

Networked Robotics Systems, Cooperative Control and Swarming (ROB-GY 6333 Section A)

- **Today's lecture:**
 - Nonlinear Formation Control

Standard Recipe

- This is the power of algebraic _____!
 - Algebraic graph theory
 - Algebraic geometry
- Laplacian \rightarrow Consensus Protocol
- ????????? \rightarrow Formation Control
- We will meet a lot of old friends

What might a decentralized linear formation control law look like?

- Assume a 2D formation defined by a framework (graph: $\mathcal{G} = (\mathcal{V}, \mathcal{E})$) and function $p(\mathcal{V})$ to locate vertices

$$\begin{bmatrix} \dot{x}_i \\ \dot{y}_i \end{bmatrix} = -k \sum_{j \in \mathcal{N}_i} \left(\begin{bmatrix} x_i(t) \\ y_i(t) \end{bmatrix} - \begin{bmatrix} x_j(t) \\ y_j(t) \end{bmatrix} \right) - \left(\begin{bmatrix} p_{x,i} \\ p_{x,i} \end{bmatrix} - \begin{bmatrix} p_{y,j} \\ p_{y,j} \end{bmatrix} \right)$$

- Is it local/decentralized?
- Is it stable?
- What is the behavior like?

Formation that can only translate (no rotation)

- Assume a formation defined by a framework with graph $\mathcal{G}_f = (\mathcal{V}, \mathcal{E}_f)$ and function $p(\mathcal{V})$ to locate vertices.
- Assume that agents are connected through a communication graph $\mathcal{G}_{com} = (\mathcal{V}, \mathcal{E}_{com})$
- We want agents to be at the desired locations up to a common translation

$$\dot{\mathbf{x}}_i = \mathbf{p}(v_i) + \boldsymbol{\tau}, \text{ where the position of each agent } \mathbf{x}_i \in \mathbb{R}^d$$

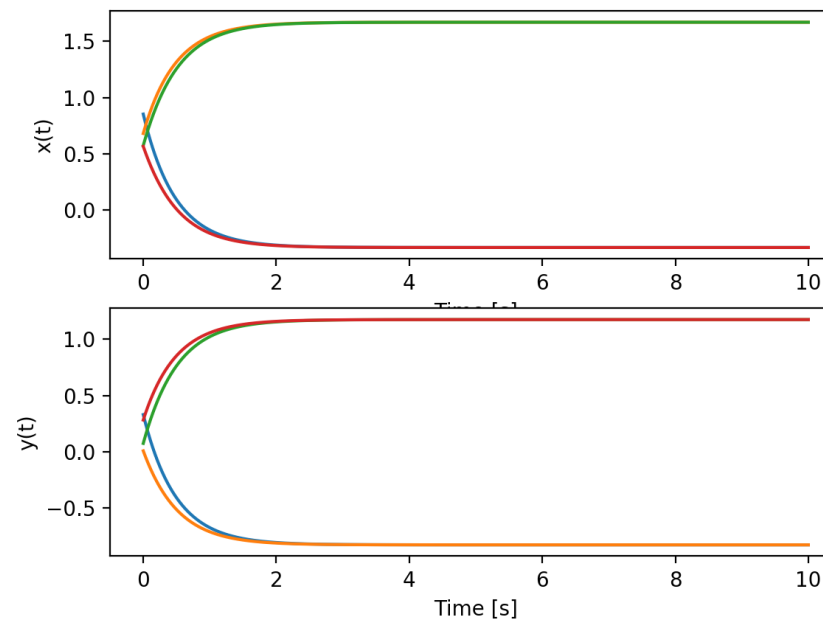
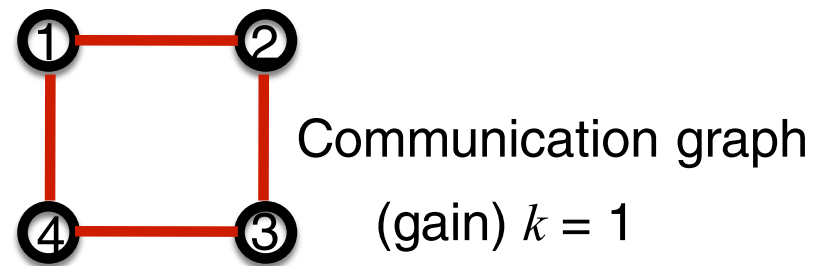
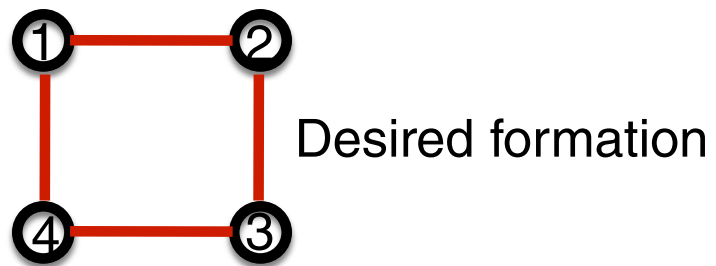
- Define variables $\tau_i(t) = x_i - p_i$.

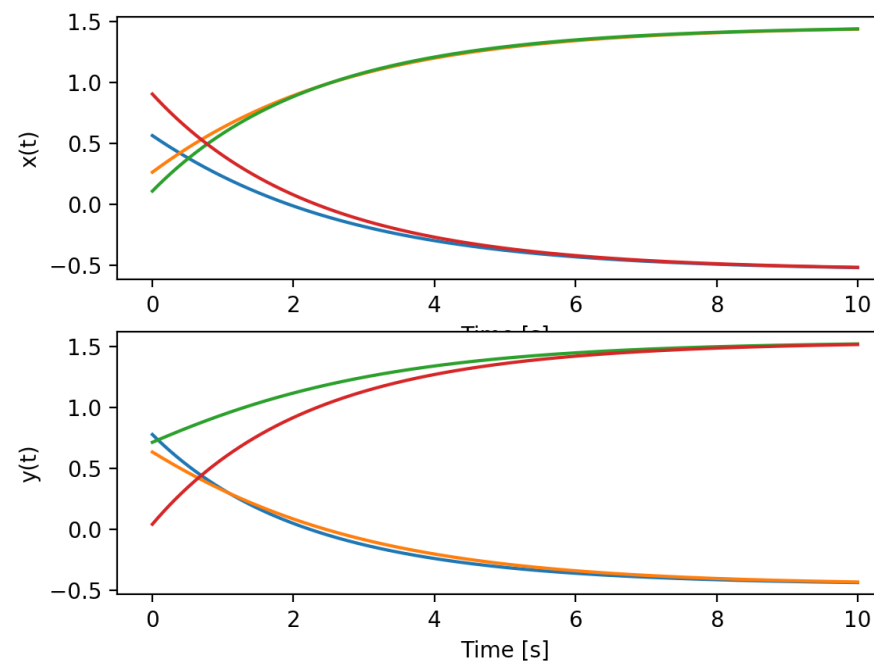
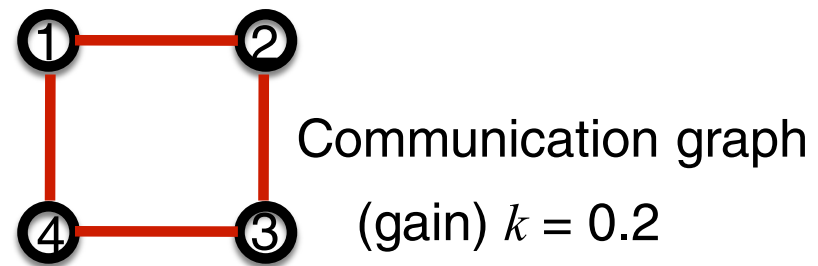
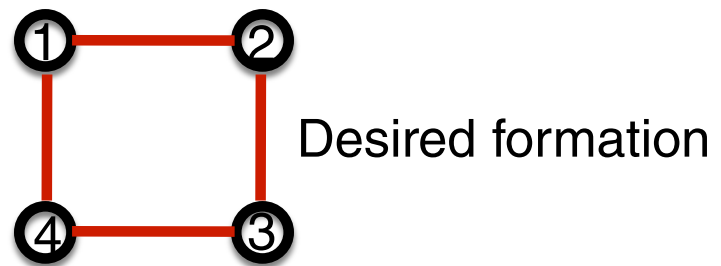
The formation is achieved when there is **consensus on translation** $\forall i, j, \tau_i = \tau_j$

