Lecture: Introduction to Graph Theory

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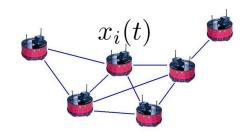


Review Distributed Consensus

✓ Objective: $x_1(t) = x_2(t) = \cdots = x_m(t) = x^*$

✓ Update:
$$x_i(t+1) = \sum_{j \in \mathcal{N}_i} w_{ij} x_j(t)$$

 w_{ij} : the weight assigned by agent \emph{i} to agent \emph{j}



agent's dynamics $x_i(t+1) = u_i$

distributed control: $u_i = f_i(x_j(t), j \in \mathcal{N}_i)$

Consensus Goals	Choices of Weights
x^st is an unknown constant	$w_{ij} = \begin{cases} > 0, \ j \in \mathcal{N}_i & \sum_{j=1}^m w_{ij} = 1 \\ 0, \ \text{otherwise.} \end{cases}$
x^* is the global averag $\sum_{i=1}^{n} \sum_{i=1}^{m} x_i(0)$	$w_{ij} = \begin{cases} \min\{\frac{1}{d_i}, \frac{1}{d_j}\} & j \in \mathcal{N}_i, \ j \neq i; \\ 1 - \sum_{j \in \mathcal{N}_i, \ j \neq i} w_{ij} & j = i \\ 0, \text{ otherwise.} \end{cases}$
x^* is a specific convex combination $\sum_{i=1}^m \gamma_i x_i(0)$	

$x_i(t)$ **Distributed Algorithm for Consensus** $x_i(t+1) = \sum w_{ij} x_j(t)$ $j \in \mathcal{N}_i$ $w_{ij} = \begin{cases} > 0, \ j \in \mathcal{N}_i \\ 0, \ \text{otherwise.} \end{cases} \quad \sum_{j=1}^m w_{ij} = 1$ **Network** connectivity x(t+1) = Ax(t)**Graph Theory** A is row If A is also stochastic **Primitive Perron - Frobenius** Gershgorin **Theorem Circle Theorem** 1 is a simple all the other eigenvalues are with 1 is the largest eigenvalue magnitude strictly less than 1 eigenvalue in magnitude. $\lim A^t = \mathbf{1}w'$ $t\rightarrow\infty$ $\lim_{t \to \infty} x(t) = \mathbf{1}w'x(0)$

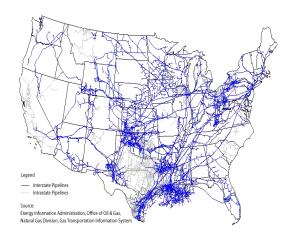
What is a **graph?** Graph: A set of vertices(nodes) connected by a group of edges.

Why Graphs? Graphs can model many types of **relations** in many real-world systems.

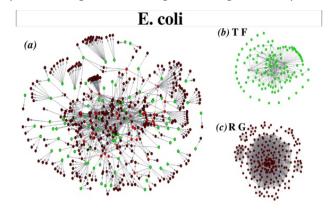
Social Networks (Nodes: People; Edges: Friendship)



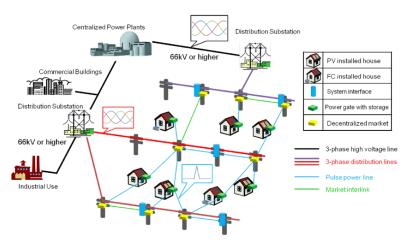
Transportation Networks (Nodes: cities; Edges: transportation flow)



Biological Networks (Nodes: genes; Edges: Regulation)



Power Networks (Nodes: cites; Edges: power flow)



Graph and Its Elements

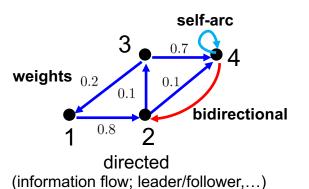


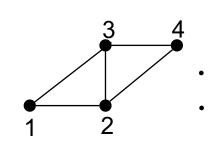
Graph: A set of vertices (nodes) connected by a group of edges.



 $\mathcal{V} = \{1, 2, 3, 4\}$

$$\mathcal{E} = \{(1, 2), (2, 3), (3, 1), (2, 4), (3, 4)\}$$





Path: (1,2),(2,3),(3,4) of length 3.

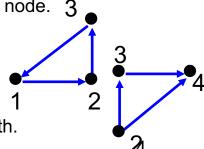
Cycle: (1,2),(2,3),(3,1) of length 3. (4,4) of length 1.

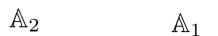
undirected (distance, bidirectional communications,...)

