

# Networked Control Modeling, Design, and Applications

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Outline:

1. Graph-Based Control
2. Multi-Agent Networks
3. Control of Robot Teams
4. Sensor Networks





## A Mood Picture

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### Automatic Deployment and Assembly of Persistent Multi-Robot Formations

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January, 2009



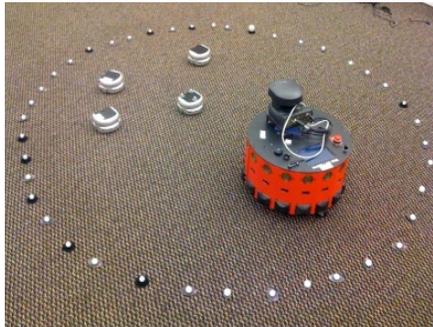
# Ruining the Mood...

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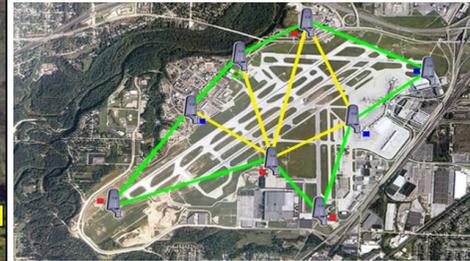
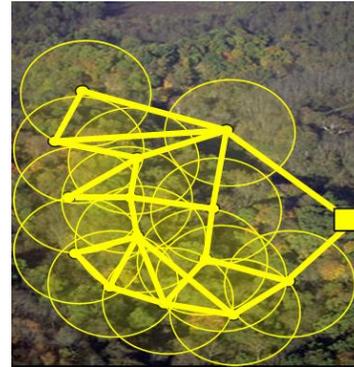


# Application Domains

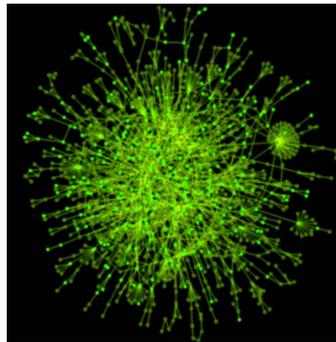
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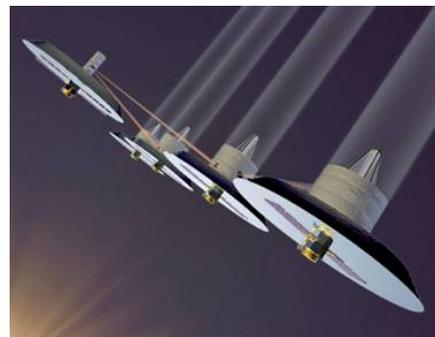
Multi-agent robotics



Sensor and communications networks



Biological networks



Coordinated control



## The Mandatory Bio-Slide

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- As sensor webs, large-scale robot teams, and networked embedded devices emerge, algorithms are needed for inter-connected systems with *limited communication, computation, and sensing capabilities*



- How to effectively control such systems?
  - What is the correct model?
  - What is the correct mode of interaction?
  - Does every individual matter?