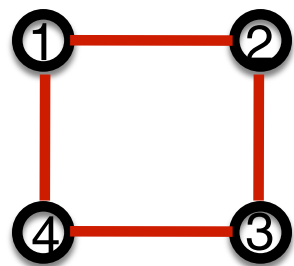
- Results in control law

$$\boxed{\dot{\mathbf{x}}_i = -k \sum_{j \in \mathcal{N}_i} (\mathbf{x}_i(t) - \mathbf{x}_j(t)) - (\mathbf{p}_i - \mathbf{p}_j)}$$

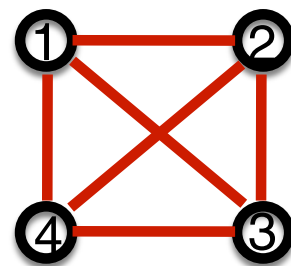
- Convergence is guaranteed as long as $\mathcal{E}_f \subseteq \mathcal{E}_{com}$ and \mathcal{G}_{com} is connected



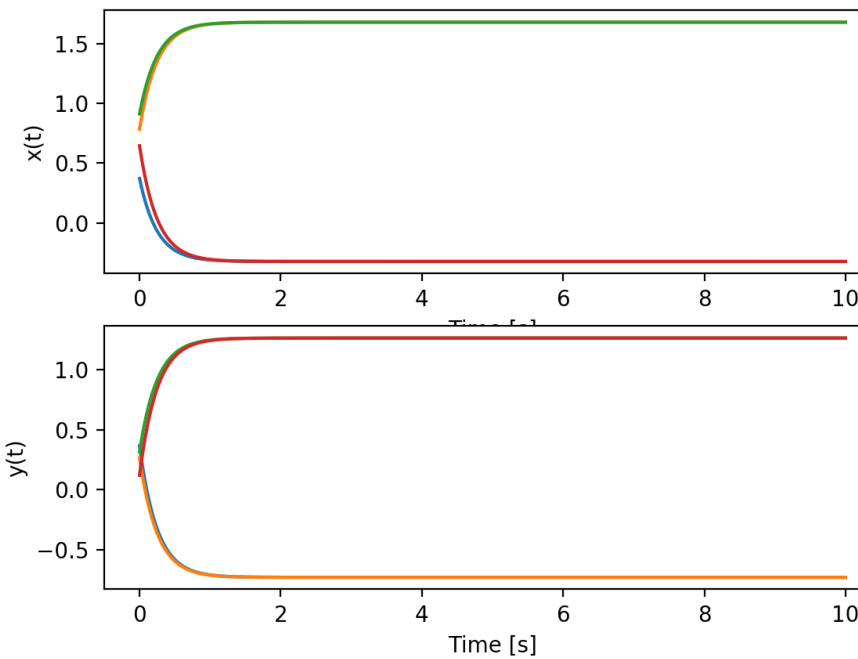


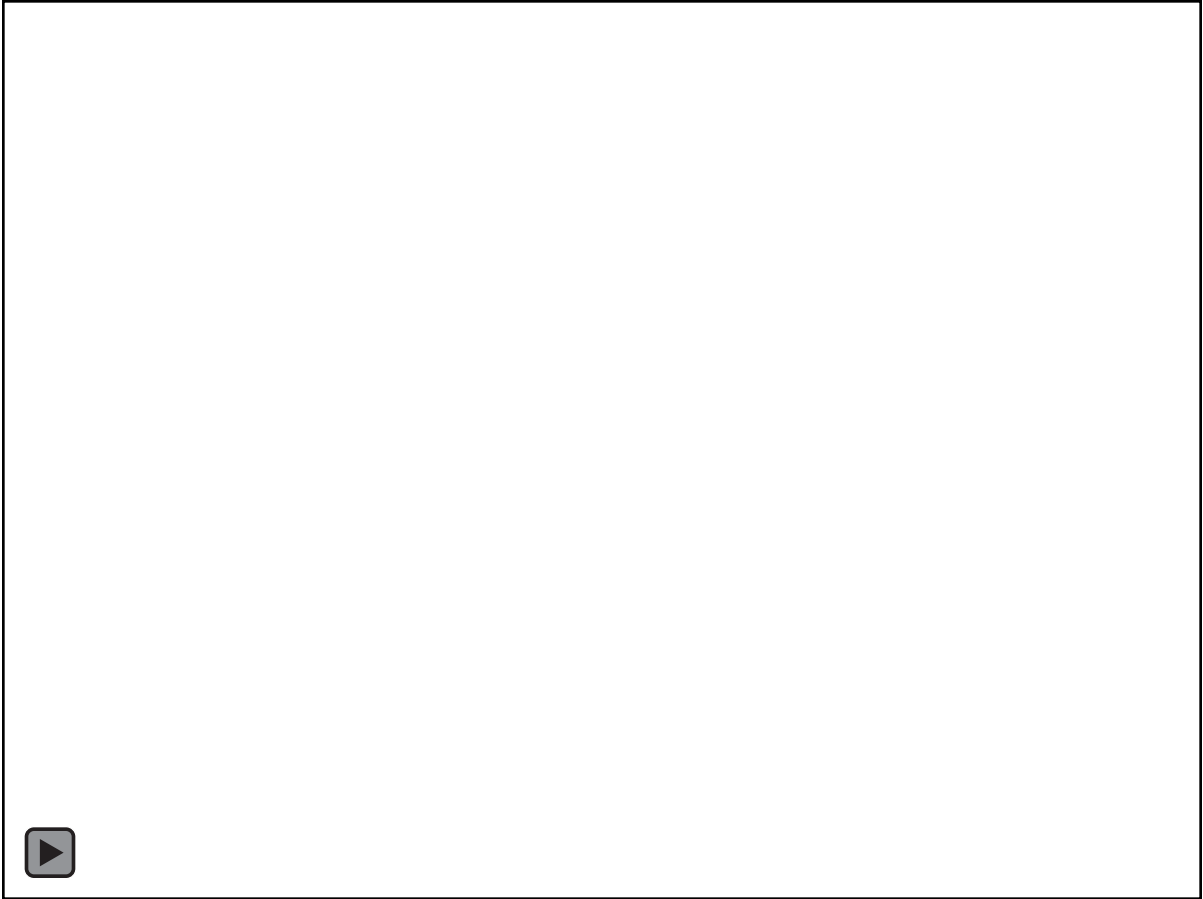


Desired formation



Communication graph
(gain) $k = 0.2$





Issues

- Needs the absolute position of all agents (e.g. using GPS)
- Does not allow rotations
- No guarantees that connectivity can be maintained at all times (especially with linear control)
- Quick fix when original connectivity graph is not good enough
 - First agreement protocol to get all agents to form a fully connected graph
 - Then apply formation control

Next level

- In rendezvous, the agents converge to something like a “center-of-mass”
- **Idea:** For formations without rotation, easy to define the “center-of-mass” motion, and then figure out the relative position of each agent relative to the “center-of-mass”
 - No need for absolute coordinates (GPS)
- As usual, first look for a simple linear control law

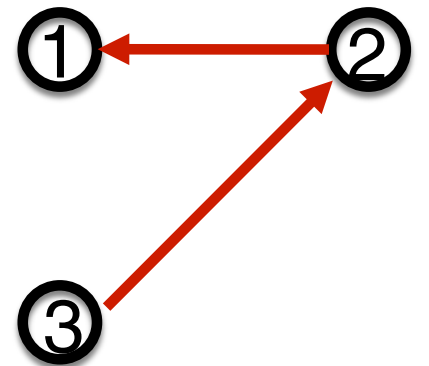
Formation control: relative state-based control

