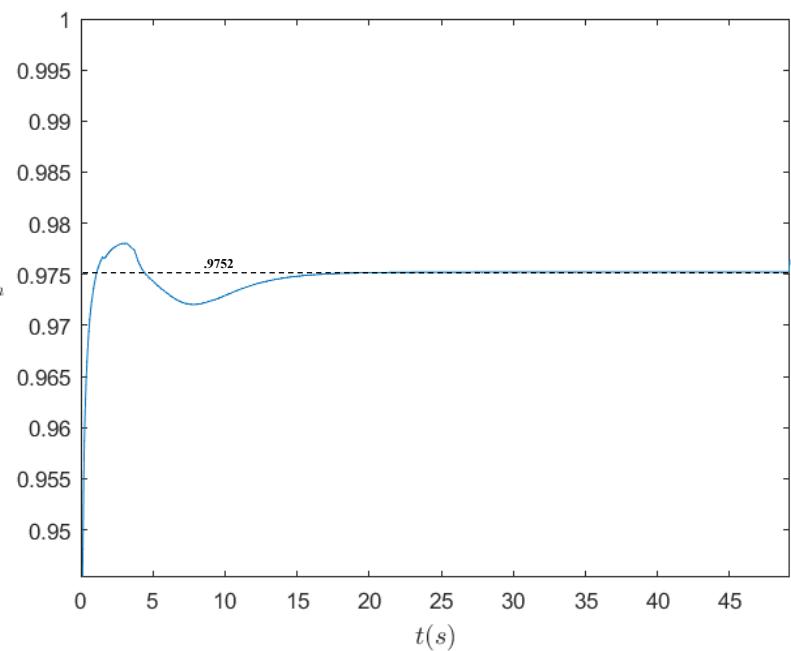
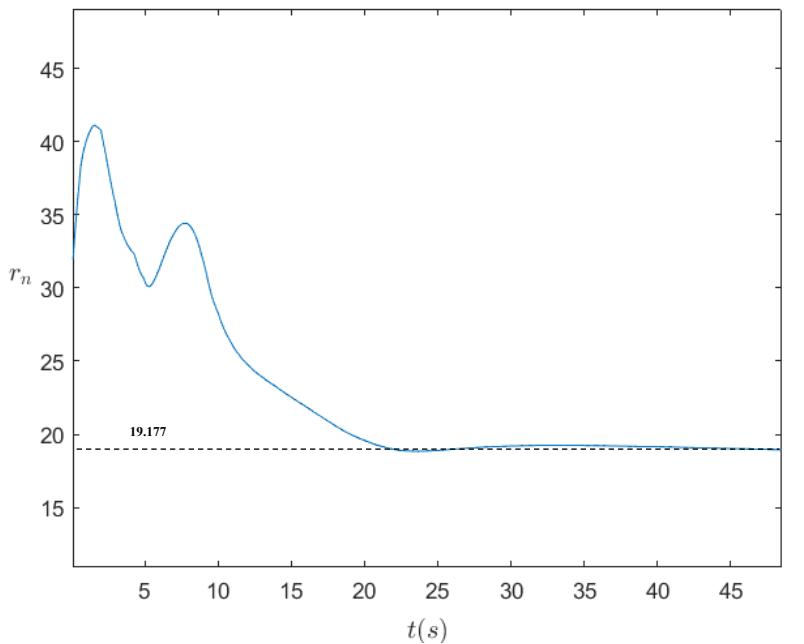


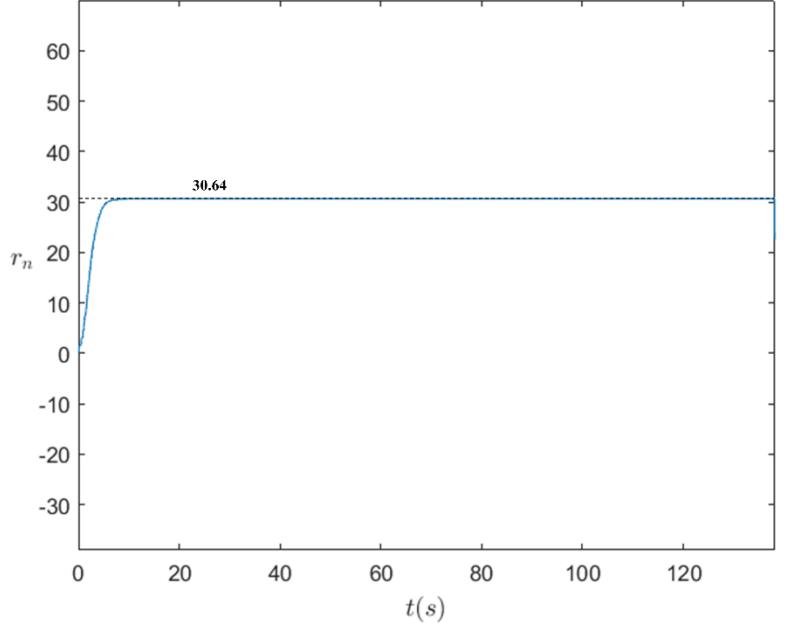
$J_n$ : Average Communication Performance Indicator

$$J_n = \frac{\sum_{i=1}^n \sum_{j \in N_i} \phi(r_{ij})}{2n|N_i|}$$

$r_n$ : Average Neighboring Distance Indicator

$$r_n = \frac{\sum_{i=1}^n \sum_{j \in N_i} r_{ij}}{2n|N_i|}$$





Agents Traveling in E Direction

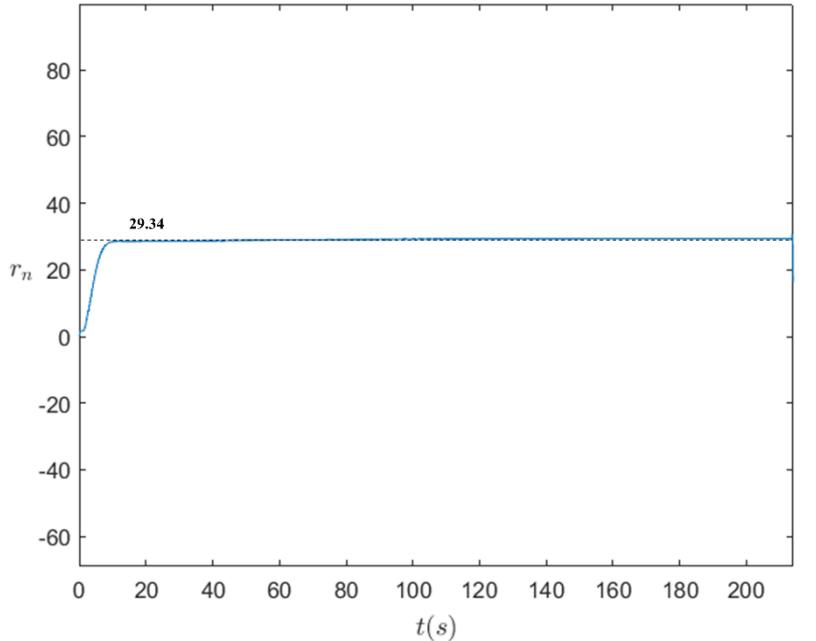
Before Proposed

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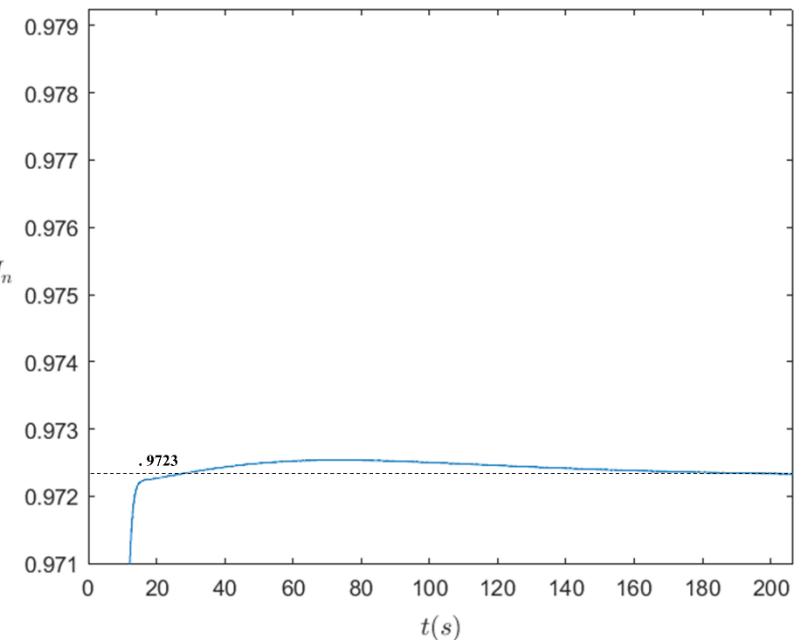
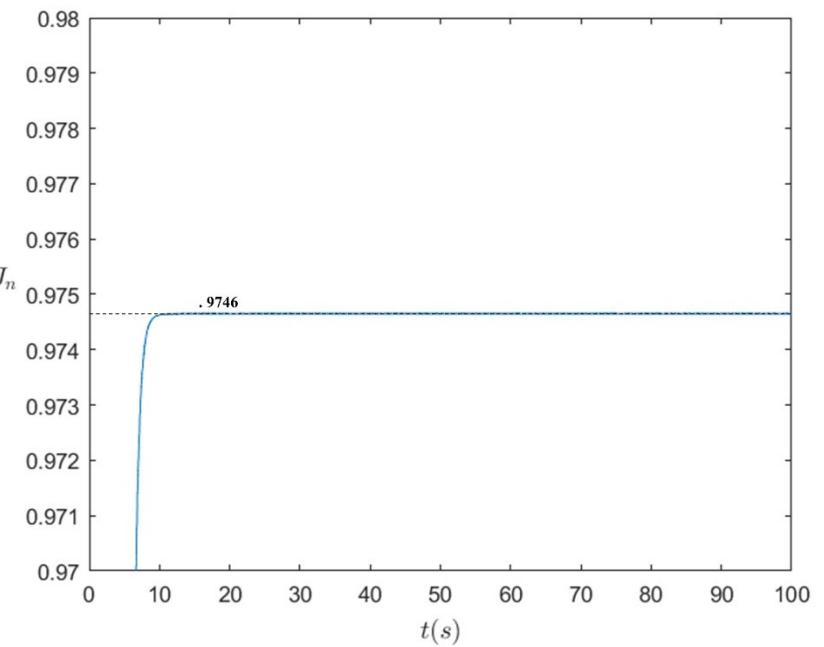
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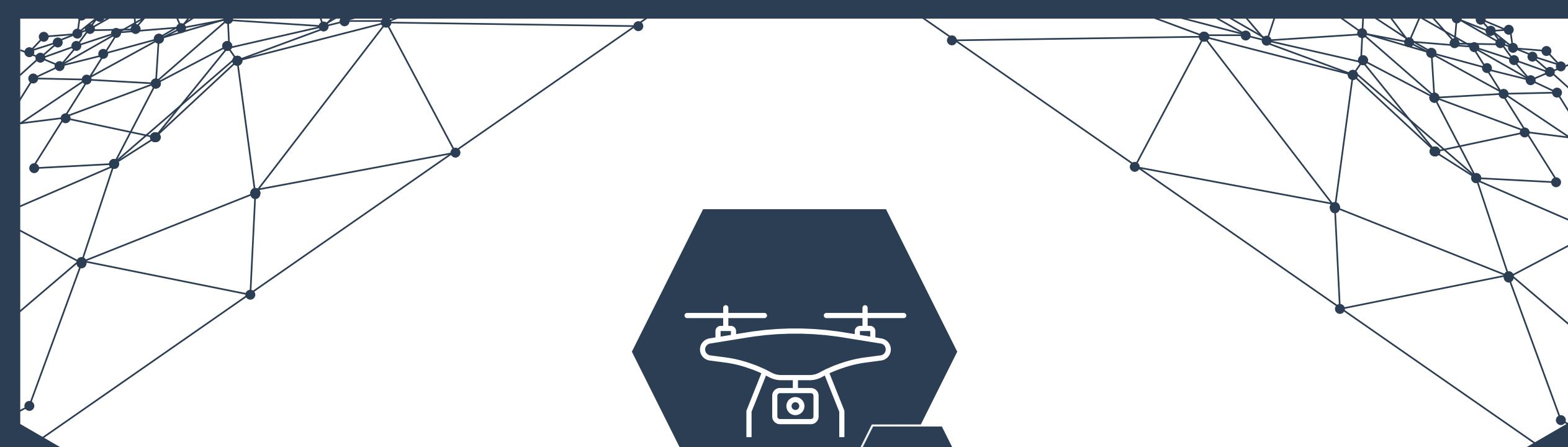
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After Proposed



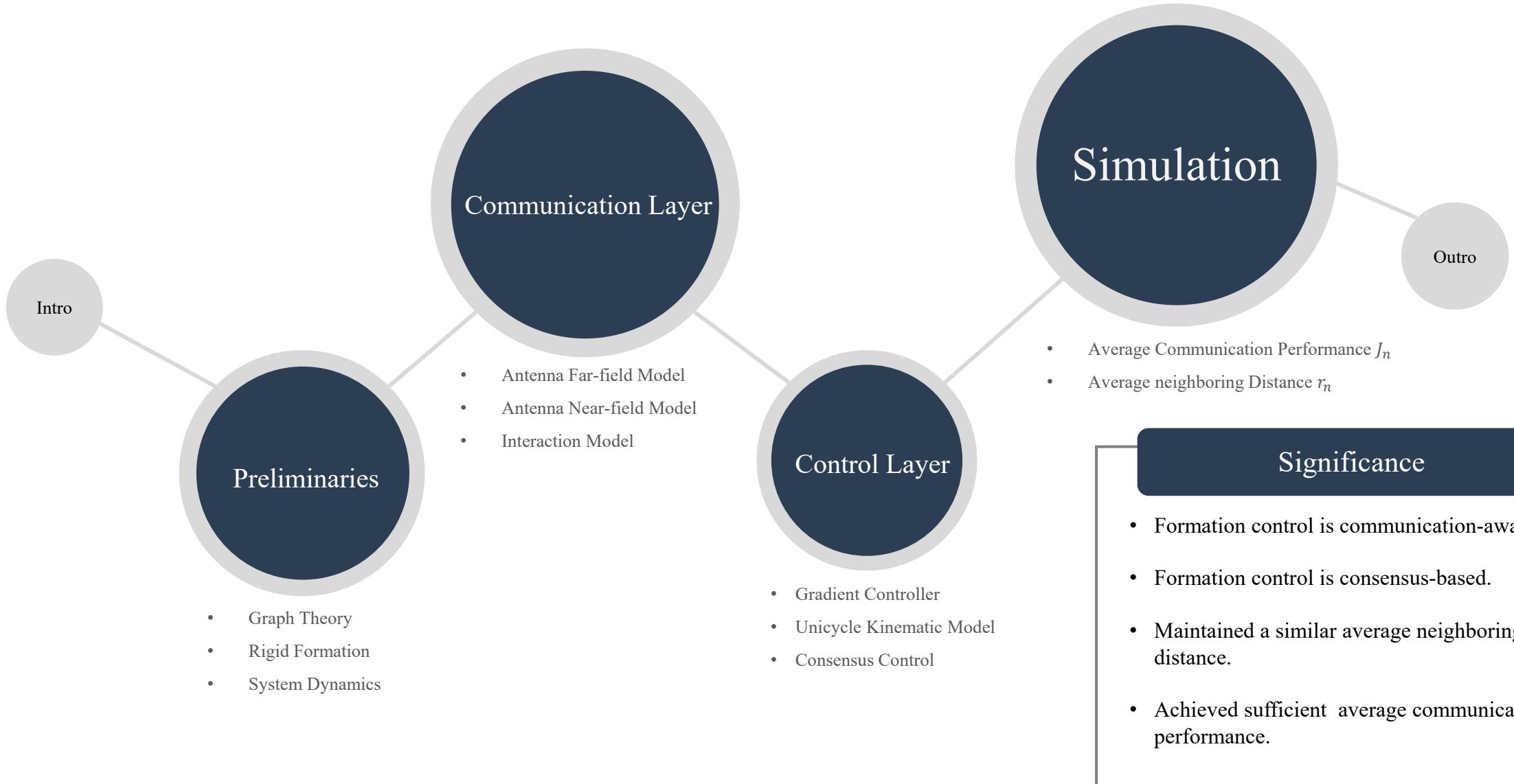


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## Conclusion

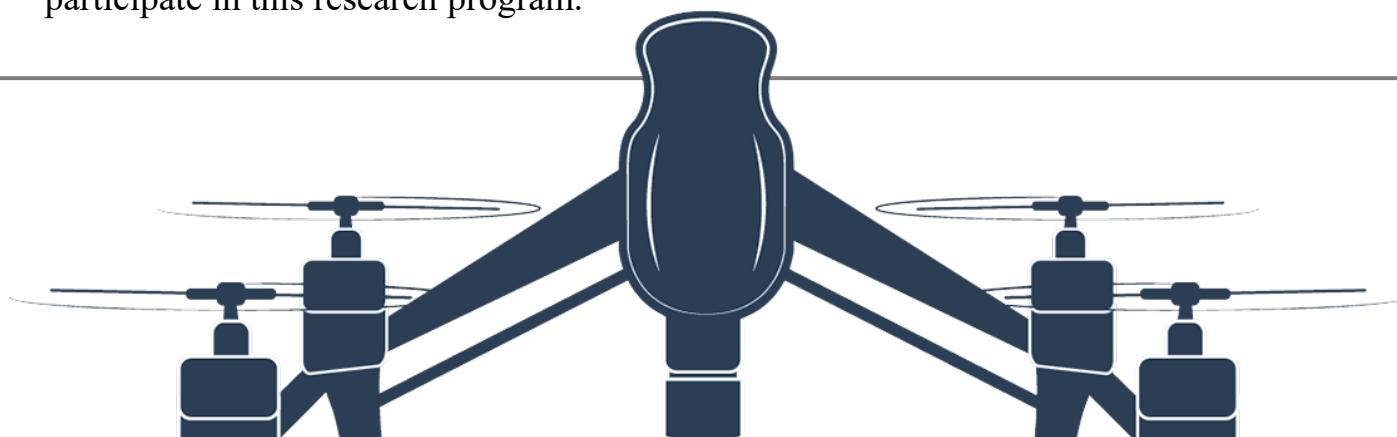


# Outline



## A c k n o w l e d g e m e n t s

- This research was supported by the National Science Foundation under Grant No. 2150213.
- I'd like to thank you Dr. Song and Dr. Stansbury for considering me into this research program.
- I'd also like to thank you Dr. Yang's guidance throughout the project.
- Last but not the least, I would thank to Dr. Jayasena Dr. Ofori-Boateng, and Dr. Taylor-Rodriguez for taking the time to write me a letter of recommendation which allowed me to participate in this research program.

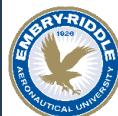
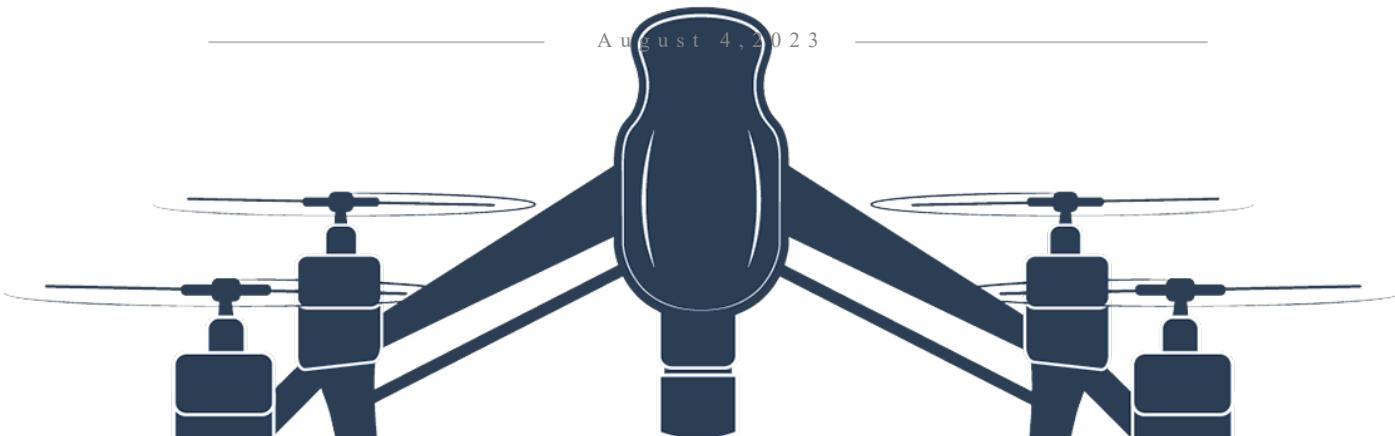


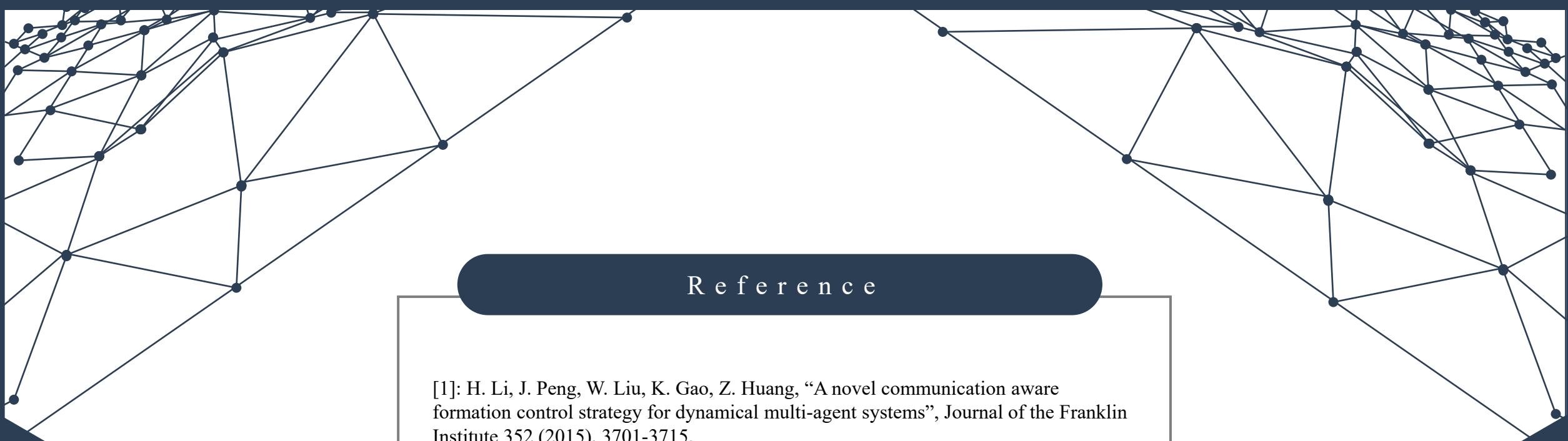


# THANKS

Q u e s t i o n s ?

