

# SIGNED NONLINEAR NETWORKS

A PASSIVITY AND ELECTRICAL CIRCUIT THEORY APPROACH

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Daniel Zelazo

January 20, 2020

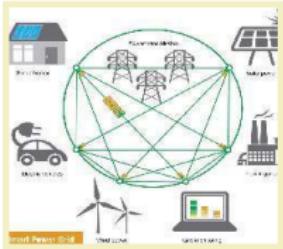
Technion CST Seminar



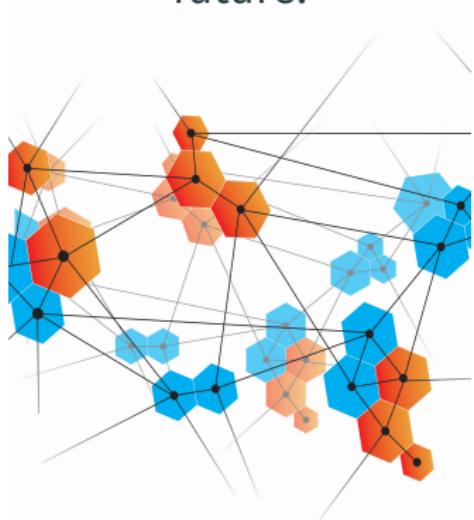
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# NETWORKED DYNAMIC SYSTEMS



Networks of dynamical systems are one of **the** enabling technologies of the future.



# COOPERATIVE V. ANTAGONISTIC

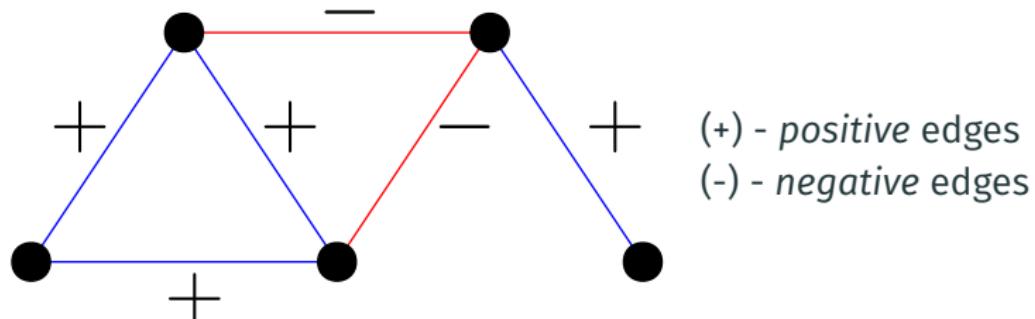
In many *biological* and *social* networks, there are generally **two** kinds of interactions:

- **trustful** versus **distrustful**
- **cooperative** versus **competitive**



## COOPERATIVE V. ANTAGONISTIC

*Signed Networks* provide an abstraction for modeling cooperative or antagonistic interactions.



$$\mathcal{G} = (\mathcal{V}, \mathcal{E}, \mathcal{W})$$

$$\mathcal{W} : \mathcal{E} \rightarrow \mathbb{R}$$

$$\mathcal{E}_+ = \{e \in \mathcal{E} : \mathcal{W}(e) > 0\}$$

$$\mathcal{E}_- = \{e \in \mathcal{E} : \mathcal{W}(e) < 0\}$$

$$\mathcal{G} = \mathcal{G}_+ \cup \mathcal{G}_-$$

$$\mathcal{E} = \mathcal{E}_+ \cup \mathcal{E}_-$$

## GRAPHS IN NETWORKED DYNAMIC SYSTEMS

- A graph can describes how agents interact with each other
- Node  $i$  acquires information from its neighbors  $\mathcal{N}(i)$

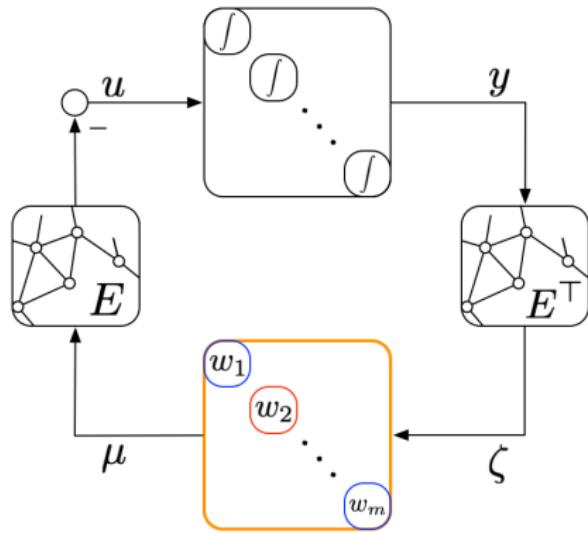


- Node  $i$  has a state  $x_i(t)$  and neighbor information  
 $I_i(t) = \{x_j(t) | j \in \mathcal{N}(i)\}$
- Provides a naturally distributed dynamics over  $\mathcal{G}$

$$\dot{x}_i(t) = f_i(x_i(t), I_i(t))$$

- some of the earlier works in distributed decision-making include: DeGroot ('74), Borkar and Varaiya ('82), Tsitsiklis ('84) ...

# LINEAR CONSENSUS AND SIGNED NETWORKS



Edge weights can be **positive** or **negative** to indicate the cooperative or antagonistic interactions

## Agent Dynamics

$$\dot{x}_i = u_i$$

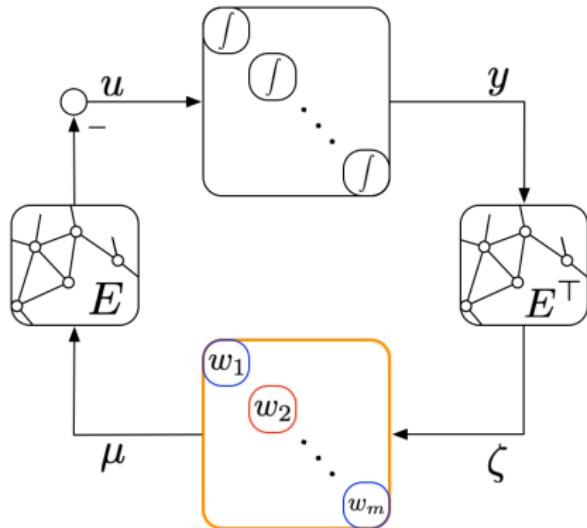
$$u_i \rightarrow \int \rightarrow x_i$$

## Consensus Protocol

$$u_i = \sum_{j \sim i} w_{ij}(x_j - x_i)$$

$$\dot{x} = -L(\mathcal{G})x$$

# LINEAR CONSENSUS AND SIGNED NETWORKS



## Consensus Protocol

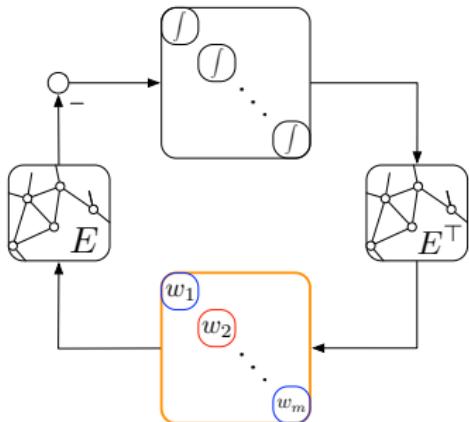
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## Theorem

Let  $\mathcal{G} = (\mathcal{V}, \mathcal{E}, \mathcal{W})$  be a weighted and connected graph with positive edge weights  $\mathcal{W}(k) > 0$  for  $k = 1, \dots, |\mathcal{E}|$ . Then the consensus dynamics reach agreement, i.e.,  $\lim_{t \rightarrow \infty} x_i(t) = \beta$  for  $i = 1, \dots, |\mathcal{V}|$ .