- Something closer to the original “center-of-mass” idea we started with:
 - Consider three agents and a fully connected formation, compute the relative states $z(t)$ and impose a reference z_{ref}

$$z(t) = \begin{bmatrix} x_1(t) - x_2(t) \\ x_2(t) - x_3(t) \end{bmatrix}$$

- The **incidence matrix** returns here as the matrix that maps between the relative and actual state

$$z(t) = E^T x(t)$$



Formation control: relative state-based control

- Inspired by the Laplacian-based consensus protocol, you may be tempted to pick a control law of the form

$$\dot{\mathbf{x}} = -kL(\mathbf{x} - \mathbf{x}_{ref})$$

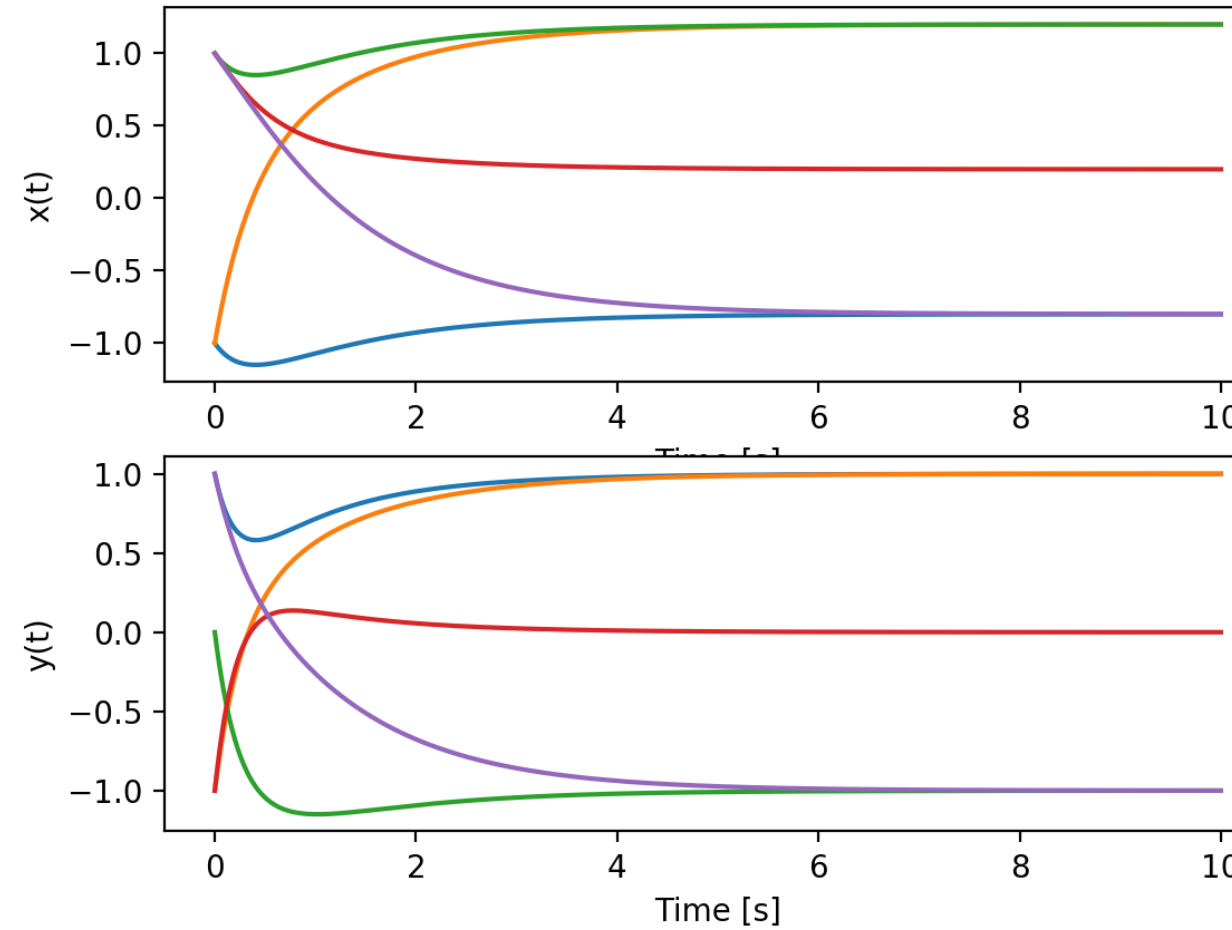
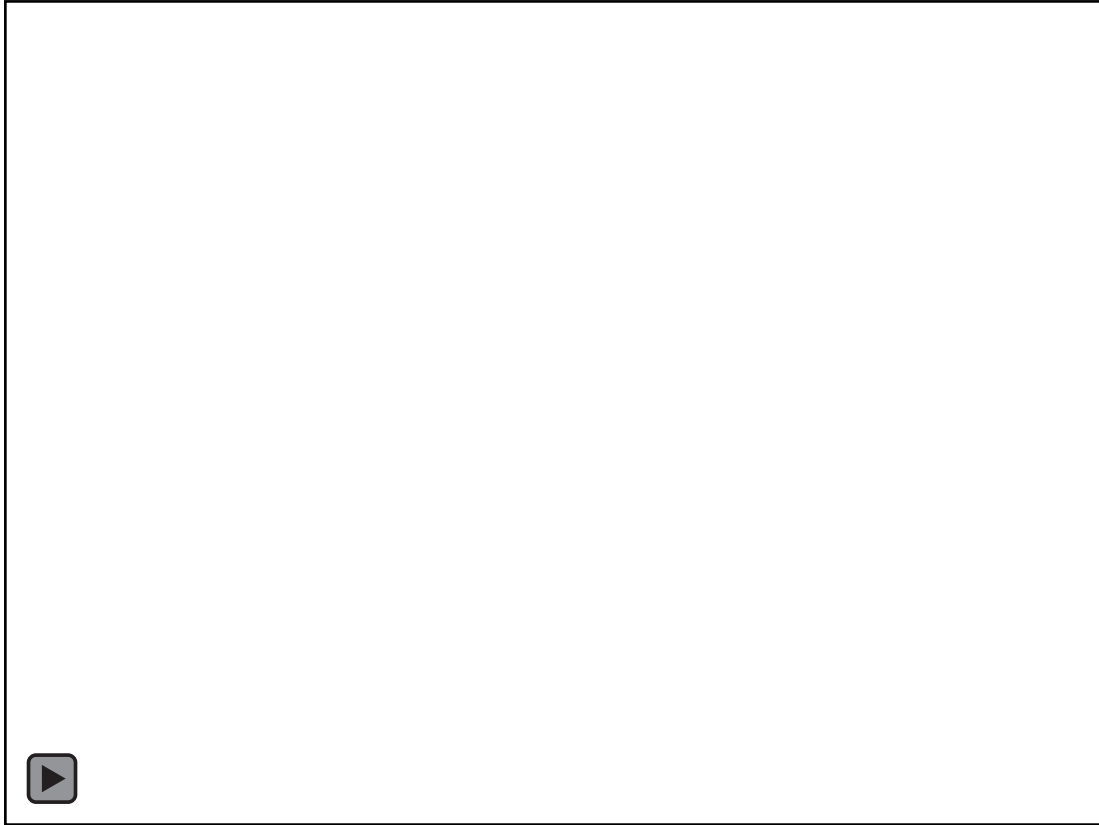
- However, \mathbf{x}_{ref} doesn't really exist, because there is no absolute position we must converge. But if we play around with the idea and recall that $L = EE^T$

$$E(\mathbf{z}_{ref}) = E(E^T \mathbf{x}_{ref}) = L\mathbf{x}_{ref}$$

- You can obtain a local linear control law that only needs relative positions to converge onto \mathbf{z}_{ref}

$$\dot{\mathbf{x}} = -kL\mathbf{x} + kE\mathbf{z}_{ref}$$

Formation control: relative state-based control



Formation control: relative state-based control

- Previous slide: Single integrator dynamics
(trivial dynamics, only control input)