— August 4, 2023 —



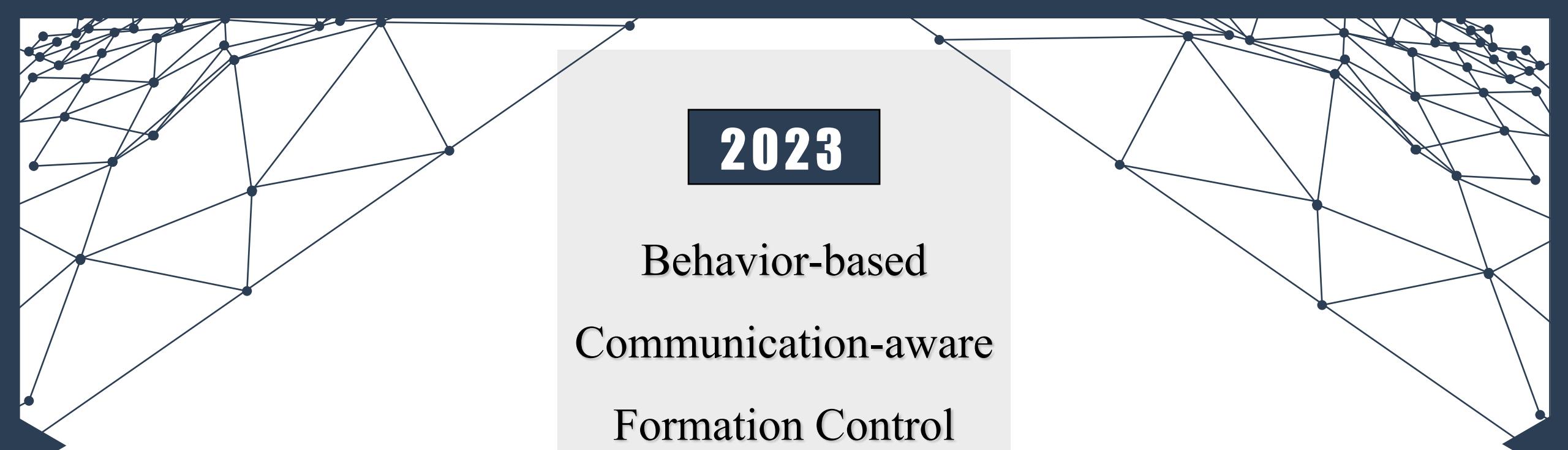


## R e f e r e n c e

[1]: H. Li, J. Peng, W. Liu, K. Gao, Z. Huang, “A novel communication aware formation control strategy for dynamical multi-agent systems”, Journal of the Franklin Institute 352 (2015), 3701-3715.

[2]: K. Fathian, T. H. Summers, and N. R. Gans, “Distributed formation control and navigation of fixed-wing UAVs at constant altitude,” in 2018 International Conference on Unmanned Aircraft Systems (ICUAS), 2018.





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# Behavior-based Communication-aware Formation Control

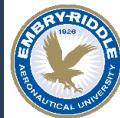
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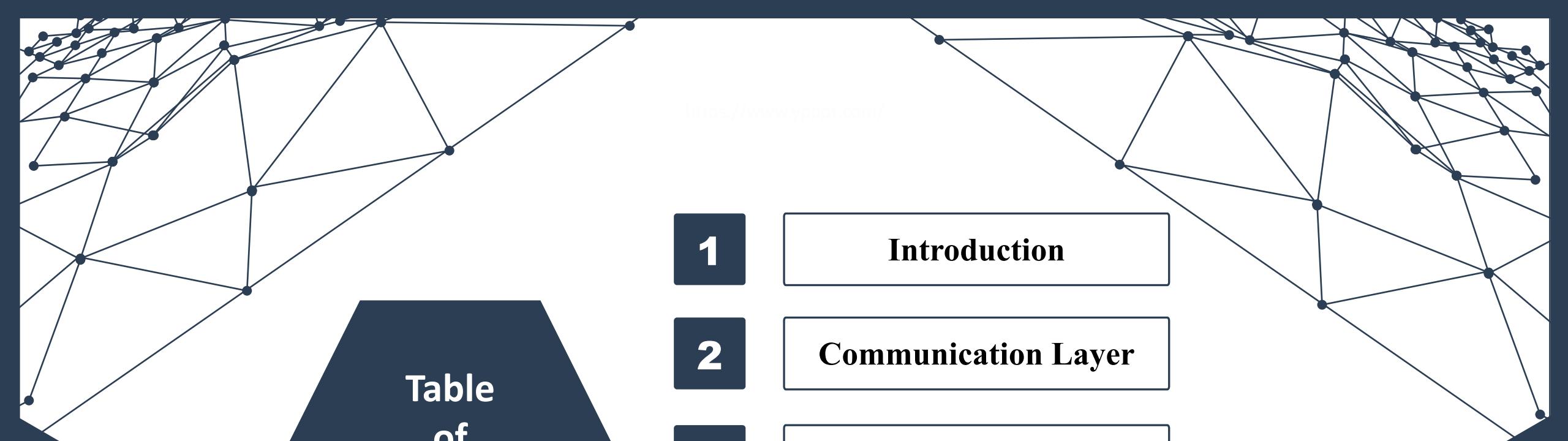
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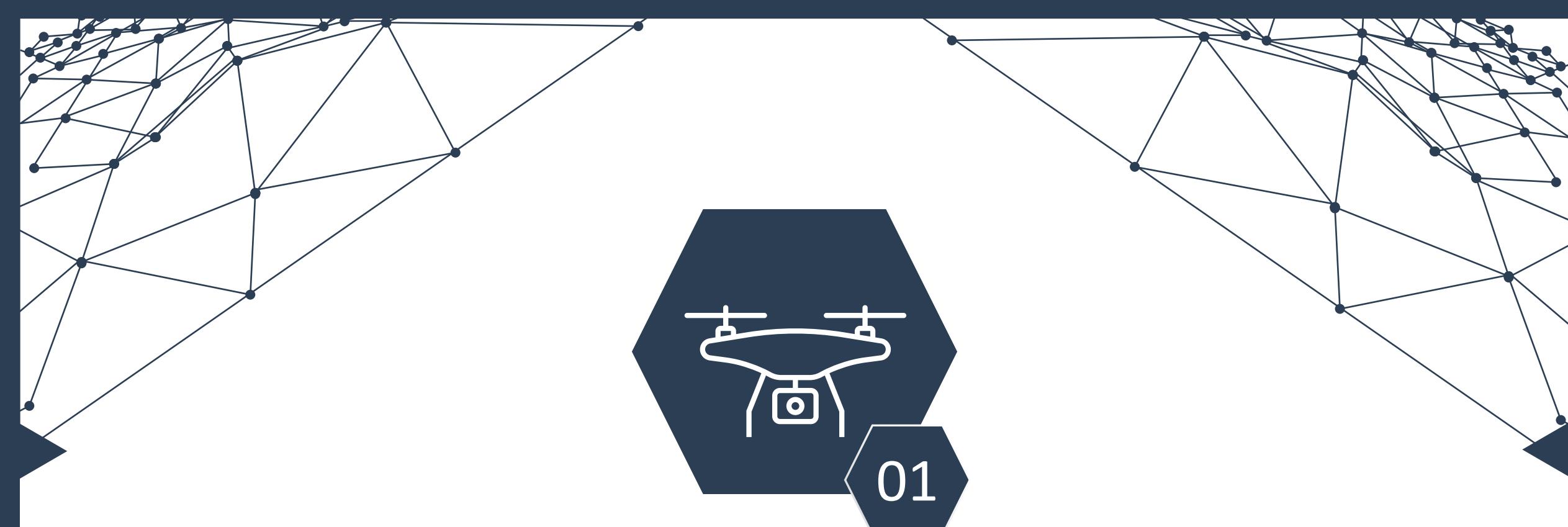




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# Introduction

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# Formation Control

## Tasks

1. How to navigate a swarm towards a destination?
2. How to avoid a jamming area without prior knowledge of its position?
3. How to achieve the above 2 tasks while maximizing communication quality?



# Formation Control

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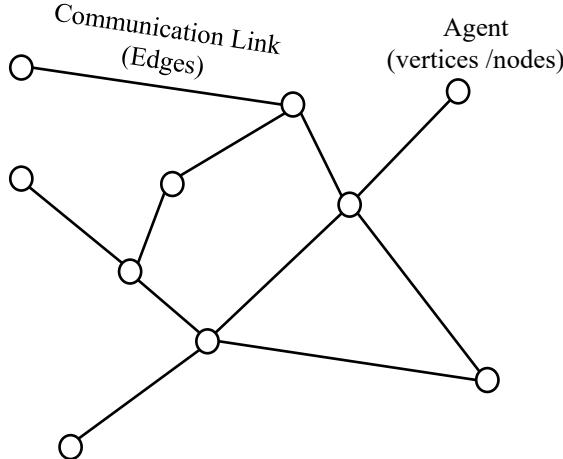


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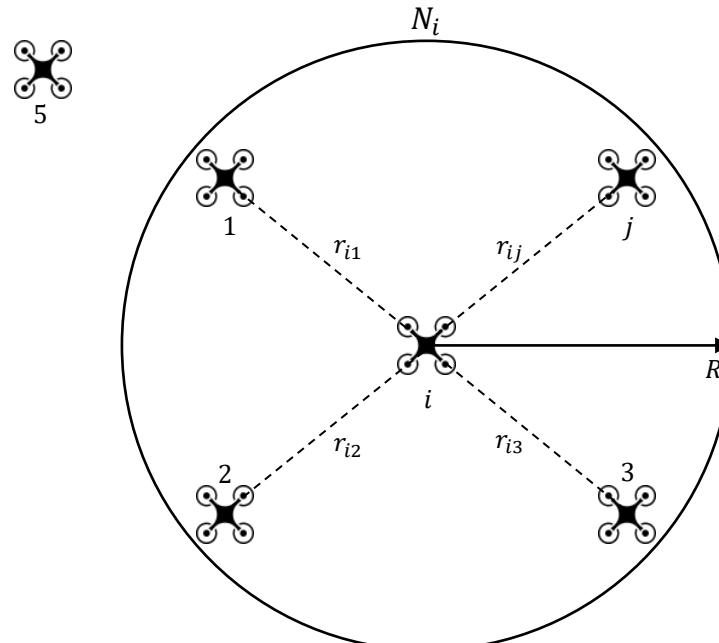
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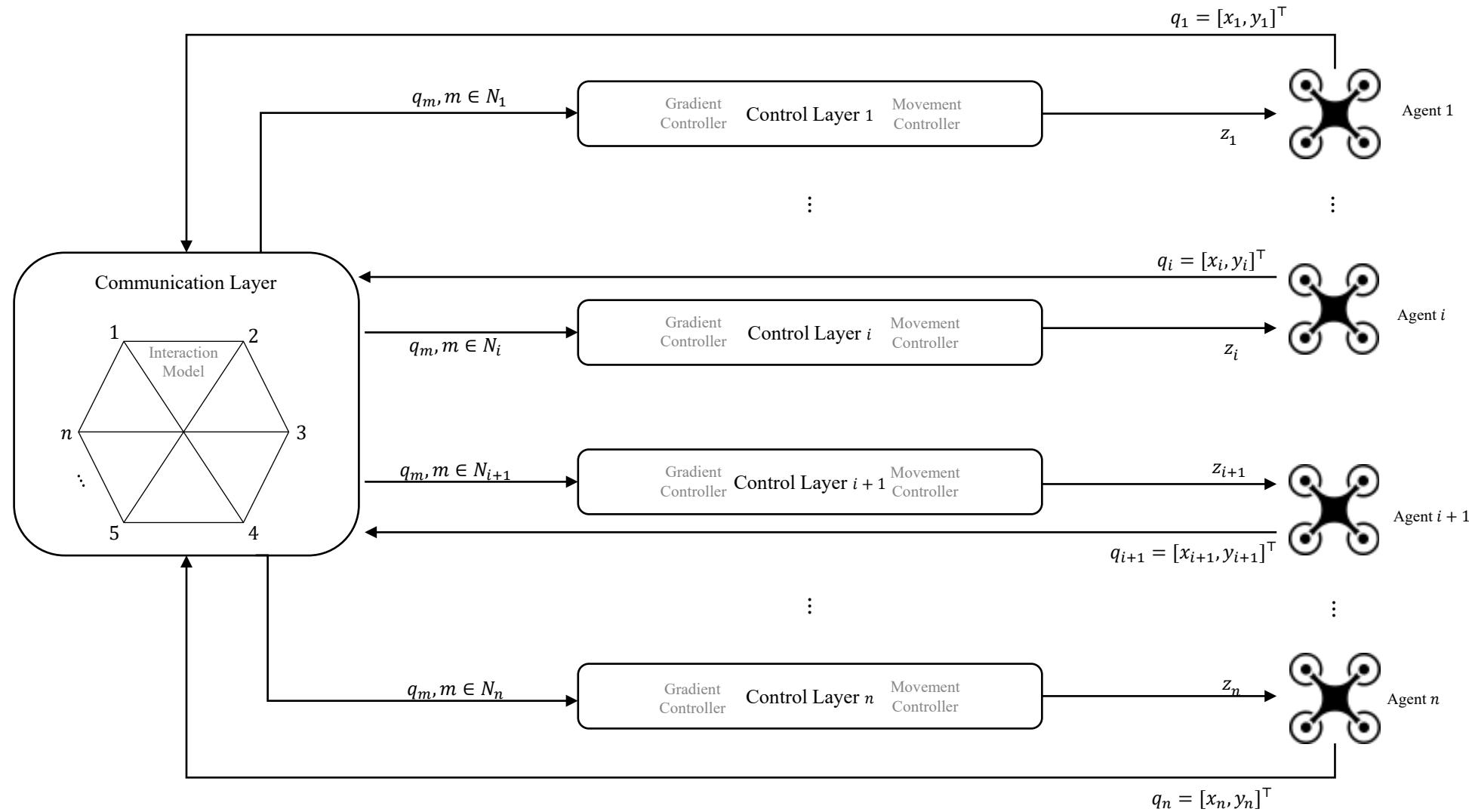
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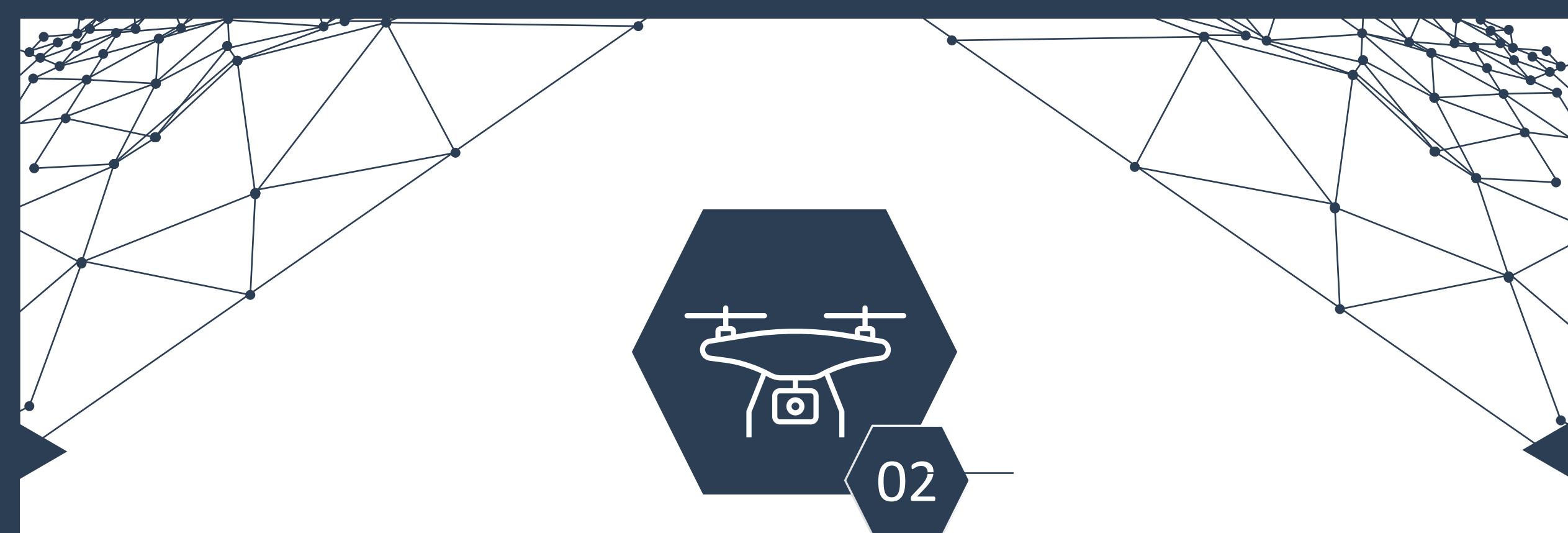
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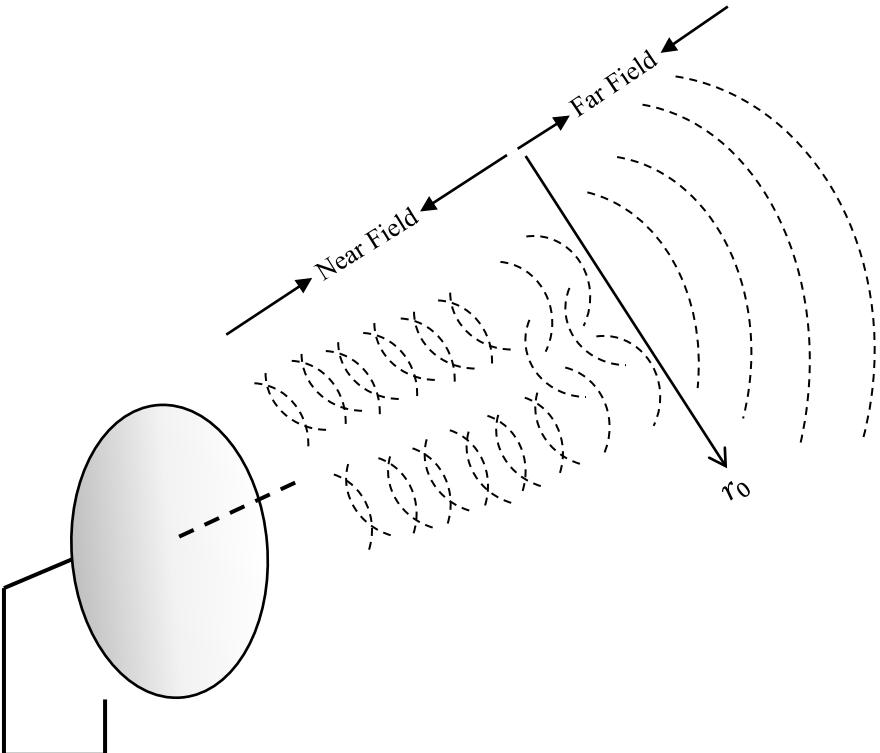
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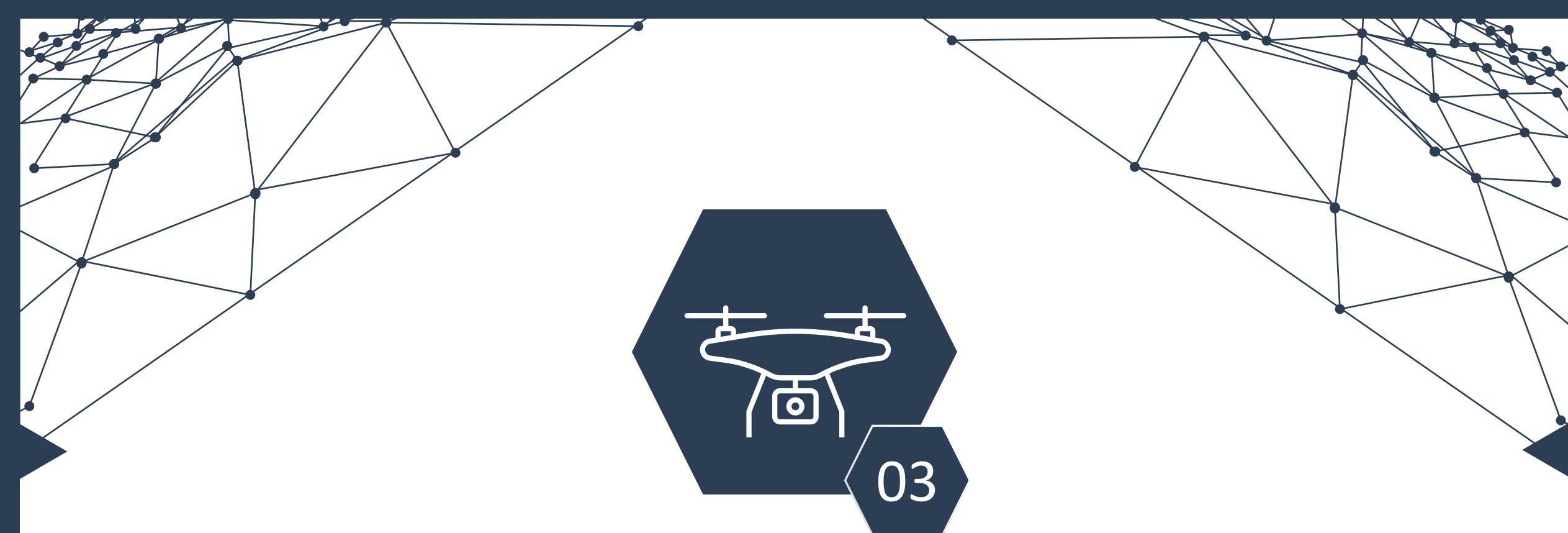
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## Control Layer

- Gradient Controller
- Movement Controller
  - Reach Goal Behavior
  - Jamming Avoidance Behavior
  - Edge Following Behavior



## Gradient Controller

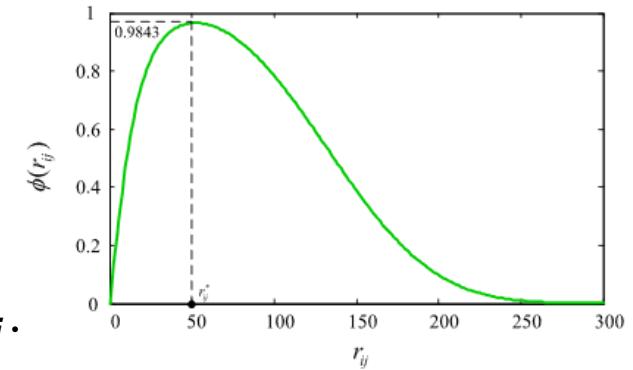
In order to optimize the communication performance, the interaction model is designed to maximize its communication performance by taking the first-order derivative of interaction model we denote

$$\frac{d\phi}{dr_{ij}} = \varphi(r_{ij}) = \frac{-\beta v (r_{ij})^{v+2} - \beta v r_0^2 (r_{ij})^v + r_0^{v+2}}{\sqrt{(r_{ij}^2 + r_0^2)^3}} \cdot \exp\left(-\beta \left(\frac{r_{ij}}{r_0}\right)^v\right),$$

where  $\beta = \alpha(2^\delta - 1)$ .

We find that interaction model has the best communication performance  $\phi^*$  at  $r_{ij}^*$ .

