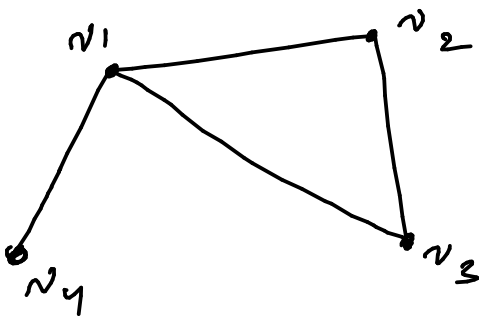


# ELL805

## Lecture - 14

→ Consensus in Discrete-time Agent Dynamics.

Agent Dynamics:  $x_i(k+1) = x_i(k) + u_i(k)$       $x_i(k) \in \mathbb{R}$



$$d(v_1) = 3$$

$$d(v_2) = 2$$

$$d(v_3) = 2$$

$$d(v_4) = 1$$

Nonnormalized Adjacency matrix

$$A_N = \begin{bmatrix} 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix} \rightarrow \text{not symmetric}$$

Define Laplacian matrix as follows

$$\mathcal{L}_N := I - A_N$$

$$\mathcal{L}_N = \begin{bmatrix} 1 & -\frac{1}{3} & -\frac{1}{3} & -\frac{1}{3} \\ -\frac{1}{2} & 1 & -\frac{1}{2} & 0 \\ -\frac{1}{2} & -\frac{1}{2} & 1 & 0 \\ -1 & 0 & 0 & 1 \end{bmatrix} \begin{array}{l} \rightarrow \text{Not symmetric} \\ \rightarrow \text{row sum is 0.} \end{array}$$

For agent dynamics  $x_i(k+1) = x_i(k) + u_i(k)$

Define the control protocol as follows

$$u_i(k) = \frac{1}{d(v_i)} \left[ \sum_{j \in \mathcal{N}(i)} (x_j(k) - x_i(k)) \right]$$

$\mathcal{N}(i)$ : Neighborhood set of  $i$ th agent.

For the considered example:

$$u_1(k) = \frac{1}{3} \left[ (x_2(k) - x_1(k)) + (x_3(k) - x_1(k)) + (x_4(k) - x_1(k)) \right]$$

$$u_2(k) = \frac{1}{2} \left[ (x_1(k) - x_2(k)) + (x_3(k) - x_2(k)) \right]$$

$$u_3(k) = \frac{1}{2} \left[ (x_2(k) - x_3(k)) + (x_1(k) - x_3(k)) \right]$$

$$u_4(k) = x_1(k) - x_4(k)$$

$$\begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \end{bmatrix} = \underbrace{\begin{bmatrix} -1 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & -1 & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & -1 & 0 \\ 1 & 0 & 0 & -1 \end{bmatrix}}_{-L_N} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