$$\dot{\mathbf{x}} = \mathbf{u}$$

$$\mathbf{u} = -kL\mathbf{x} + kE\mathbf{z}_{ref}$$

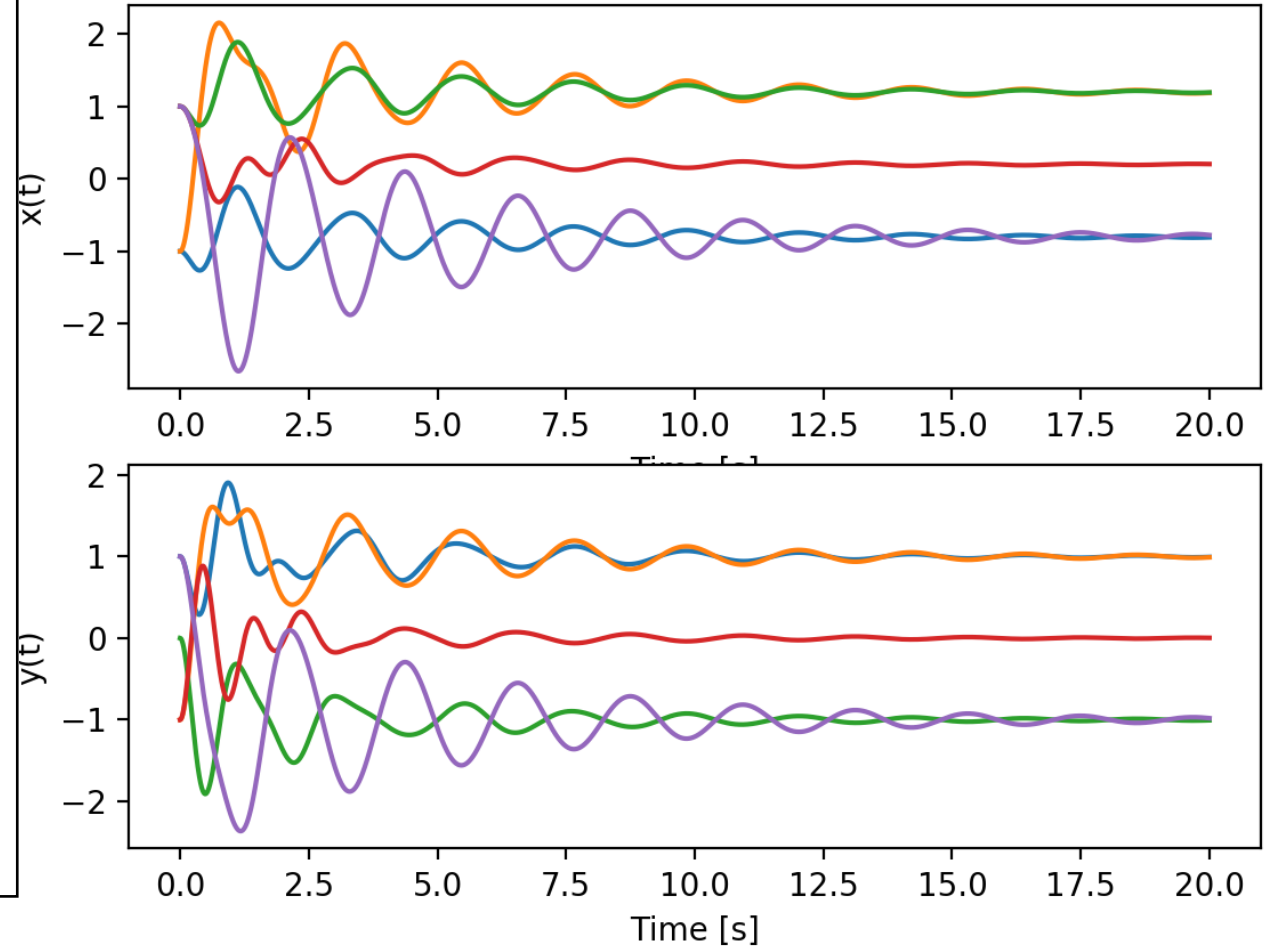
- Double integrator dynamics

$$\ddot{\mathbf{x}} = \mathbf{u}$$

$$\mathbf{u} = -kL\mathbf{x} + kE\mathbf{z}_{ref} - dL\dot{\mathbf{x}} + dE\dot{\mathbf{z}}_{ref}$$

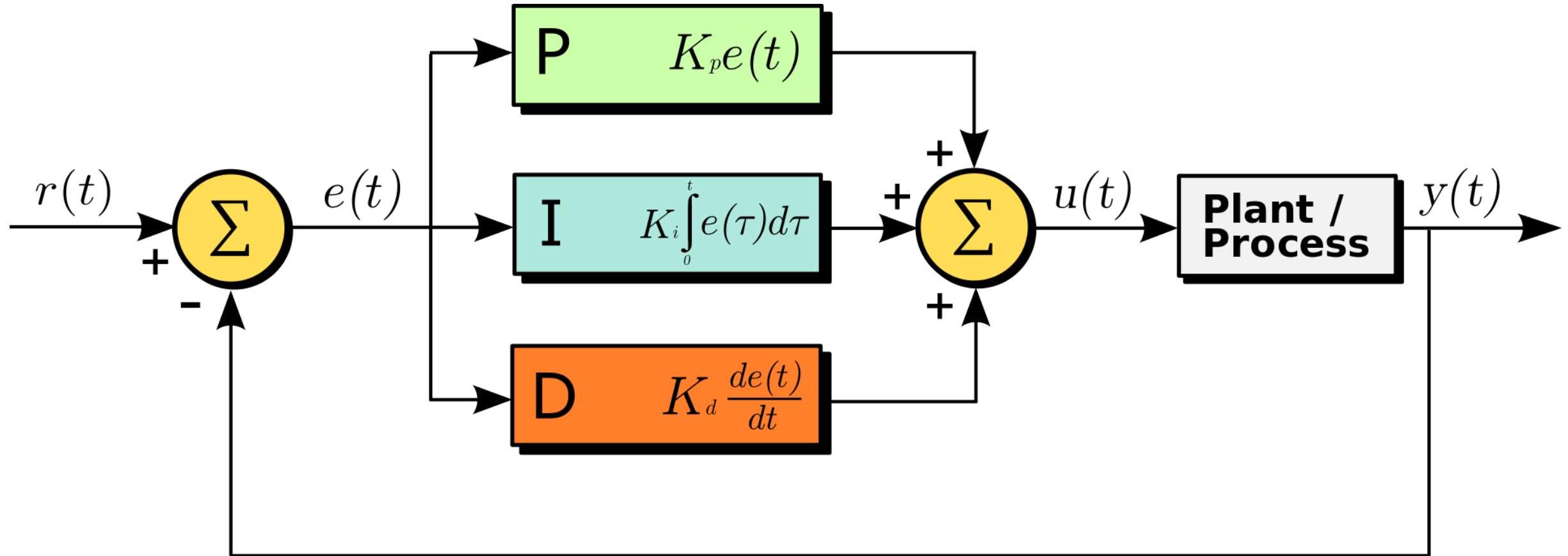
Converges to $\mathbf{z}(t) = \mathbf{z}_{ref}(t)$, where $k, d > 0$, if the connectivity undirected graph contains a spanning tree (similar to rooted out-branching, an acyclic graph that reaches all vertices, but no notion of root)

Formation control: relative state-based control



$$k = 10, d = 0.5$$

PID control and Integrator Dynamics



Rigidity-based control law

$$\dot{\mathbf{p}} = \mathbf{u}$$

$$\mathbf{u} = \mathbf{R}_{\mathcal{G}}^T(\mathbf{p})(\rho_{ref} - \rho_{\mathcal{G}}(\mathbf{p}))$$