A gradient controller can be designed for agents converge in the formation with the maximized communication performance of function  $\phi(r_{ij})$ .



### Gradient Control Model

The gradient control model of agent  $i$  is denoted by

$$\mathcal{G}_i = \sum_{j \in N_i} [\nabla_{q_i} \phi(r_{ij})] = \sum_{j \in N_i} [\varphi(r_{ij}) \cdot e_{ij}]$$

where  $e_{ij} = (q_i - q_j)/\sqrt{r_{ij}}$ .



# Movement Controller: Reach Goal Behavior

## Description

Navigating the agents towards the destination.

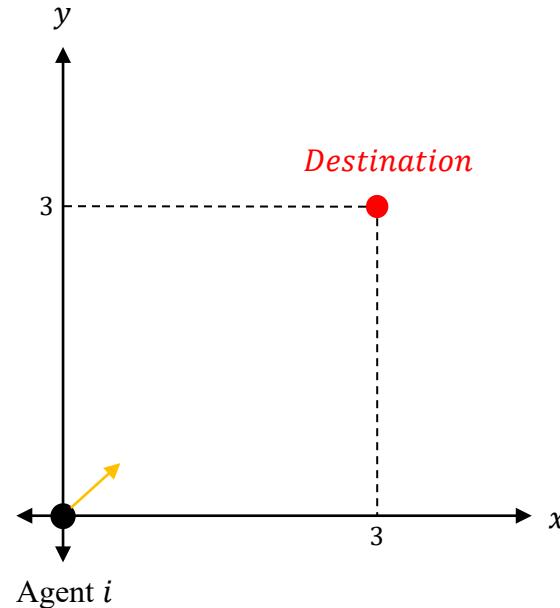
The behavior vector is calculated based on the current agent's coordinates and the destination coordinates.

The controlling parameter adjusts the agent's movement speed based on the distance between the agent and the target point.

## Formula

$$V_{navigation} = \frac{1}{\sqrt{(x_{dest} - x_i)^2 + (y_{dest} - y_i)^2}} [x_{dest} - x_i, y_{dest} - y_i]$$

$$f_1(d) = \begin{cases} a_m, & d > b_m \\ a_m \frac{d}{b_m}, & d \in [0, b_m] \end{cases}$$



$$\begin{aligned} V_{navigation} &= \frac{1}{\sqrt{(x_{dest} - x_i)^2 + (y_{dest} - y_i)^2}} [x_{dest} - x_i, y_{dest} - y_i] \\ &= \frac{1}{\sqrt{(3 - 0)^2 + (3 - 0)^2}} [3 - 0, 3 - 0] = \frac{1}{\sqrt{18}} [3, 3] = \left[ \frac{3}{\sqrt{18}}, \frac{3}{\sqrt{18}} \right] = [0.707, 0.707] \end{aligned}$$



# Movement Controller: Reach Goal Behavior

## Description

Navigating the agents towards the destination.

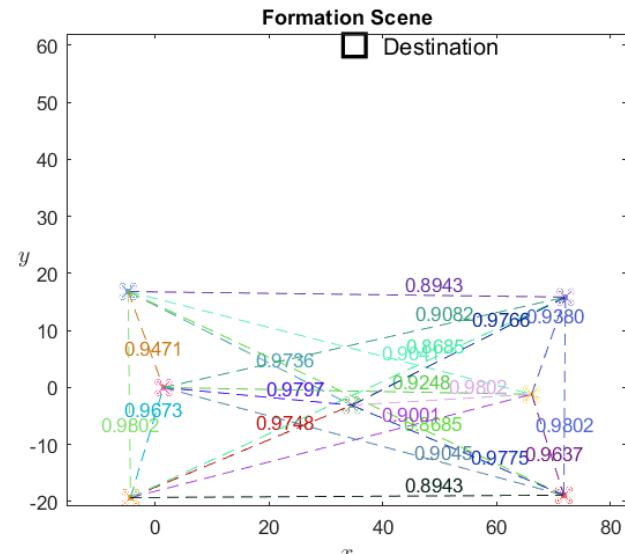
The behavior vector is calculated based on the current agent's coordinates and the destination coordinates.

The controlling parameter adjusts the agent's movement speed based on the distance between the agent and the target point.

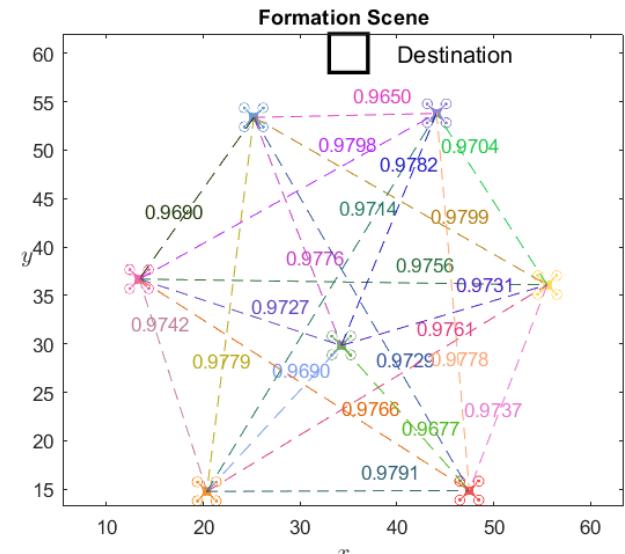
## Formula

$$V_{navigation} = \frac{1}{\sqrt{(x_{dest} - x_i)^2 + (y_{dest} - y_i)^2}} [x_{dest} - x_i, y_{dest} - y_i]$$

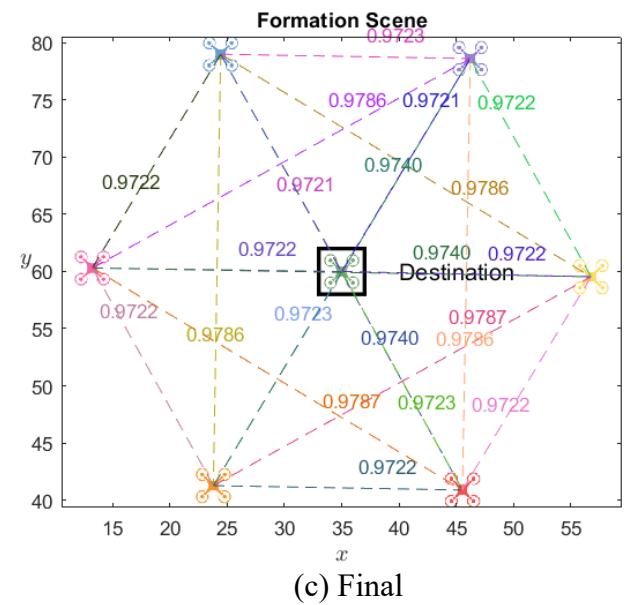
$$f_1(d) = \begin{cases} a_m, & d > b_m \\ a_m \frac{d}{b_m}, & d \in [0, b_m] \end{cases}$$



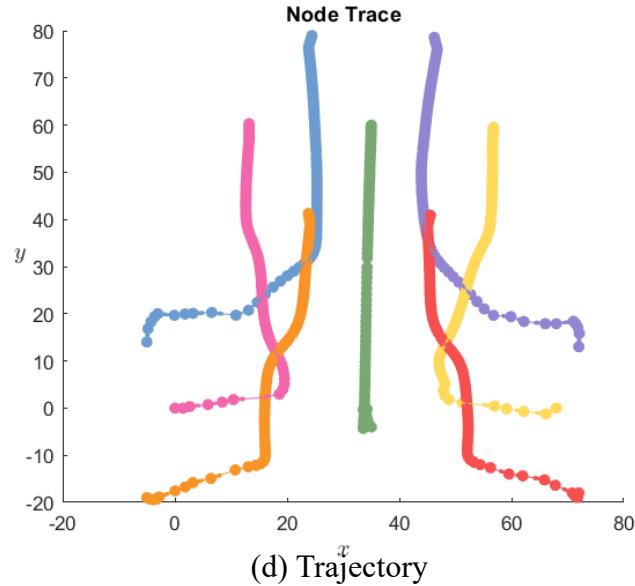
(a) Initial



(b) Halfway



(c) Final



(d) Trajectory



# Movement Controller: Jamming Avoidance Behavior

## Description

Enables agents to avoid jamming area in its path.

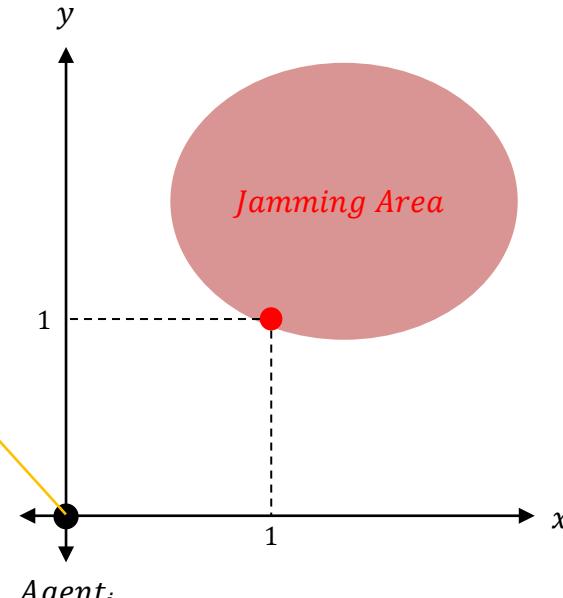
The behavior vector is determined by the distance between the agent and the nearest obstacle.

The controlling parameter adjusts the behavior to steer the agent away from the obstacle, with the sign of the vector determined by the obstacle's position relative to the agent.

## Formula

$$V_{avoidance} = \frac{1}{\sqrt{(x_{jam} - x_i)^2 + (y_{jam} - y_i)^2}} \begin{bmatrix} \pm y_{jam} - y_i \\ \mp x_{jam} - x_i \end{bmatrix}$$

$$f_2(d) = \begin{cases} a_0 \left( \frac{d}{b_f - b_0} + \frac{b_0}{b_0 - b_f} \right), & d \in [b_f, b_0] \\ 0, & otherwise \end{cases}$$



$$\begin{aligned} V_{avoidance} &= \frac{1}{\sqrt{(x_{jam} - x_i)^2 + (y_{jam} - y_i)^2}} \begin{bmatrix} \pm y_{jam} - y_i \\ \mp x_{jam} - x_i \end{bmatrix} \\ &= \frac{1}{\sqrt{(1 - 0)^2 + (1 - 0)^2}} \begin{bmatrix} +(1 - 0) \\ -(1 - 0) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 0.707 \\ -0.707 \end{bmatrix} \end{aligned}$$



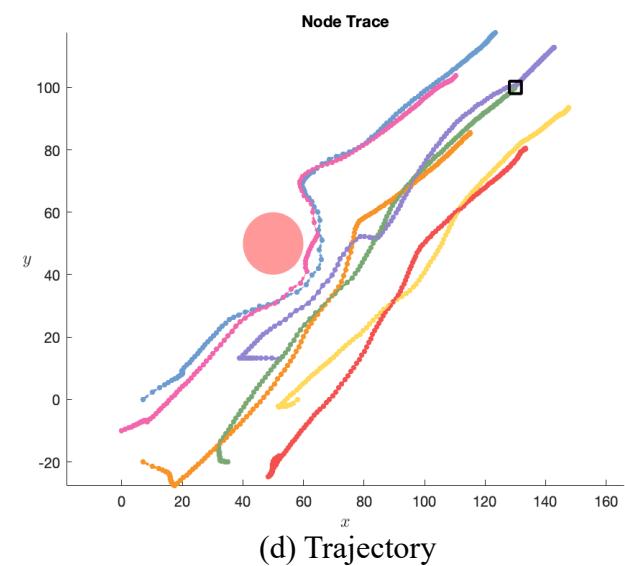
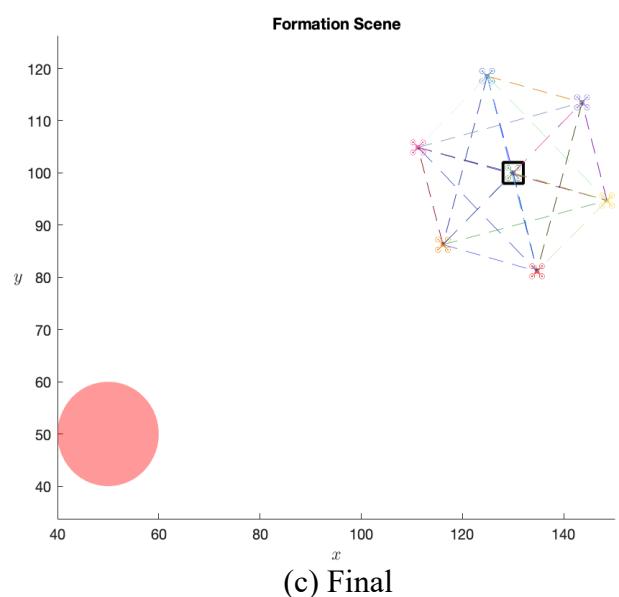
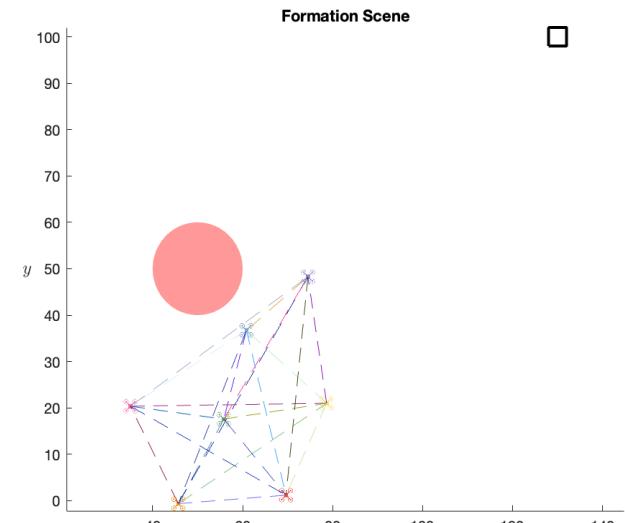
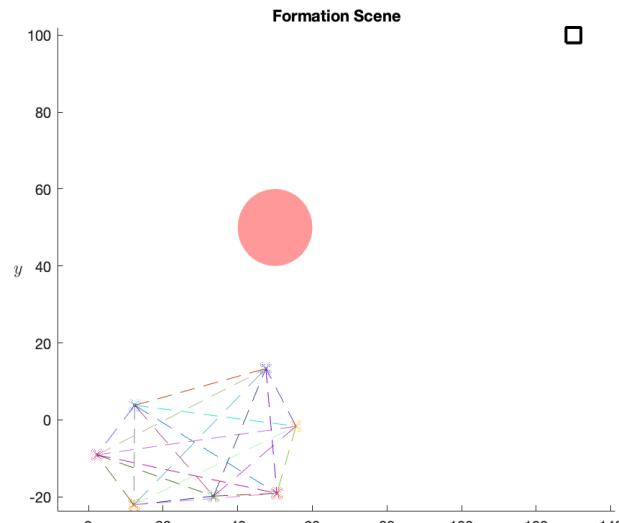
# Movement Controller: Jamming Avoidance Behavior

## Description

Enables agents to avoid jamming area in its path.

The behavior vector is determined by the distance between the agent and the nearest obstacle.

The controlling parameter adjusts the behavior to steer the agent away from the obstacle, with the sign of the vector determined by the obstacle's position relative to the agent.



## Formula

$$V_{avoidance} = \frac{1}{\sqrt{(x_{jam} - x_i)^2 + (y_{jam} - y_i)^2}} \begin{bmatrix} \pm y_{jam} - y_i \\ \mp x_{jam} - x_i \end{bmatrix}$$

$$f_2(d) = \begin{cases} a_0 \left( \frac{d}{b_f - b_0} + \frac{b_0}{b_0 - b_f} \right), & d \in [b_f, b_0] \\ 0, & otherwise \end{cases}$$



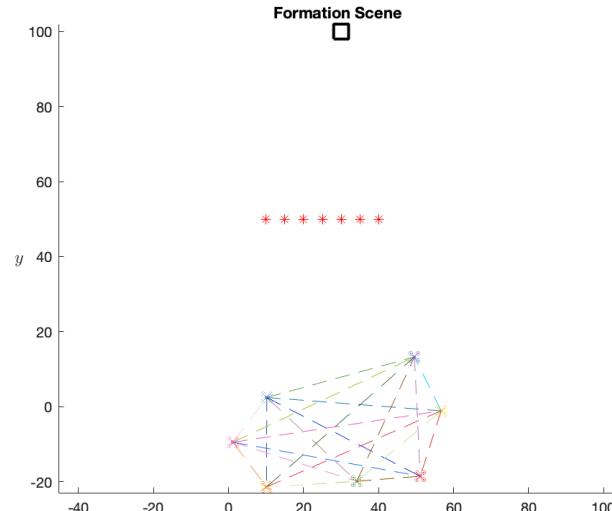
# Movement Controller: Edge Following Behavior

## Description

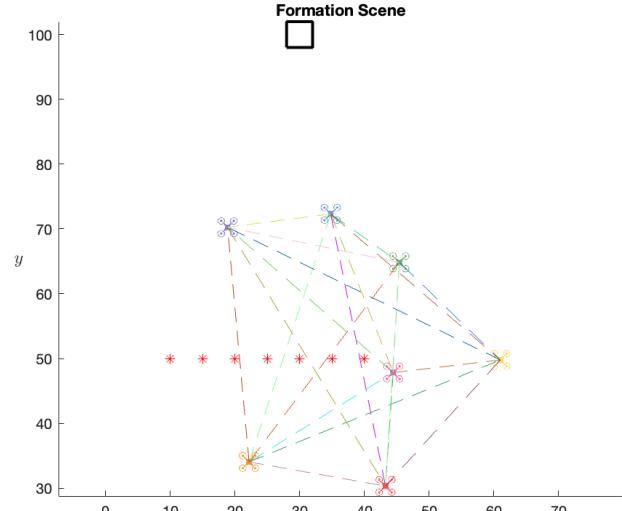
Helps the agents navigate by following jamming area edges.

The behavior vector is calculated based on the coordinates of the nearest obstacle and the agent's current position.

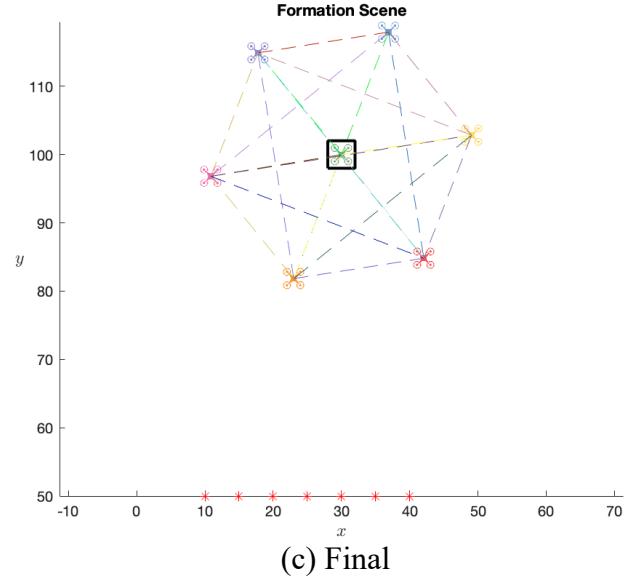
The controlling parameter activates the behavior and determines the direction of edge following based on the obstacle's position relative to the agent.



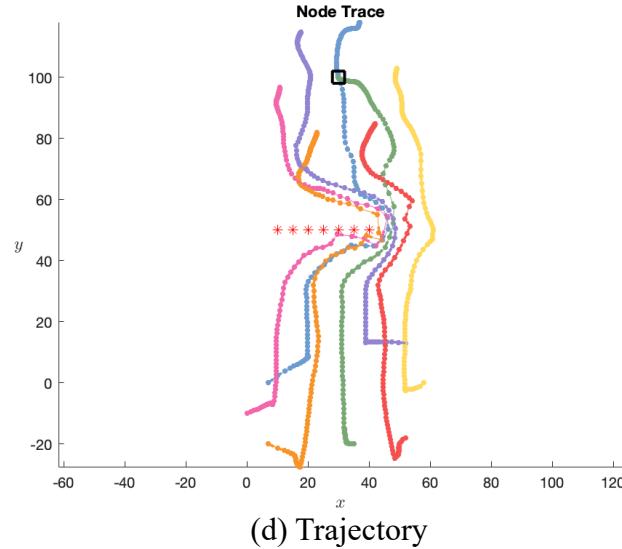
(a) Initial



(b) Halfway



(c) Final



(d) Trajectory

## Formula

$$V_{edge-following} = \frac{1}{\sqrt{(x_{jam} - x_i)^2 + (y_{jam} - y_i)^2}} \begin{bmatrix} \pm y_{jam} - y_i \\ \mp x_{jam} - x_i \end{bmatrix}$$