# The Starting Point

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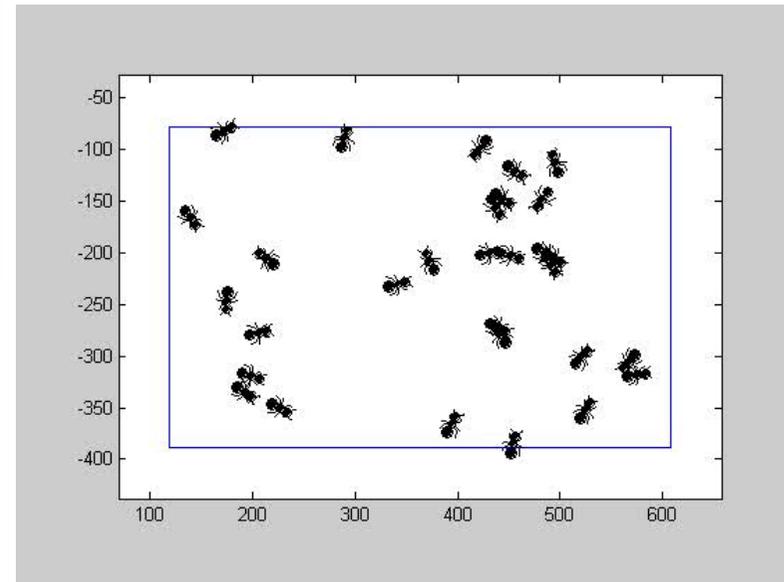
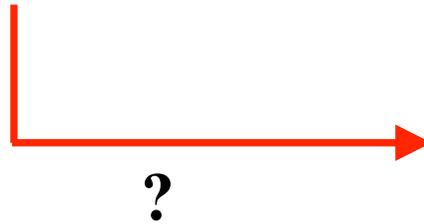
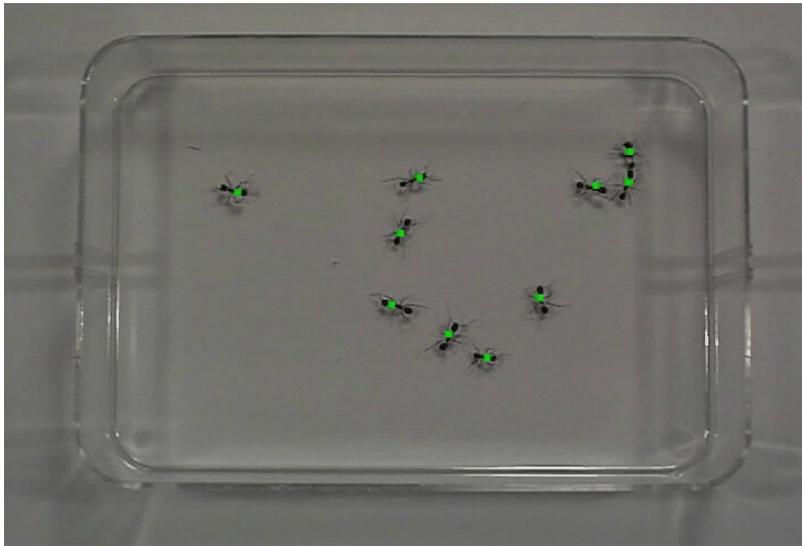


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# SESSION 1

# GRAPH-BASED CONTROL

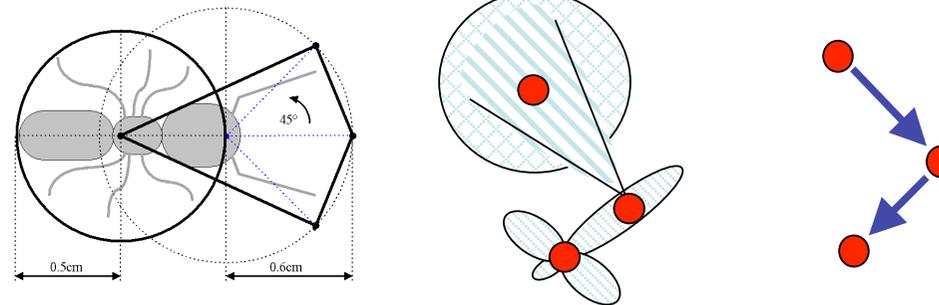
# Why I Started Caring About Multi-Agent Systems