Rigidity-based control law

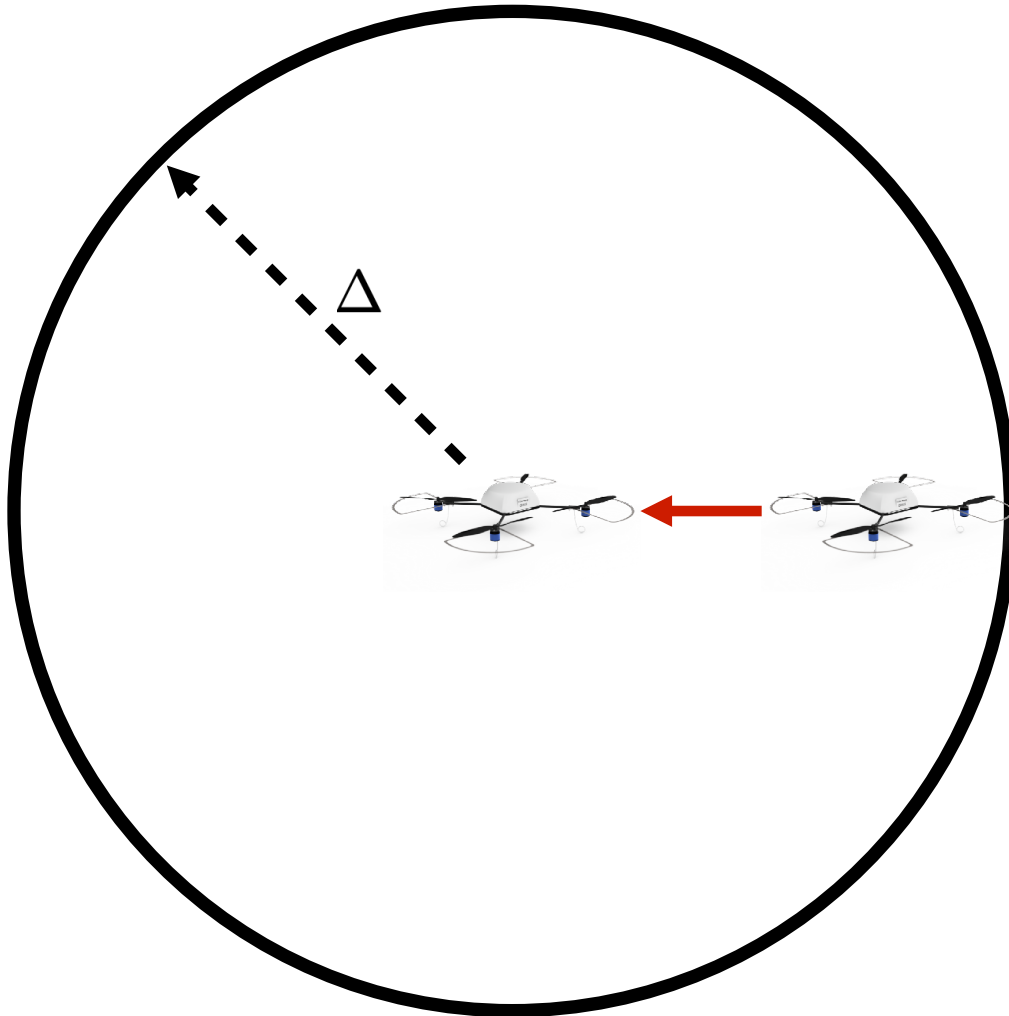
$$\dot{\mathbf{p}} = \mathbf{u}$$

$$\mathbf{u} = \mathbf{R}_{\mathcal{G}}^T(\mathbf{p})(\rho_{ref} - \rho_{\mathcal{G}}(\mathbf{p}))$$

- Is local
- Translation and rotation invariant
- Nonlinear control law based on rigidity
 - Nonlinear is not necessarily always better than linear, but it allows for more complicated behavior
- Can be problematic when rigidity matrix is singular

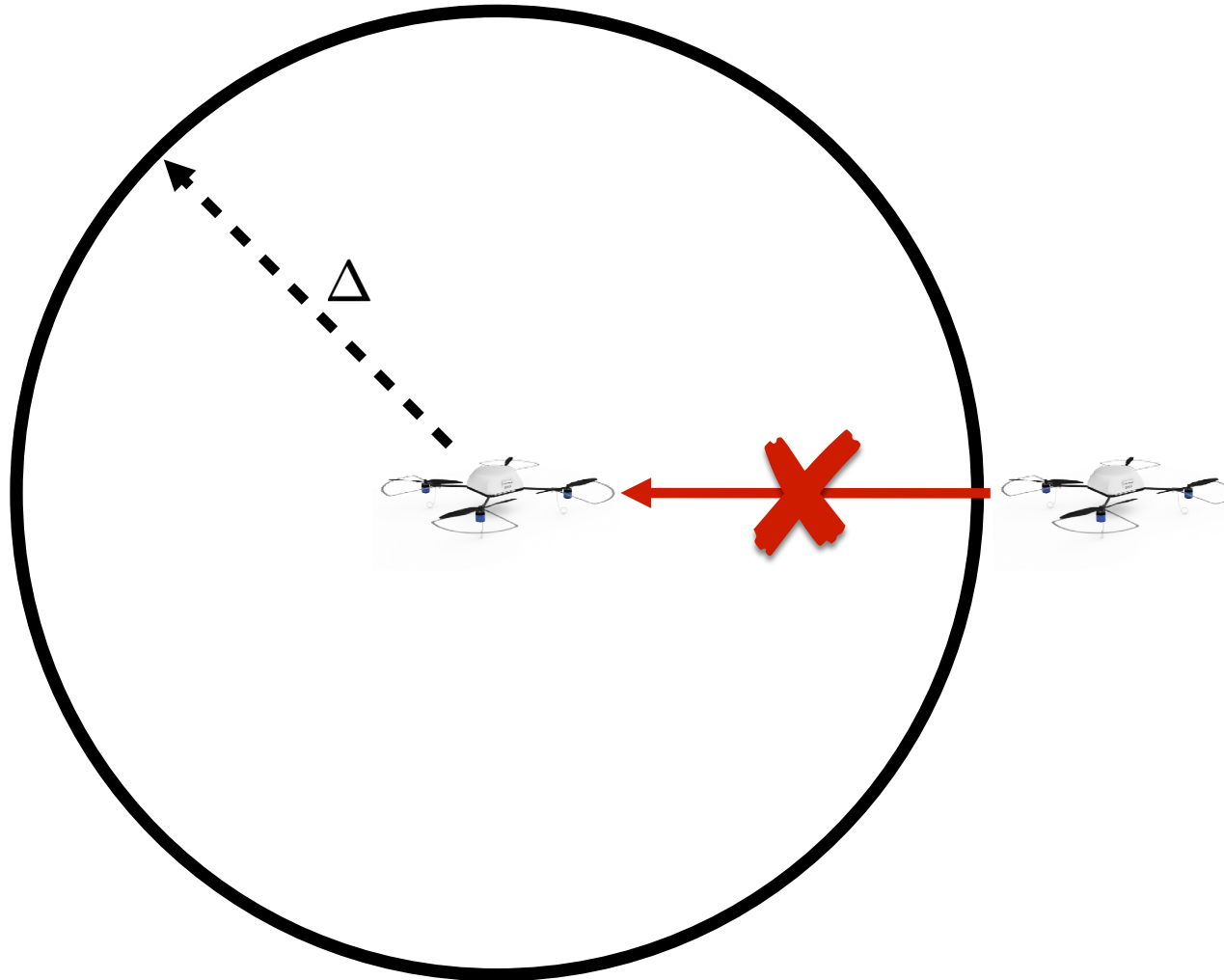
Formation control and real-world constraints