$$f_3(d) = \begin{cases} a_f, & d \in [0, e_f] \\ 0, & otherwise \end{cases}$$



# Final Controller

## System Dynamics

The dynamics of this multi-agent system is denoted by

$$\dot{q}_i = z_i, \quad i = 1, 2, \dots, n,$$

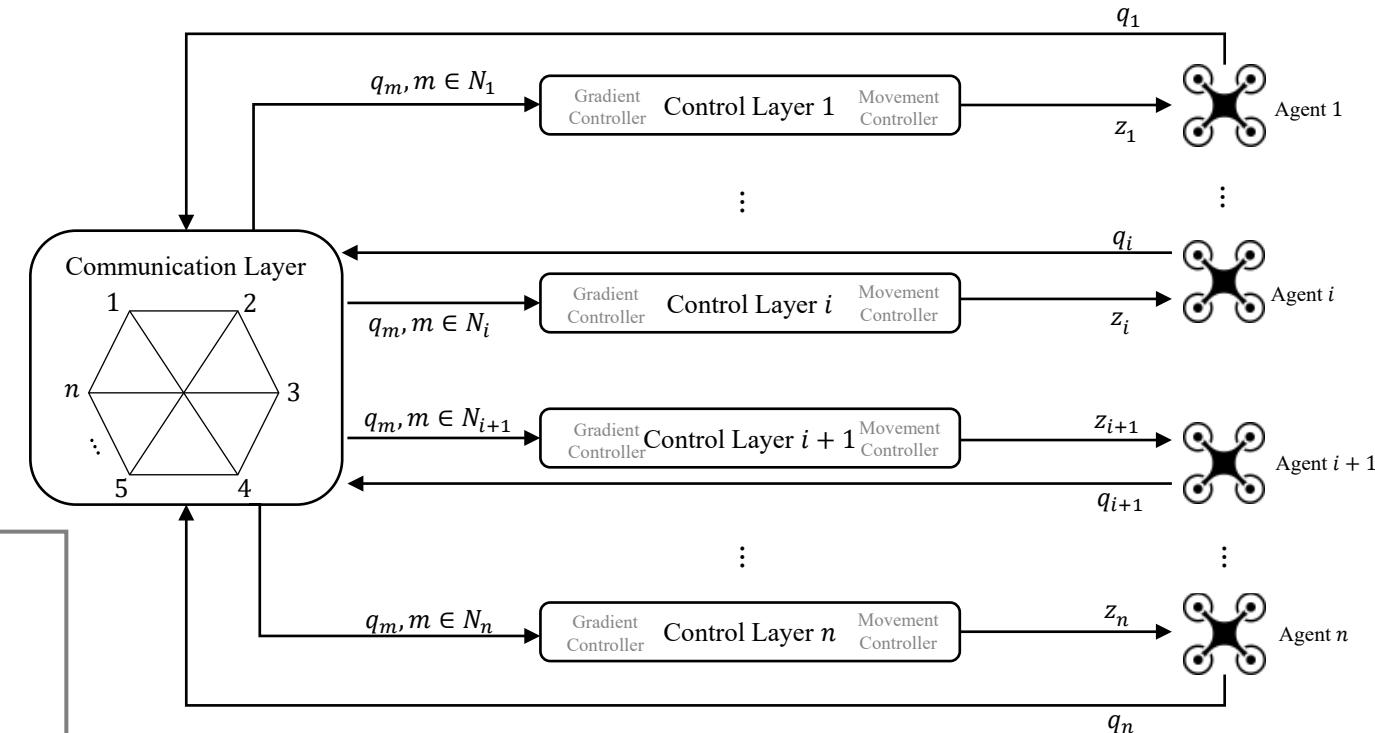
where

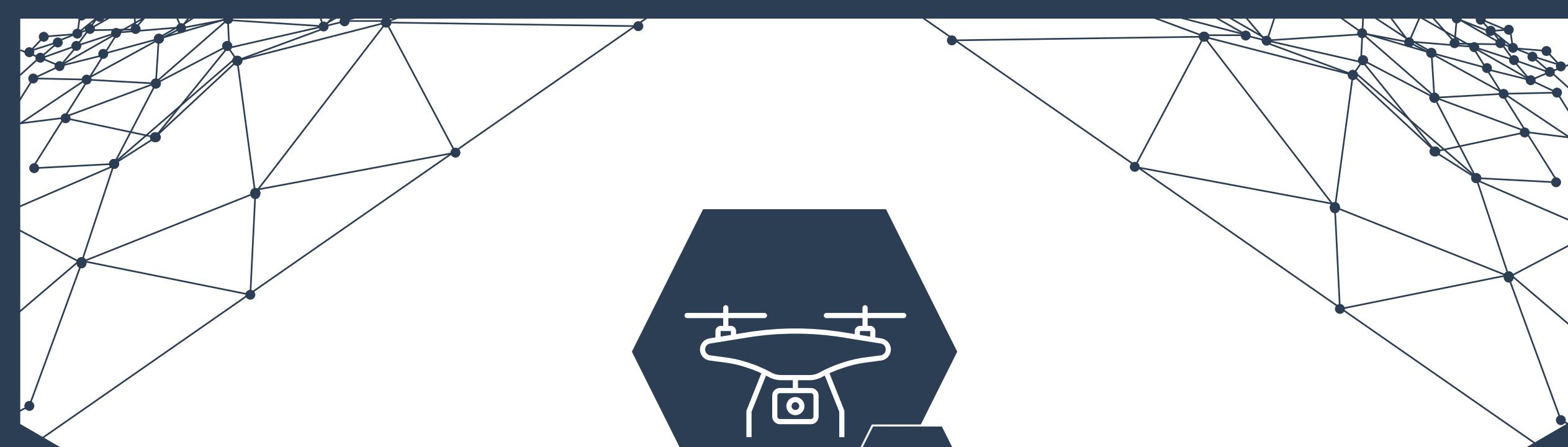
$q$ : position input of agents,  
 $z$ : control input of agents.

## Final Formation Controller

$$z_i = \mathcal{G}_i + \mathcal{M}_i$$

$$= \sum_{j \in N_i} [\varphi(r_{ij}) \cdot e_{ij}] + [f_1(\cdot) \ f_2(\cdot) \ f_3(\cdot)] \begin{bmatrix} V_{navigation} \\ V_{avoidance} \\ V_{edge-following} \end{bmatrix}$$

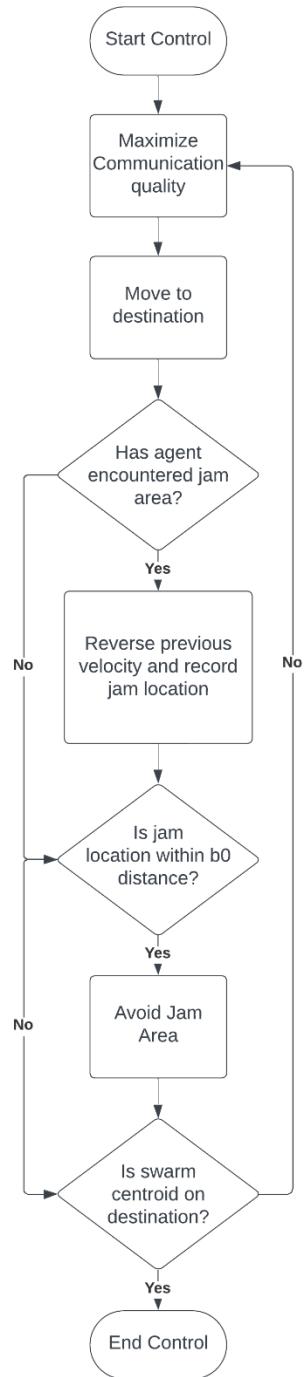


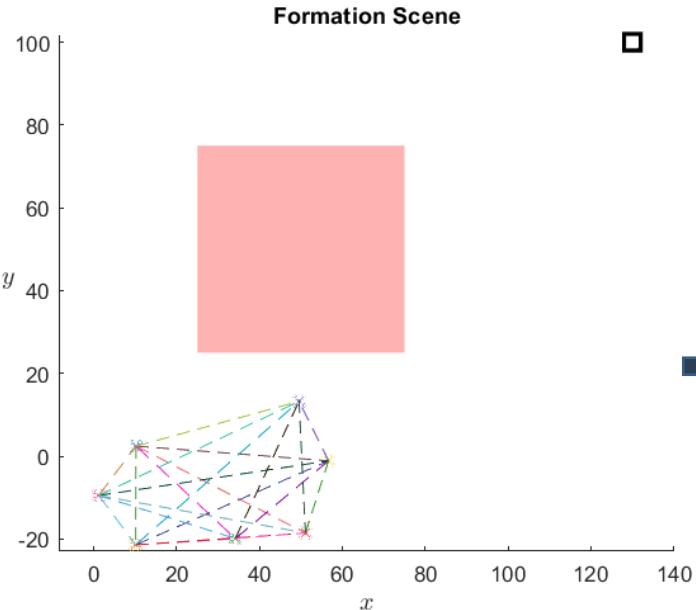


## Simulations

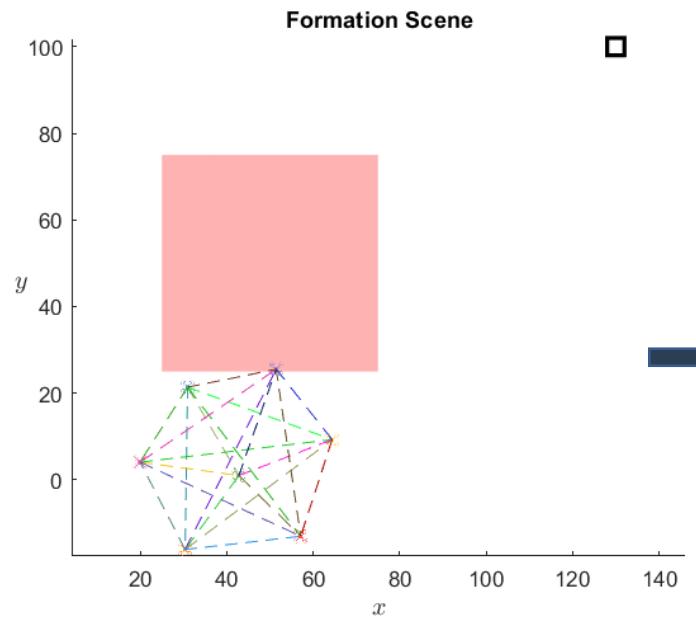


# Workflow

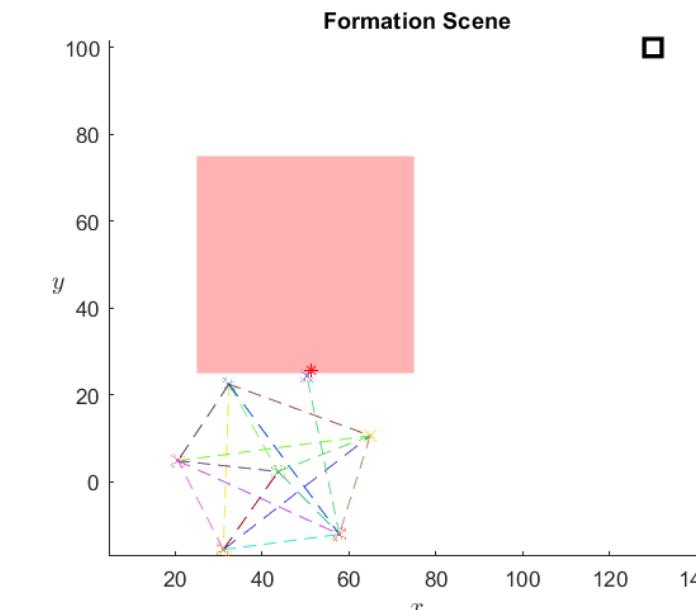




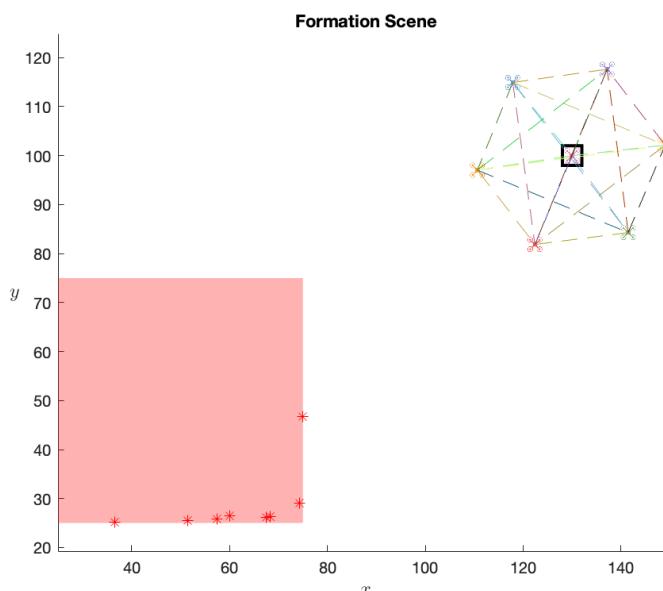
$t = 0s$



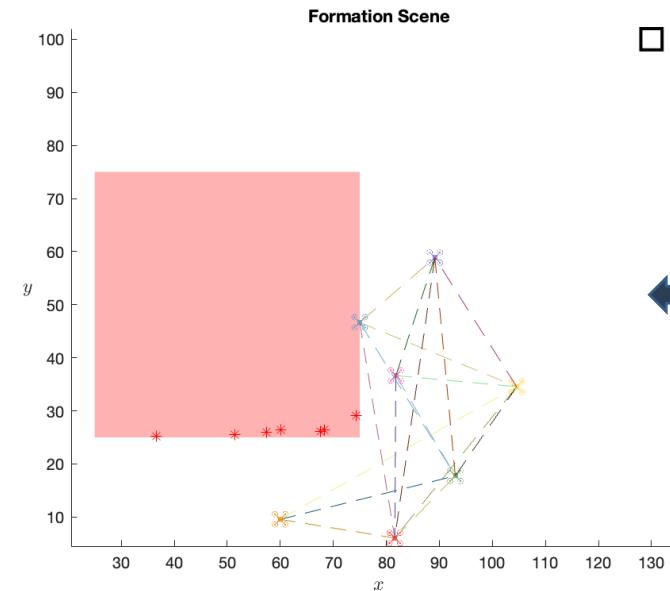
$t = 4s$



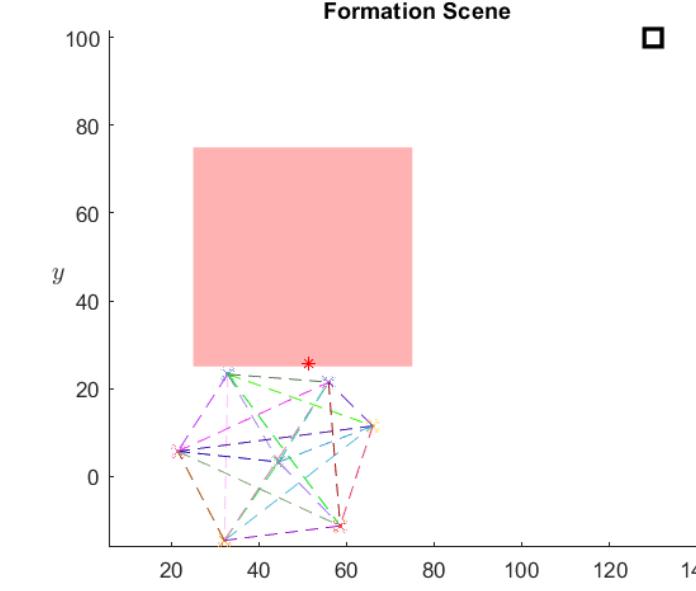
$t = 5s$



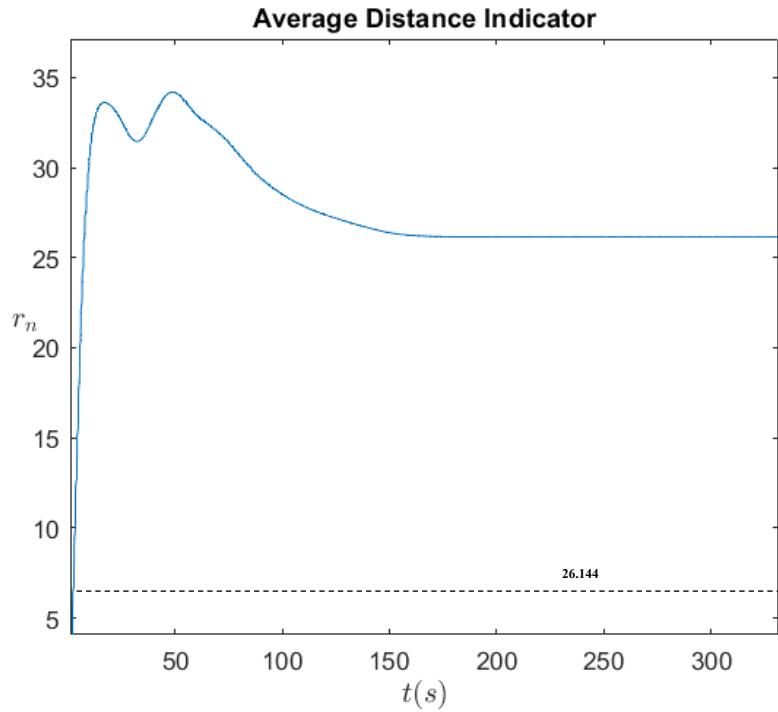
$t = 300s$



$t = 24s$



$t = 6s$

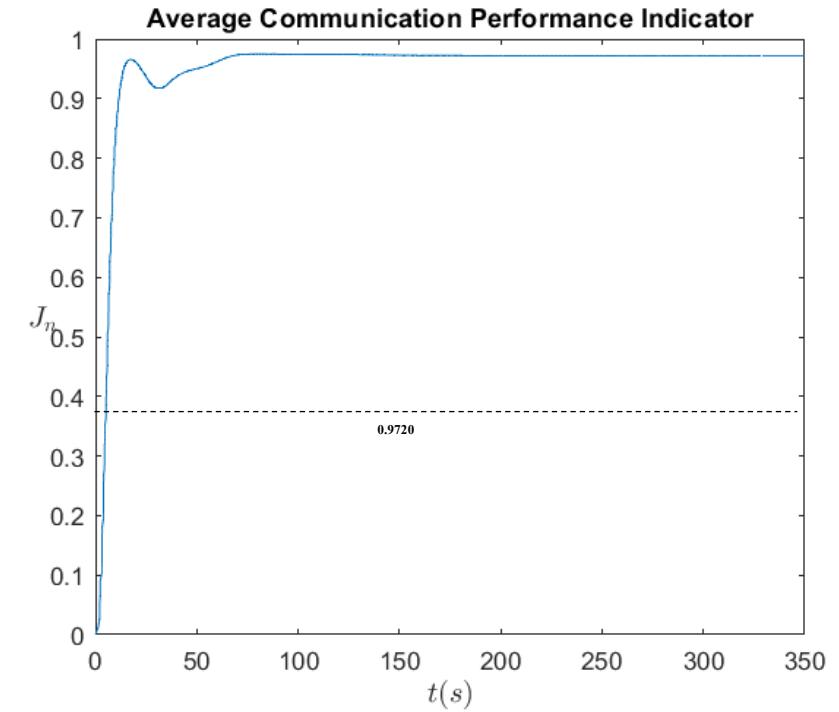


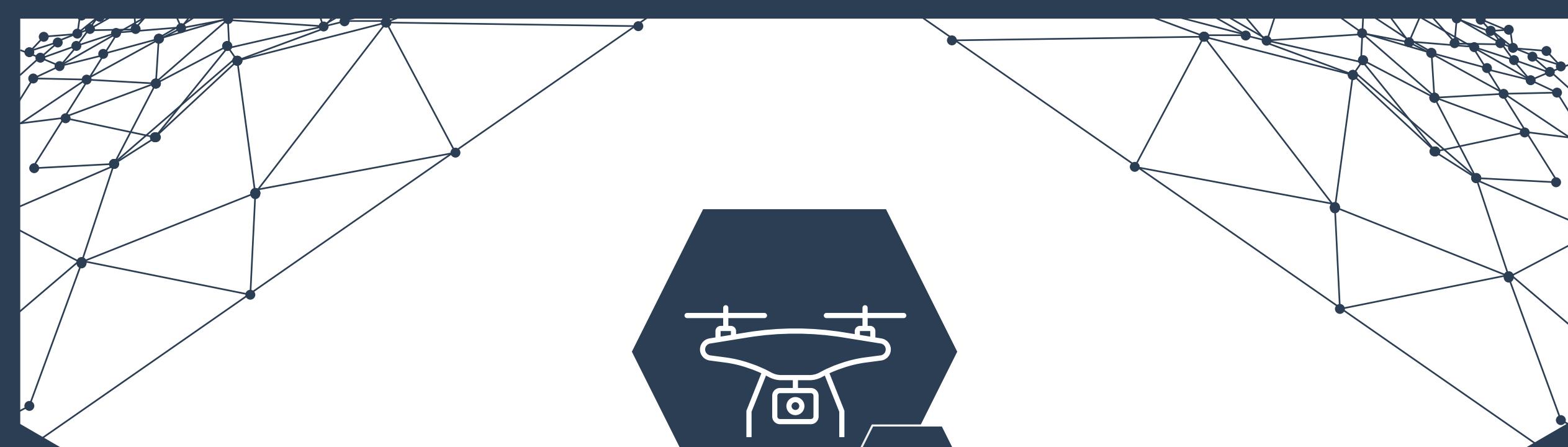
$$J_n: \text{Average Communication Performance Indicator}$$

$$J_n = \frac{\sum_{i=1}^n \sum_{j \in N_i} \phi(r_{ij})}{2n|N_i|}$$

$$r_n: \text{Average Neighboring Distance Indicator}$$

$$r_n = \frac{\sum_{i=1}^n \sum_{j \in N_i} r_{ij}}{2n|N_i|}$$



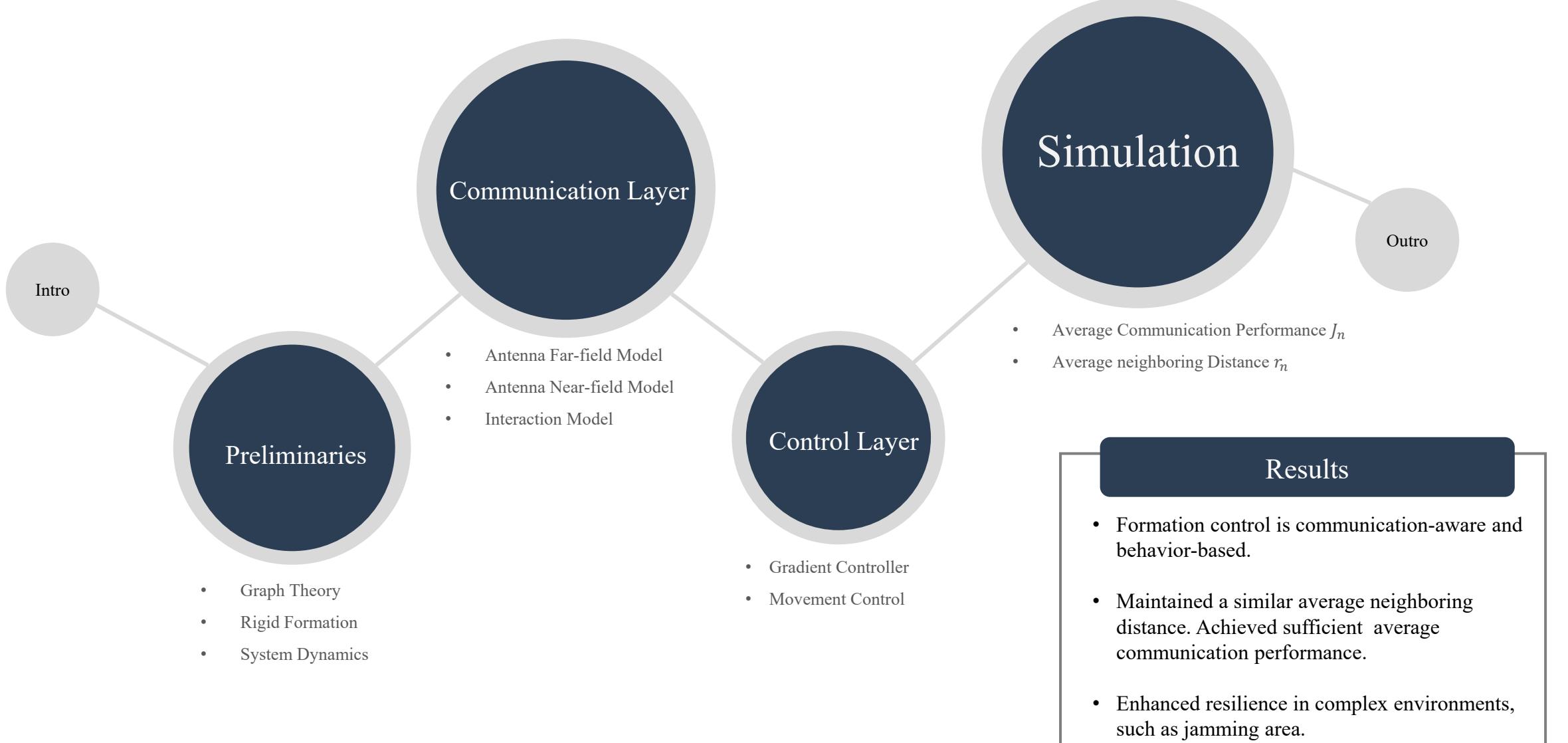


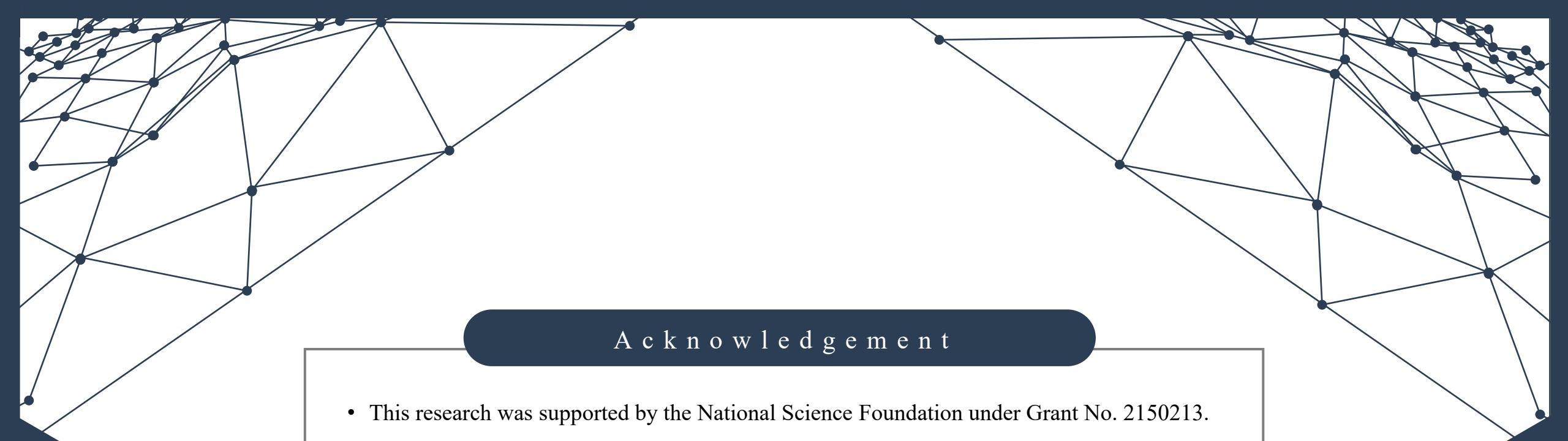
05

## Conclusion



# Review





## A c k n o w l e d g e m e n t

- This research was supported by the National Science Foundation under Grant No. 2150213.
- We'd like to thank you Dr. Stansbury for considering us into this research program.
- We'd like to thank you Dr. Yang's guidance throughout this research project.