**“They look like ants.”**  
 – Stephen Pratt, Arizona State University

## Graphs as Network Abstractions

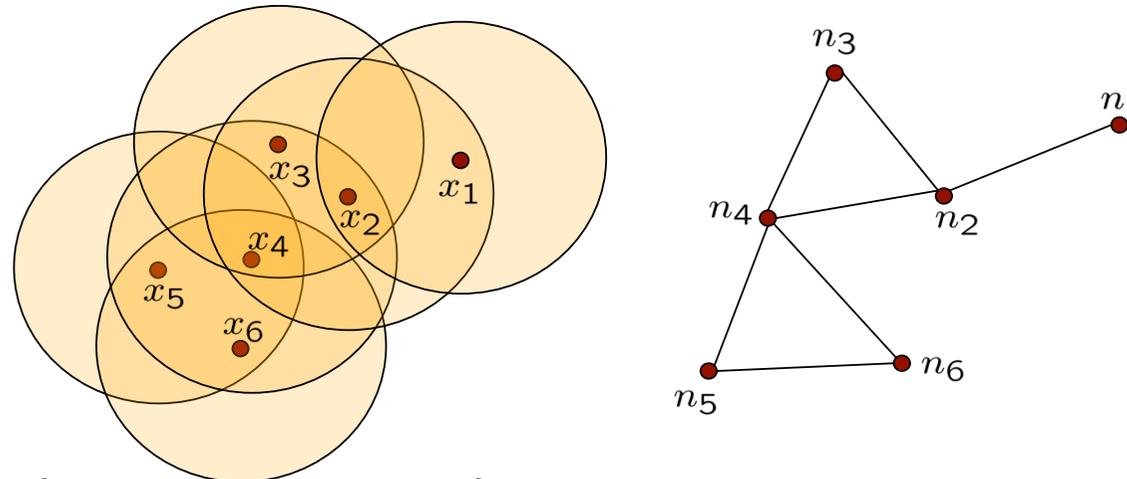
- A networked sensing and actuation system consists of
  - **NODES** - physical entities with limited resources (computation, communication, perception, control)
  - **EDGES** - virtual entities that encode the flow of information between the nodes



- The “right” mathematical object for characterizing such systems at the network-level is a **GRAPH**
  - Purely *combinatorial* object (no *geometry* or *dynamics*)
  - The characteristics of the information flow is abstracted away through the (possibly weighted and directed) edges

## Graphs as Network Abstractions

- The connection between the combinatorial graphs and the geometry of the system can for instance be made through geometrically defined edges.
- Examples of such proximity graphs include **disk-graphs**, **Delaunay graphs**, **visibility graphs**, and **Gabriel graphs** [1].



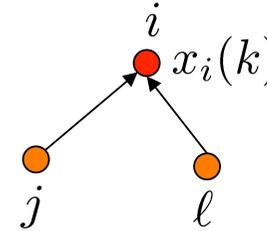
$$\mathcal{N} = \{n_1, n_2, n_3, n_4, n_5, n_6\}$$

$$\mathcal{E} = \{(n_1, n_2), (n_2, n_3), (n_3, n_4), (n_2, n_4), (n_4, n_5), (n_4, n_6), (n_5, n_6)\}$$



# The Basic Setup

- $x_i(k)$  = “state” at node  $i$  at time  $k$
- $N_i(k)$  = “neighbors” to agent  $i$



- Information “available to agent  $i$ ”

$$I_i^c(k) = \{x_j(k) \mid j \in N_i(k)\} \longleftarrow \text{common ref. frame (comms.)}$$

or

$$I_i^r(k) = \{x_i(k) - x_j(k) \mid j \in N_i(k)\} \longleftarrow \text{relative info. (sensing)}$$

- Update rule:

$$x_i(k+1) = F_i(x_i(k), I_i(k)) \longleftarrow \text{discrete time}$$

or

$$\dot{x}_i(t) = F_i(x_i(t), I_i(t)) \longleftarrow \text{continuous time}$$

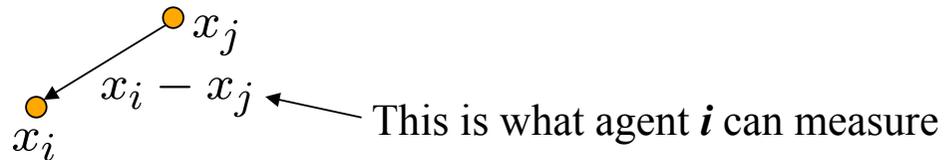
- *How pick the update rule?*



## Rendezvous – A Canonical Problem

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- Given a collection of mobile agents who can only measure the relative displacement of their neighbors (no global coordinates)