- Real robots will typically be able to communicate only locally



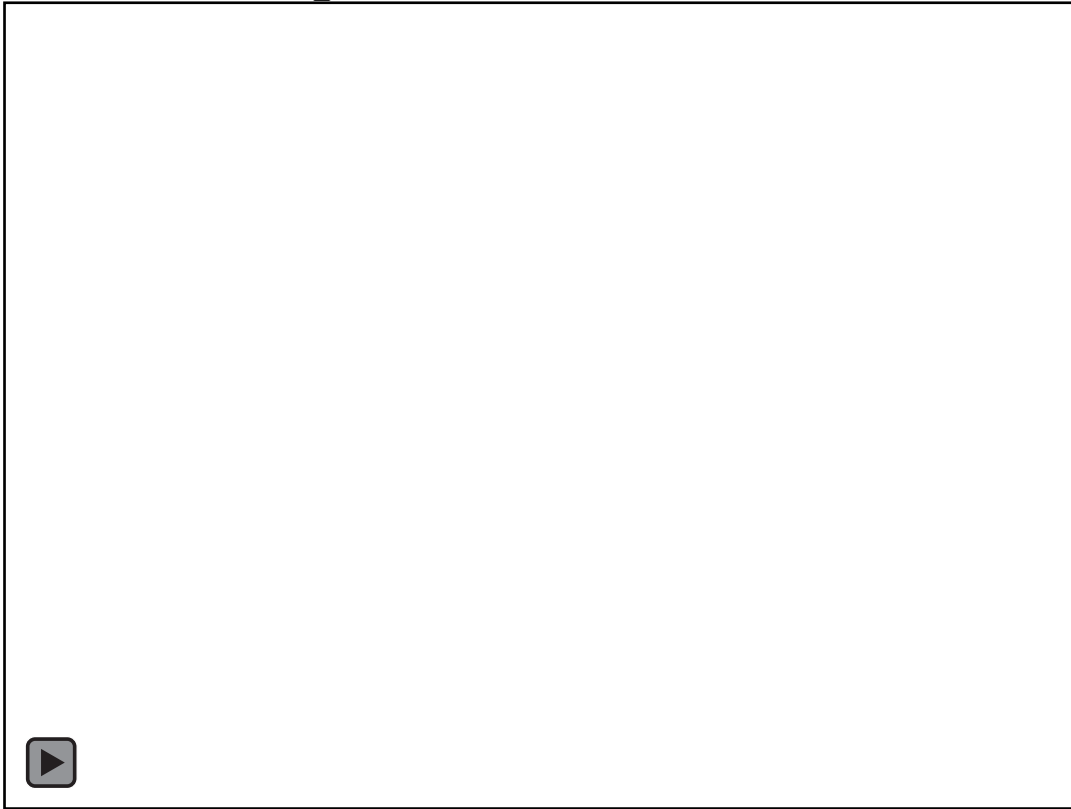
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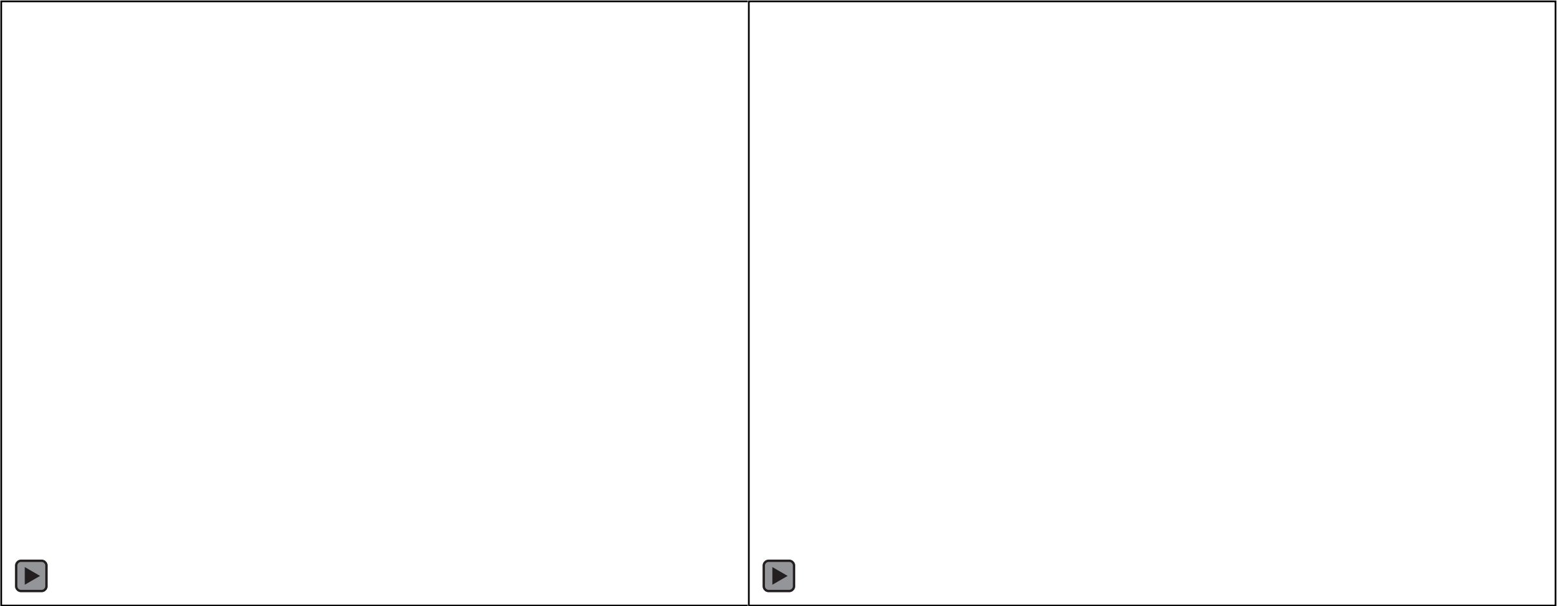
Formation control and real-world constraints

- Can we guarantee that during consensus or formation control the distance between neighbors does not exceed a certain radius Δ ?



$$\Delta = 1.5$$

Formation control and real-world constraints



$$\Delta = 1.5$$

Weighted graph-based feedback

- Idea: extend formation control to take into account maximum distance constraints
 - Control law with single integrator dynamics:

$$\begin{aligned}\dot{\mathbf{x}}_i &= \mathbf{u}_i \\ \mathbf{u}_i &= - \sum_{j \in \mathcal{N}_i} f(\mathbf{x}_i - \mathbf{x}_j)\end{aligned}$$

where f is antisymmetric $f(\mathbf{x}_i - \mathbf{x}_j) = -f(\mathbf{x}_j - \mathbf{x}_i)$

- Local (depends only on neighbors) and homogenous (same for all agents) control law
- Only relative information is needed
- Centroid of the states is guaranteed to be constant! $\bar{\mathbf{x}} = \frac{1}{N} \sum_i \mathbf{x}_i$

Weighted graph-based feedback: Rendezvous

- How to pick the function $f(\cdot)$?
- **Idea:** Connection strength changes as a function of distance between agents

$$f(\mathbf{x}_i - \mathbf{x}_j) = -\omega(\mathbf{x}_i - \mathbf{x}_j) \cdot (\mathbf{x}_i - \mathbf{x}_j)$$

where $\omega(\mathbf{x}) : \mathbb{R}^p \rightarrow \mathbb{R}$ is a symmetric positive definite function

- This choice will guarantee a Laplacian-like control law

$$\dot{\mathbf{x}} = -E(\mathcal{G})W(x)E(\mathcal{G})^T \mathbf{x}$$

where $L_W = -E(\mathcal{G})W(x)E(\mathcal{G})^T \geq 0$ as long the G is connected (with only one zero eigenvalue)

Weighted graph-based feedback: Rendezvous

- $\mathcal{D}_{\mathcal{G},\delta}$ is the set of all swarm configurations (recall points $p \in \mathbb{R}^{Nd}$) where robots are always within a given distance $\delta > 0$ from each other.

$$\mathcal{D}_{\mathcal{G},\delta} = \{p \in \mathbb{R}^{Nd} \mid \|p_i - p_j\| \leq \delta, \forall v_i, v_j \in \mathcal{E}\}$$

Weighted graph-based feedback: Rendezvous

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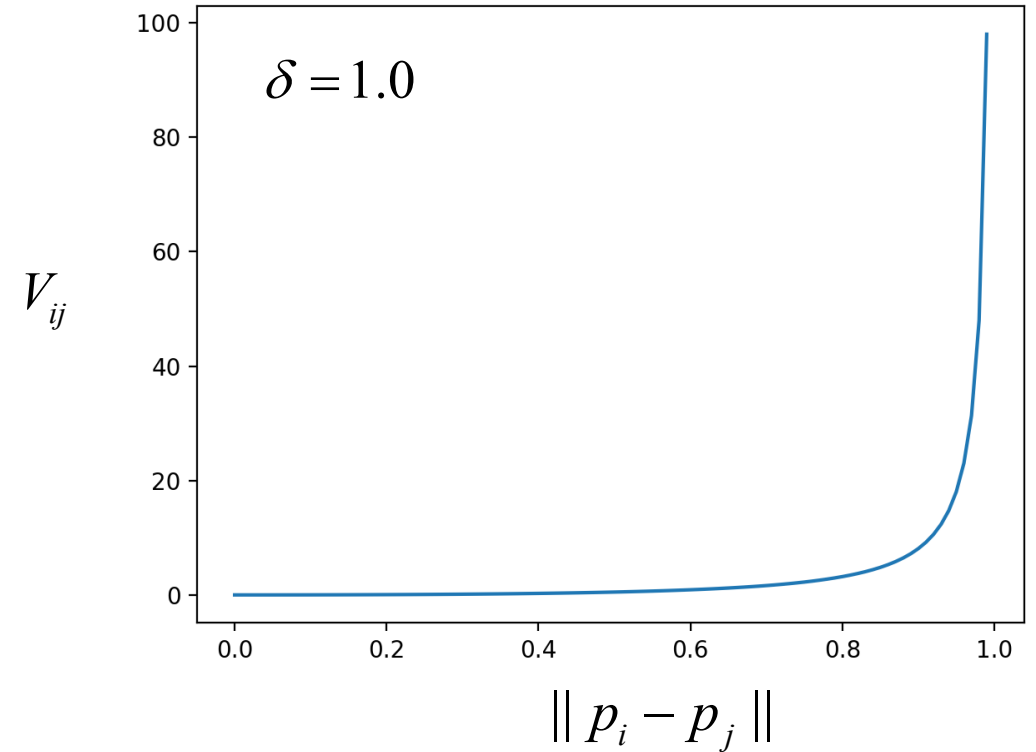
- $\mathcal{D}_{\mathcal{G},\delta}^{\varepsilon}$ is the set of all swarm configurations where robots are always within a given distance $\delta - \varepsilon > 0$ from each other. For now, think of ε as a bit of extra margin.