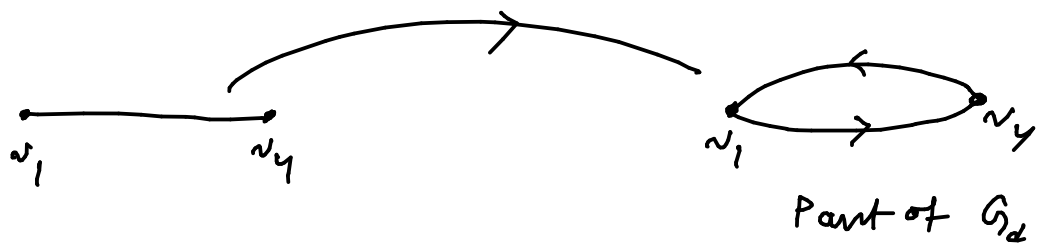
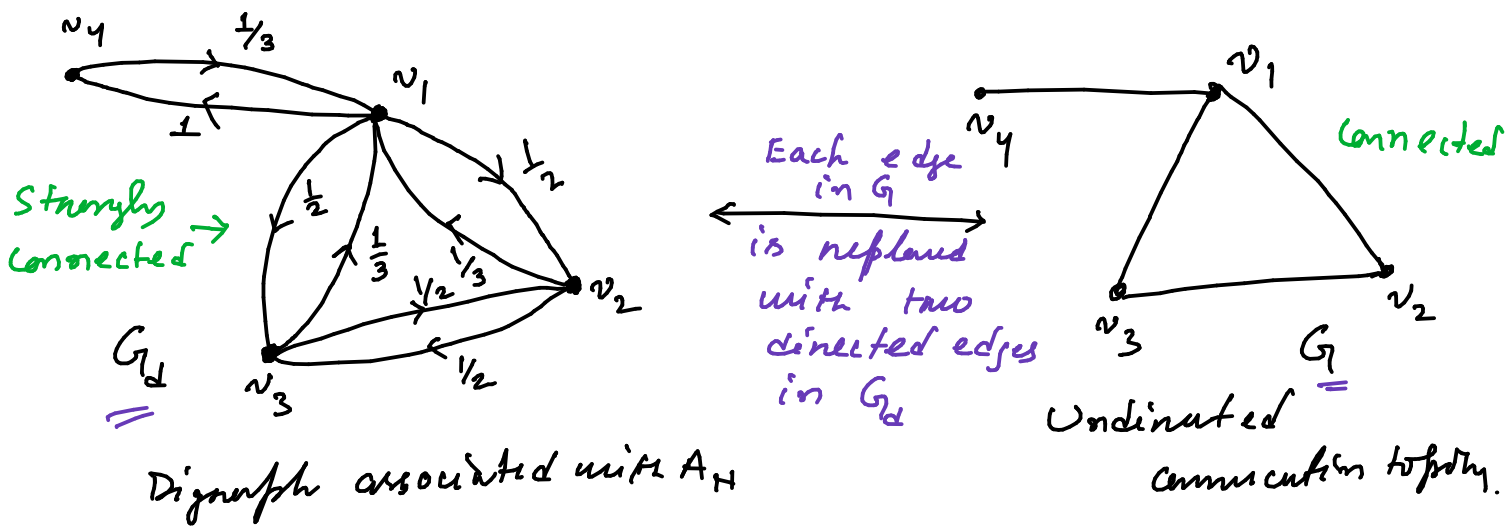
$$\begin{bmatrix} x_1(k+1) \\ x_2(k+1) \\ \vdots \\ x_n(k+1) \end{bmatrix} = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{bmatrix} \begin{bmatrix} x_1(k) \\ x_2(k) \\ \vdots \\ x_n(k) \end{bmatrix} + \underbrace{\begin{bmatrix} & & & \\ & -L_N & & \\ & & & \end{bmatrix}}_{u(k)} \begin{bmatrix} x_1(k) \\ x_2(k) \\ \vdots \\ x_n(k) \end{bmatrix}$$

The closed loop system :

$$\begin{aligned} x(k+1) &= (I - L_N) x(k) \\ &= A_N x(k) \end{aligned}$$

The properties of  $A_N$  :

- It is Non-negative and stochastic matrix
- Since  $A_N$  is stochastic matrix, the spectral radius  $\rho(A_N) = 1$
- let us construct a digraph corresponding to the normalized adjacency matrix  $A_N$  :



- The digraph associated with  $A_N(G_N)$  is "strongly connected" iff the network topology  $G$  is "connected".

$\Downarrow$

$A_N$  has "strongly connected" property.

$\Downarrow$

$A_N$  is irreducible.  
 $\Leftarrow$  also  $A_N$  is non-negative

$\Downarrow$  Perron-Frobenius Results

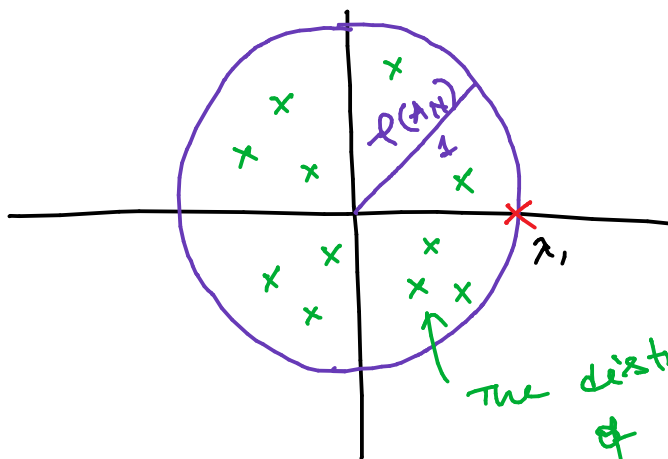
- $\rho(A_N) = 1$  is an eigenvalue of  $A_N$ .
- The Algebraic multiplicity of 1 is one.