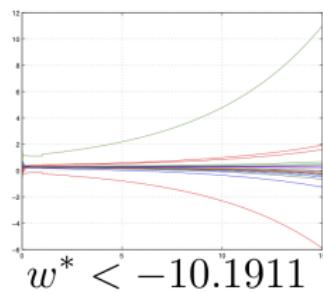
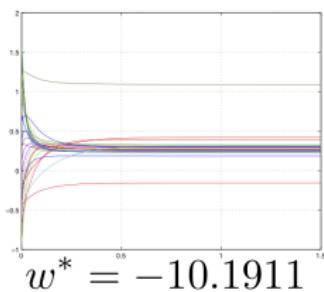
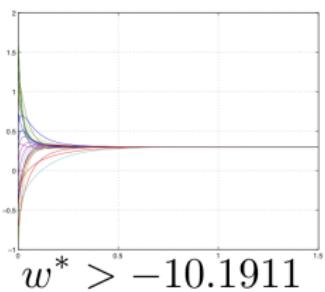
# POSITIVE WEIGHTS IS A SUFFICIENT CONDITION



## Consensus Protocol

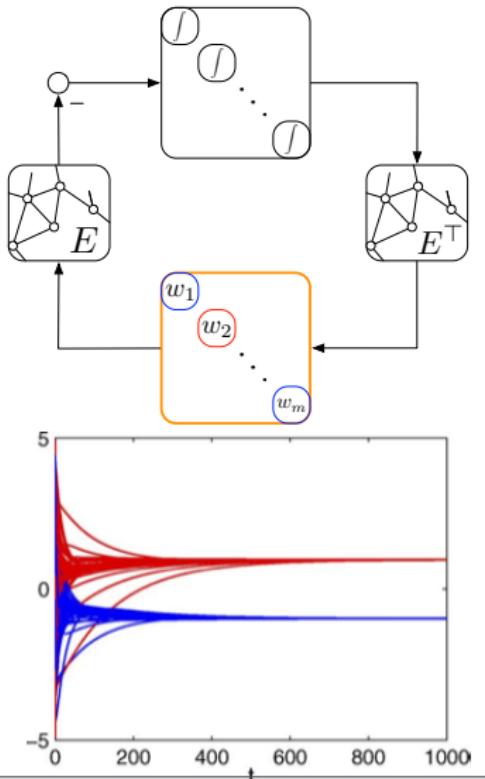
$$u_i = \sum_{i \sim j} w_{ij}(x_j - x_i)$$

$$\dot{x} = -L(\mathcal{G})x$$



$|\mathcal{V}| = 25, |\mathcal{E}| = 98 \quad w^*$  is weight of predefined edge

# POSITIVE WEIGHTS IS A SUFFICIENT CONDITION



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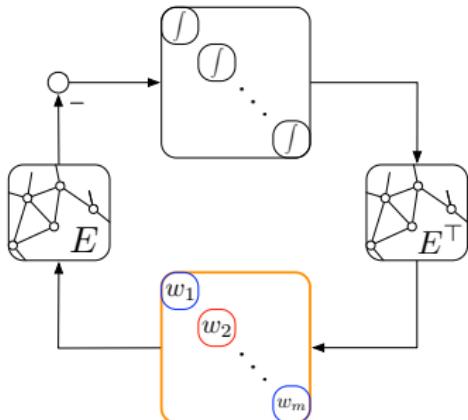
$$\dot{x} = -L(\mathcal{G})x$$

The **bipartite consensus** problem over signed graphs

- signed graph is structurally balanced

<sup>1</sup>C. Altafini, Consensus Problems on Networks with Antagonistic Interactions, IEEE Transactions on Automatic Control, 58(4):935-946, 2013.

# POSITIVE WEIGHTS IS A SUFFICIENT CONDITION



## Consensus Protocol

$$u_i = \sum_{i \sim j} w_{ij}(x_j - x_i)$$

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How can we understand dynamic networks over signed graphs?

Introduction

## Circuit Analogies

From Linear to Nonlinear Networks

Signed Nonlinear Edges

Convergence Analysis

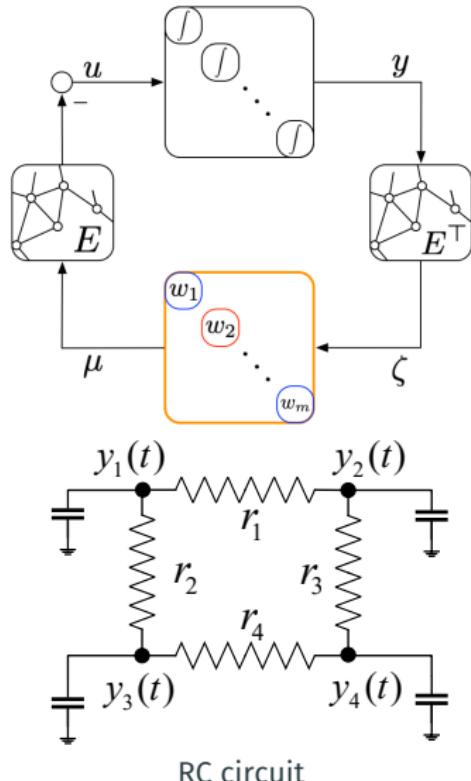
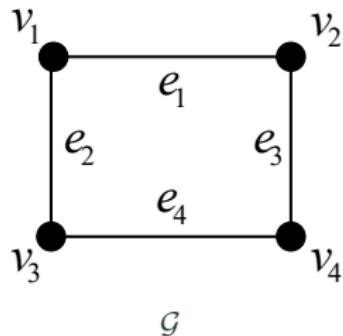
Conclusion

Thanks

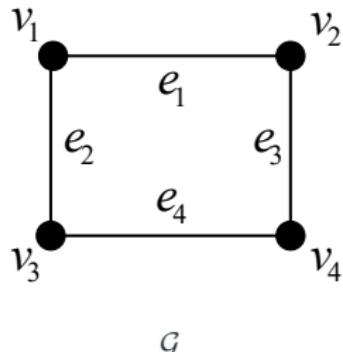
# CIRCUIT INTERPRETATIONS

## Linear Consensus as an RC-Circuit

Capacitors  $\Leftrightarrow$  Node Dynamics (integrators)  
Resistors  $\Leftrightarrow$  Edge Dynamics (linear gain)

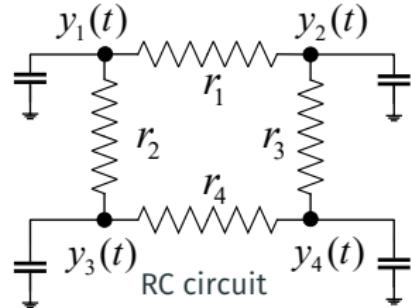


# CIRCUIT INTERPRETATIONS



Edge Functions:

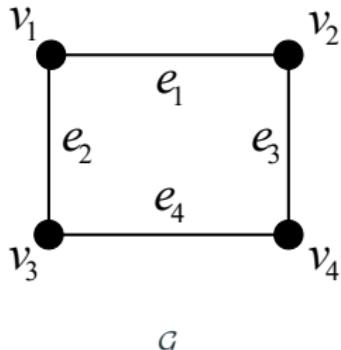
$$\mu_k = w_k \zeta_k = w_k(y_i - y_j)$$



Control :  $\mathbf{u} = -EWE^T \mathbf{y}$

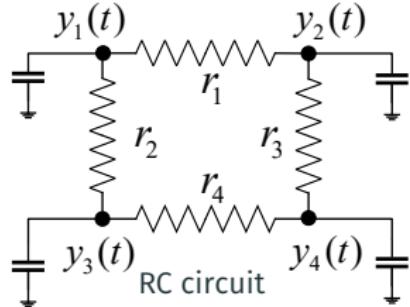
$$w_k = \frac{1}{r_k}$$

# CIRCUIT INTERPRETATIONS



Edge Functions:

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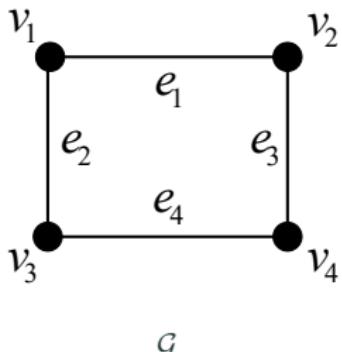


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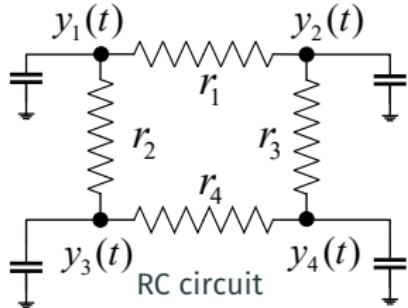
In steady-state,  $\mathbf{u} = 0$ , and the network corresponds to a **resistive circuit**.