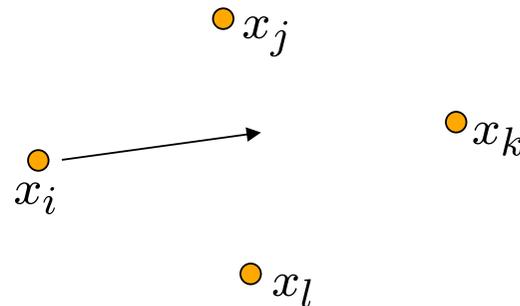
- Problem: Have all the agents meet at the same (unspecified) position
- If there are only two agents, it makes sense to have them drive towards each other, i.e.

$$\begin{aligned}\dot{x}_1 &= -\gamma_1(x_1 - x_2) \\ \dot{x}_2 &= -\gamma_2(x_2 - x_1)\end{aligned}$$

- If  $\gamma_1 = \gamma_2$  they should meet halfway

## Rendezvous – A Canonical Problem

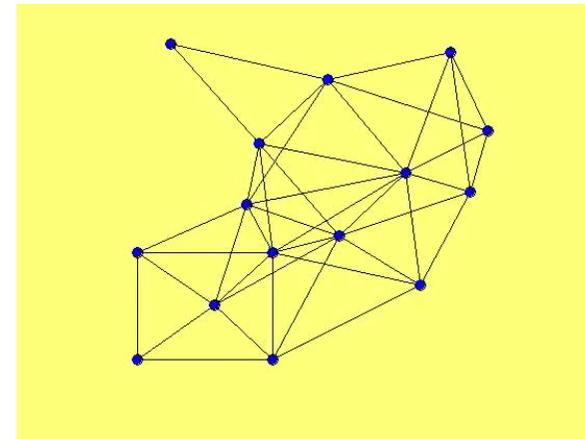
- If there are more than two agents, they should probably aim towards the centroid of their neighbors (or something similar)



$$\dot{x}_i = -\gamma \sum_{j \in \mathcal{N}_i} (x_i - x_j)$$

**Fact [2-4]:** As long as the graph is connected (iff), the *consensus equation* drives all agents to the same state value

$$\lim_{t \rightarrow \infty} x_i(t) = \bar{x} = \frac{1}{N} \sum_{j=1}^N x_j(0)$$





# Algebraic Graph Theory

- To show this, we need some tools...
- Algebraic graph theory provides a bridge between the combinatorial graph objects and their matrix representations

- **Degree matrix:**

$$D = \text{diag}(\text{deg}(n_1), \dots, \text{deg}(n_N))$$

- **Adjacency matrix:**

$$A = [a_{ij}], \quad a_{ij} = \begin{cases} 1 & \text{if } n_i \text{ --- } n_j \\ 0 & \text{o.w.} \end{cases}$$

- **Incidence matrix** (directed graphs):

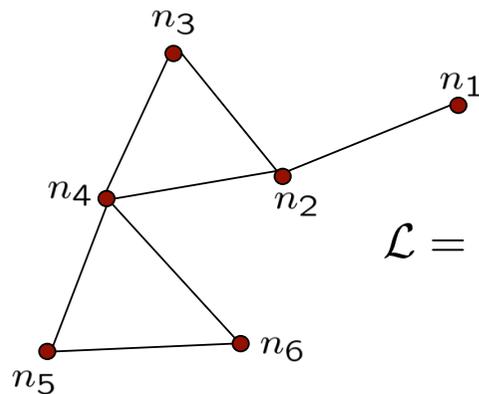
$$\mathcal{I} = [\iota_{ij}], \quad \iota_{ij} = \begin{cases} 1 & \text{if } n_i \xrightarrow{e_j} n_j \\ -1 & \text{if } n_i \xleftarrow{e_j} n_j \\ 0 & \text{o.w.} \end{cases}$$