• Since  $A_N$  is stochastic

$$A_N \cdot \mathbb{1} = \mathbb{1}$$

$\Downarrow$

$\mathbb{1}$  is an eigenvector of  $A_N$  corresponding to the eigenvalue 1.



The distribution of eigenvalues of  $A_N$  which are less than 1.

$$x(k+1) = A_N x(k) \dots \dots \dots (*)$$

The initial condition  $x(0)$

$$x(1) = A_N x(0)$$

$$x(2) = A_N x(1) = A_N^2 x(0)$$

$$x(3) = A_N x(2) = A_N^3 x(0)$$

⋮

$$\boxed{x(k) = A_N^k x(0)} \leftarrow \text{solution of } (*)$$

• let us assume that  $A_N$  has 'n' distinct eigenvalues.

constituted by using the eigenvectors of A

$$V^{-1} A V = \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix} \Rightarrow A = V \Lambda V^{-1}$$

→ show that  $A^k = V \Lambda^k V^{-1}$

$$\Lambda^k = \begin{bmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \ddots & \\ & & & \lambda_n^k \end{bmatrix}$$

$A v_i = \lambda_i v_i$   
 $A^2 v_i = \lambda_i^2 v_i$

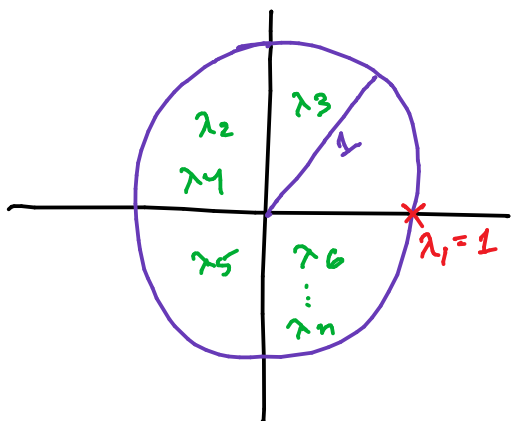
From the above we can write

$$\boxed{A_N^k = \lambda_1^k v_1 q_1^T + \lambda_2^k v_2 q_2^T + \dots + \lambda_n^k v_n q_n^T} \quad (A_N^0 = I)$$

where  $v_i$  &  $q_i$  are the right & left eigenvectors, respectively, of  $A_N$  corresponding to eigenvalue  $\lambda_i$ .

The root

$$\text{let } \lambda_1 > \lambda_2 \geq \lambda_3 \geq \dots \geq \lambda_n$$



The complex number  $\lambda_i$  can be represented as

$$\lambda_i = r e^{i\theta}$$

$$\text{where } r = |\lambda_i|$$

$$\Rightarrow \lambda_i^k = r^k e^{ik\theta}$$

The solution

$$x(k) = A_N^k x(0)$$

$$= \left[ \lambda_1^k v_1 q_1^T + \lambda_2^k v_2 q_2^T + \dots + \lambda_n^k v_n q_n^T \right] x(0)$$

$\lambda_i$ 's satisfy  $|\lambda_i| < 1$

$$\Rightarrow \lim_{k \rightarrow \infty} x(k) = \lim_{k \rightarrow \infty} \left[ \lambda_1^k v_1 q_1^T + \lambda_2^k v_2 q_2^T + \dots + \lambda_n^k v_n q_n^T \right] x(0)$$

$\downarrow$   $v_1 q_1^T x(0)$        $\rightarrow 0$

Recall that  $v_1$  is an eigenvector of  $A_N$  corresponding to the eigenvalue 1.  $\Rightarrow v_1 = \mathbb{1}$

$$\text{Hence } x(k) \xrightarrow{k \rightarrow \infty} v_1 q_1^T x_0 = \underbrace{(q_1^T x_0)}_{\in \text{span}\{\mathbb{1}\}}$$

Hence  $x(k)$  converges to the Agreement subspace.  
 $\Rightarrow$  Consensus achieved.