# CIRCUIT INTERPRETATIONS



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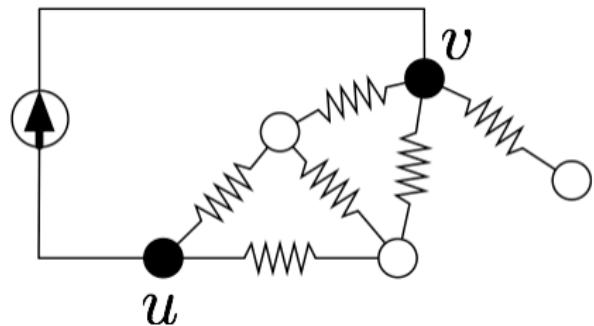
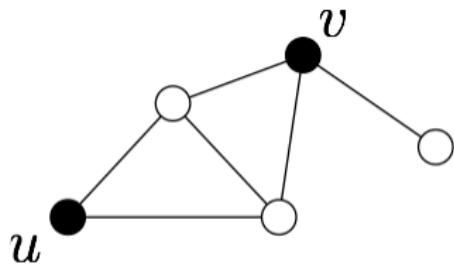
$$w_k = \frac{1}{r_k}$$

In steady-state,  $\mathbf{u} = 0$ , and the network corresponds to a **resistive circuit**.

What happens if we add a **negative resistor**?

## EFFECTIVE RESISTANCE

The **effective resistance** between two nodes  $u$  and  $v$  is the electrical resistance measured across the nodes when the graph represents a resistive circuit.



### Effective Resistance Calculation [Klein and Randić 1993]

$$r_{uv} = [L^\dagger(\mathcal{G})]_{uu} + 2[L^\dagger(\mathcal{G})]_{uv} + [L^\dagger(\mathcal{G})]_{vv}$$

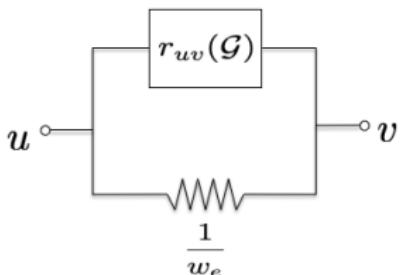
# EFFECTIVE RESISTANCE AND SIGNED LINEAR NETWORKS

## Theorem [Z, Bürger CDC2014]

Let  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_>)$  be a strictly positive network with edge functions  $\mu_k = w_k \zeta_k$  (i.e.,  $w_k > 0$  for all  $k \in \mathcal{E}$ ) and let  $\bar{\mathcal{G}} = (\mathcal{V}, \mathcal{E}_> \cup e)$  where  $e = (u, v)$  is a negative edge with  $\mu_e = w_e \zeta_e$  and  $w_e < 0$ . Then the signed consensus network reaches agreement if and only if

$$|w_e| \leq r_{uv}^{-1}(\mathcal{G}),$$

where  $r_{uv}(\mathcal{G})$  is the effective resistance in  $\mathcal{G}$  between nodes  $u$  and  $v$ .



$$r_{uv}(\bar{\mathcal{G}}) = \frac{\frac{1}{w_e} r_{uv}(\mathcal{G})}{r_{uv}(\mathcal{G}) + \frac{1}{w_e}} \Rightarrow \begin{cases} > 0, & \text{cooperative} \\ < 0, & \text{antagonistic} \\ = \infty, & \text{no interaction} \end{cases}$$

A matched negative edge weights effectively creates an open circuit

Introduction

Circuit Analogies

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Thanks

In many applications, **nonlinear protocols** are used to achieve the desired behavior of the network.

- **Kuramoto model**<sup>1</sup>

$$\mu_k = w_k \sin \zeta_k.$$

- **Finite-time consensus protocol**<sup>2</sup>

$$\mu_k = w_k \cdot \text{sign}(\zeta_k) \cdot |\zeta_k|^{\alpha_k}, \quad 0 < \alpha_k < 1$$

## Motivation

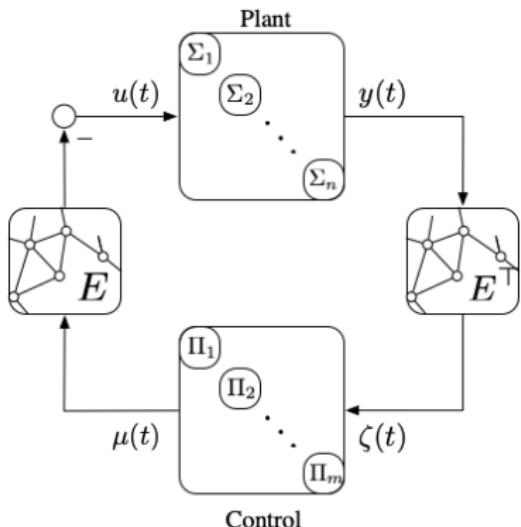
Generalize the notion of **signed networks** to the nonlinear case, and analyze its convergence properties.

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<sup>1</sup>J. A. Acebrn, L. L. Bonilla, C. J. P. Vicente, et al., The Kuramoto model: A simple paradigm for synchronization phenomena, *Reviews of Modern Physics*, vol. 77, no. 1, pp. 137–185, 2005.

<sup>2</sup>L. Wang and F. Xiao, Finite-time consensus problems for networks of dynamic agents, *IEEE Transactions on Automatic Control*, vol. 55, no. 4, pp. 950–955, 2010.

# NONLINEAR NETWORK MODEL



The network is denoted by the triple  $(\mathcal{G}, \Sigma, \Pi)$ .

## Node dynamics

$$\begin{aligned}\Sigma_i : \dot{x}_i(t) &= f_i(x_i(t), u_i(t)), \\ y_i(t) &= h_i(x_i(t), u_i(t)).\end{aligned}$$

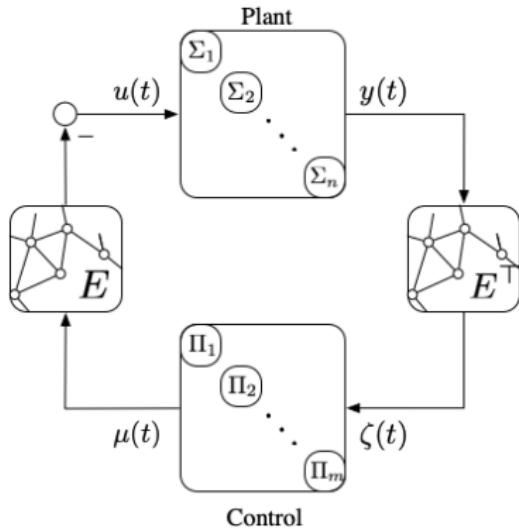
## Edge functions

$$\Pi_k : \mu_k(t) = \psi_k(\zeta_k(t)),$$

“Equilibrium Set”

$$I_k = \{\zeta_k \mid \psi_k(\zeta_k) = 0\}$$

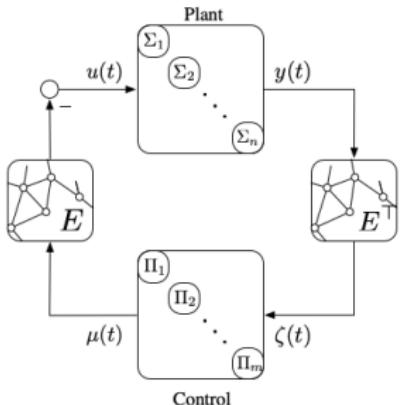
# NETWORK MODEL



Convergence analysis on these non-linear networks relies on:

- a variant of **passivity theory** (Maximal Equilibrium Independent Passivity - MEIP)
- the notion of **steady-state input output relations**

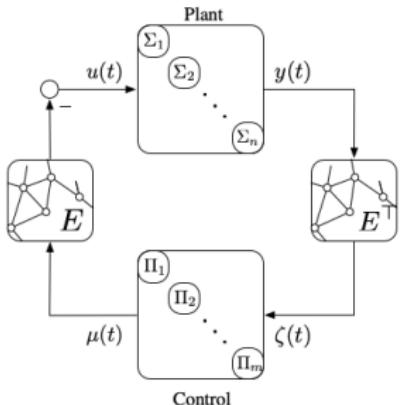
# STEADY-STATE INPUT-OUTPUT MAPS



## Assumption 1

Each agent  $\Sigma_i$  and controller  $\Pi_e$  admit forced steady-state solutions.