- **Graph Laplacian:**

$$\mathcal{L} = D - A = \mathcal{I}\mathcal{I}^T$$



## Algebraic Graph Theory - Example



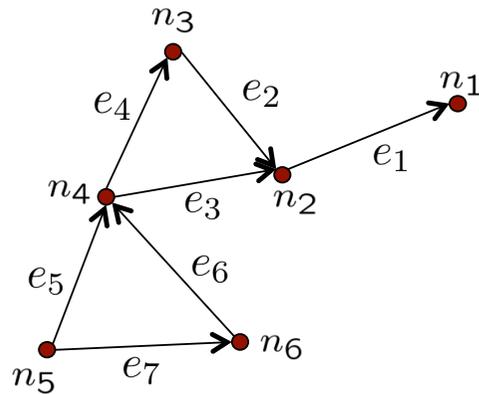
$$\mathcal{L} = D - A = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ -1 & 3 & -1 & -1 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & -1 & -1 & 4 & -1 & -1 \\ 0 & 0 & 0 & -1 & 2 & -1 \\ 0 & 0 & 0 & -1 & -1 & 2 \end{bmatrix}$$

$$D = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 4 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 2 \end{bmatrix}$$

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}$$



## Algebraic Graph Theory - Example



$$\mathcal{I} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ -1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & -1 & -1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & -1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 1 \end{bmatrix}$$

$$\mathcal{L} = \mathcal{I}\mathcal{I}^T = \begin{bmatrix} 1 & -1 & 0 & 0 & 0 & 0 \\ -1 & 3 & -1 & -1 & 0 & 0 \\ 0 & -1 & 2 & -1 & 0 & 0 \\ 0 & -1 & -1 & 4 & -1 & -1 \\ 0 & 0 & 0 & -1 & 2 & -1 \\ 0 & 0 & 0 & -1 & -1 & 2 \end{bmatrix}$$



## The Consensus Equation

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- The reason why the graph Laplacian is so important is through the already seen “consensus equation”

$$\dot{x}_i = - \sum_{j \in \mathcal{N}_i} (x_i - x_j), \quad i = 1, \dots, N$$

or equivalently (W.L.O.G. scalar agents)

$$\left. \begin{array}{l} \dot{x}_i = -\deg(n_i)x_i + \sum_{j=1}^N a_{ij}x_j \\ x = \begin{bmatrix} x_1 & x_2 & \cdots & x_N \end{bmatrix}^T \end{array} \right\} \Rightarrow \dot{x} = -\mathcal{L}x$$

- This is an autonomous LTI system whose convergence properties depend purely on the spectral properties of the Laplacian.



## Graph Laplacians: Useful Properties

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- It is orientation independent
- It is symmetric and positive semi-definite
- If the graph is *connected* then

$$\text{eig}(\mathcal{L}) = \{\lambda_1, \dots, \lambda_N\}, \text{ with } 0 = \lambda_1 < \lambda_2 \leq \dots \leq \lambda_N$$

$$\text{eigv}(\mathcal{L}) = \{\nu_1, \dots, \nu_N\}, \text{ with } \text{null}(\mathcal{L}) = \text{span}\{\nu_1\} = \text{span}\{\mathbf{1}\}$$