$$\mathcal{D}_{\mathcal{G},\delta}^{\varepsilon} = \{p \in \mathbb{R}^{Nd} \mid \|p_i - p_j\| \leq \delta - \varepsilon, \forall v_i, v_j \in \mathcal{E}\}$$

Weighted graph-based feedback: Rendezvous

- Define edge tension as a function

$$V_{ij}(\delta, p) = \begin{cases} \frac{\|p_i - p_j\|^2}{\delta - \|p_i - p_j\|} & \text{for } v_i, v_j \in \mathcal{E} \\ 0 & \text{otherwise} \end{cases}$$



- Define the total tension energy as $V(\delta, p) = \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N V_{ij}(\delta, p)$
 - Looks weird because the units are not in Joules

Weighted graph-based feedback: Rendezvous

- Recall: $\mathcal{D}_{\mathcal{G},\delta}^{\varepsilon} = \{p \in \mathbb{R}^{Nd} \mid \|p_i - p_j\| \leq \delta - \varepsilon, \forall v_i, v_j \in \mathcal{E}\}$ $V(\delta, p) = \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N V_{ij}(\delta, p)$

- To make things simpler for control, I will define the system state \mathbf{x} as the point p

$$\mathbf{x} \triangleq p$$

- Given an initial position $\mathbf{x}_0 \in \mathcal{D}_{\mathcal{G},\delta}^{\varepsilon}$ for $\varepsilon \in (0, \delta)$, if the graph \mathcal{G} is connected, then

$$\Omega(\delta, \mathbf{x}_0) = \{\mathbf{x} \mid V(\delta, \mathbf{x}) \leq V(\delta, \mathbf{x}_0)\}$$

is an invariant set under the control law
$$\mathbf{u}_i = - \sum_{j \in \mathcal{N}_i} \frac{2\delta - \|\mathbf{x}_i - \mathbf{x}_j\|}{(\delta - \|\mathbf{x}_i - \mathbf{x}_j\|)^2} (\mathbf{x}_i - \mathbf{x}_j)$$

where $x(t)$ will converge to the centroid $\bar{x} = \frac{1}{N} \sum_i \mathbf{x}_i$

Weighted graph-based feedback: Rendezvous

- Reflect: We just designed a swarm where everybody stays in touch
 - Not obvious from just looking at a control law
 - Required thinking back about invariant sets and Lyapunov functions

Why the margin ε ?

- Graph dynamics based on proximity function

$$v_i, v_j \in \mathcal{E} \Leftrightarrow \|p_i - p_j\| \leq \Delta$$

- We can try to set $\delta = \Delta$ but we cannot switch the controller on as soon as the connection is made

Edge created

$$f(\mathbf{x}_i - \mathbf{x}_j) = 0$$

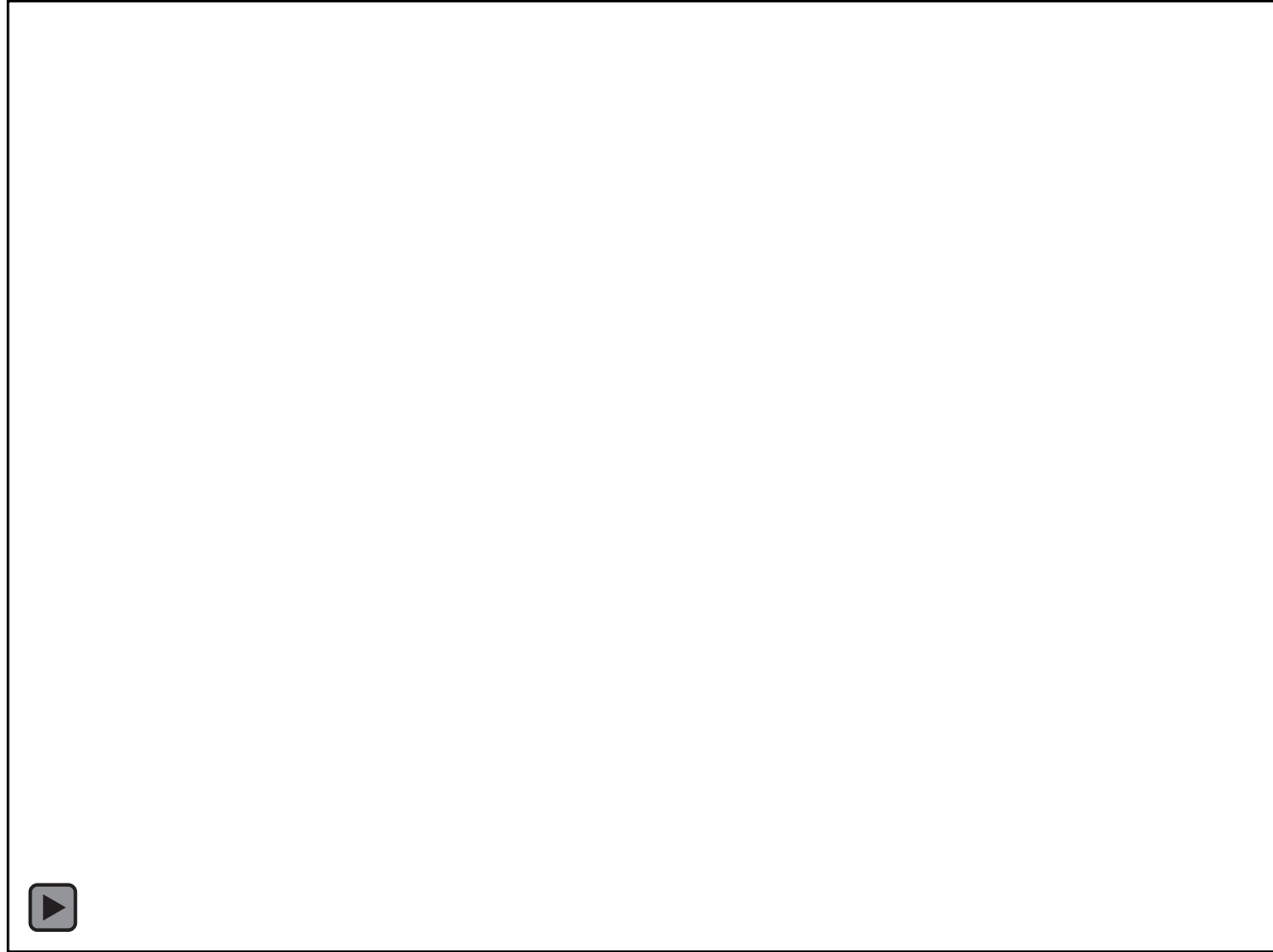
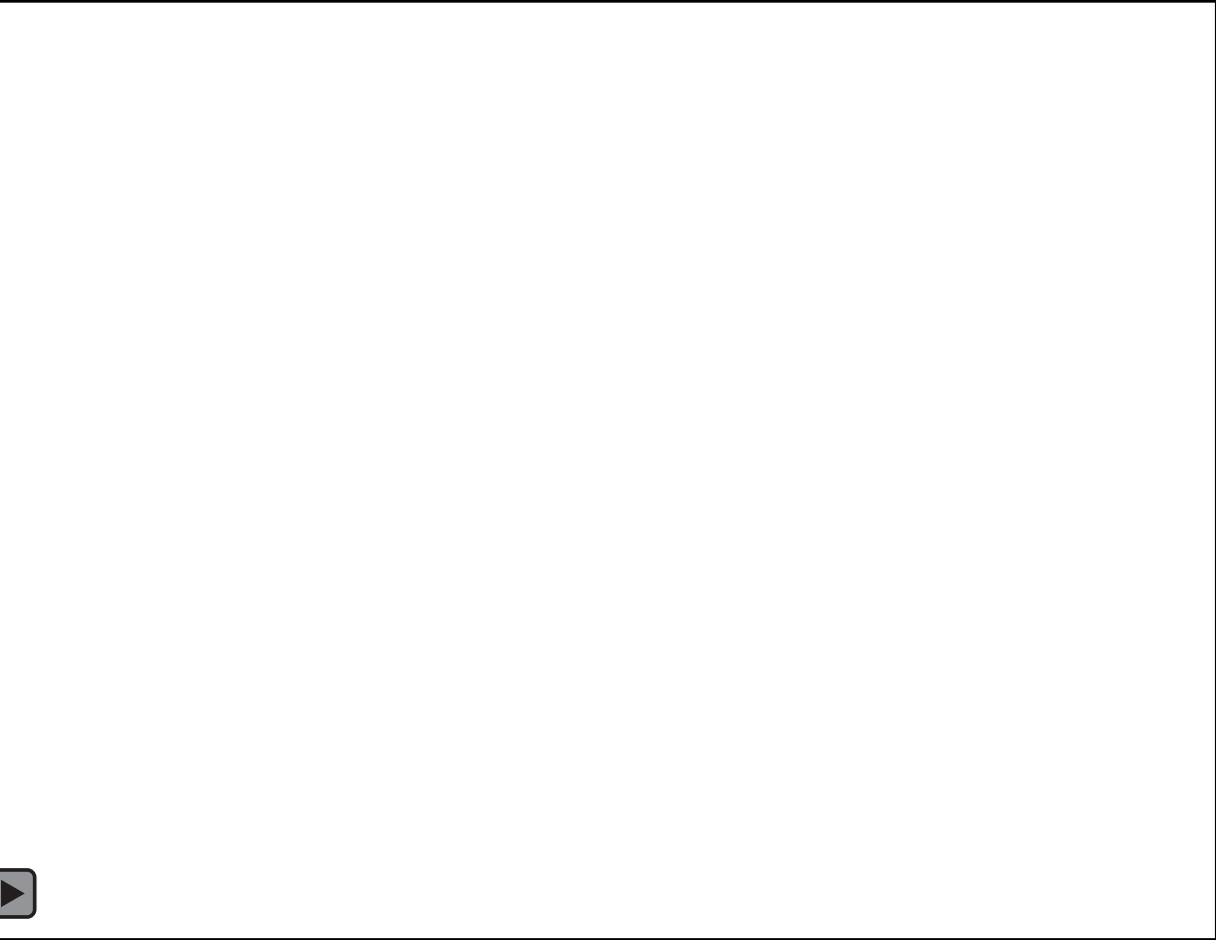
$$\|p_i - p_j\| \leq \Delta - \varepsilon$$