# STEADY-STATE INPUT-OUTPUT MAPS



## Assumption 1

Each agent  $\Sigma_i$  and controller  $\Pi_e$  admit forced steady-state solutions.

## Input-Output Maps

The steady-state input-output map  $k : \mathcal{U} \rightarrow \mathcal{Y}$  associated with  $\Sigma$  is the set consisting of all steady-state input-output pairs  $(u, y)$  of the system.

$$u_i \xrightarrow{y_i \in k_i(u_i)} y_i$$

$$\zeta_e \xrightarrow{\mu_e \in \gamma_e(\zeta_e)} \mu_e$$

$$u_i \rightarrow \Sigma_i \rightarrow y_i$$

$$\zeta_e \rightarrow \Pi_e \rightarrow \mu_e$$

$$u_i \leftarrow y_i \xleftarrow{u_i \in k_i^{-1}(y_i)}$$

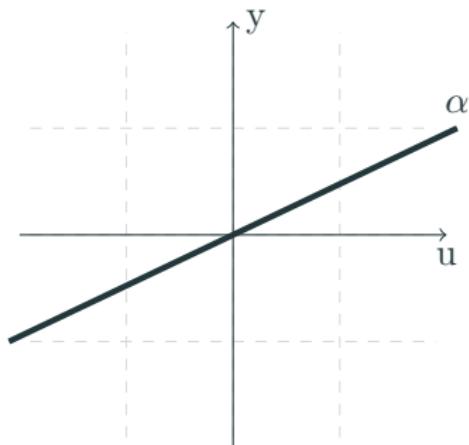
$$\zeta_e \leftarrow \mu_e \xleftarrow{\zeta_e \in \gamma_e^{-1}(\mu_e)}$$

## INPUT-OUTPUT RELATIONS

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

$$\Rightarrow k(u) = \{y \mid \underbrace{(-CA^{-1}B + D)}_{\alpha} u\}$$



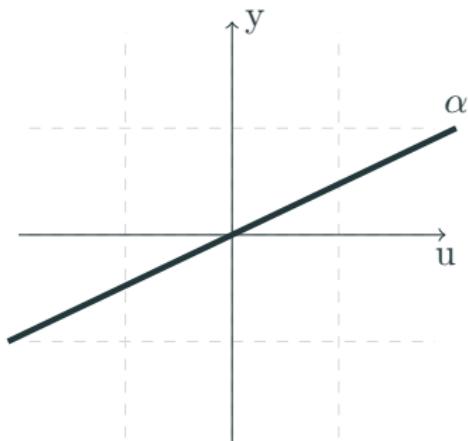
SISO and stable linear system

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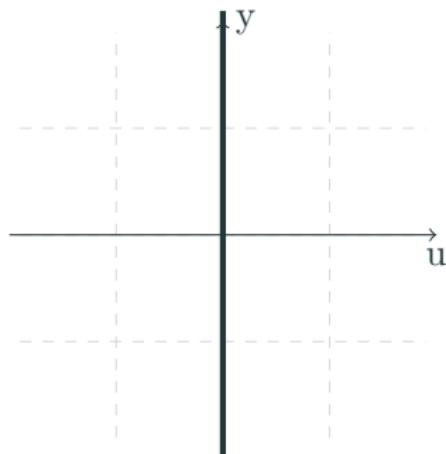


SISO and stable linear system

$$\dot{x} = u$$

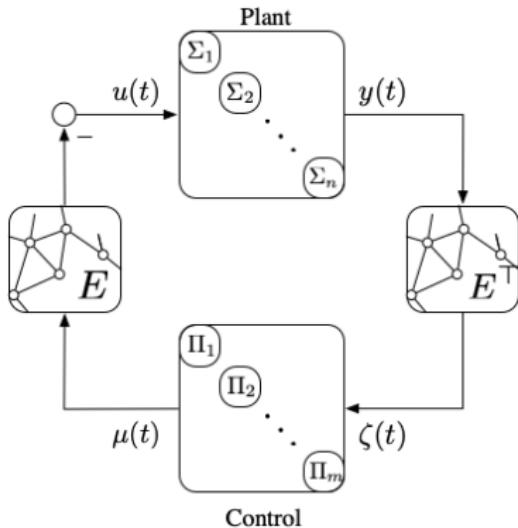
$$y = x$$

$$\Rightarrow k = \{(0, y), y \in \mathbb{R}\}$$



simple integrator

# NETWORK MODEL



Convergence analysis on these non-linear networks relies on:

- a variant of **passivity theory** (Maximal Equilibrium Independent Passivity - MEIP)
- the notion of **steady-state input output relations**

## Assumption 1: Feasibility of Consensus

$\Sigma_i$  is MEIP, and the equilibrium input-output relations of the nodes satisfy  $k(\mathbf{0}) \cap \mathcal{N}(E^T) \neq \emptyset$ .

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**Signed Nonlinear Edges**

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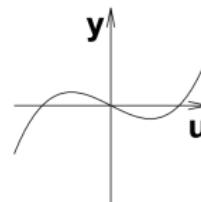
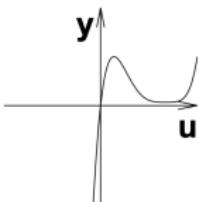
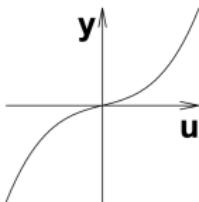
Thanks

## Passivity

A system  $\eta = \pi(t, \xi)$ , where  $\xi, \eta$  are the system input vector and system output vector, respectively, is

- i) **passive** if  $\xi^T \eta \geq 0$ ;
- ii) **input strictly passive** if there exists  $\epsilon > 0$ , such that  $\xi^T \eta \geq \epsilon \xi^T \xi$ ;
- iii) **active** if  $\xi^T \eta \leq 0$ ;
- iv) **input strictly active** if there exists  $\epsilon > 0$ , such that  $\xi^T \eta \leq -\epsilon \xi^T \xi$ .

In all above cases, the inequality should hold for all  $(t, \xi)$ .

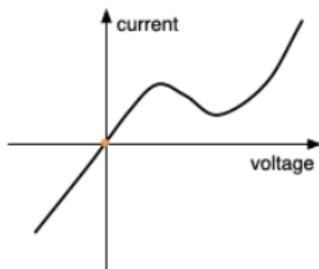


# SIGNED NONLINEAR EDGES

## Definition

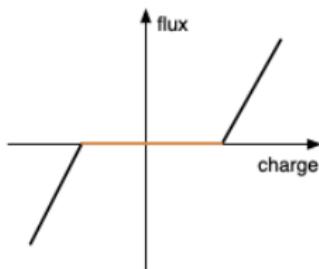
Suppose  $\Pi_k$  is a map from  $\mathbb{R}$  to  $\mathbb{R}$  with  $\psi_k(0) = 0$ . Then edge  $k$  is termed

- i) (strictly) positive if  $\Pi_k$  is (input strictly) passive;
- ii) (strictly) negative if  $\Pi_k$  is (input strictly) active.



