

diffusion

Laplacian

consensus

SVSM 6302

CLASS 20

# Processes on Networks

(Dynamics & Control)

- A state that evolves over time

(everything we  
need to know to describe  
the system)

follows a rule to find the  
state at the next moment  
in time

differential or difference equation

- The state typically describes the nodes of the network, but sometimes represents edges.
- The topology of the network (the location of edges between nodes), indicates which nodes/states have an effect on other nodes/states
- In simple models, nodes pass their state to their neighbors. More sophisticated models have nodes pass some function of their state, or have a way to preserve their own state

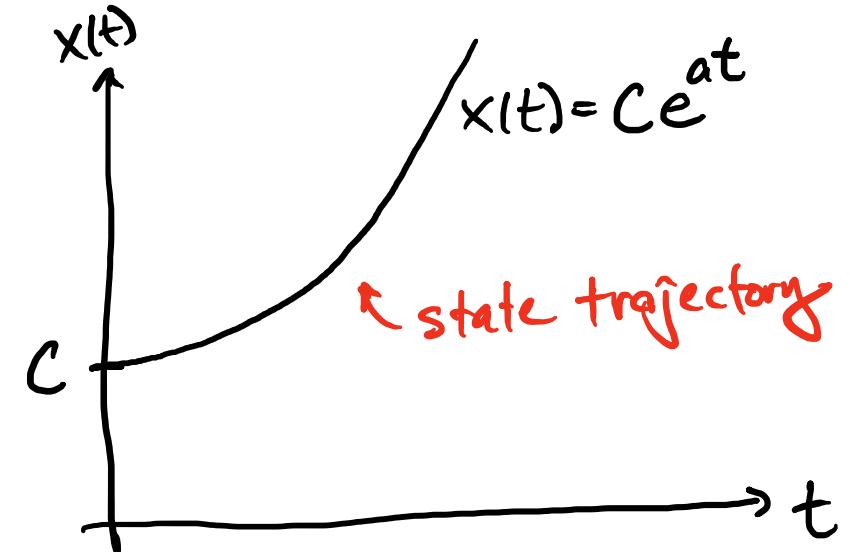
# Solutions to differential Equations

$$\dot{x}(t) = ax(t) \Rightarrow \frac{dx}{dt} = ax \Rightarrow \int \frac{dx}{x} = \int a dt \Rightarrow \ln(x) = at + \tilde{c}$$

$t \in \mathbb{R}$   
 $x(t) \in \mathbb{R}$   
 $a \in \mathbb{R}$

$$\Rightarrow x(t) = e^{at + \tilde{c}} = Ce^{at}$$

$$x(0) = Ce^{a \cdot 0} = C$$



# General Network Dynamics

Note: we are assuming an undirected graph

$$\dot{x}_i = f_i(x_i) + \sum_{j=1}^n A_{ij} g_{ij}(x_i, x_j)$$

due to connectivity

$f_i(x_i)$   
 ↑  
 the time rate of  
 change of the state  
 of node  $i$   
 intrinsic  
 dynamics  
 of the node  
 itself

$A_{ij}$  term only includes  
 contribution from node  $j$  if  
 there is an edge between  $i$  and  $j$

→ Often we assume the same form for all nodes:

$$\dot{x}_i = f(x_i) + \sum_{j=1}^n A_{ij} g(x_i, x_j)$$

## Diffusion

→ Physical diffusion follows a pressure differential

→ Rate of exchange between nodes  $\sim x_j - x_i$

if  $x_i = x_j$  nothing flows

if  $x_i > x_j$  flow is out, so negative

no intrinsic dynamics

← simple passing of state ← conservation of "state"

$$\dot{x}_i = \sum_{j=1}^n A_{ij} c(x_j - x_i)$$

Sum of row  $i = k_i$

$$= c \sum_j A_{ij} x_j - c x_i \sum_j A_{ij}$$

$$\delta_{ij} = \begin{cases} 1 & i=j \\ 0 & \text{otherwise} \end{cases}$$

$$= c \sum_j A_{ij} x_j - c x_i k_i = c \sum_j (A_{ij} - \delta_{ij} k_i) x_j$$

$$\dot{x}_i = c \sum_j (A_{ij} - \delta_{ij} k_i) x_j$$

$$\dot{x} = c \underbrace{(A - D)}_{= -L} x, \quad D = \begin{pmatrix} k_1 & & \\ & k_2 & \\ & & \ddots \\ & & & k_n \end{pmatrix}$$