Edge influence starts

$$f(\mathbf{x}_i - \mathbf{x}_j) = \frac{2\delta - \|\mathbf{x}_i - \mathbf{x}_j\|}{(\delta - \|\mathbf{x}_i - \mathbf{x}_j\|)^2} (\mathbf{x}_i - \mathbf{x}_j)$$

Weighted graph-based feedback: Rendezvous



Formation control with weights

- The translation invariant formation control law can be modified to ensure connectivity is maintained as

$$\dot{\mathbf{x}}_i(t) = - \sum_{j \in \mathcal{N}_{\mathcal{G}_d}(i)} \frac{2(\Delta - \|\boldsymbol{\tau}_i - \boldsymbol{\tau}_j\|) - \|\mathbf{x}_i(t) - \mathbf{x}_j(t) - (\boldsymbol{\tau}_i - \boldsymbol{\tau}_j)\|}{(\Delta - \|\boldsymbol{\tau}_i - \boldsymbol{\tau}_j\| - \|\mathbf{x}_i(t) - \mathbf{x}_j(t) - (\boldsymbol{\tau}_i - \boldsymbol{\tau}_j)\|)^2} (\mathbf{x}_i(t) - \mathbf{x}_j(t) - (\boldsymbol{\tau}_i - \boldsymbol{\tau}_j))$$

where $\boldsymbol{\tau}_i - \boldsymbol{\tau}_j$ is the desired translation vector between two vertices i and j
for all i, j such that $v_i, v_j \in \mathcal{E}_d$

Formation control with weights

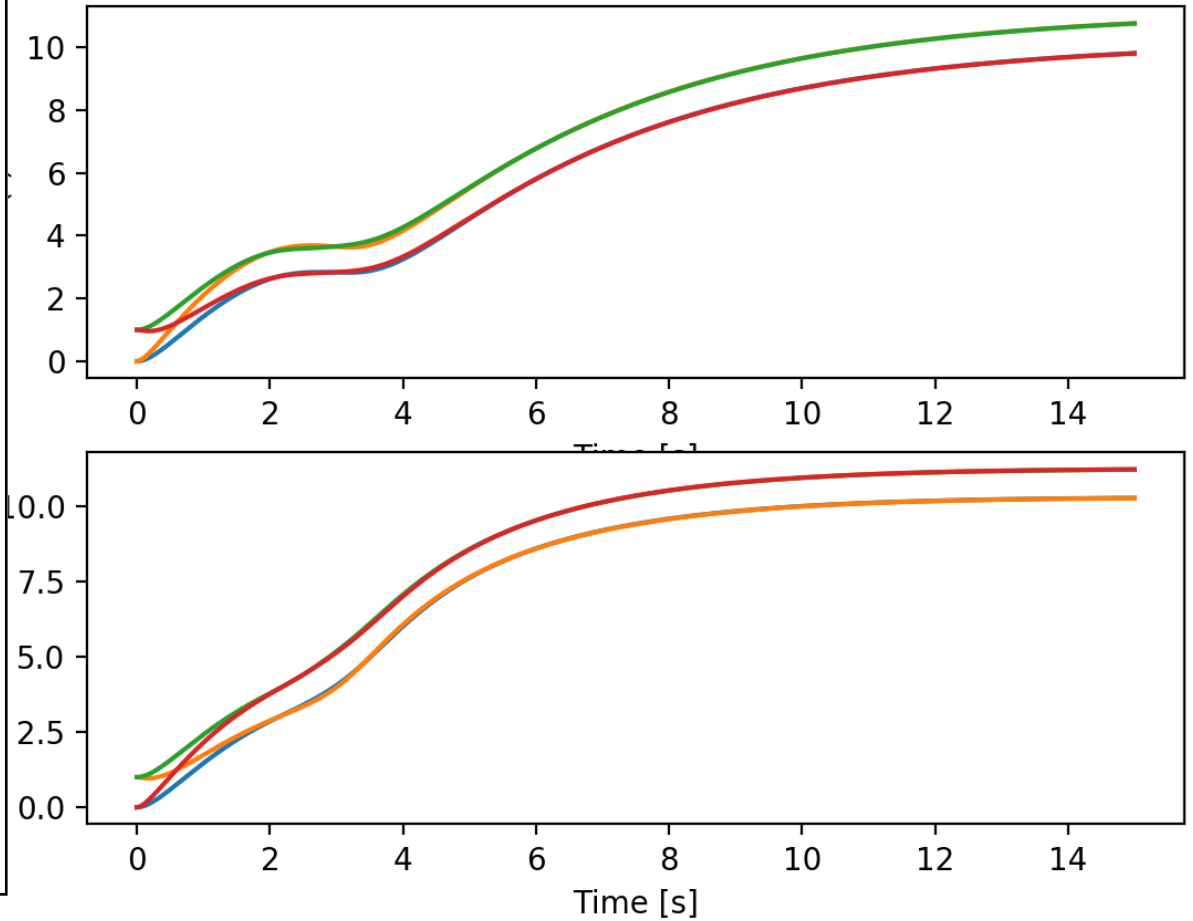
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where $\mathbf{\tau}_i - \mathbf{\tau}_j$ is the desired translation vector for the formation between two vertices i and j for all i, j such that $v_i, v_j \in \mathcal{E}_d$

- Think about how to make it rotation invariant

Control of formation + target + obstacle avoidance



Control Law Example

- Why is a second derivative an appropriate choice here?

$$\begin{bmatrix} \ddot{x} \\ \ddot{y} \end{bmatrix} = F_{formation,i} + F_{target,i} + F_{obstacle,i}$$