- Distance between i and j: the length of the shortest path connecting them. Distance between 1 and 4=2
- **Diameter:** the largest one among all distances.
- In-neighbors/out-neighbors of vertex i: For a directed edge from 1 to 2
- 1 is an in-neighbor of 2; 2 is an out-neighbor of 1.
- In-degree/out-degree of vertex i: the number of incoming/outgoing edges.
- (degree for undirected graphs. Maximum degree)
- A directed graph is **strongly connected** if there is a path from any node to any other node. 3
- A directed graph is **weakly connected** if its undirected version is connected.

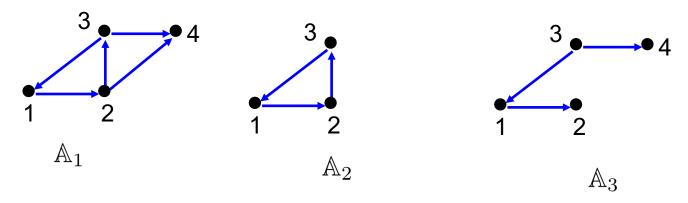
 Connected for undirected graphs.
- A globally reachable node if this node can be reached from any other node by a path.

Every node is globally reachable=strongly connected.

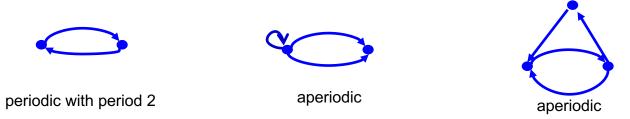




- A graph $\mathbb{G}_1=\{\mathcal{V}_1,\mathcal{E}_1\}$ is a subgraph of $\mathbb{G}_2=\{\mathcal{V}_2,\mathcal{E}_2\}$ if $\mathcal{V}_1\subset\mathcal{V}_2,\mathcal{E}_1\subset\mathcal{E}_2$
- A graph $\mathbb{G}_1=\{\mathcal{V}_1,\mathcal{E}_1\}$ is a spanning subgraph of $\mathbb{G}_2=\{\mathcal{V}_2,\mathcal{E}_2\}$ if $\mathcal{V}_1=\mathcal{V}_2,\mathcal{E}_1\subset\mathcal{E}_2$



- Period of graphs: For a strongly connected directed graph, let k denote the greatest common divisor of lengths of all its cycles.
- A strongly connected directed graph is periodic if k>1; is aperiodic, otherwise.



Any strongly connected directed graph with a self-loop is aperiodic.

Graph Operations

Given two graphs with the same vertex set $\mathbb{G}_1 = \{\mathcal{V}, \mathcal{E}_1\}$

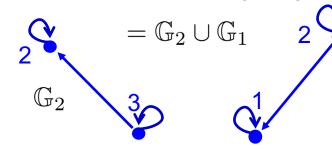
$$\mathbb{G}_1 = \{\mathcal{V}, \mathcal{E}_1\}$$

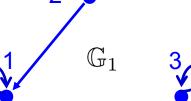
$$\mathbb{G}_2 = \{\mathcal{V}, \mathcal{E}_2\}$$

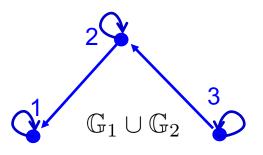
Graph Union: $\mathbb{G}_1 \cup \mathbb{G}_2 = \{\mathcal{V}, \mathcal{E}\}$

$$\mathcal{E}$$
} $\mathcal{E} = \mathcal{E}_1 \cup \mathcal{E}_2$

Union: direct impacts







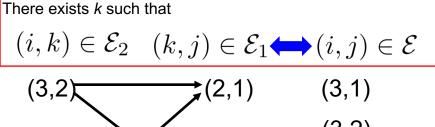
$$\mathcal{E}_1 = \{(3,2), (1,1), (2,2), (3,3)\}$$

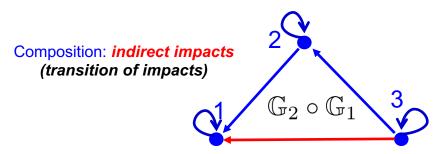
(1,1)

$$\mathcal{E}_2 = \{(2,1), (1,1), (2,2), (3,3)\}$$

$$\mathcal{E} = \{(3,2), (2,1), (1,1), (2,2), (3,3)\}$$

Graph Composition: $\mathbb{G}_2 \circ \mathbb{G}_1 = \{\mathcal{V}, \mathcal{E}\}$





(3,2)If all graphs are with **self-arcs** at every vertex, then (1,1)

 $\mathbb{G}_1 \cup \mathbb{G}_2 \subset \mathbb{G}_2 \circ \mathbb{G}_1$

Proof?