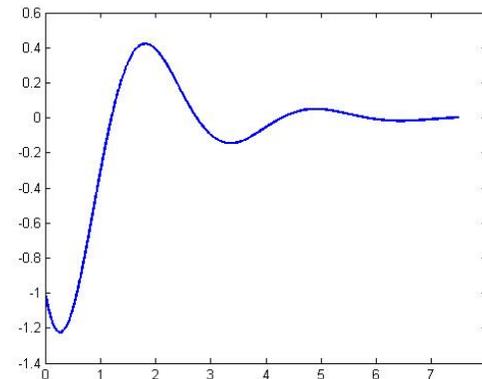
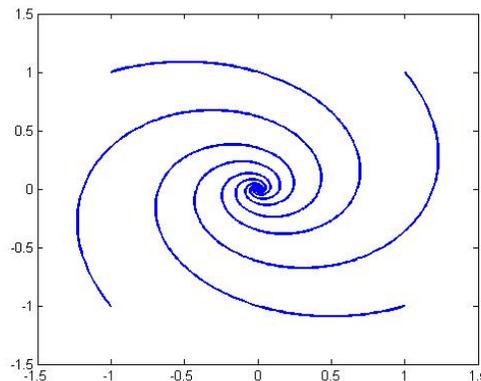


# Stability - Basics

- The stability properties (what happens as time goes to infinity?) of a linear, time-invariant system is completely determined by the eigenvalues of the system matrix

$$\dot{x} = Ax \quad (\dot{x} = -Lx)$$

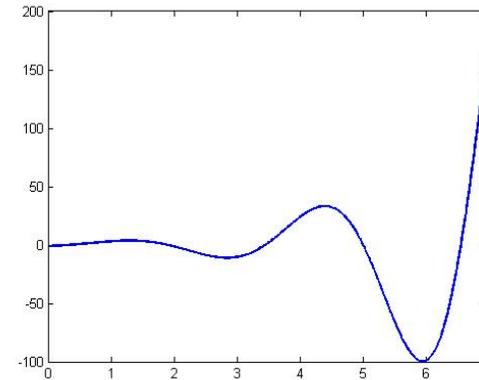
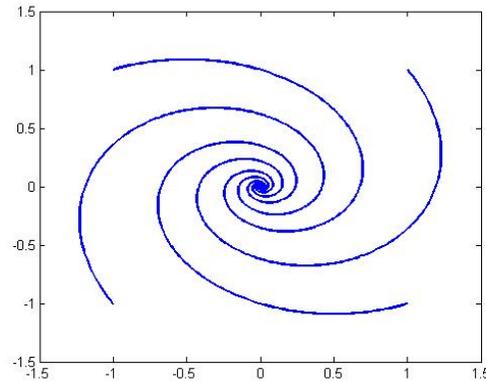
- Eigenvalues  $\lambda(A) = \lambda_1, \dots, \lambda_n$
- Asymptotic stability:  $\text{Re}(\lambda_i) < 0, i = 1, \dots, n \Rightarrow \lim_{t \rightarrow \infty} x(t) = 0$





# Stability - Basics

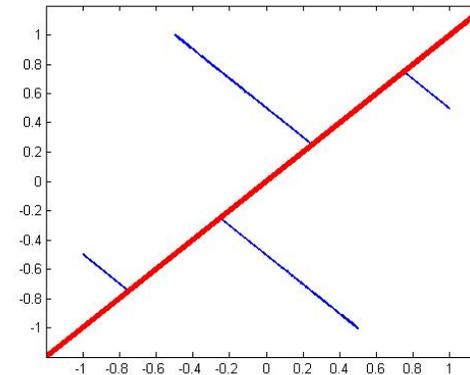
- Unstable:  $\exists i$  s.t.  $\text{Re}(\lambda_i) > 0$ ,  $\Rightarrow \lim_{t \rightarrow \infty} x(t) = \infty$



- Critically stable:

$$0 = \lambda_1 > \lambda_2 \geq \dots \geq \lambda_n, \Rightarrow \lim_{t \rightarrow \infty} x(t) \in \text{null}(A)$$

This is the case for the consensus equation





## Static Consensus

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- If the graph is static and connected, under the consensus equation, the states will reach  $\text{null}(L)$

- Fact (again):

$$\text{null}(L) = \text{span}\{\mathbf{1}\}, x \in \text{null}(L) \Leftrightarrow x = \begin{bmatrix} \alpha \\ \alpha \\ \vdots \\ \alpha \end{bmatrix}, \alpha \in \mathfrak{R}$$

