

Continuous-time Averaging Systems

AU7036: Introduction to Multi-agent Systems

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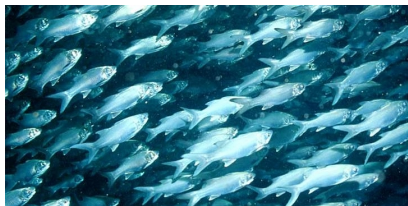
March 19, 2024

- Definitions of Laplacian matrices
- Properties of Laplacian matrices
 - ① Spectrum
 - ② Multiplicity of eigenvalue 0 v.s. graph properties
- Symmetric Laplacian and algebraic connectivity
 - Multiplicity of eigenvalue 0 and number of connected components
 - Algebraic connectivity (Fiedler eigenvalue and Fiedler eigenvector)
- Laplacian systems and pseudoinverses

- 1 Example systems
- 2 Continuous-time linear systems and their convergence properties
- 3 The Laplacian flow
- 4 Design of weight-balanced digraphs

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Example #1: Flocking dynamics



Flocking dynamics: a simple alignment rule

- Each animal steers towards the average heading of its neighbors

$$\dot{\theta}_i = \begin{cases} (\theta_j - \theta_i), & \text{if one neighbor} \\ \frac{1}{2}(\theta_{j_1} - \theta_i) + \frac{1}{2}(\theta_{j_2} - \theta_i), & \text{if two neighbors} \\ \frac{1}{m}(\theta_{j_1} - \theta_i) + \cdots + \frac{1}{m}(\theta_{j_m} - \theta_i), & \text{if } m \text{ neighbors} \end{cases}$$

$= \text{average}(\{\theta_j, \text{ for all neighbors } j\}) - \theta_i$

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$$= \text{average}(\{\theta_j, \text{ for all neighbors } j\}) - \theta_i$$

In matrix form

$$\dot{\theta} = A\theta - \theta = (A - I)\theta = -L\theta$$

Example #2: Continuous-time opinion dynamics

- Discrete-time opinion dynamics

$$x(k+1) = Ax(k)$$

- A continuous-time counterpart

$$\begin{aligned}x_i(k+1) &= \sum_{j=1}^n a_{ij}x_j \\&= a_{ii}x_i(k) + \sum_{j \neq i} a_{ij}x_j(k) \\&= (1 - \sum_{j \neq i} a_{ij})x_i(k) + \sum_{j \neq i} a_{ij}x_j(k) \\&= x_i(k) + \sum_{j \neq i} a_{ij}(x_j(k) - x_i(k))\end{aligned}$$

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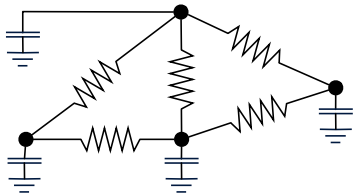
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- If the edge weights a_{ij} are of the form $a_{ij} = \bar{a}_{ij}\tau$

$$\dot{x}(t) = -\bar{L}x(t)$$

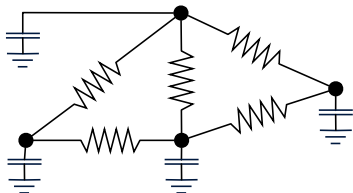
Example #3: A simple RC circuit



- The injected currents at nodes satisfy

$$C_{\text{injected}} = L v,$$

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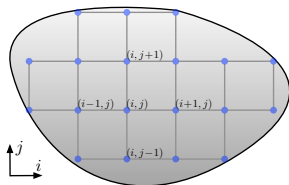
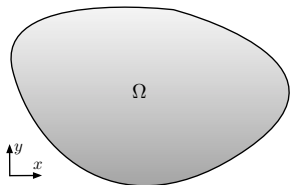
- With capacitors at node i

$$C_i \frac{d}{dt} v_i = -C_{\text{injected at } i}$$

- The overall dynamics

$$\frac{d}{dt} v = -C^{-1} L v$$

Example #4: Discretization of partial differential equations



- Let $u(t, x, y)$ be temperature at $(x, y) \in \Omega$ at time t , heat equation

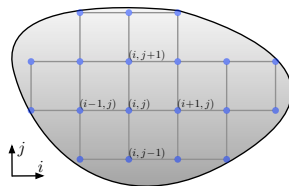
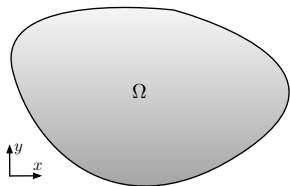
$$\frac{\partial u}{\partial t} = c \Delta u,$$

where

- 1 c is thermal diffusivity
- 2 Δ is Laplacian differential operator

$$\Delta u(t, x, y) = \frac{\partial^2 u}{\partial x^2}(t, x, y) + \frac{\partial^2 u}{\partial y^2}(t, x, y)$$

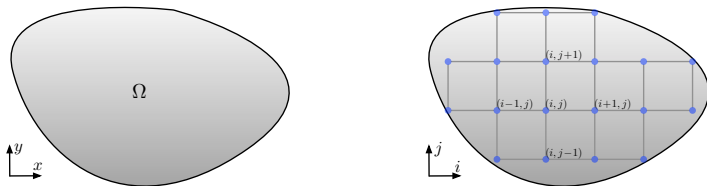
Example #4: Discretization of partial differential equations