$\leftarrow$  the **Laplacian**

$$\frac{dx}{dt} + c L x = 0$$

$\leftarrow$  Analogous to diffusion  
of a gas

time derivative  $\downarrow$       spatial derivative  $\downarrow$

$$\frac{dx}{dt} + c \nabla^2 x = 0$$

# Properties of the Laplacian

① Let  $x(t) = \sum_{i=1}^n a_i(t) v_i$  ← eigenvectors of  $L$

recall  $a_i(t) \in \mathbb{R}$

$v_i \in \mathbb{R}^n$

$$\frac{dx}{dt} + cLx = 0 \rightarrow \sum_i \frac{da_i}{dt} v_i + cL \sum a_i v_i = 0$$

$$\sum_i \left[ \frac{da_i}{dt} + cL a_i \right] v_i = 0$$

$$\sum_i \left[ \frac{da_i}{dt} + c\lambda_i a_i \right] v_i = 0$$

$$\left\langle \sum_i v_i, v_j \right\rangle = \left\langle 0, v_j \right\rangle$$

since  $L$  is symmetric

$$\left\langle v_i, v_j \right\rangle = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$$

$$a_i(t) = a_i(0) e^{-c\lambda_i t}$$



$$\frac{da_i}{dt} + c\lambda_i a_i = 0 \quad \text{for } i=1, 2, \dots, n$$

② It is possible to show that  $L$  is positive semi-definite

$$\Rightarrow \lambda_i \geq 0 \quad (\text{eigenvalues of } L \text{ are non negative})$$

① + ②:  $a_i(t) = a_i(0) e^{-c\lambda_i t}$ ,  $\lambda_i \geq 0 \rightarrow$  all  $a_i(t)$  converge

$$\Rightarrow x(t) = \sum_i a_i(t) v_i \rightarrow \text{equilibrium}$$

let  $x^* = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} \Rightarrow L x^* = (D - A)x^* = \left[ \begin{pmatrix} k_1 & & & \\ & k_2 & & \\ & & \ddots & \\ & & & k_n \end{pmatrix} - \begin{pmatrix} & & & \\ A_{11} & A_{12} & \cdots & A_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ A_{n1} & A_{n2} & \cdots & A_{nn} \end{pmatrix} \right] \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$

$$\rightarrow \sum_j (D_{ij} - A_{ij}) x_j^* = k_i - A_{i1} - A_{i2} - \cdots - A_{in} = 0 \quad \forall i$$

$\Rightarrow x^*$  is the eigenvector of  $L$  with  $\lambda = 0$

Thus  $x(t) = \sum_{i=1}^n a_i(t) v_i \rightarrow x^*$  as  $t \rightarrow \infty$