- So all the agents state values will end up at the same value, i.e. the consensus/rendezvous problem is solved!

$$\dot{x}_i = - \sum_{j \in N_i} (x_i - x_j) \Rightarrow \lim_{t \rightarrow \infty} x_i(t) = \frac{1}{n} \sum_{j=1}^n x_j(0) = \frac{1}{n} \mathbf{1}^T x(0)$$



# Formation Control

- Being able to reach consensus goes beyond solving the rendezvous problem.
- Formation control:

$$\begin{array}{ccc} x_1, \dots, x_N & \longrightarrow & y_1, \dots, y_N \\ \text{agent positions} & & \text{target positions} \end{array}$$

- But, formation achieved if the agents are in any translated version of the targets, i.e.,

$$x_i = y_i + \tau, \quad \forall i, \text{ for some } \tau$$

- Enter the consensus equation [5]:

$$e_i = x_i - y_i$$

$$\dot{e}_i = - \sum_{j \in N_i} (e_i - e_j)$$

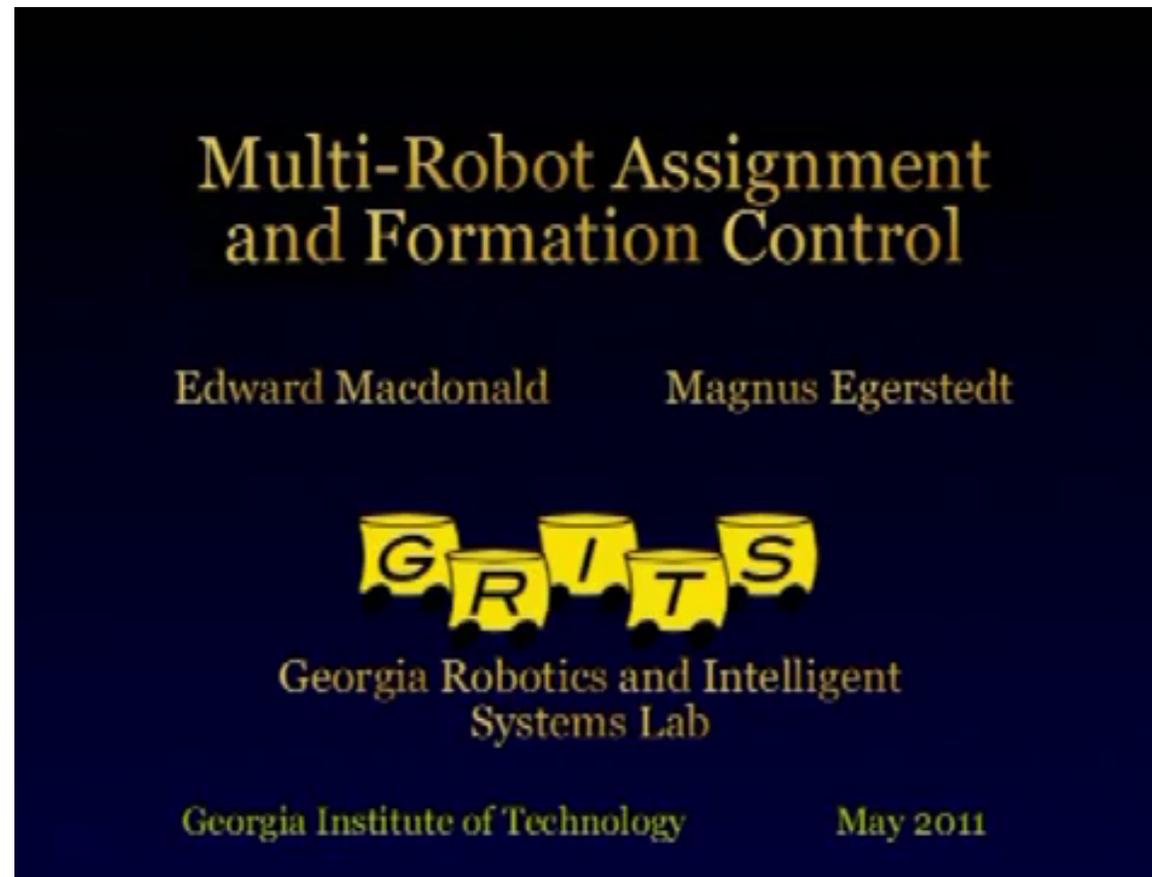
$$e_i(\infty) = e_j(\infty) = \tau$$

$$\begin{aligned} \dot{x}_i &= \sum_{j \in N_i} [(x_i - x_j) - (y_i - y_j)] \\ x_i(\infty) &= y_i + \tau, \quad \forall i \end{aligned}$$



# Formation Control

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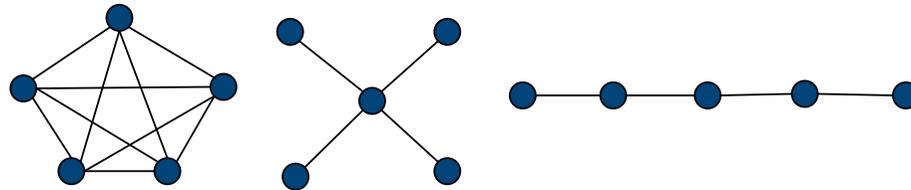
# Convergence Rates

- The second smallest eigenvalue of the graph Laplacian is really important!
- Algebraic Connectivity (= 0 if and only if graph is disconnected)
- Fiedler Value (robustness measure)
- **Convergence Rate:**

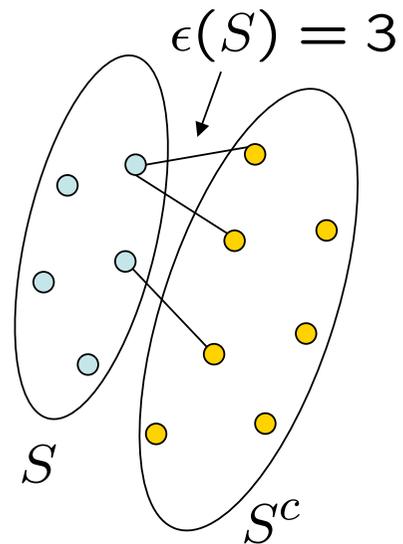
$$\|x(t) - \frac{1}{n} \mathbf{1}^T x(0)\| \leq C e^{-\lambda_2 t}$$