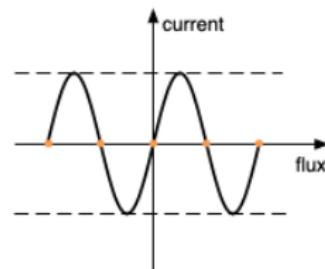
tunnel diode

(strictly) positive



memristor

positive



Josephson junction

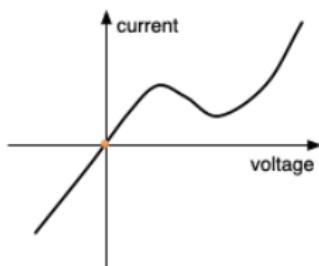
neither positive  
nor negative

# SIGNED NONLINEAR EDGES

## Definition: Signed Nonlinear Networks

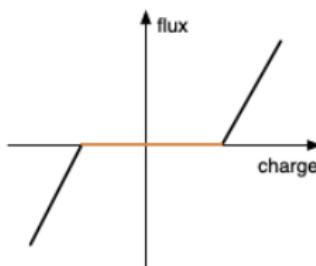
A networked system  $(\mathcal{G}, \Sigma, \Pi)$  is a

- **positive network** if all the edges are positive, and  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_{\geq})$ ;
- **strictly positive network** if all the edges are strictly positive, and  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_{>})$ ;
- **signed network** if not all the edges are positive.



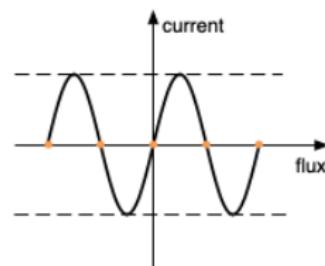
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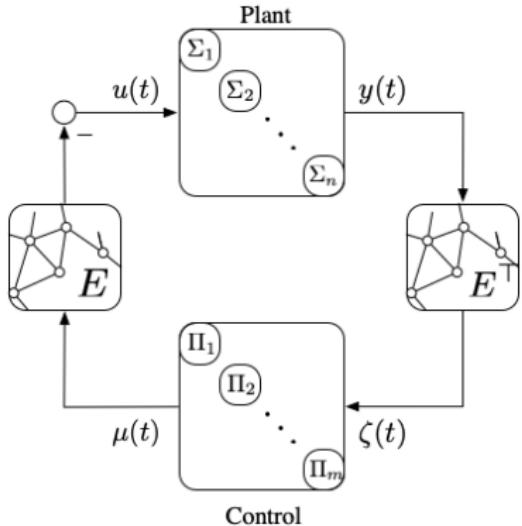
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## “Signed” Edge functions

$$\Pi_k : \mu_k(t) = \psi_k(\zeta_k(t)),$$

## Assumption 1: Feasibility of Consensus

$\Sigma_i$  is MEIP, and the equilibrium input-output relations of the nodes satisfy  $k(\mathbf{0}) \cap \mathcal{N}(E^T) \neq \emptyset$ .

## Theorem

Consider a **positive** network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_\geq)$  and suppose Assumption 1 holds. Then  $\lim_{t \rightarrow \infty} \zeta(t) = \tilde{\zeta}$  exists, and  $\tilde{\zeta} \in \mathbf{I}$ .

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## PROOF

Consider the Lyapunov function  $V(\mathbf{x}(t)) = \sum_{i=1}^{|\mathcal{V}|} S_i(x_i(t))$

$$\dot{V} = \sum_{i=1}^{|\mathcal{V}|} \dot{S}_i \leq (\mathbf{u}(t) - \mathbf{0})^T (\mathbf{y}(t) - \tilde{\mathbf{y}}) = -\zeta(t)^T \boldsymbol{\mu}(t) \leq 0.$$

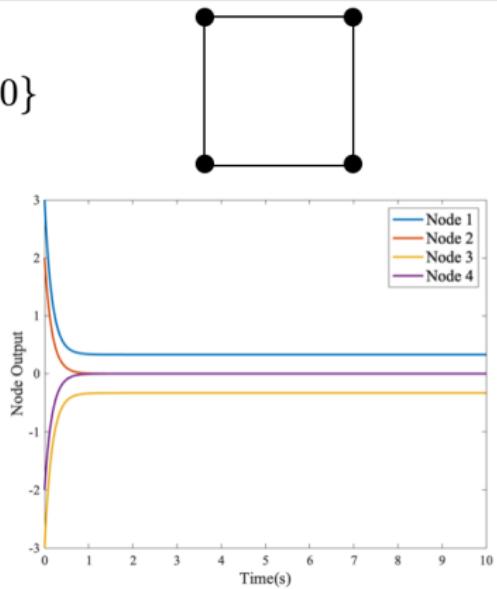
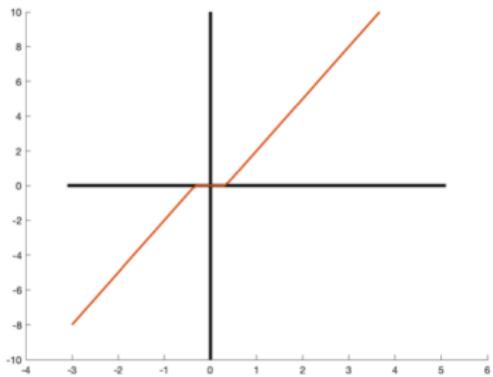
By using LaSalle's invariance principle,  $\lim_{t \rightarrow \infty} \zeta(t)^T \boldsymbol{\mu}(t) = \mathbf{0}$ , meaning  $\lim_{t \rightarrow \infty} \boldsymbol{\mu}(t) = \mathbf{0}$ , and  $\lim_{t \rightarrow \infty} \zeta(t) \in \mathbf{I}$ . As a result,  $\lim_{t \rightarrow \infty} \mathbf{u}(t) = \mathbf{0}$ , and  $\lim_{t \rightarrow \infty} \mathbf{y}(t) = \tilde{\mathbf{y}} \in k(\mathbf{0})$ , therefore  $\lim_{t \rightarrow \infty} \zeta(t) = \tilde{\zeta} = E^T \tilde{\mathbf{y}}$  exists.

# POSITIVE NETWORKS

## Theorem

Consider a **positive** network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_{\geq})$  and suppose Assumption 1 holds. Then  $\lim_{t \rightarrow \infty} \zeta(t) = \tilde{\zeta}$  exists, and  $\tilde{\zeta} \in \mathbf{I}$ .

$$\mu_k = \text{sign}(\zeta_k) \max\{3|\zeta_k| - 1, 0\}$$

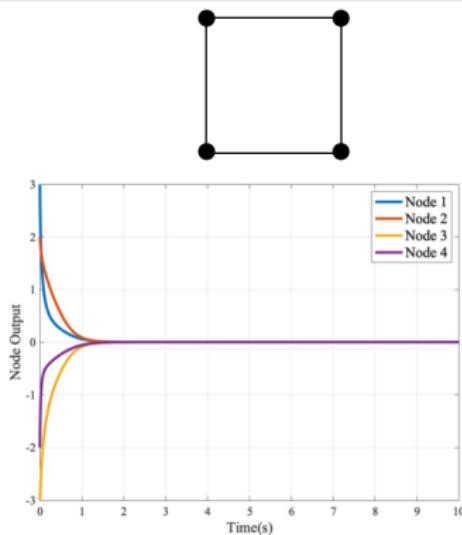
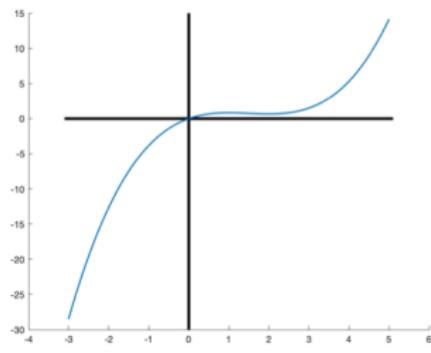


# STRICTLY POSITIVE NETWORKS

## Corollary

Consider a **strictly positive** network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_>)$  and suppose Assumption 1 holds. Then  $\lim_{t \rightarrow \infty} \zeta(t) = 0$ , and  $\lim_{t \rightarrow \infty} y(t) = \beta \mathbf{1}$ ,  $\beta \in \mathbb{R}$ .

$$\mu_k = \frac{1}{3}\zeta_k^3 - \frac{3}{2}\zeta_k^2 + 2\zeta_k$$



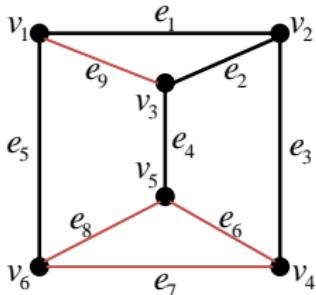
## Corollary

Consider a **positive** network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_\geq)$  and suppose Assumption 1 holds. If there exists a **connected** subgraph spanning **all nodes and strictly positive edges**, then

$$\lim_{t \rightarrow \infty} \zeta(t) = \mathbf{0} \text{ and } \lim_{t \rightarrow \infty} \mathbf{y}(t) = \beta \mathbf{1}, \quad \beta \in \mathbb{R}.$$

A strictly positive spanning subgraph  
is needed to guarantee consensus!