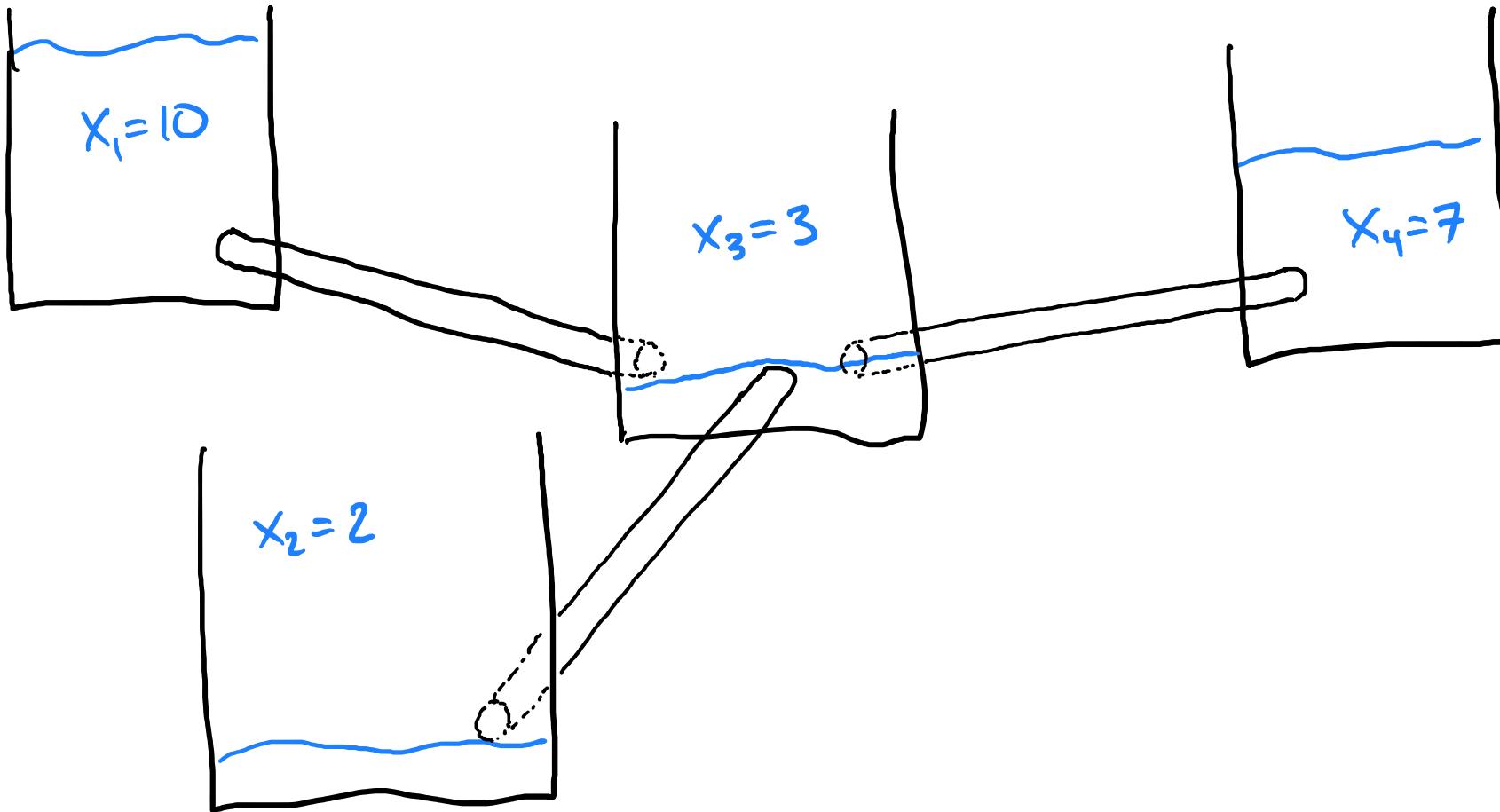
$$a_i(t) \rightarrow 0 \quad \forall \lambda_i > 0$$

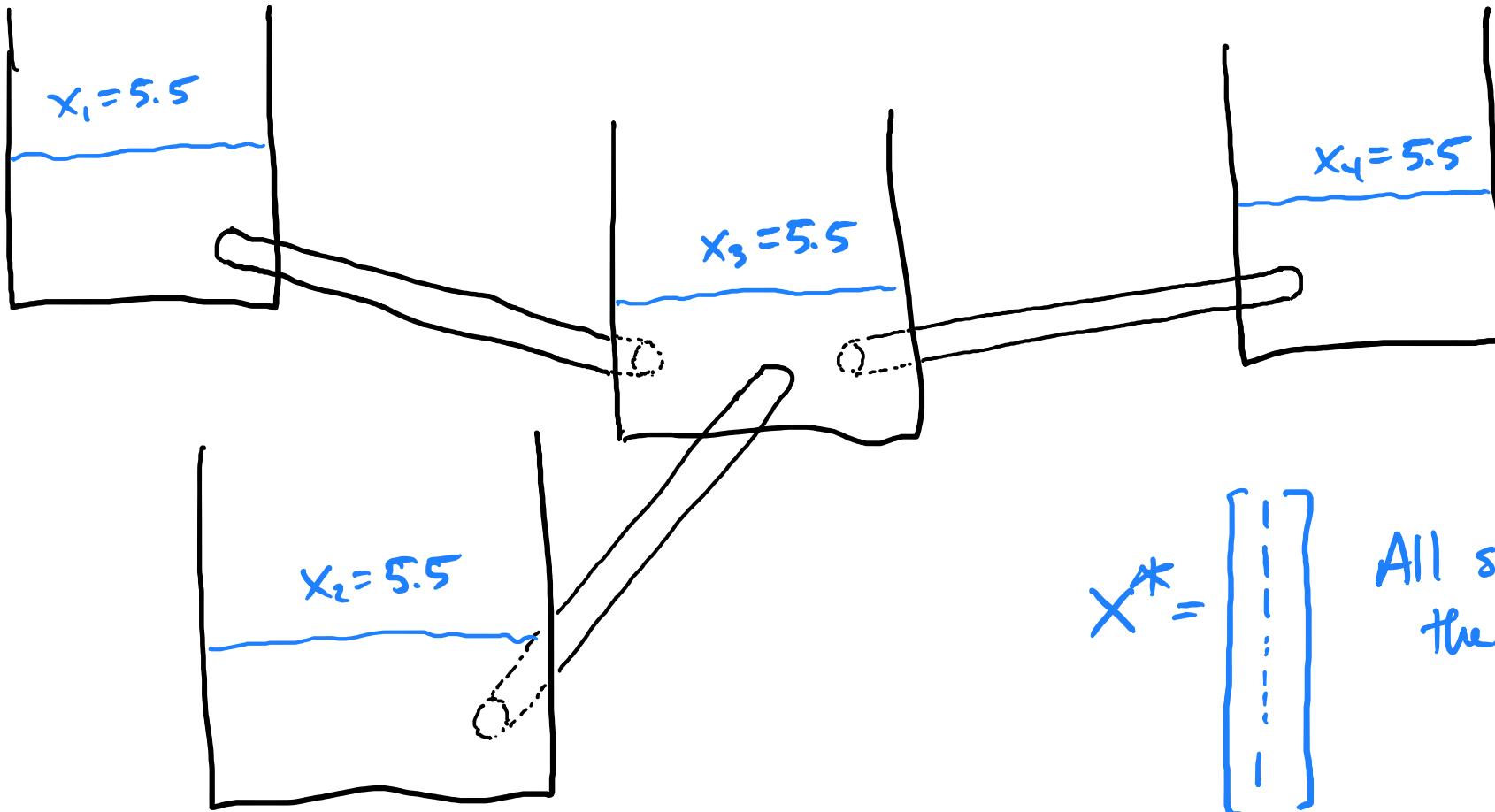
$$a_i(t) \rightarrow a_i(0) \quad \forall \quad \lambda_i = 0$$

