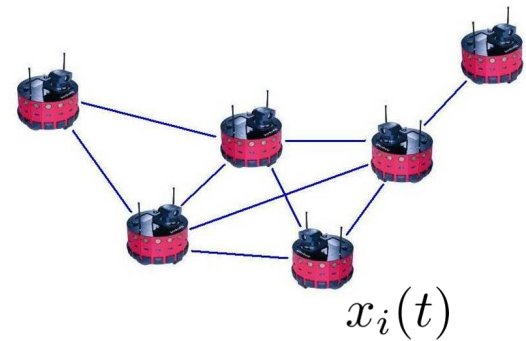


# Lecture: Distributed Algorithms for General Consensus

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# Review of Consensus Algorithms



## Consensus:

just **a** constant

- Objective:  $x_1(t) = x_2(t) = \dots = x_m(t) = x^*$

- Update:

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

$$w_{ij} = \begin{cases} > 0, & j \in \mathcal{N}_i \\ 0, & \text{otherwise.} \end{cases} \quad \sum_{j=1}^m w_{ij} = 1$$

Local Weighted Average  
(convex combination)

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

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

distributed

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

- Compact Form:  $x(t+1) = Ax(t)$

$$A \in \mathbb{R}^{m \times m} \text{ with entries: } A_{ij} = \begin{cases} w_{ij} & \text{if } j \in \mathcal{N}_i \\ 0 & \text{otherwise} \end{cases}$$

$$A\mathbf{1} = \mathbf{1} \quad \mathbf{1} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

- Analysis:

$$q \in \mathbb{R}^m \text{ (unknown)}$$

$$A^t \rightarrow \mathbf{1} q' \rightarrow \text{Consensus is reached.}$$

$$x(t) \rightarrow \mathbf{1} \underbrace{q' x(0)}_{x^*} = \begin{bmatrix} x^* \\ x^* \\ \vdots \\ x^* \end{bmatrix}$$

## Consensus to the Global Average

- Objective:**  $x_1(t) = x_2(t) = \dots = x_m(t) = x^*$ 

$$x^* = \frac{1}{m} \sum_{i=1}^m x_i(0) = \frac{1}{m} \mathbf{1}' x(0)$$

$$\mathbf{1} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

consensus                      to a specific value (the global average)
- Distributed Update 1:** *Local weighted average*                      Metropolis Weights

$$w_{ij} = \begin{cases} > 0, & j \in \mathcal{N}_i \\ 0, & \text{otherwise.} \end{cases} \quad \sum_{j=1}^m w_{ij} = 1$$

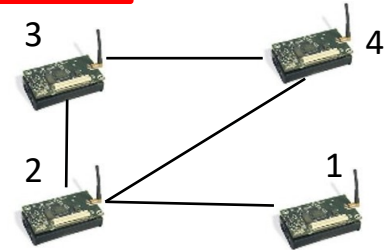
$$x_i(t+1) = \sum_{j \in \mathcal{N}_i} w_{ij} x_j(t)$$

$$w_{ij} = \begin{cases} \min\{\frac{1}{d_i}, \frac{1}{d_j}\} & j \in \mathcal{N}_i, j \neq i; \\ 1 - \sum_{j \in \mathcal{N}_i, j \neq i} w_{ij} & j = i \end{cases}$$

- Distributed Update 2:** Periodic Gossiping.

(1,2), (3,4), (2,3), (2,4)    (1,2), (3,4), (2,3), (2,4), ....

$$x(4) = M_{24} M_{23} M_{34} M_{12} x(0) \quad x(4(t+1)) = A x(4t)$$



- Analysis:**  $A^t \rightarrow \frac{1}{m} \mathbf{1} \mathbf{1}'$  ➡ Consensus to the global average is reached.
- $x^*$

$$x(t) \rightarrow \mathbf{1} \left( \frac{1}{m} \mathbf{1}' x(0) \right) = \begin{bmatrix} x^* \\ x^* \\ \vdots \\ x^* \end{bmatrix}$$

## Consensus to a more general value

consensus

to a specific  
convex combination

$$\gamma_i > 0, \sum_{i=1}^m \gamma_i = 1$$

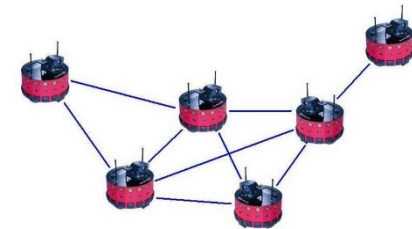
- Objective:**  $x_1(t) = x_2(t) = \dots = x_m(t) = x^*$   $x^* = \sum_{i=1}^m \gamma_i x_i(0) = \gamma' x(0)$

Why do we care about such specific convex combination?

$$\gamma = \begin{bmatrix} \gamma_1 \\ \gamma_2 \\ \vdots \\ \gamma_m \end{bmatrix}, \mathbf{1} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

When we fuse data in a large sensor network, accurate (inaccurate) measurements should be with higher (lower) weights.