# AN EXAMPLE



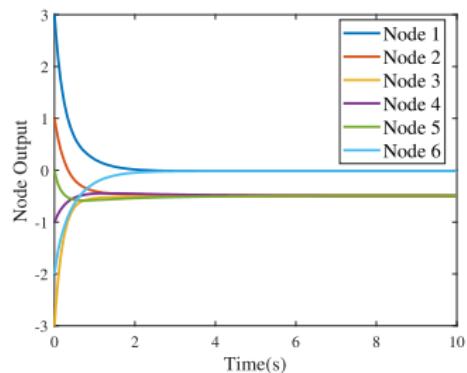
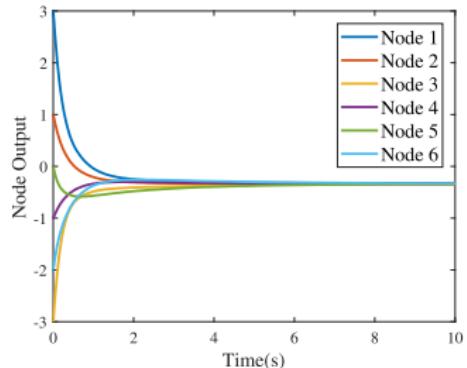
Edges  $e_k$  ( $k = 1, \dots, 5$ ) form a spanning tree.

Two possible edge functions:

$$\mu_k(t) = \text{sign}(\zeta_k(t)) \cdot \max\{\zeta_k(t) - 1, 0\}, \quad \begin{cases} (1) \\ (2) \end{cases}$$

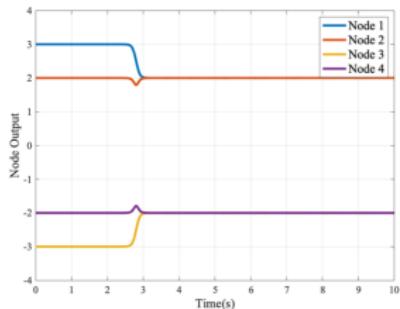
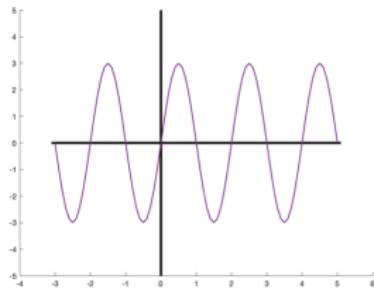
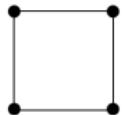
Two scenarios:

- Edges  $e_k$  ( $k = 1, \dots, 5$ ) use (1), and edges  $e_k$  ( $k = 6, \dots, 9$ ) use (2).
- Edges  $e_k$  ( $k = 2, \dots, 5$ ) use (1), and edges  $e_k$  ( $k = 1, 6, \dots, 9$ ) use (2).

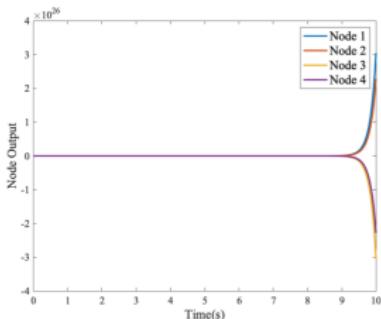
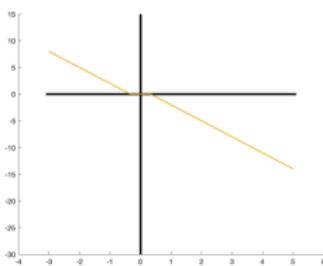


# AN EXAMPLE: NON-POSITIVE NETWORKS

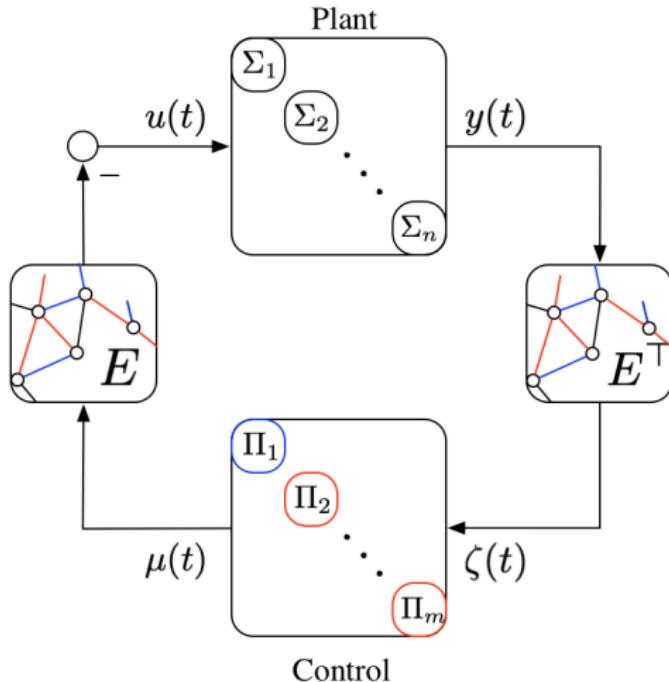
$$\mu_k = 3 \sin(\pi \zeta_k)$$



$$\mu_k = -\text{sign}(\zeta_k) \max\{3|\zeta_k| - 1, 0\}$$



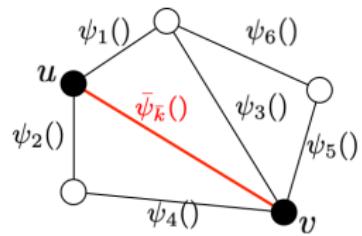
# NON-POSITIVE NETWORKS



How do we analyze nonlinear networks  
with positive and negative edges?

# EQUIVALENT EDGE FUNCTION

Generalizations for non-linear resistors



## Equivalent (non-linear) Edge Functions

Consider a network  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  and  $\bar{\mathcal{G}} = (\mathcal{V}, \mathcal{E} \cup \bar{k})$ , with  $\bar{k} = (u, v)$ .

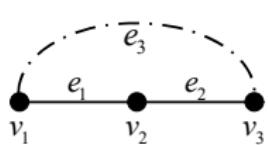
For each given  $\zeta_{\bar{k}}$ , if there exists a unique  $(\bar{\zeta}, \bar{\mu})$  and some  $\mathbf{y} \in \mathbb{R}^{|\mathcal{V}|}$  such that

$$\begin{cases} E(\bar{\mathcal{G}})^T \mathbf{y} &= \bar{\zeta} \quad \text{KVL} \\ \bar{\mu} &= \Psi(\bar{\zeta}) \\ E(\bar{\mathcal{G}})\bar{\mu} &= \mathbf{0} \quad \text{KCL} \end{cases},$$

then the flow  $\mu_{\bar{k}}$  on the edge  $\bar{k}$  can be represented as a function of  $\zeta_{\bar{k}}$ , which we denote as  $\mu_{\bar{k}} = -\bar{\psi}_{\bar{k}}(\zeta_{\bar{k}})$ .

$\bar{\psi}_{\bar{k}}(\cdot) : \mathbb{R} \rightarrow \mathbb{R}$  is termed the equivalent edge function between nodes  $u$  and  $v$ .