③ The multiplicity of  $\lambda=0$  is equal to the number of components of A.

$$L_{x_i^*} = (D - A)x_i^* = \begin{bmatrix} k_1 & & & \\ & \ddots & & \\ & & k_l & \\ & & & \vdots \\ & & & k_{l+1} \\ & & & & \ddots & \\ & & & & & k_n \end{bmatrix} - \begin{pmatrix} A_1 & & & & & & \\ & \ddots & & & & & \\ & & 0 & & & & \\ & & & \ddots & & & \\ & & & & 0 & & \\ & & & & & A_2 & \\ & & & & & & \ddots & \\ & & & & & & & 0 \\ & & & & & & & & \ddots & \\ & & & & & & & & & 0 \end{pmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

↑ we can find a permutation to order the nodes by component



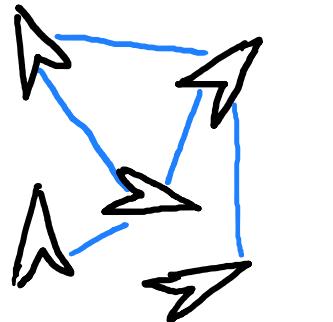


$$x^* = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

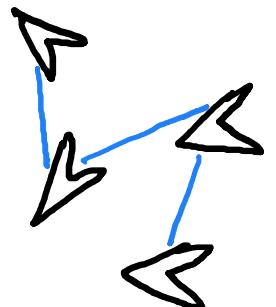
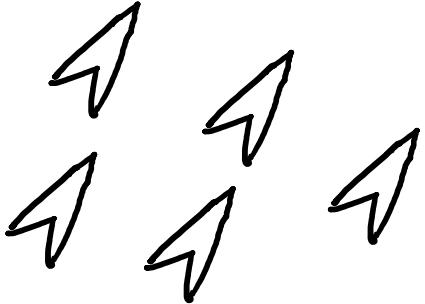
All states are  
the same!

## Consensus

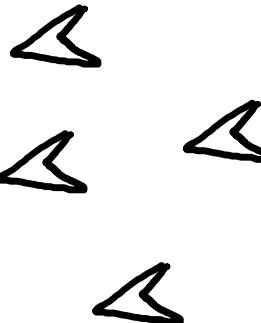
$$\dot{x}_i = \sum_{j=1}^n A_{ij} (x_j - x_i) \Rightarrow \dot{x} = -Lx$$



$t$



$t$



Directed graphs require having a directed spanning tree to reach consensus.

For switching topologies, the union of graphs must be jointly connected (undirected) or have a minimum spanning tree (directed), in some finite time interval.

Designation of leader / follower