The global average  $\frac{1}{m} \sum_{i=1}^m x_i(0)$  give equal weights to all sensors' measurements. (works for heterogeneous networks)



Let each sensor  $i$  be additionally assigned with a parameter  $\gamma_i$

(The larger  $\gamma_i$  is, the more accurate the sensor is)

- For consensus, we need

$$x_i(t+1) = \sum_{j \in \mathcal{N}_i} w_{ij} x_j(t) \quad w_{ij} = \begin{cases} > 0, & j \in \mathcal{N}_i \\ 0, & \text{otherwise.} \end{cases} \quad \sum_{j=1}^m w_{ij} = 1$$

- For consensus to the global average, we **further** need Metropolis weights:

$$w_{ij} = \begin{cases} \min\{\frac{1}{d_i}, \frac{1}{d_j}\} & j \in \mathcal{N}_i, j \neq i; \\ 1 - \sum_{j \in \mathcal{N}_i, j \neq i} w_{ij} & j = i \end{cases}$$

- How to achieve  $\sum_{i=1}^m \gamma_i x_i(0)$  ? **Any ideas?**

How to choose  $w_{ij}$  to achieve a specific convex combination  $\gamma'x(0)$ ?

- **A Distributed Update to achieve  $\gamma'x(0)$  :**

Consensus: 
$$x_i(t+1) = \sum_{j \in \mathcal{N}_i} w_{ij} x_j(t) \quad \sum_{j \in \mathcal{N}_i} w_{ij} = 1 \quad w_{ij} > 0, j \in \mathcal{N}_i$$

Furthermore:  
Corless' weights

$$w_{ij} = \begin{cases} \frac{1}{\gamma_i} \min\left\{\frac{\gamma_i}{d_i}, \frac{\gamma_j}{d_j}\right\} & j \in \mathcal{N}_i, j \neq i; \\ 1 - \sum_{j \in \mathcal{N}_i, j \neq i} w_{ij} & j = i \end{cases} \quad \text{bi-directional Networks} \quad d_i = |\mathcal{N}_i|$$

❖ L. Coduti, M. Corless. *A decentralized algorithm for assigning the weight parameters in a general synchronous consensus network. IEEE Conference on Decision and Control. 2012.*

$$x_1(t+1) = \left(1 - \frac{0.05}{0.44}\right)x_1(t) + \frac{1}{0.44} \cdot \min\{0.05, 0.22\} \cdot x_2(t)$$

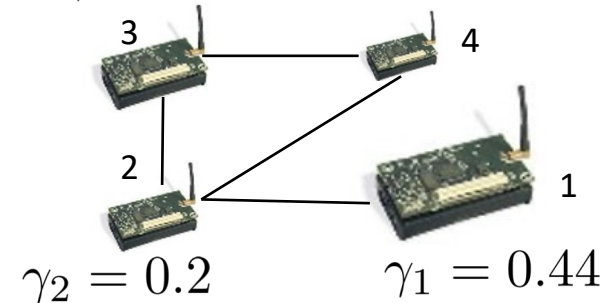
$$x_2(t+1) = \frac{1}{0.2} \cdot 0.05 \cdot x_1(t) + \left(1 - \frac{0.14}{0.2}\right)x_2(t) + \frac{1}{0.2} \cdot 0.05 \cdot x_3(t) + \frac{1}{0.2} \cdot 0.04 \cdot x_4(t)$$

$$x_3(t+1) = \frac{1}{0.24} \cdot 0.05 \cdot x_2(t) + \left(1 - \frac{1}{0.24} \cdot 0.09\right)x_3(t) + \frac{1}{0.24} \cdot 0.04 \cdot x_4(t)$$

$$x_4(t+1) = \frac{1}{0.12} \cdot 0.04 \cdot x_2(t) + \frac{1}{0.12} \cdot 0.04 \cdot x_3(t) + \left(1 - \frac{0.08}{0.12}\right)x_4(t)$$

$$\frac{\gamma_3}{d_3} = \frac{0.24}{3} = 0.08 \quad \frac{\gamma_4}{d_4} = \frac{0.12}{3} = 0.04$$

$$\gamma_3 = 0.24 \quad \gamma_4 = 0.12$$



$$\gamma_2 = 0.2$$

$$\gamma_1 = 0.44$$

$$\frac{\gamma_2}{d_2} = \frac{0.2}{4} = 0.05$$

$$\frac{\gamma_1}{d_1} = \frac{0.44}{2} = 0.22$$

- **Compact Form:**

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

$$A = \begin{bmatrix} \frac{0.39}{0.44} & \frac{0.05}{0.44} & 0 & 0 \\ \frac{0.05}{0.2} & \frac{0.06}{0.2} & \frac{0.05}{0.2} & \frac{0.04}{0.2} \\ 0 & \frac{0.05}{0.24} & \frac{0.15}{0.24} & \frac{0.04}{0.24} \\ 0 & \frac{0.04}{0.12} & \frac{0.04}{0.12} & \frac{0.04}{0.12} \end{bmatrix}$$

Observations on A

- All entries are non-negative.
- Row sum is 1.
- $\gamma' A = \gamma'$

$$A \mathbf{1} = \mathbf{1} \quad \mathbf{1} = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

$$\gamma = \begin{bmatrix} 0.44 \\ 0.2 \\ 0.24 \\ 0.12 \end{bmatrix}$$

- **Analysis:**

$$A^t \rightarrow \mathbf{1} \gamma'$$

Consensus to a specific convex combination  $\sum_{i=1}^m \gamma_i x_i(0)$  is reached.