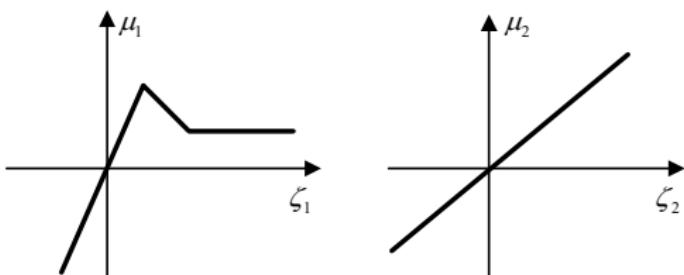
# DOES EQUIVALENT EDGE FUNCTION ALWAYS EXIST?



$$\bar{E} = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \\ 0 & -1 & -1 \end{bmatrix}$$

For each  $\zeta_3$ , there should exist a unique  $\mu_3$  such that

$$\begin{cases} \bar{E}^T \mathbf{y} = \bar{\zeta} \\ \mu = \Psi(\zeta) \\ \bar{E}\bar{\mu} = 0 \end{cases}$$



Suppose  $\zeta_3 = 3$ ,

$$\mu_1 = \begin{cases} 2\zeta_1, & \zeta_1 \in (-\infty, 1]; \\ 3 - \zeta_1, & \zeta_1 \in (1, 2]; \\ 1, & \zeta_1 \in (2, \infty). \end{cases}$$

$$\mu_2 = \zeta_2.$$

- if  $\zeta_1 = 1$ , then  $\zeta_2 = 2$ ,  $\mu_1 = \mu_2 = 2$ ,  
 $\mu_3 = -2$ ;
- if  $\zeta_1 = 2$ , then  $\zeta_2 = 1$ ,  $\mu_1 = \mu_2 = 1$ ,  
 $\mu_3 = -1$ .

Not unique!

# EXISTENCE OF EQUIVALENT EDGE FUNCTIONS

## Lemma [L. O. Chua (1969)]

Consider a circuit containing only resistors and independent voltage sources. If the current-voltage function of each resistor is strictly monotonically increasing with the current tending to  $\pm\infty$  as the voltage tends to  $\pm\infty$ , and if the circuit contains no cycles of voltage sources, then for each pair of the voltage source values, the current and voltage drop on each resistor, as well as the current on each source, are unique.

## Proposition

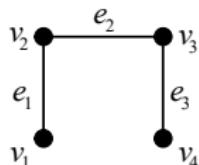
Consider a strictly positive network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_>)$ . Identify  $p, q \in \mathcal{V}$  as two terminals of interest. If for each  $k \in \mathcal{E}_>$ , its edge function  $\mu_k(t) = \psi_k(\zeta_k(t))$  is strictly monotonically increasing, and  $\mu_k(t) \rightarrow \pm\infty$  as  $\zeta_k(t) \rightarrow \pm\infty$ , then the equivalent edge function of the two-terminal network between nodes  $p$  and  $q$  exists.

### Algorithm 1 Computation of Equivalent Edge Functions

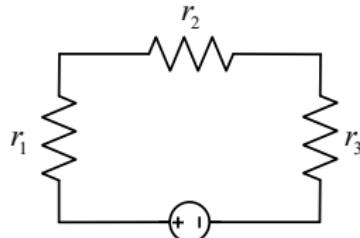
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- 1: Obtain the corresponding resistive circuit, and add a voltage source between  $p$  and  $q$ ;
  - 2: Obtain a finite set  $\{\zeta_{pq}\}$  by sampling the interval  $[-N, N]$ ;
  - 3: **for all**  $\zeta_{pq}$  **do**
  - 4:     Set  $\zeta_{pq}$  as the value of voltage source;
  - 5:     Calculate the flow on the voltage source  $\bar{\mu}_{pq}$  while satisfying **KVL** and **KCL**;
  - 6: **end for**
  - 7: Approximate the equivalent edge function by interpolation based on  $\{(\zeta_{pq}, \bar{\mu}_{pq})\}$ .
-

# EQUIVALENT EDGE FUNCTION

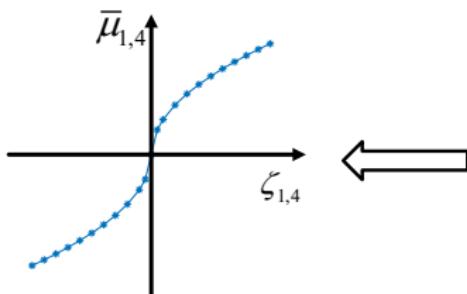


Original underlying graph.

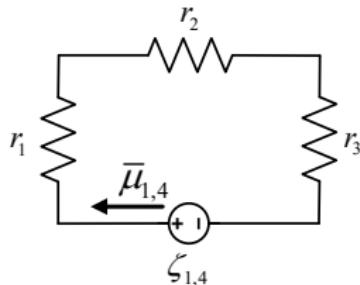


Obtain the corresponding resistive circuit, and add a voltage source.

Obtain a finite set  $\{\zeta_{1,4}\}$  by sampling the interval  $[-N, N]$ .

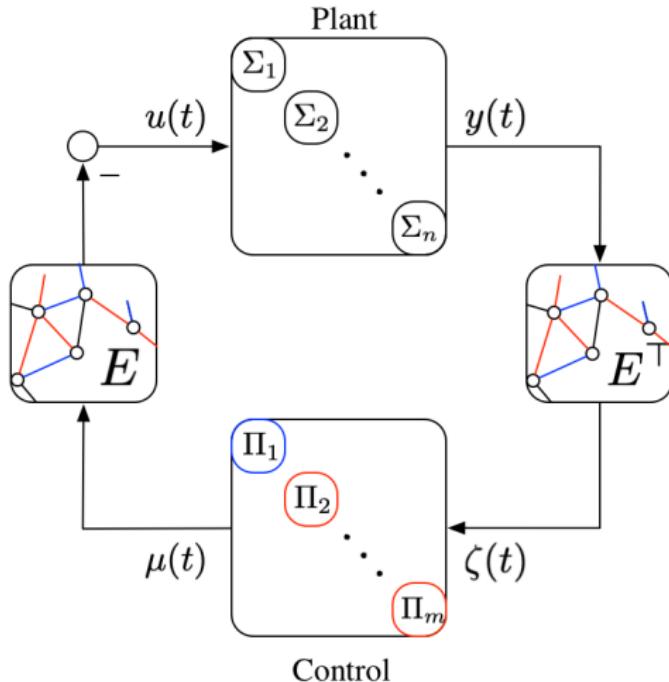


Approximate the equivalent edge function by interpolation based on  $\{(\zeta_{1,4}, \bar{\mu}_{1,4})\}$ .



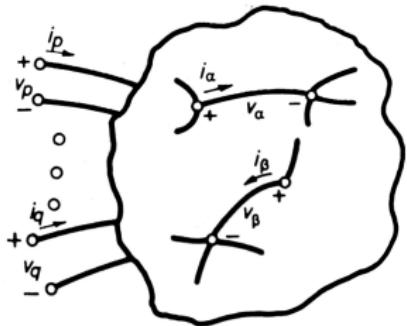
Calculating  $\bar{\mu}_{1,4}$  corresponding to each  $\zeta_{1,4}$ .

# NON-POSITIVE NETWORKS



How do we analyze networks with positive and negative edges?

# TELLEGEN'S THEOREM AND PASSIVITY



**Telegén's Theorem** states that the instantaneous powers in all elements in a circuit sum to the power injected into the network,

$$\sum_p i_p v_p = \sum_\alpha i_\alpha v_\alpha$$

## Proposition

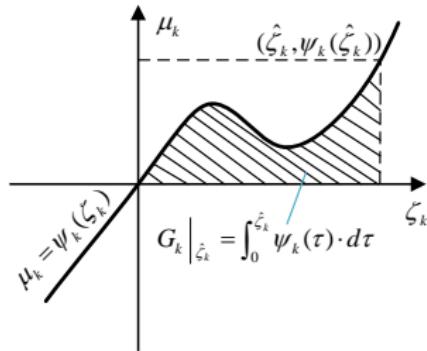
Arbitrary interconnections of inductors and capacitors with passive resistors verify the energy-balance inequality

$$\int_0^t i_p^T(\tau) v_p(\tau) d\tau \geq E[\varphi_L(t), q_C(t)] - E[\varphi_L(0), q(0)].$$

## COCONTENT FUNCTION

The **cocontent**<sup>1</sup> of a nonlinear resistor (edge  $k$ ) when its tension  $\zeta_k$  is specified, is defined as

$$G_k|_{\zeta_k} = \int_0^{\zeta_k} \psi_k(\tau) \cdot d\tau.$$



Cocontent is area under curve. Content  $(G_k^*)|_{\mu_k}$  is area above curve.

$$\text{Total Power: } G_k|_{\zeta_k} + G_k^*|_{\mu_k} = \mu_k \zeta_k$$

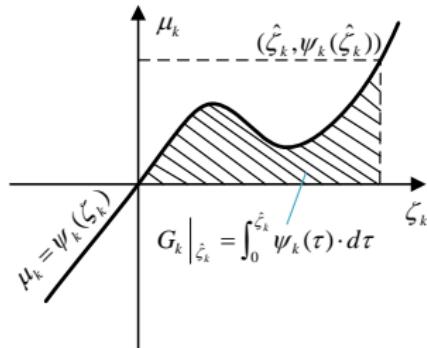
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<sup>1</sup> M. Parodi and M. Storace, Linear and Nonlinear Circuits: Basic and Advanced Concepts. Cham, Switzerland: Springer International Publishing, 2018.