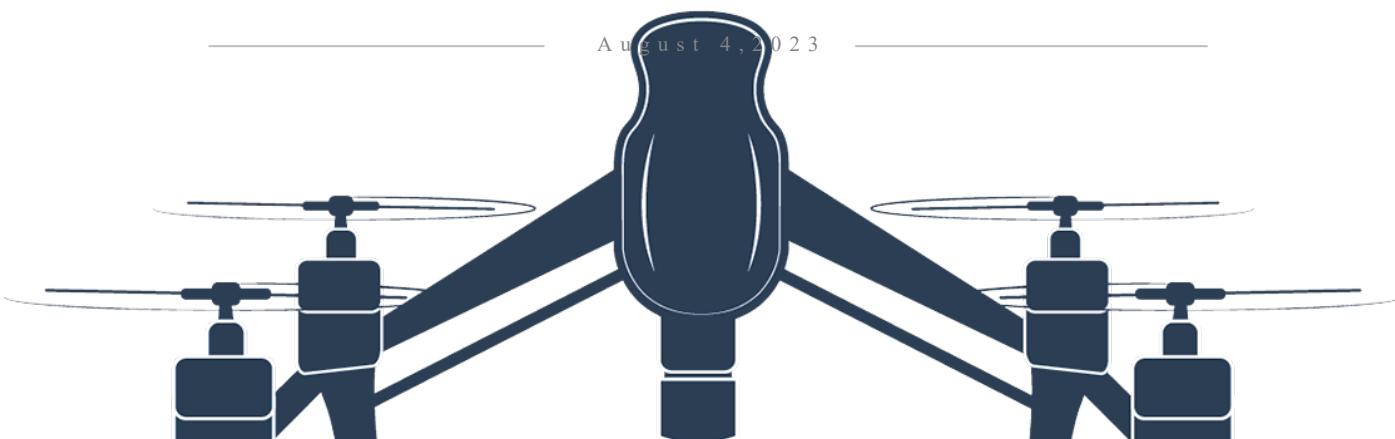


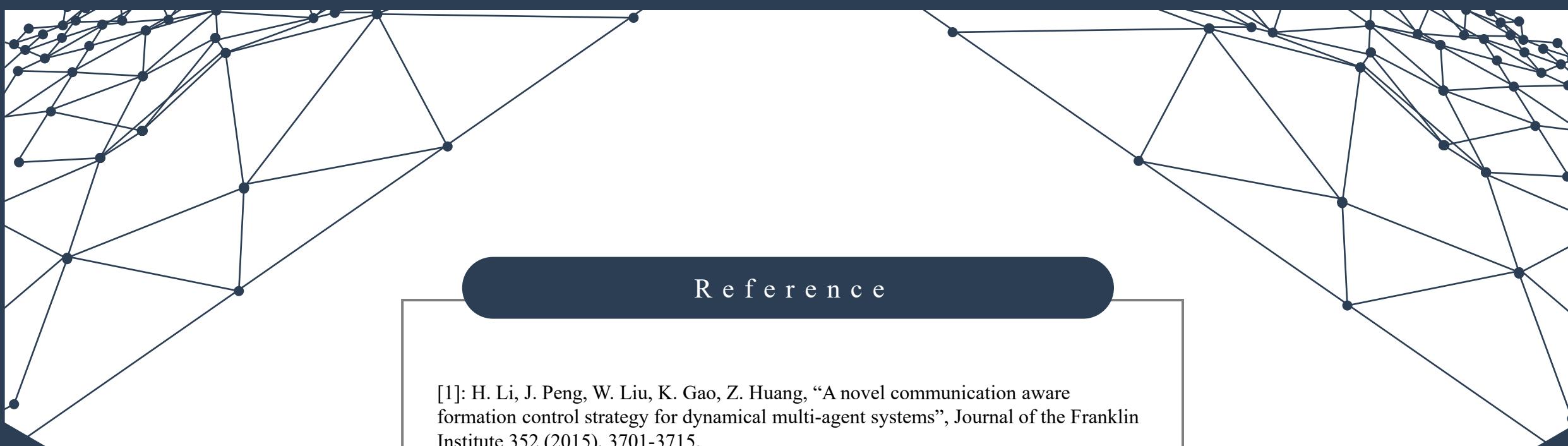


# THANKS

Q u e s t i o n s ?

— August 4, 2023 —

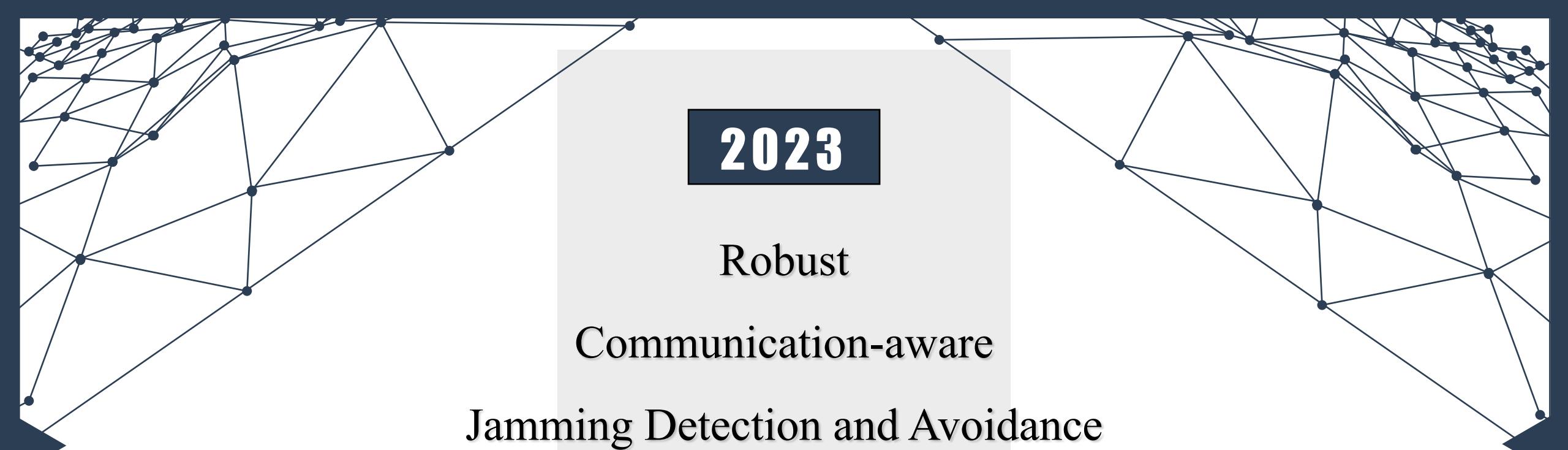




## Reference

- [1]: H. Li, J. Peng, W. Liu, K. Gao, Z. Huang, “A novel communication aware formation control strategy for dynamical multi-agent systems”, Journal of the Franklin Institute 352 (2015), 3701-3715.
- [2]: D. Xu, X. Zhang, Z. Zhu, C. Chen, and P. Yang, “Behavior-based formation control of Swarm Robots,” Mathematical Problems in Engineering, vol. 2014, pp. 1–13, 2014.  
doi:10.1155/2014/205759





2023

Robust  
Communication-aware  
Jamming Detection and Avoidance

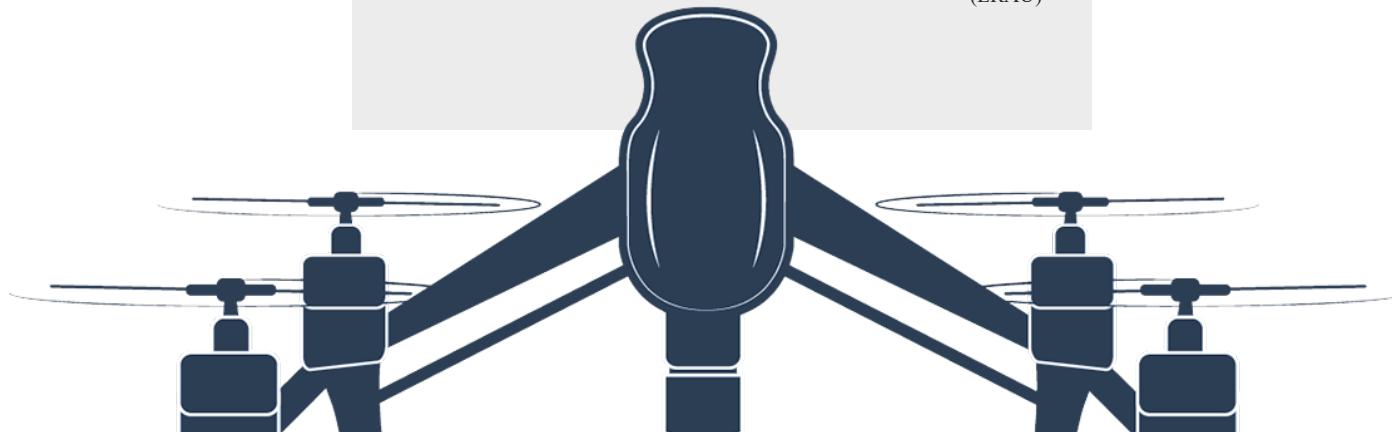
ERAU NSF REU

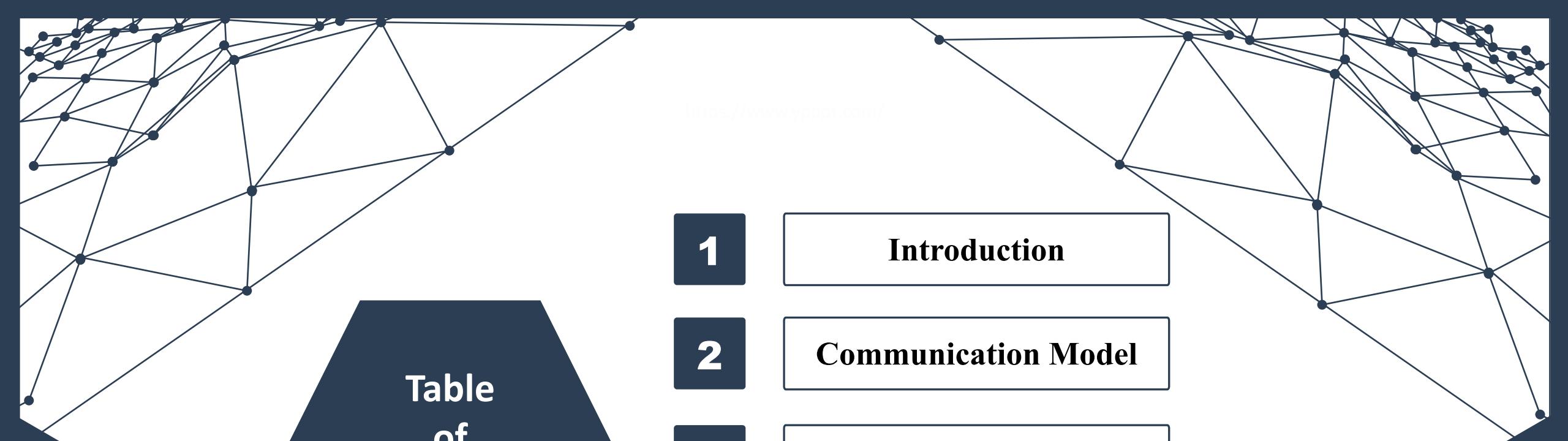
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Samuel  
Peccoud (CSU)

Sang Xing  
(PSU)

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(ERAU)





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- 1      Introduction**
- 2      Communication Model**
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# Introduction

- Research Questions
- Background
- Preliminaries
- Schematic Diagram

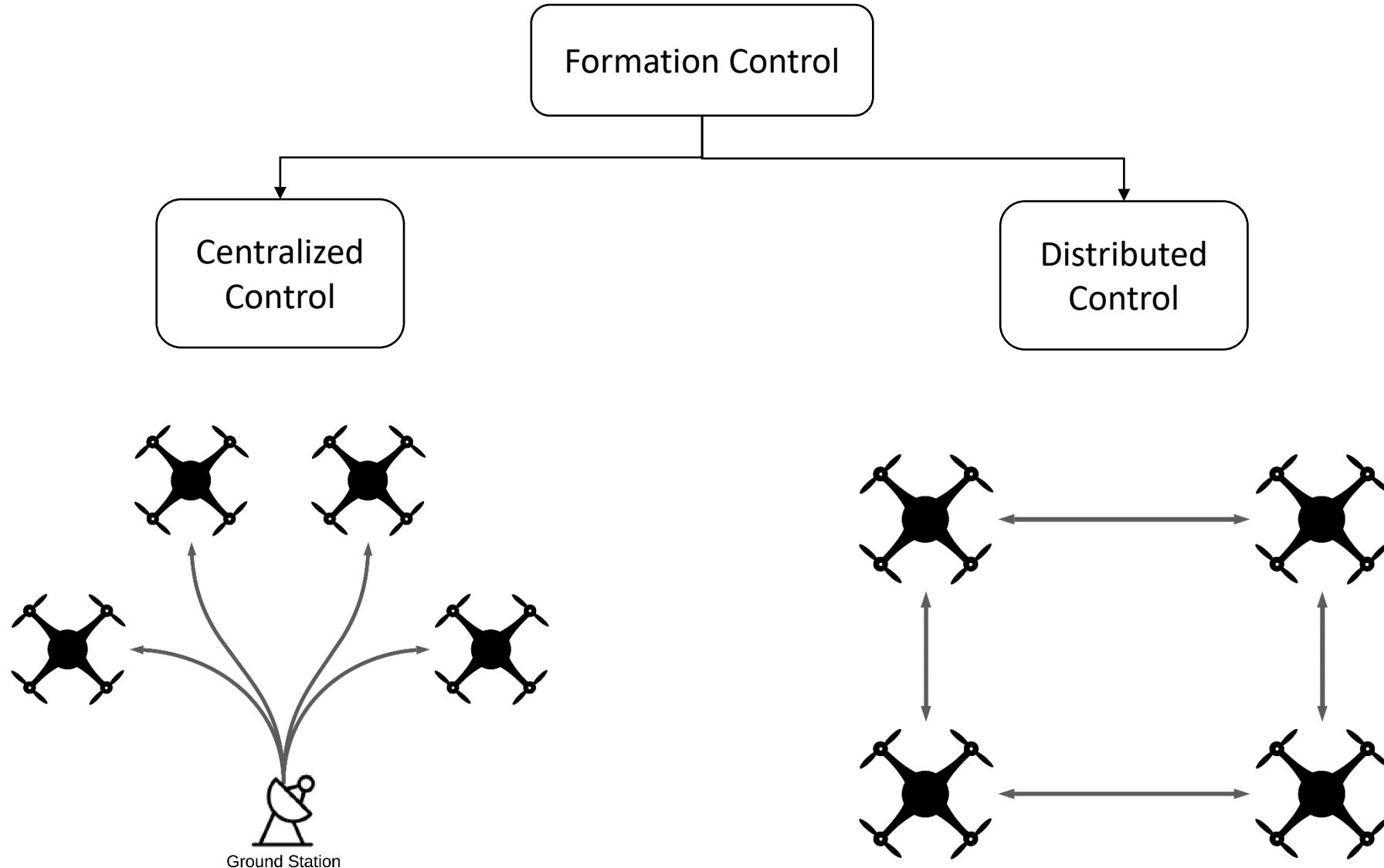


## Research Questions

1. How to navigate a swarm around jamming areas?
  - Particle Swarm Optimization (PSO) Algorithm + Path Planning Algorithm
2. How to maximizing communication quality between agents?
  - Communication-aware formation control

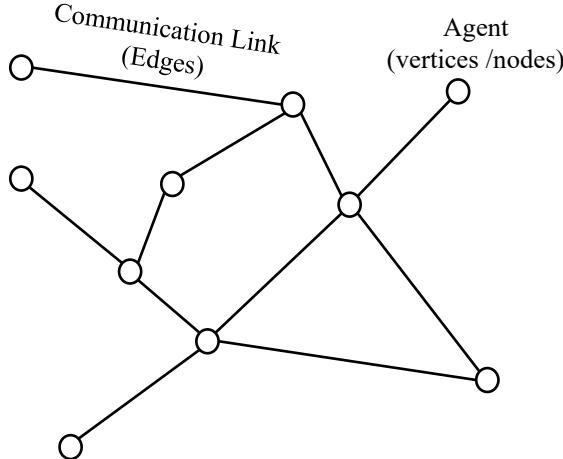


## Background





# Preliminaries



## Rigid Formation

The formation of groups of mobile agents in which all inter-agent distances remain constant is called **rigid**.

The **relative distance** between agent  $i$  and agent  $j$  is denoted by

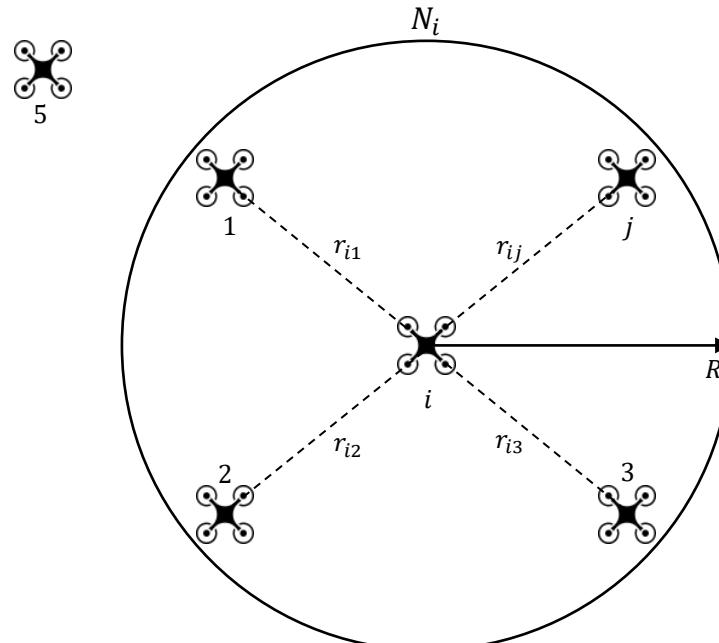
$$r_{ij} = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2} = \|q_i - q_j\|.$$