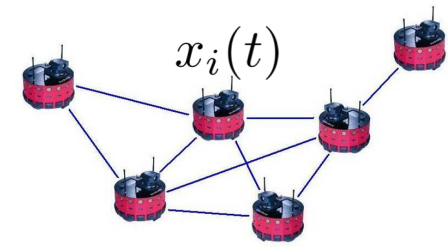
$$x(t) \rightarrow \mathbf{1} \underbrace{\gamma' x(0)}_{x^*} = \begin{bmatrix} x^* \\ x^* \\ \vdots \\ x^* \end{bmatrix}$$

## Summary of Distributed Consensus

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

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

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



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

✓ **Analysis:**  $x(t+1) = Ax(t)$   $A \in \mathbb{R}^{m \times m}$  with entries  $A_{ij} = w_{ij}$

$$x(t) \rightarrow A^t x(0)$$

# Stochastic Matrices

Consensus

$$A_1 = \begin{bmatrix} 1/2 & 1/2 & 0 & 0 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 0 & 1/3 & 1/3 & 1/2 \\ 0 & 1/3 & 1/3 & 1/3 \end{bmatrix}$$

Consensus for global average

$$A_2 = \begin{bmatrix} 3/4 & 1/4 & 0 & 0 \\ 1/4 & 1/4 & 1/4 & 1/4 \\ 0 & 1/4 & 5/12 & 1/3 \\ 0 & 1/4 & 1/3 & 5/12 \end{bmatrix}$$

Consensus for convex combination

$$A = \begin{bmatrix} \frac{0.39}{0.44} & \frac{0.05}{0.44} & 0 & 0 \\ \frac{0.05}{0.2} & \frac{0.06}{0.2} & \frac{0.05}{0.2} & \frac{0.04}{0.2} \\ 0 & \frac{0.05}{0.24} & \frac{0.15}{0.24} & \frac{0.04}{0.24} \\ 0 & \frac{0.04}{0.12} & \frac{0.04}{0.12} & \frac{0.04}{0.12} \end{bmatrix}$$

What do you observe something in common among these matrices??

➤ All entries are **non-negative**

➤ All matrices are **row stochastic**.  $A\mathbf{1} = \mathbf{1}$

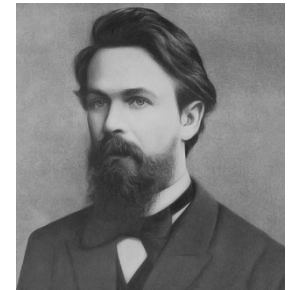
- A row stochastic matrix  $P$  describes a **Markov chain** with the  $ij$ th entry as the transition probability, namely, probability of moving from state  $i$  to  $j$  in one-time step.

$$P = \begin{bmatrix} P_{1,1} & P_{1,2} & \dots & P_{1,j} & \dots & P_{1,s} \\ P_{2,1} & P_{2,2} & \dots & P_{2,j} & \dots & P_{2,s} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ P_{i,1} & P_{i,2} & \dots & P_{i,j} & \dots & P_{i,s} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ P_{s,1} & P_{s,2} & \dots & P_{s,j} & \dots & P_{s,s} \end{bmatrix}.$$

$$\sum_{j=1}^n P_{ij} = 1$$

**Long-time probability distribution**

Stationary probability vector is a row vector such that  $\pi P = \pi$

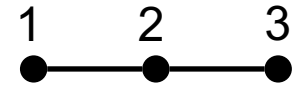


Andrey Markov

➤  $A_2$  is **doubly stochastic**.  $A\mathbf{1} = \mathbf{1}$ ,  $\mathbf{1}'A = \mathbf{1}'$



## Summary by Exercise



Write a distributed update for:

➤ Consensus:

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

➤ Consensus to global average:

- Distributed Update with Metropolis Weights:

$$w_{ij} = \begin{cases} \min\{\frac{1}{d_i}, \frac{1}{d_j}\} & j \in \mathcal{N}_i, j \neq i; \\ 1 - \sum_{j \in \mathcal{N}_i, j \neq i} w_{ij} & j = i \\ 0, & \text{otherwise.} \end{cases}$$

- Periodic Gossiping:

➤ Consensus to  $\gamma'x(0)$       $\gamma = \begin{bmatrix} 1/2 \\ 1/4 \\ 1/4 \end{bmatrix}$

- Distributed Update with Metropolis Weights:

$$w_{ij} = \begin{cases} \frac{1}{\gamma_i} \min\left\{\frac{\gamma_i}{d_i}, \frac{\gamma_j}{d_j}\right\} & j \in \mathcal{N}_i, j \neq i; \\ 1 - \sum_{j \in \mathcal{N}_i, j \neq i} w_{ij} & j = i \\ 0, & \text{otherwise.} \end{cases}$$