- Finite difference approximation for Laplacian operator

$$\Delta u(t, x_i, y_j) \approx \frac{u(t, x_{i-1}, y_j) + u(t, x_{i+1}, y_j) + u(t, x_i, y_{j-1}) + u(t, x_i, y_{j+1}) - 4u(t, x_i, y_j)}{h^2}$$

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- Heat equation can be approximated by

$$\frac{d}{dt} u_{\text{discrete}} = -\frac{c}{h^2} L u_{\text{discrete}}$$

Today

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- 2 Continuous-time linear systems and their convergence properties
- 3 The Laplacian flow
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Continuous-time linear systems

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Semi-convergence and convergence

A matrix $A \in \mathbb{R}^{n \times n}$ is

- 1 continuous-time **semi-convergent** if $\lim_{t \rightarrow +\infty} \exp(At)$ exists;
- 2 **Hurwitz** or continuous-time **convergent** if $\lim_{t \rightarrow +\infty} \exp(At) = \mathbb{0}_{n \times n}$

$$\dot{x}(t) = Ax(t)$$

- The **spectral abscissa** of A is maximum of real parts of eigenvalues:

$$\alpha(A) = \max\{\Re(\lambda) \mid \lambda \in \text{spec}(A)\}$$

Convergence and spectral abscissa

For a square matrix A , the following statements hold:

- 1 A is continuous-time convergent (Hurwitz) if and only if $\alpha(A) < 0$,

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Convergence and spectral abscissa

For a square matrix A , the following statements hold:

- 1 A is continuous-time convergent (Hurwitz) if and only if $\alpha(A) < 0$,
- 2 A is continuous-time semi-convergent and not convergent iff
 - a 0 is a semisimple eigenvalue;
 - b all other eigenvalues have negative real part.

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Let G be a weighted digraph with n nodes and Laplacian matrix L

- The Laplacian flow on \mathbb{R}^n is the dynamical system

$$\dot{x} = -Lx$$

- In components

$$\dot{x}_i = \sum_{j=1}^n a_{ij}(x_j - x_i) = \sum_{j \in \mathcal{N}^{\text{out}}(i)} a_{ij}(x_j - x_i)$$

Matrix exponential of a Laplacian matrix

$$\dot{x} = -Lx$$

- The solution to the Laplacian flow is given by

$$x(t) = \exp(-Lt)x(0)$$

The matrix exponential of a Laplacian matrix

Let $L \in \mathbb{R}^{n \times n}$ be a Laplacian matrix with associated weighted digraph G and with maximum diagonal entry $l_{\max} = \max\{l_{11}, \dots, l_{nn}\}$. Then

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- 5 $\exp(-L) > 0$, iff G strongly connected (L irreducible).

$$\dot{x} = -Lx$$

Equilibria

If G contains a globally reachable node, then the set of equilibria is

$$\text{span}\{\mathbf{1}_n\} = \{\beta\mathbf{1}_n \mid \beta \in \mathbb{R}\}$$

Equilibria and convergence of the Laplacian flow

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Consensus conditions for $x(k+1) = Ax(k)$

Let A be a row-stochastic matrix and let G be its associated digraph. The following statements are equivalent:

- (A1) the eigenvalue 1 is simple and all other eigenvalues μ satisfy $|\mu| < 1$;
- (A2) A is semi-convergent and $\lim_{k \rightarrow \infty} A^k = \mathbf{1}_n w^\top$, where $w \geq 0$, $w^\top A = w^\top$ and $w^\top \mathbf{1}_n = 1$;
- (A3) G contains a globally reachable node and the subgraph of globally reachable nodes is aperiodic.

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Consensus for Laplacian matrices with a globally reachable node

Let L be a Laplacian matrix and let G be its associated digraph. If any of (A1)-(A3) holds, then

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- 3 if additionally G is weight-balanced, then G is strongly connected, $\mathbb{1}_n^\top L = 0_n^\top$, $w = \frac{1}{n} \mathbb{1}_n$, and

$$\lim_{t \rightarrow \infty} x(t) = \frac{\mathbb{1}_n^\top x(0)}{n} \mathbb{1}_n = \text{average}(x(0)) \mathbb{1}_n.$$

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Design of weight-balanced digraphs

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- Let $l_{\max} = \max\{l_{11}, \dots, l_{nn}\}$, w be left dominant eigenvector of L
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- Doubly stochastic \bar{A}

$$\bar{A} = I_n - \bar{L}$$

Upcoming

Week 1-6:

- Introduction
- Elements of matrix theory
- Elements of graph theory
- Elements of algebraic graph theory
- Discrete-time averaging systems
- The Laplacian matrix
- Continuous-time averaging systems
- Diffusively-coupled linear systems
- (*) The incidence matrix and its applications
- (*) Metzler matrices and dynamical flow systems

Week 7-14:

- Lyapunov stability theory
- Nonlinear averaging systems (Euler-Lagrangian, oscillators)
- Other advanced topics

Week 15-16:

- Project presentation