Let  $R > 0$  denote the **communication range** between two agents. The neighboring set of agent  $i$  can be denoted by

$$N_i = \{j \in \mathcal{V} \mid r_{ij} \leq R\}.$$

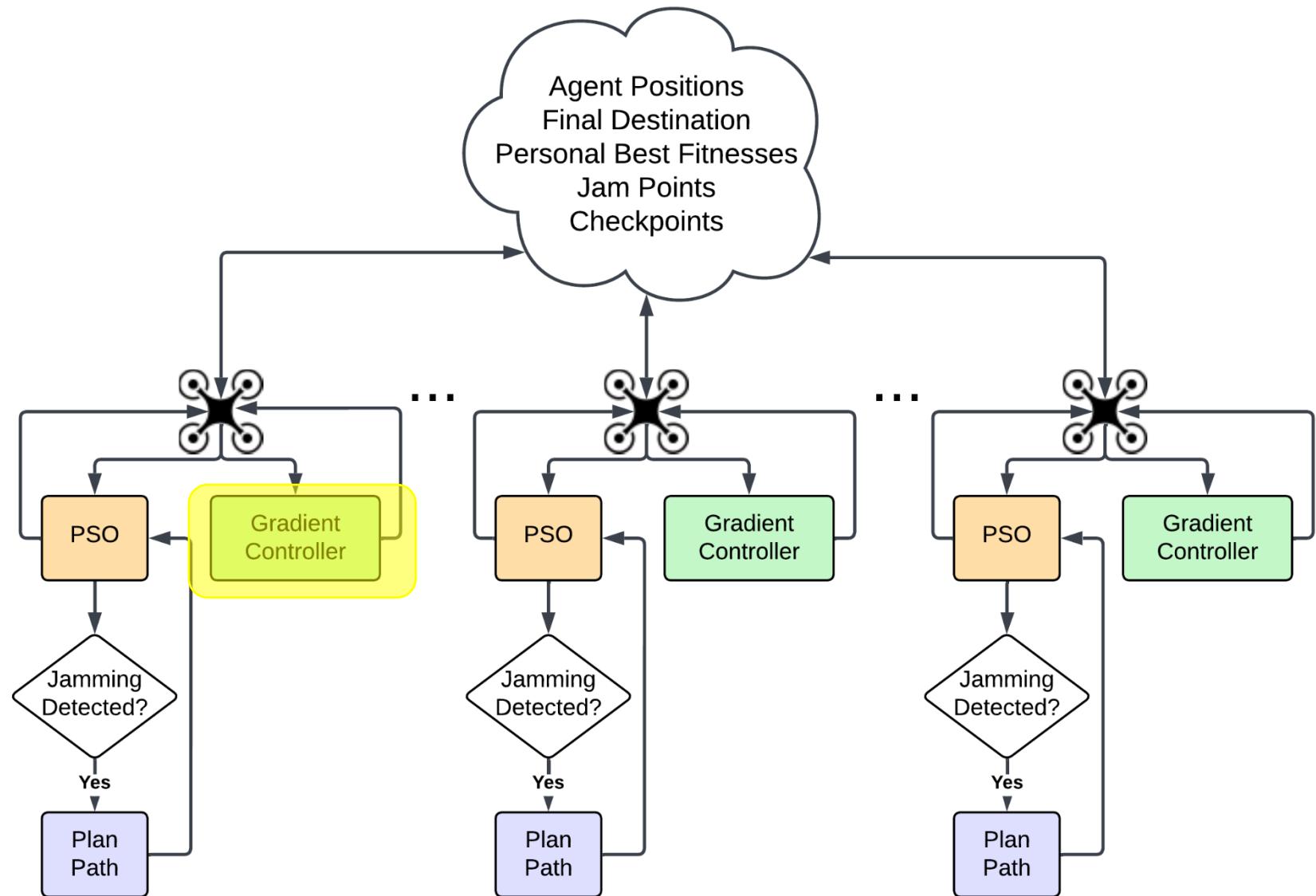
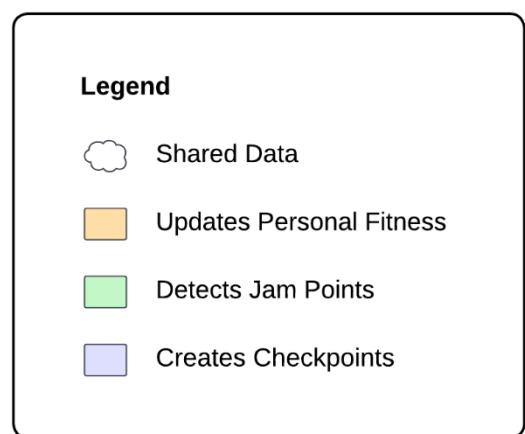
## Graph Theory

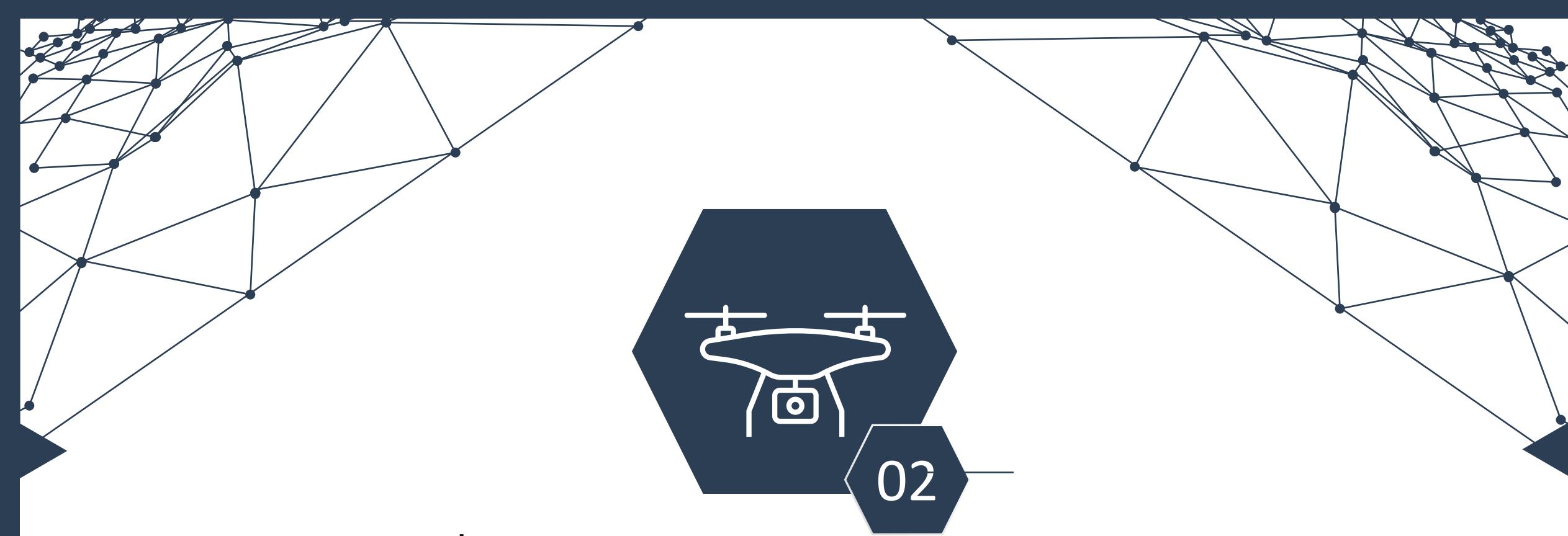
A **graph**  $G$  is a pair of  $(\mathcal{V}, \mathcal{E})$  consisting of a set of **vertices**  $\mathcal{V} = \{1, 2, \dots, i, \dots, j, \dots, n\}$  and a set of ordered pairs of the vertices called **edges**  $\mathcal{E} \subseteq \mathcal{V} \times \mathcal{V}$ . I.e.,  $\mathcal{E} = \{(i, j) \mid i, j \in \mathcal{V}, i \neq j\}$ . Here, we assume that  $G$  has no **self-edges** and **undirected**.





# Schematic Diagram



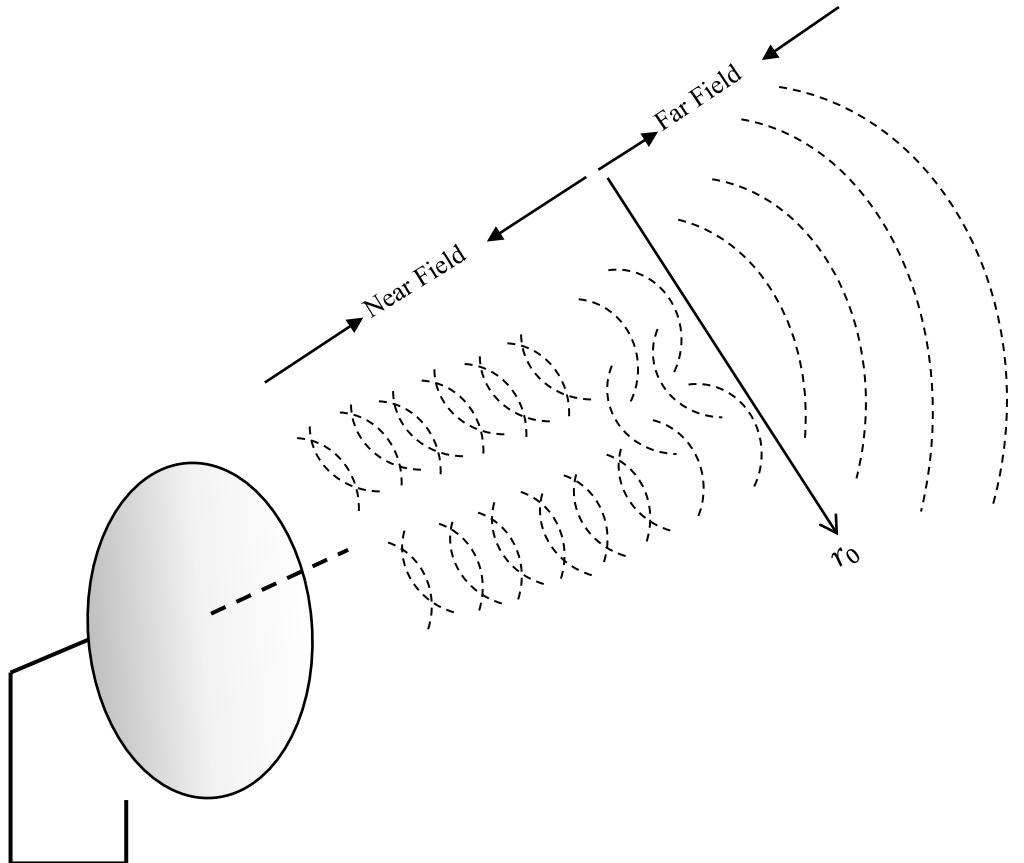


# Communication Model

- ❑ Antenna Near-field and Far-field
- ❑ Interaction Model



# Antenna Near-field and Far-field



## Far-field

The communication channel quality in antenna far-field is:

$$f_{ij} = \exp\left(-\alpha(2^\delta - 1)\left(\frac{r_{ij}}{r_0}\right)^\nu\right).$$

## Near-field

A simple model of antenna near-field communication quality is:

$$n_{ij} = \frac{r_{ij}}{\sqrt{r_{ij}^2 + r_0^2}}$$

where

$r_0$ : antenna field separator,

$r_{ij}$ : distance between agent  $i$  and agent  $j$ .



# Interaction Model

## Signal Scattering Effect

When a traveling wave encounters a change in the wave impedance, it will reflect, at least partially. If the reflection is not total, it will also partially transmit into the new impedance.

## Path Loss Effect

The reduction in power density (attenuation) of an electromagnetic wave as it propagates through space. As a result, the received signal power level is several orders below the transmitted power level.

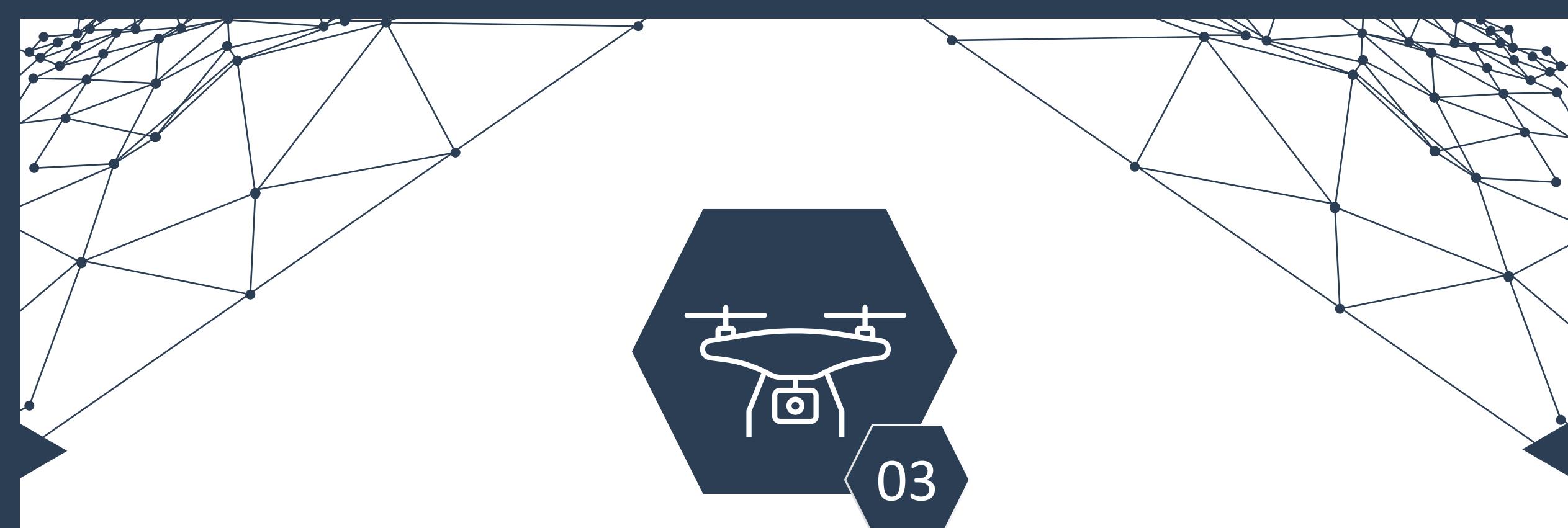
## Interference Effect

When a signal is disrupted as it travels along the communication channel between its source and receiver. It may cause only a temporary loss of a signal and may affect the quality of the communication.

## Interaction Model

The interaction model is denoted by

$$\phi(r_{ij}) = n_{ij} \cdot f_{ij} = \frac{r_{ij}}{\sqrt{{r_{ij}}^2 + {r_0}^2}} \cdot \exp\left(-\alpha(2^\delta - 1)\left(\frac{r_{ij}}{r_0}\right)^\nu\right).$$



## Control Layer

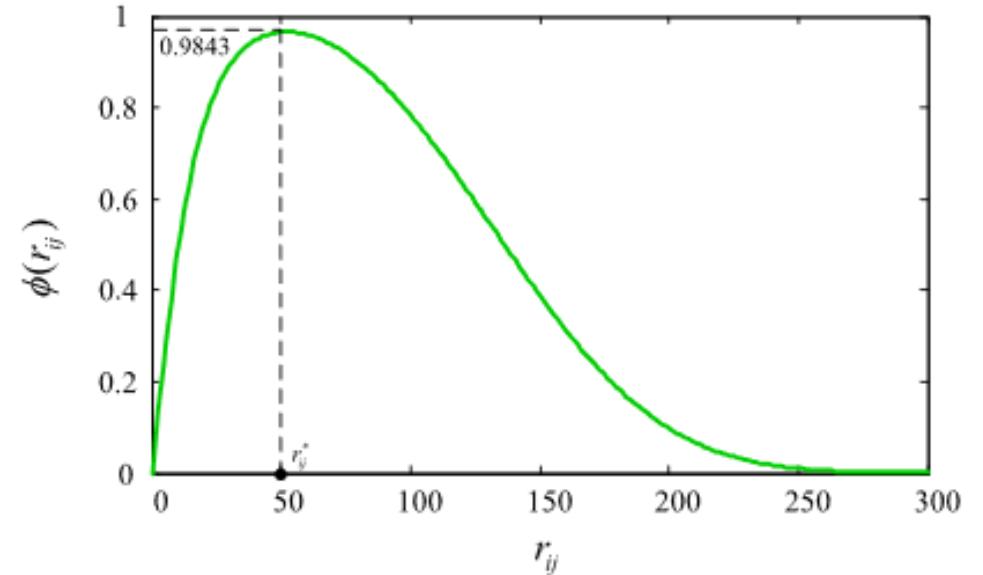
- Gradient Controller
- Movement Controller
  - Particle Swarm Optimization (PSO)
  - Path Planning



## Gradient Controller

To maximizes the communication performance, we take the first-order derivative of the interaction model,

$$\frac{d\phi}{dr_{ij}} = \varphi(r_{ij}) = \frac{-\beta\nu(r_{ij})^{\nu+2} - \beta\nu r_0^2(r_{ij})^\nu + r_0^{\nu+2}}{\sqrt{(r_{ij}^2 + r_0^2)^3}} \cdot \exp\left(-\beta\left(\frac{r_{ij}}{r_0}\right)^\nu\right).$$



A gradient controller moves agents to maximize communication performance.

### Gradient Control Model

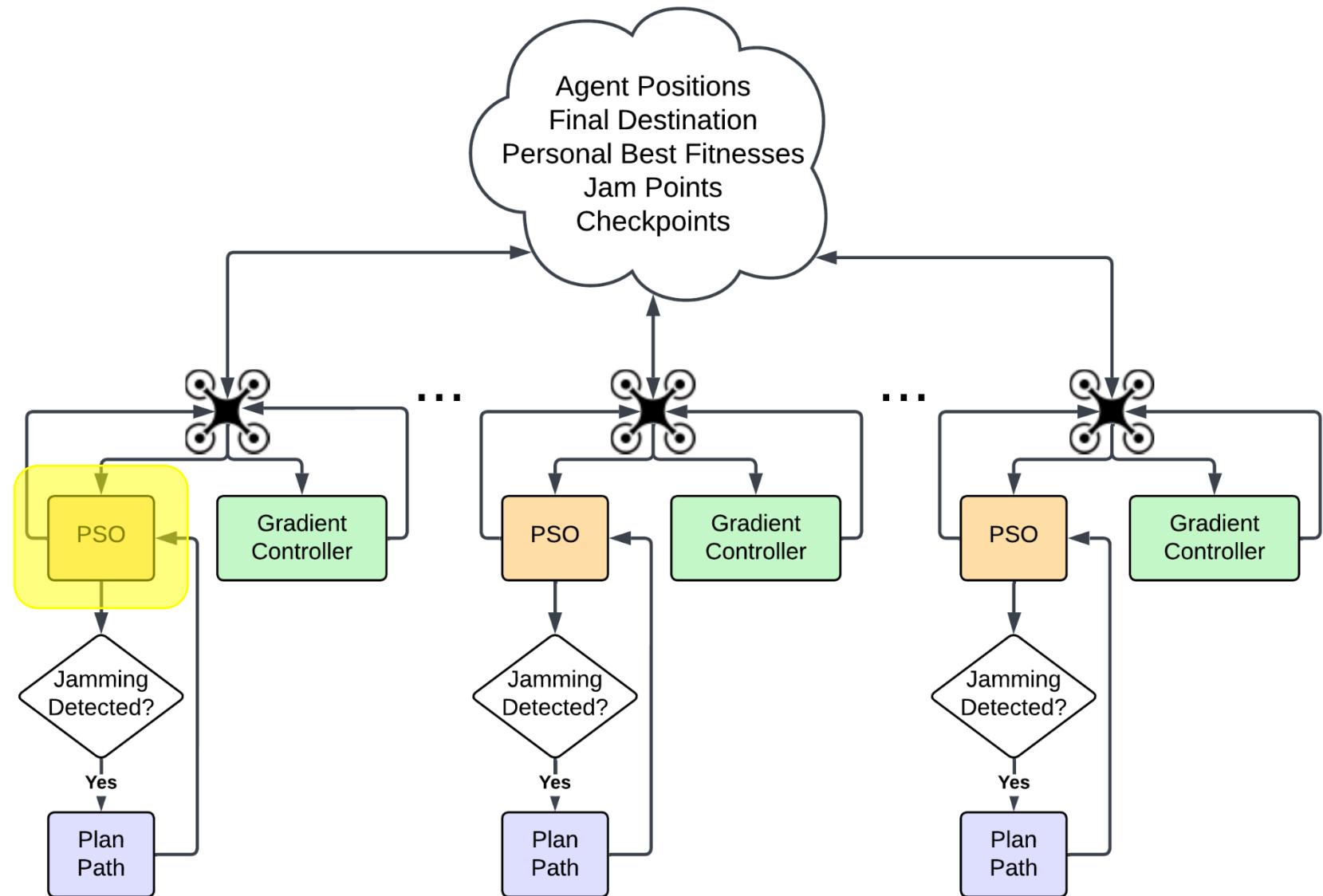
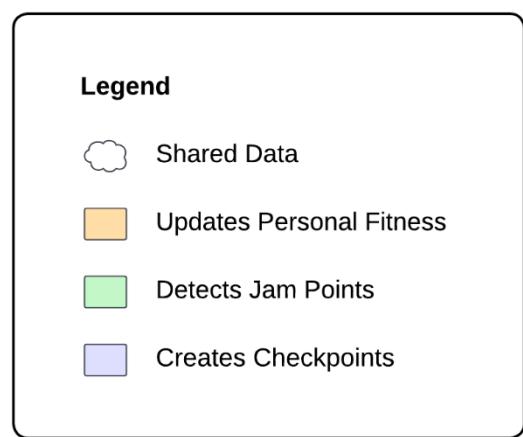
The gradient control model of agent  $i$  is denoted by

$$\mathcal{G}_i = \sum_{j \in N_i} [\nabla_{q_i} \phi(r_{ij})] = \sum_{j \in N_i} [\varphi(r_{ij}) \cdot e_{ij}]$$

where  $e_{ij} = (q_i - q_j)/\sqrt{r_{ij}}$ .



# Schematic Diagram





# Particle Swarm Optimization

Particle Swarm Optimization (PSO) is a biology inspired algorithm.

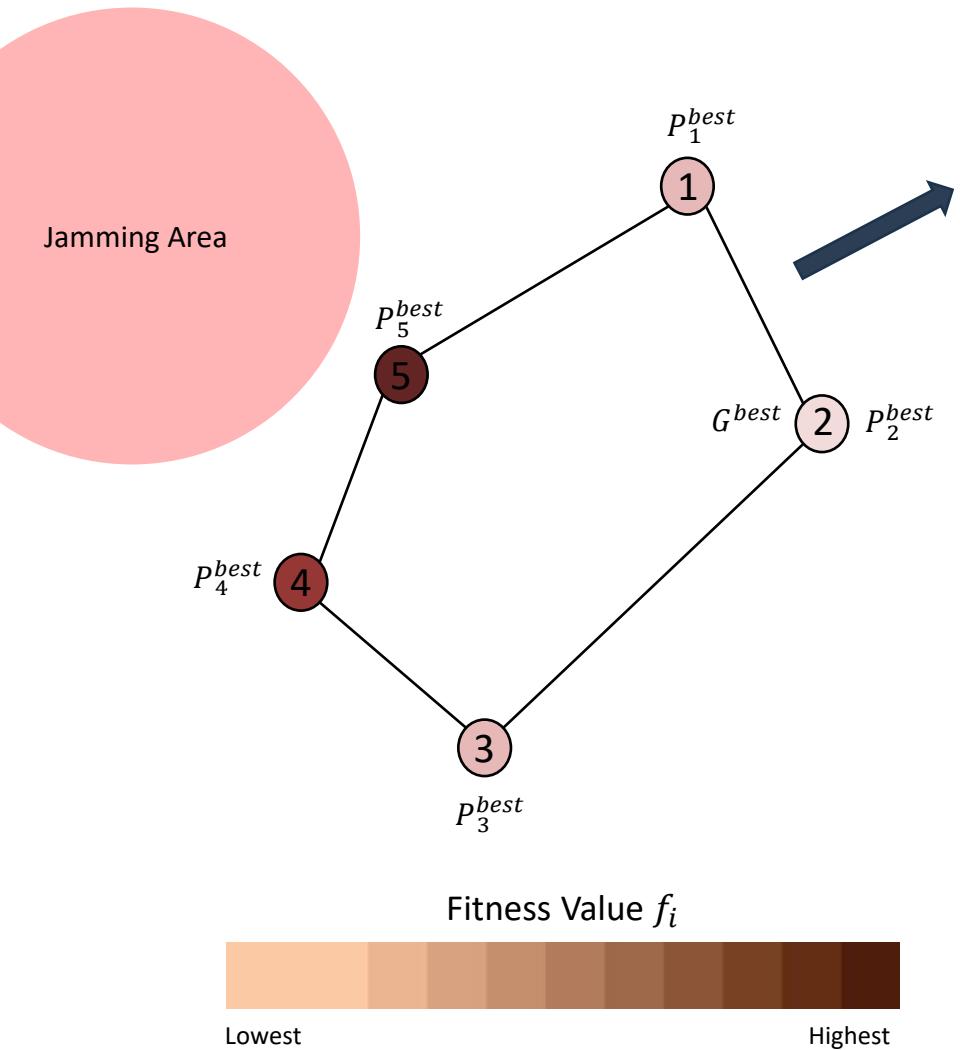
It is commonly used in multi-robot path planning.