(2,2)☐ Composition for graphs with self-arcs include both direct and indirect impacts. (3,3)

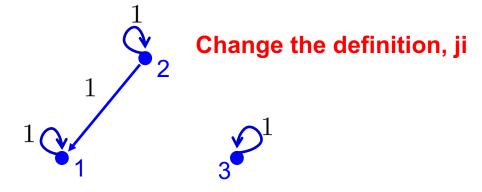
(2,1)

Graph Representation for Non-Negative Matrices

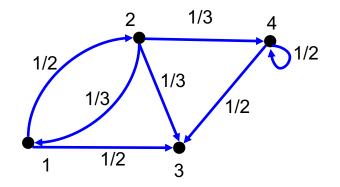
Given an non-negative matrix $A \in \mathbb{R}^{n \times n}$,

the *graph of a matrix* A *is* a directed graph of \mathbf{n} nodes such that there exists a directed **edge** $i \to j$ with the weight A_{ij} if and only if $A_{ij} > 0$.

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



$$A = \begin{bmatrix} 0 & 1/2 & 1/2 & 0 \\ 1/3 & 0 & 1/3 & 1/3 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1/2 & 1/2 \end{bmatrix}$$



Correspondingly, the matrix A is called the adjacency matrix of the graph.

Adjacency Matrix of a Graph

Algebraic Graph Theory applying algebraic methods in solving graph problems.

$$\mathbb{G} = \{\mathcal{V}, \mathcal{E}\} \qquad \mathcal{V} = \{1, 2, 3, 4\}$$

$$\mathcal{E} = \{(1, 2), (1, 3), (2, 1), (2, 3), (2, 4), (4, 3), (4, 4)\}$$

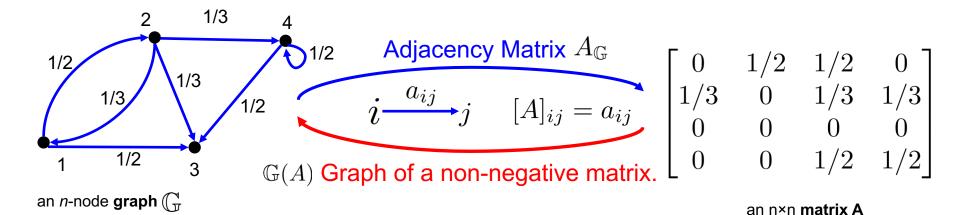
➤ Adjacency Matrix
$$A = [a_{ij}]_{n \times n}$$

$$a_{ij} = \left\{ \begin{array}{ll} w_{ij}, & i \to j; \\ 0, & \text{otherwise} \end{array} \right. A = \left[\begin{array}{lll} 0 & 1/2 & 1/2 & 0 \\ 1/3 & 0 & 1/3 & 1/3 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1/2 & 1/2 \end{array} \right] \quad \text{Matrix of a graph.}$$

Adjacency matrix of an unweighted graph $A = [a_{ij}]_{n \times n}$

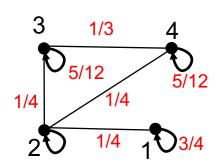
$$a_{ij} = \begin{cases} 1, & i \to j; \\ 0, & \text{otherwise} \end{cases}$$
 $A = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$

Adjacency matrix of an undirected/unweighted graph? Symmetric



Example: Distributed Averaging using Metropolis Weights

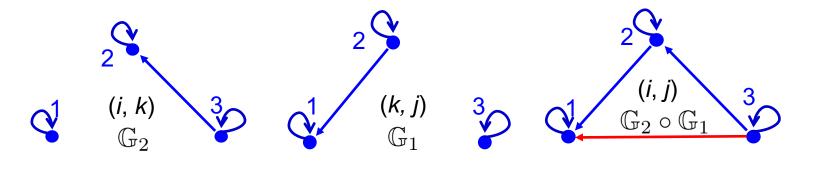
$$w_{ij} = w_{ji} \quad x(t+1) = Nx(t)$$



- Write out the weights for each edge
- Write out the weights for each edge $N = \begin{bmatrix} 3/4 & 1/4 & 0 & 0 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 0 & 1/4 & 5/12 & 1/3 \\ 0 & 1/4 & 1/3 & 5/12 \end{bmatrix}$

Graph composition $\mathbb{G}_2 \circ \mathbb{G}_1$

of graphs of the same vertex set, with **self-arcs**.



$$M_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \qquad M_1 = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \qquad M_2 M_1 = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

$$M_2 M_1 = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{bmatrix}$$

$$\mathbb{G}(M_2) \circ \mathbb{G}(M_1) = \mathbb{G}(M_2 M_1)$$

Graph Composition

Matrix Product

Seminar 2 for Extra Credit

Title: "Visual-Inertial State Estimation and Perception for Autonomous Vehicles"

Time: 9:30am on Thursday (Feb. 20), 2020

Location: ARMS B071

Speaker: Guoquan Huang (Assistant Professor, University of Delaware)

- Correspondingly, the class on Wednesday (Feb. 19) will be canceled and replaced by this seminar video.
- Please start to work on the problem formulation of your course projects. Feel free to email me or stop by my office during office hours for discussion.