- **Punch-line:** The more connected the network is, the faster it converges (and the more information needs to be shuffled through the network)

- Complete graph:  $\lambda_2 = n$
- Star graph:  $\lambda_2 = 1$
- Path graph:  $\lambda_2 < 1$



## Cheeger's Inequality



$$\phi(S) = \frac{\epsilon(S)}{\min\{|S|, |S^c|\}}$$

(measures how many edges need to be cut to make the two subsets disconnected as compared to the number of nodes that are lost)

**isoperimetric number:**

$$\phi(G) = \min_S \phi(S)$$

(measures the robustness of the graph)

$$\phi(G) \geq \lambda_2 \geq \frac{\phi(G)^2}{2\Delta(G)}$$



# Beyond Static Consensus

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- So far, the consensus equation will drive the node states to the same value if the graph is static and connected.
- But, this is clearly not the case in a number of situations:
  - **Edges = communication links**
    - Random failures
    - Dependence on the position (shadowing,...)
    - Interference
    - Bandwidth issues
  - **Edges = sensing**
    - Range-limited sensors
    - Occlusions
    - Weirdly shaped sensing regions



# Summary I

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- Graphs are natural abstractions (combinatorics instead of geometry)
- Consensus problem (and equation)
- Static Graphs:
  - Undirected: Average consensus iff  $G$  is connected
- Need richer network models!