Photo by [Don DeBold](#)



# Particle Swarm Optimization



## Fitness Function

The fitness value of agent  $i$  is:

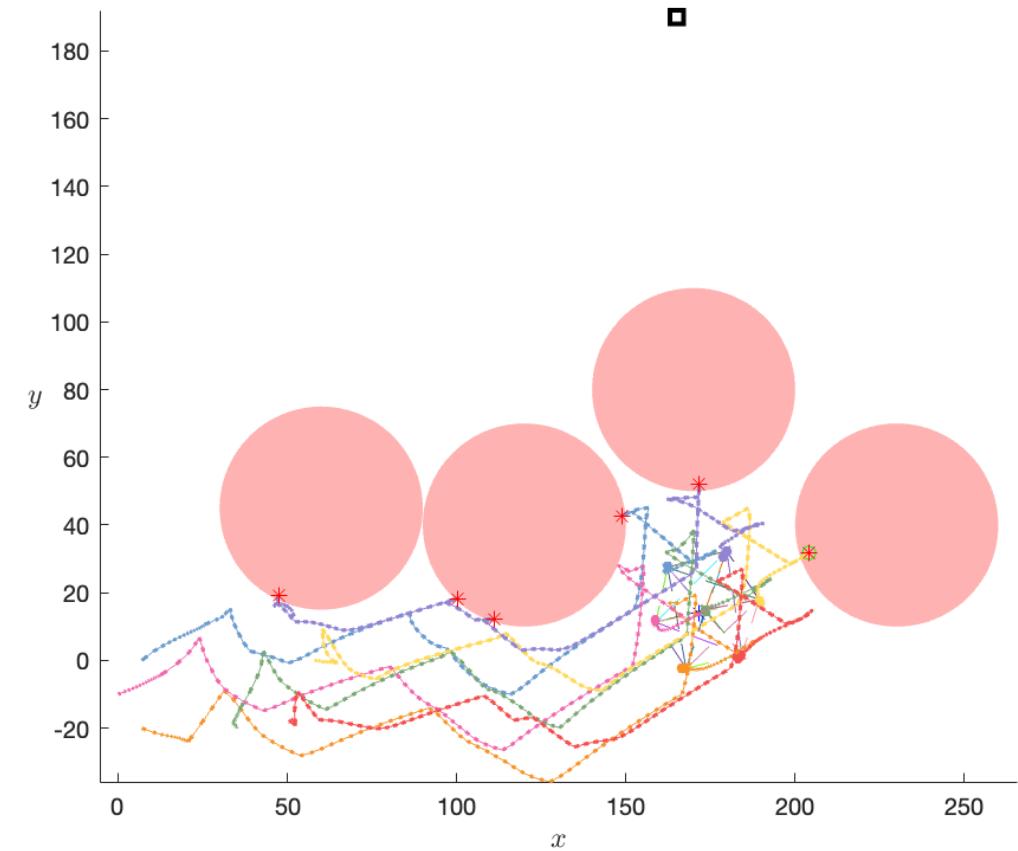
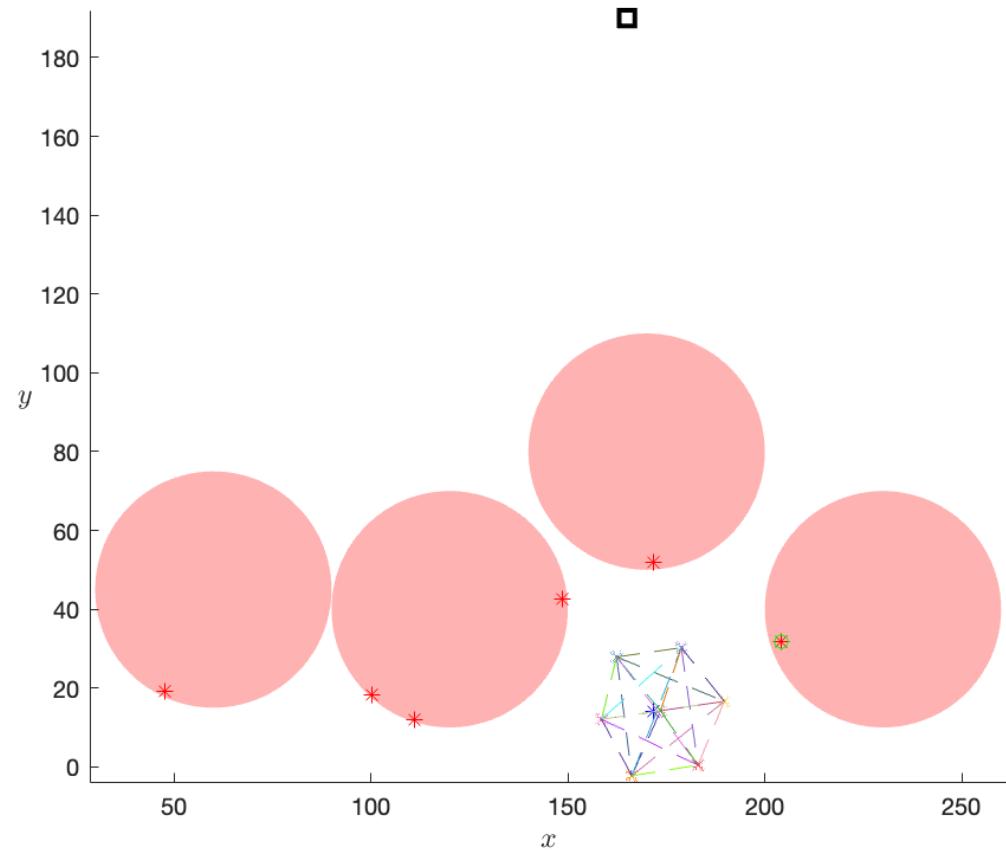
$$f_i = d_{\text{dest}} \cdot w_{\text{dest}} - \log_{10}(h_{\text{jam}}) \cdot w_{\text{jam}},$$

where

- $d_{\text{dest}}$  is the distance from agent to destination.
- $h_{\text{jam}}$  is the distance from agent to jam point.
- $w_{\text{dest}}$  and  $w_{\text{jam}}$  are adjustable weight to the distance vector.

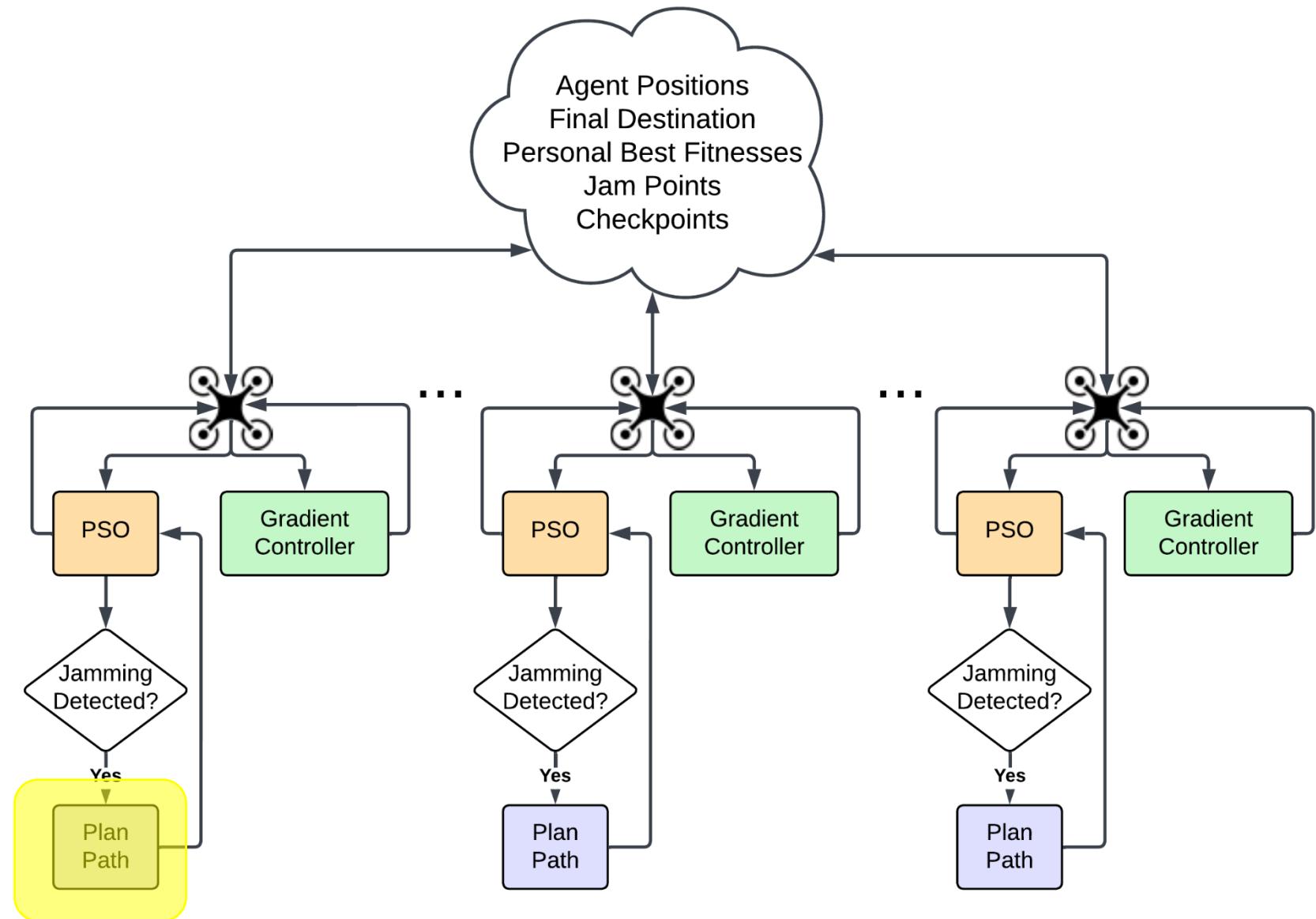
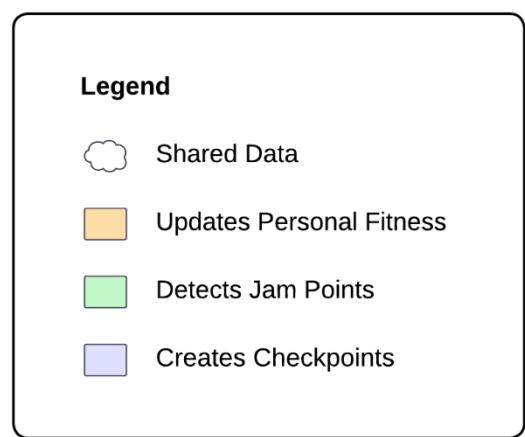


# Particle Swarm Optimization



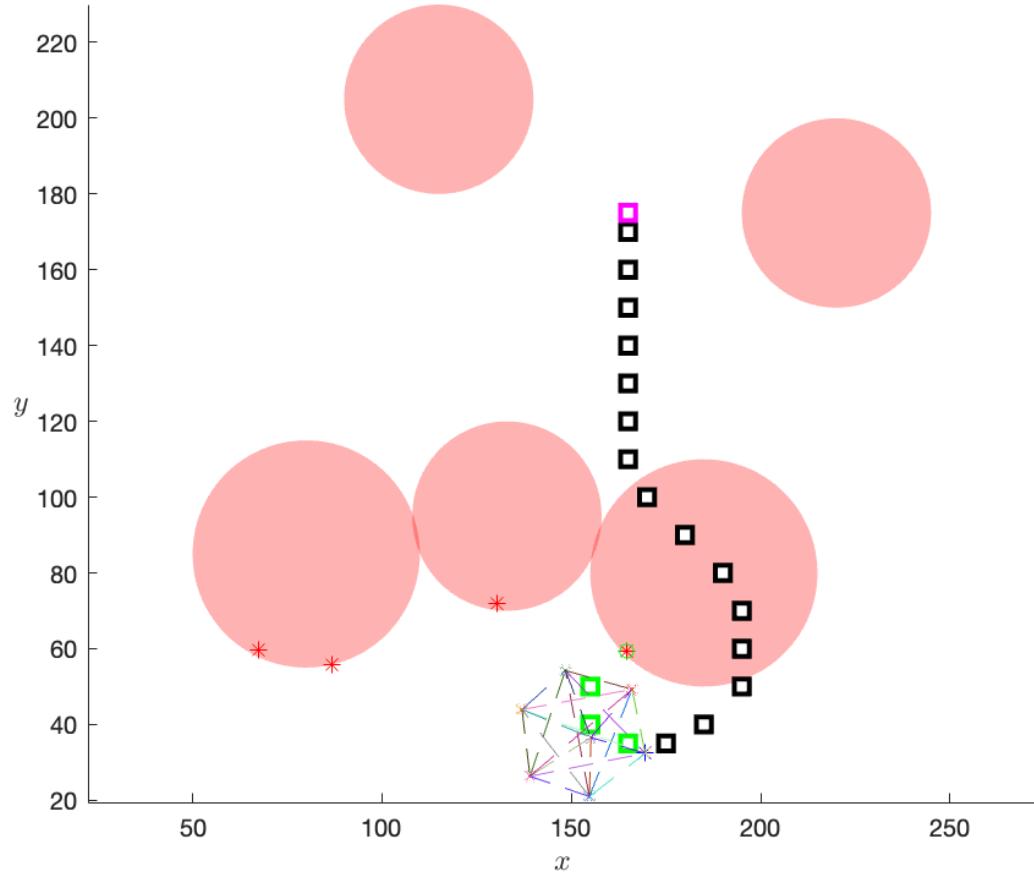


# Schematic Diagram

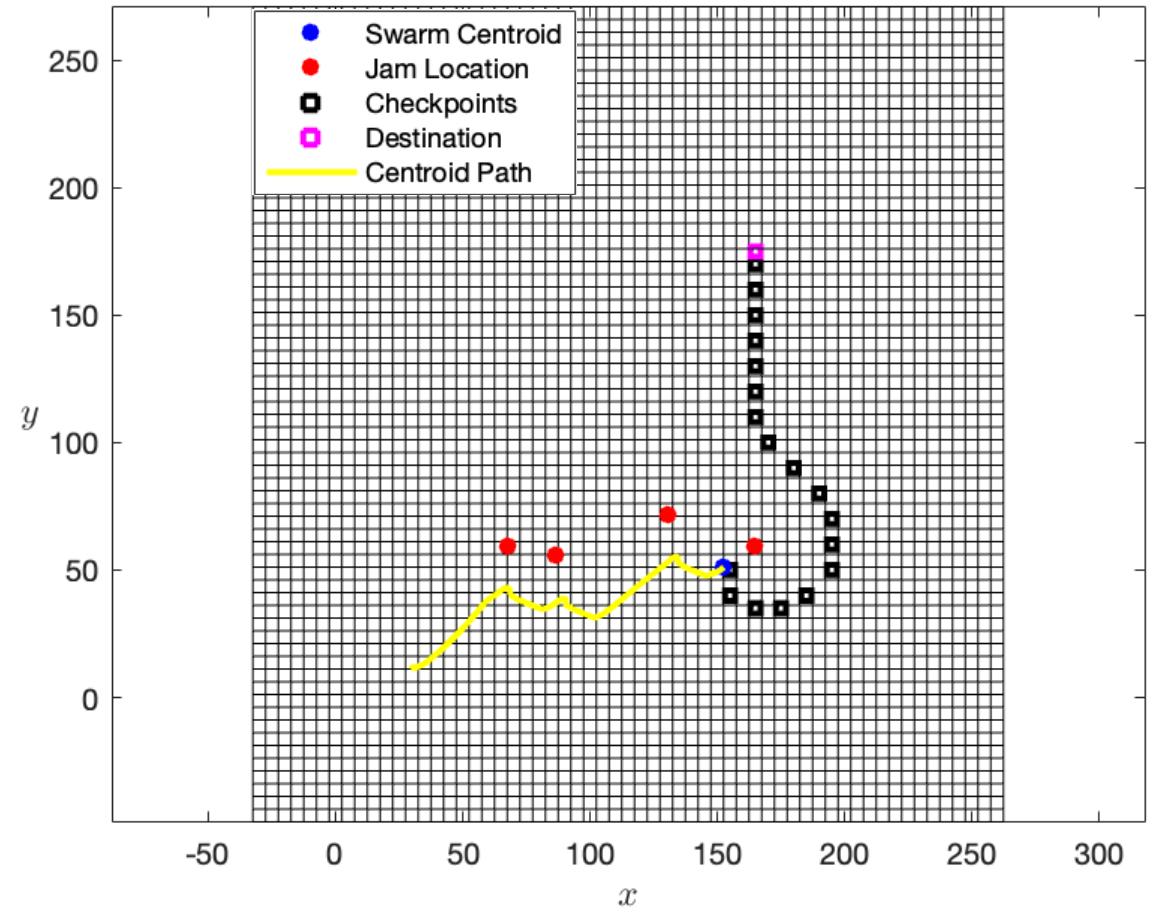




# Path Planning Algorithm

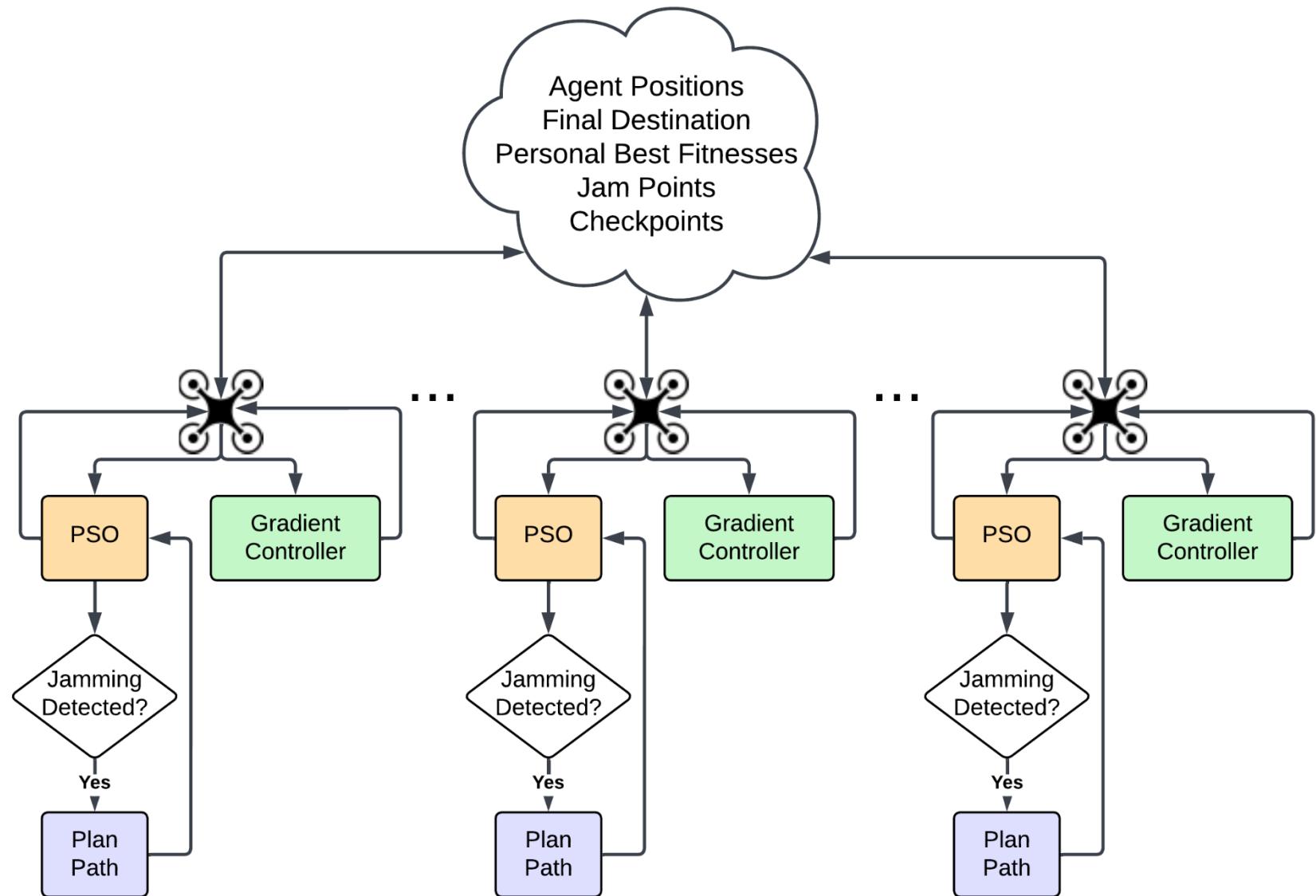
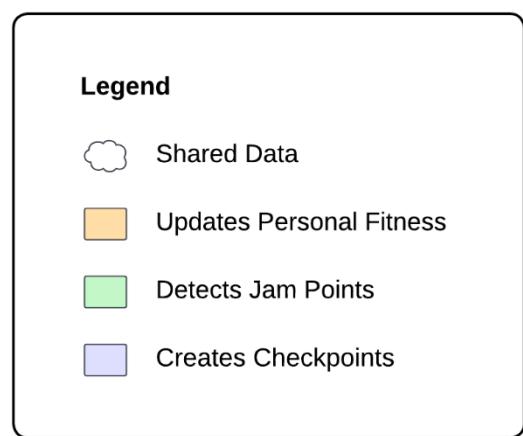


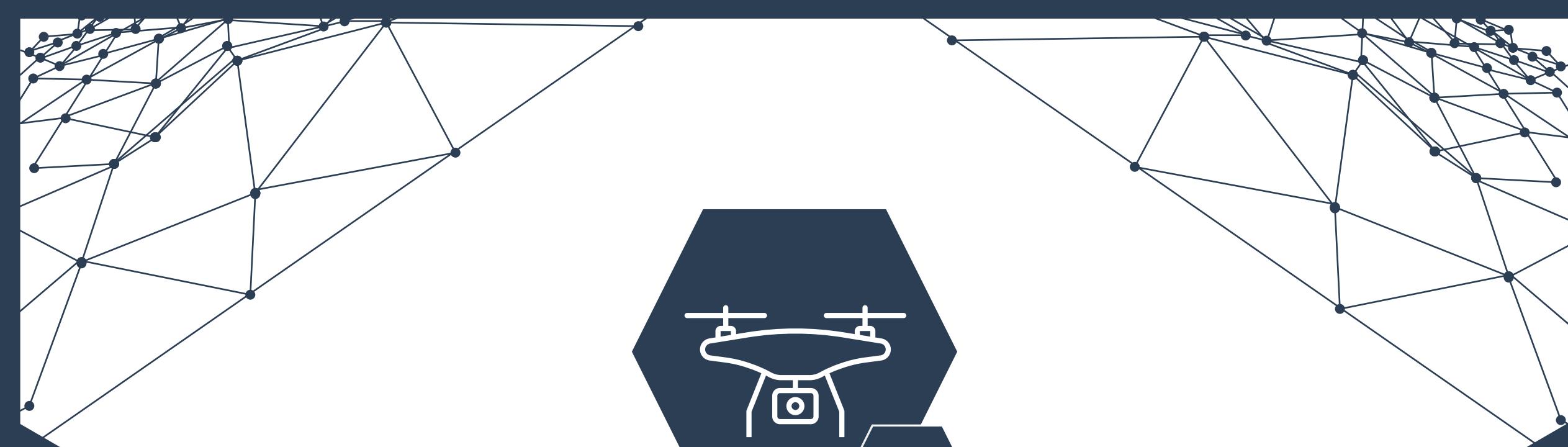
Discretized Environment





# Schematic Diagram



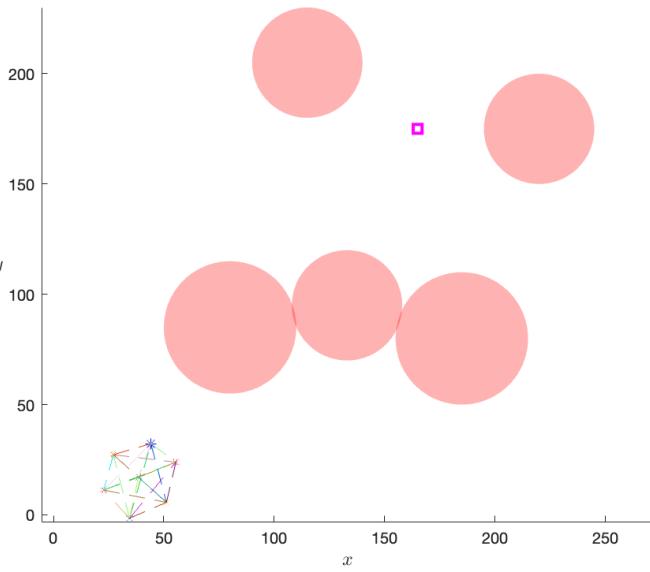
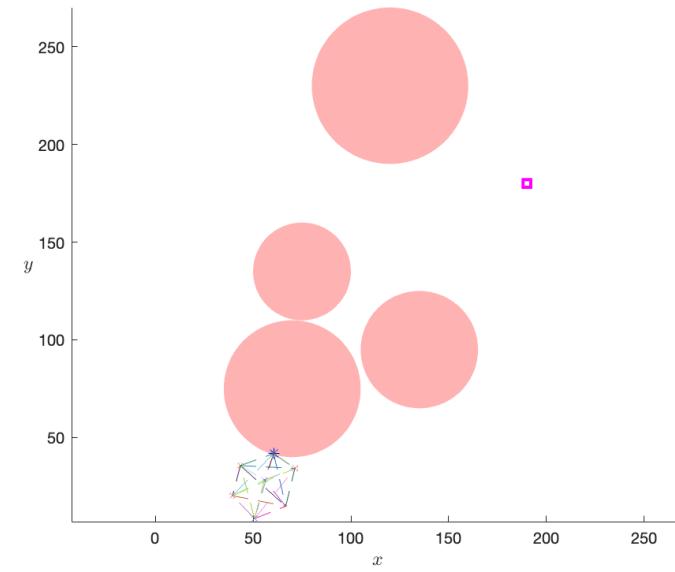
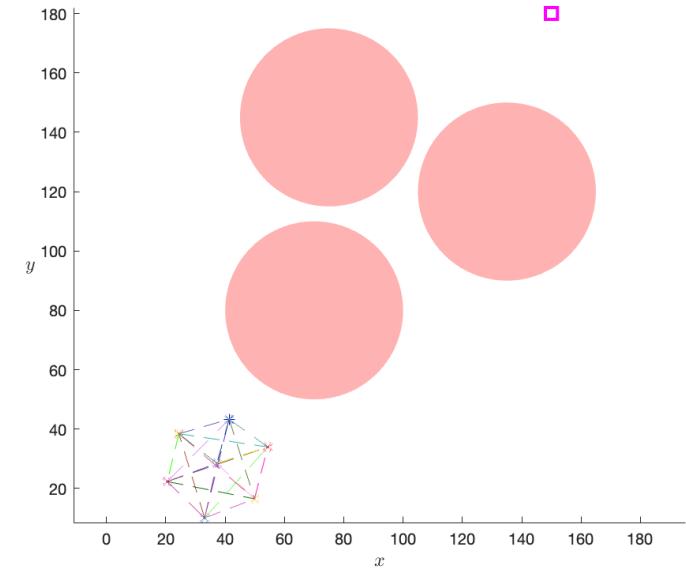
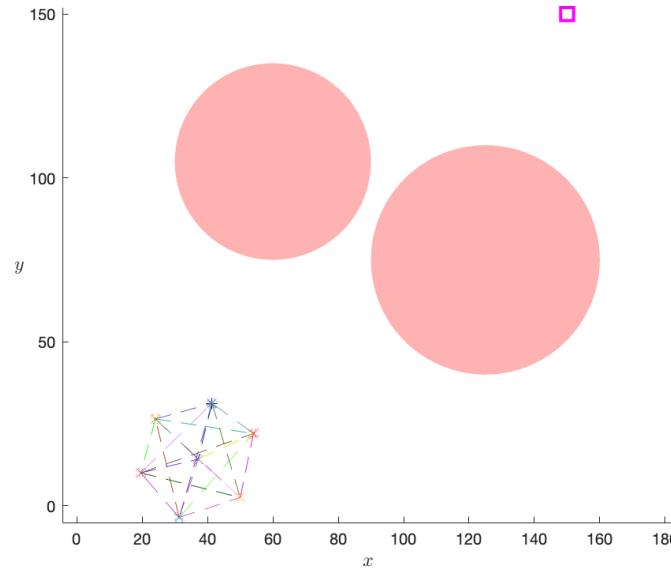
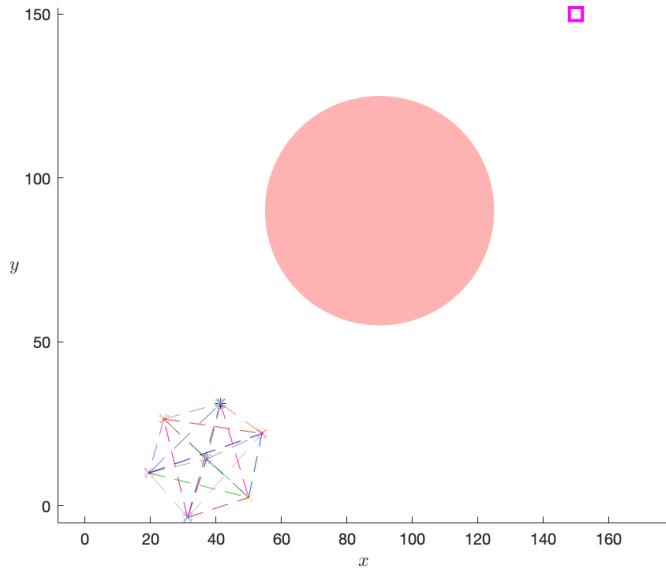


## Simulations

- Simulation Environments
- Simulation Results
- Simulation Evaluation



# Simulation Environments





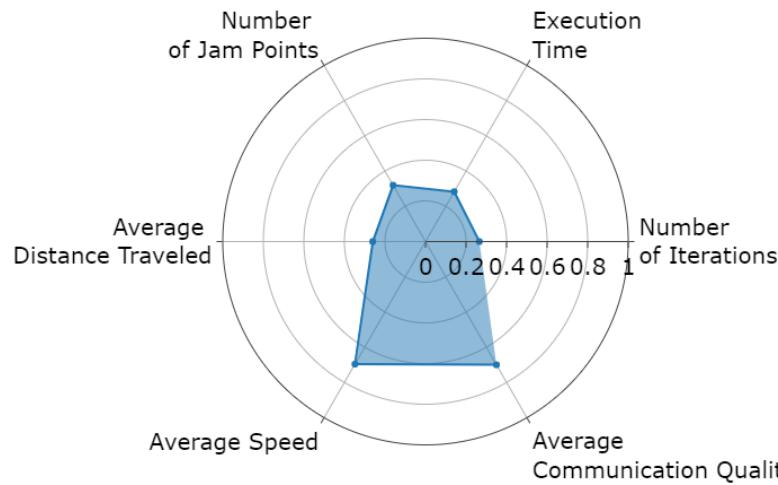
# Simulation Results



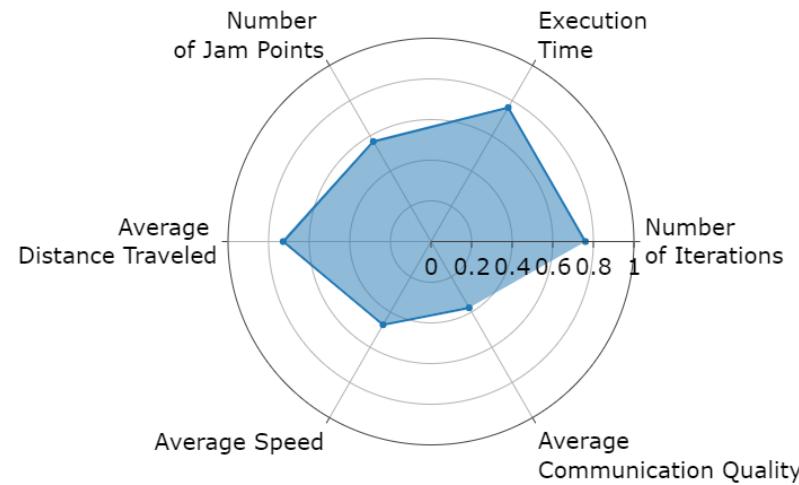


# Simulation Evaluation

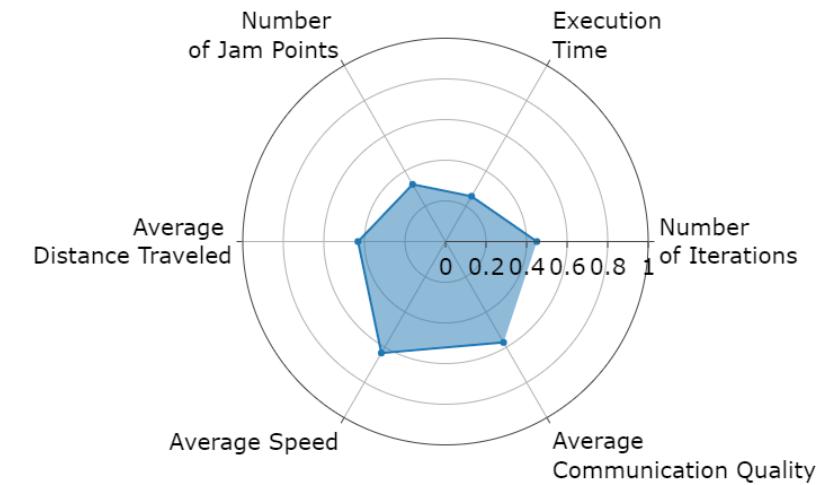
Theta\*



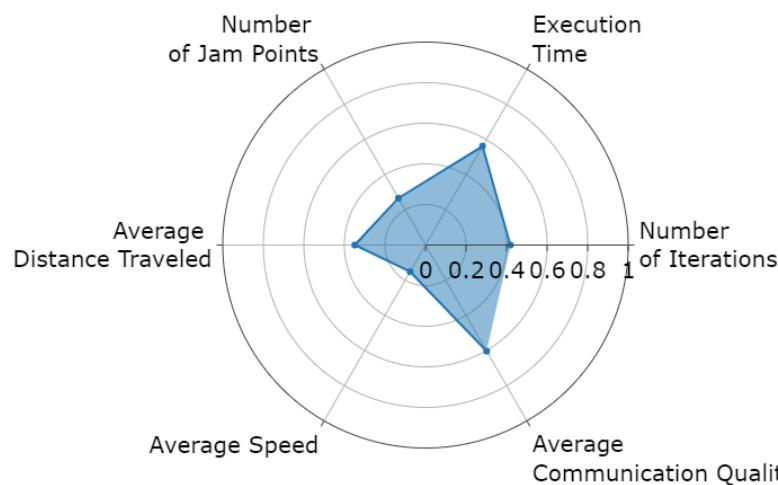
Jump Point Search



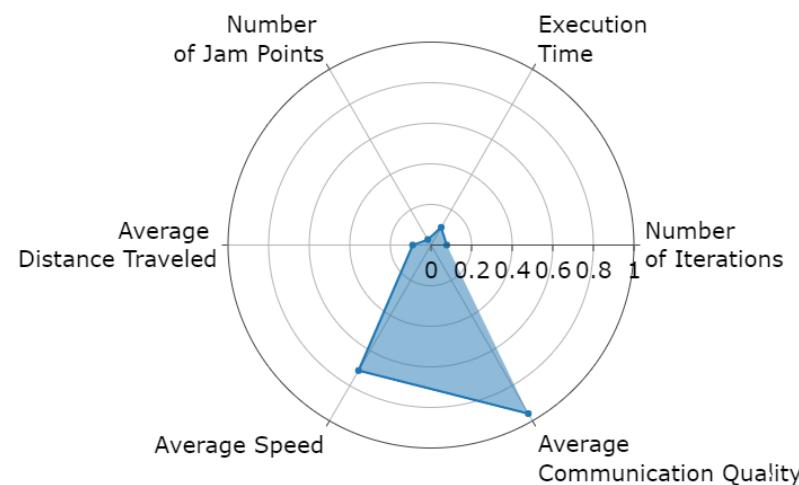
Breadth First Search



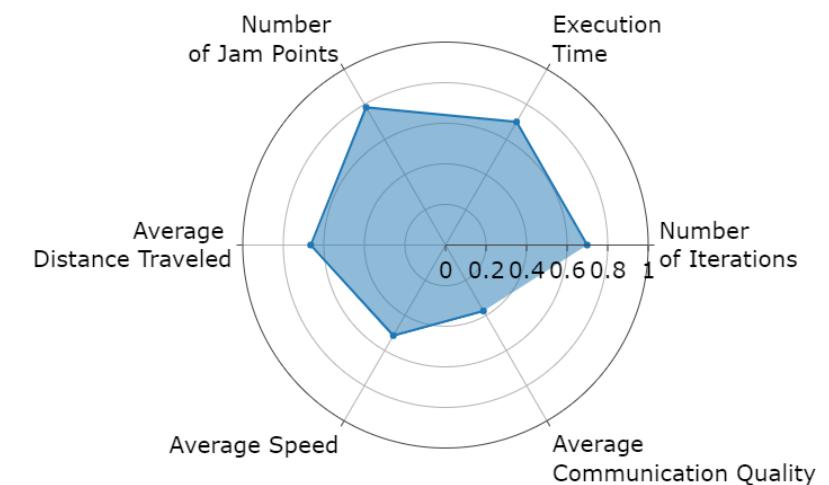
Greedy Best First Search

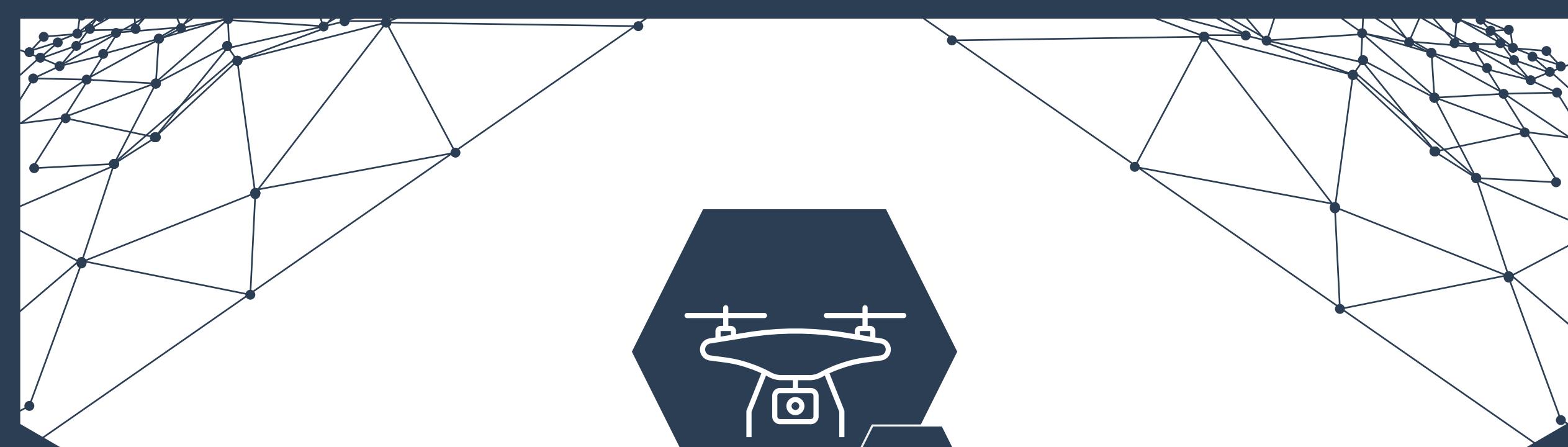


A\*



Dijkstra





05

## Conclusion

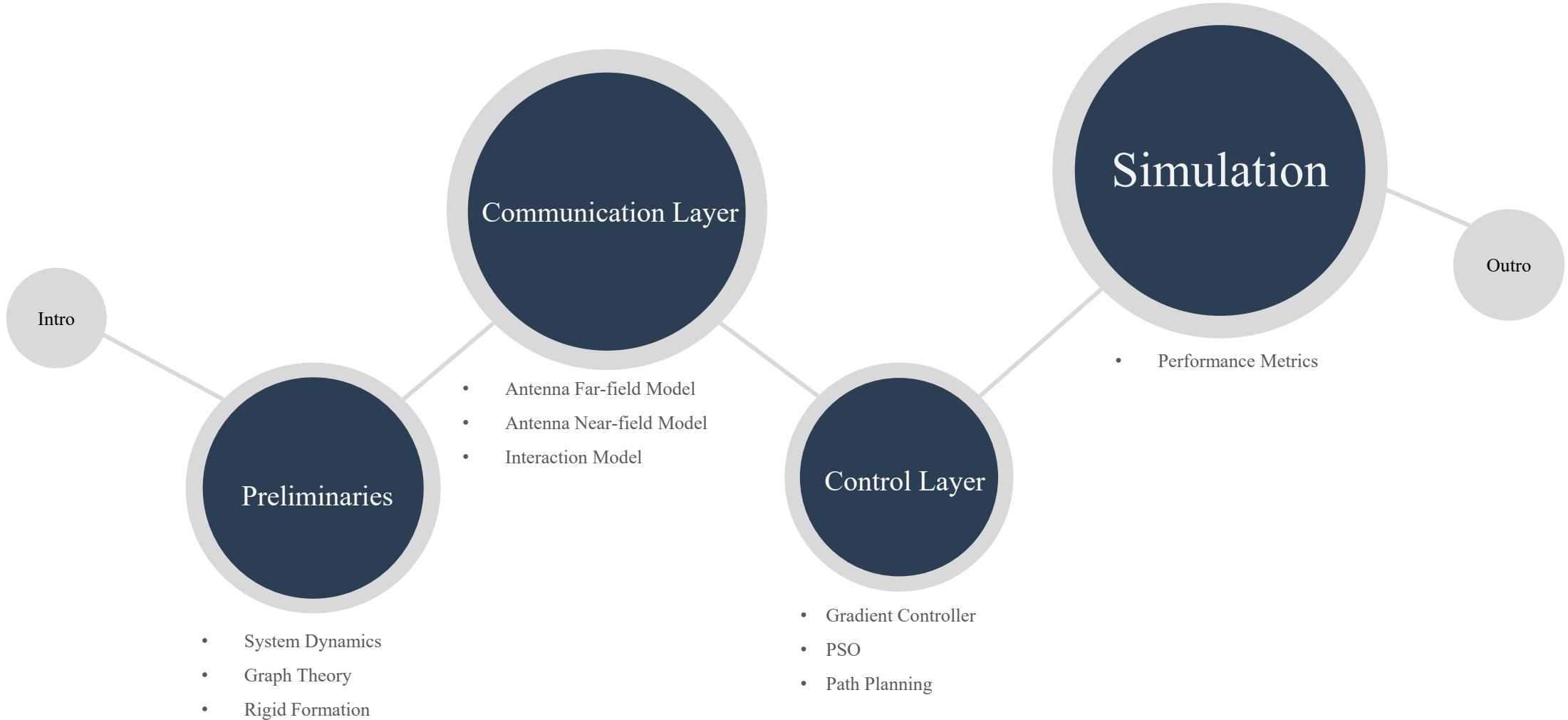


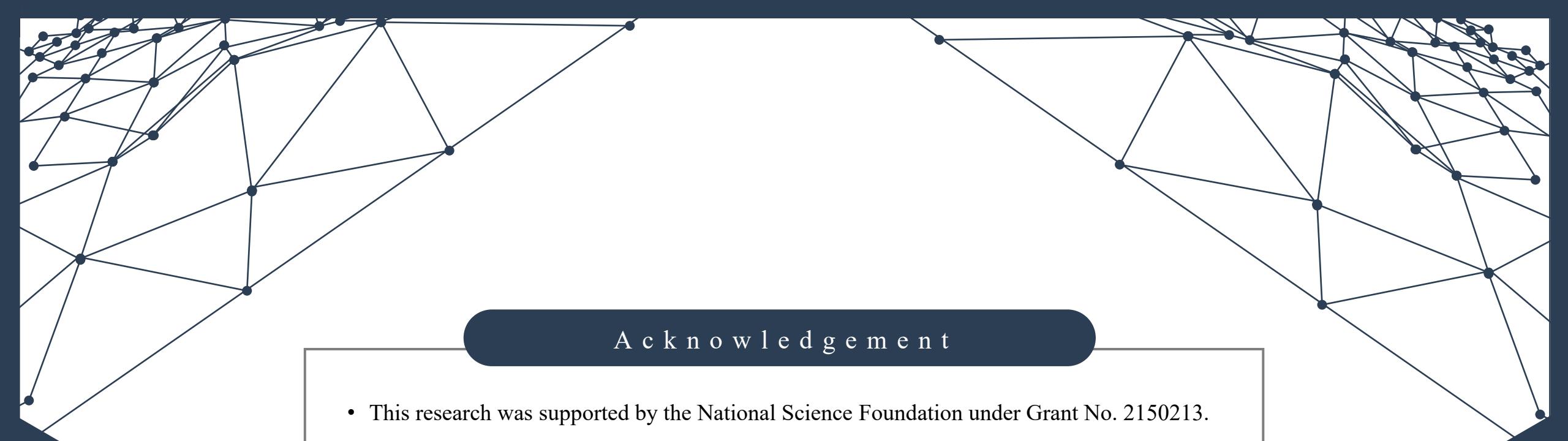
## Result

- A\* is identified as the most effective approach.
- Proposed a novel control strategy for swarm navigation in the presence of unknown jamming areas.
- Ensures efficient navigation, formation maintenance, and robust communication through extensive simulations in all 30 simulations.



# Review





## A c k n o w l e d g e m e n t

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- We'd like to thank you Dr. Stansbury for considering us into this research program.
- We'd like to thank you Dr. Yang's guidance throughout this research project.





# THANKS

Questions ?

August 4, 2023