## COCONTENT FUNCTION

The **cocontent**<sup>1</sup> of a nonlinear resistor (edge  $k$ ) when its tension  $\zeta_k$  is specified, is defined as

$$G_k|_{\zeta_k} = \int_0^{\zeta_k} \psi_k(\tau) \cdot d\tau.$$

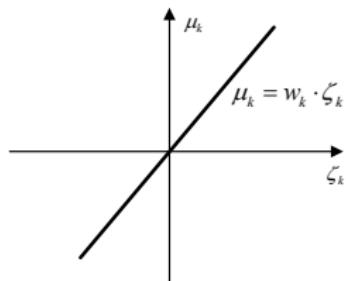


Cocontent of the network denoted as  $\mathbf{G} = \sum_{k=1}^{|\mathcal{E}|} G_k|_{\zeta_k}$ .

<sup>1</sup> M. Parodi and M. Storace, *Linear and Nonlinear Circuits: Basic and Advanced Concepts*. Cham, Switzerland: Springer International Publishing, 2018.

# COCONTENT FUNCTION

**Example 1:** A resistor with constant resistance.



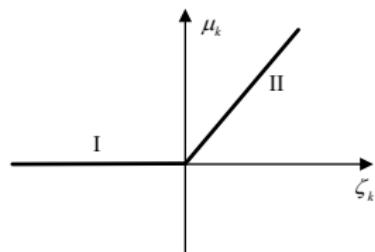
Edge function:

$$\mu_k = w_k \cdot \zeta_k.$$

Cocontent function:

$$\begin{aligned} G_k|_{\zeta_k} &= \int_0^{\zeta_k} w_k \cdot \tau d\tau \\ &= \frac{1}{2} w_k \zeta_k^2. \end{aligned}$$

**Example 2:** An ideal diode.

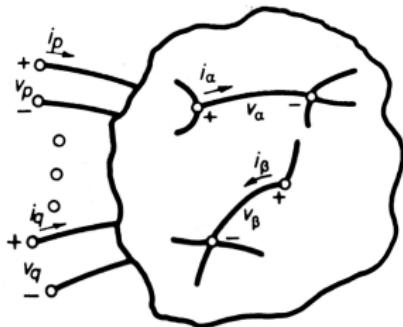


Edge function:

$$\mu_k = \begin{cases} 0, & \text{if } \zeta_k < 0. \\ w_k \cdot \zeta_k, & \text{if } \zeta_k \geq 0. \end{cases}$$

Cocontent function:

$$G_k|_{\zeta_k} = \begin{cases} 0, & \text{if } \zeta_k < 0. \\ \frac{1}{2} w_k \zeta_k^2, & \text{if } \zeta_k \geq 0. \end{cases}$$



**Telegén's Theorem** states that the instantaneous powers in all elements in a circuit sum to the power injected into the network,

$$\sum_p i_p v_p = \sum_\alpha i_\alpha v_\alpha$$

## Proposition

Arbitrary interconnections of passive capacitors with convex energy function  $E_C(q_C)$ , voltage-controlled resistors, and sources, satisfy the power-balance inequality

$$\int_0^t i_p^T(\tau) \frac{dv_p}{d\tau}(\tau) d\tau \geq \mathbf{G}.$$

<sup>1</sup> D. Jeltsema, R. Ortega, and J. Scherpen, *On Passivity and Power-Balance Inequalities of Nonlinear RLC Circuits*, IEEE Tran. Circuits and Systems 50(9):1174-1179, 2003.

# MINIMUM OF COCONTENT FUNCTION

## Lemma

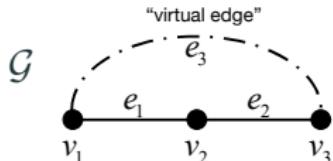
Consider a strictly positive network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_>)$ , and the augmented graph  $\bar{\mathcal{G}} = (\mathcal{V}, \bar{\mathcal{E}})$  obtained by adding the virtual edge  $\bar{k}$  ( $\bar{\mathcal{E}} = \mathcal{E}_> \cup \{\bar{k}\}$ ). Suppose for each  $k \in \mathcal{E}_>$ ,  $\mu_k = \psi_k(\zeta_k)$  is monotonically increasing. For any fixed  $\zeta_{\bar{k}}$ , if there exists  $\bar{\zeta}^0, \bar{\mu}^0, \bar{y}^0$ , such that

$$\begin{cases} E(\bar{\mathcal{G}})^T \bar{y}^0 &= \bar{\zeta}^0 \quad \text{KVL} \\ \bar{\mu}^0 &= \Psi(\bar{\zeta}^0) \\ E(\bar{\mathcal{G}})\bar{\mu}^0 &= \mathbf{0} \quad \text{KCL}, \end{cases}$$

then  $\sum_{k=1}^{|\mathcal{E}|} G_k|_{\zeta_k}$  reaches its minimum at  $(\bar{\zeta}^0, \bar{\mu}^0, \bar{y}^0)$ .

consequence of Tellegen's Theorem and Minimum Heat Theorem

# MINIMUM OF COCONTENT FUNCTION



$$E = \begin{bmatrix} 1 & 0 \\ -1 & 1 \\ 0 & -1 \end{bmatrix} \quad \bar{E} = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \\ 0 & -1 & -1 \end{bmatrix}$$

## Edge Functions

$$\begin{aligned}\mu_1 &= \frac{1}{2}\zeta_1 \\ \mu_2 &= \zeta_2\end{aligned}$$

## Network Co-Content

$$\begin{aligned}\mathbf{G} &= \int_0^{\zeta_1} \frac{1}{2}\tau \cdot d\tau + \int_0^{\zeta_2} \tau \cdot d\tau \\ &= \frac{1}{4}\zeta_1^2 + \frac{1}{2}\zeta_2^2\end{aligned}$$

## Example 1

$$\zeta_1 = 2 \quad \text{potential drop between } v_1-v_2$$

$$\zeta_2 = 1 \quad \text{potential drop between } v_2-v_3$$

$$\zeta_3 = 3 \quad \text{potential drop between } v_1-v_3$$

network equations

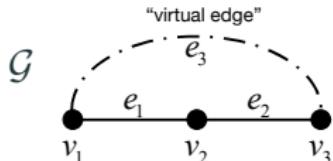
$$\bar{E}^T y = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} 2 \\ 0 \\ -1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix} = \bar{\zeta}$$

$$\mu = \Psi(\zeta) = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$\bar{E}\bar{\mu} = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \\ 0 & -1 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathbf{G} = 1.5$$

# MINIMUM OF COCONTENT FUNCTION



$$E = \begin{bmatrix} 1 & 0 \\ -1 & 1 \\ 0 & -1 \end{bmatrix} \quad \bar{E} = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \\ 0 & -1 & -1 \end{bmatrix}$$

## Edge Functions

$$\mu_1 = \frac{1}{2}\zeta_1$$

$$\mu_2 = \zeta_2$$

## Network Co-Content

$$\begin{aligned} \mathbf{G} &= \int_0^{\zeta_1} \frac{1}{2}\tau \cdot d\tau + \int_0^{\zeta_2} \tau \cdot d\tau \\ &= \frac{1}{4}\zeta_1^2 + \frac{1}{2}\zeta_2^2 \end{aligned}$$

## Example 2

$$\zeta_1 = 1 \quad \text{potential drop between } v_1-v_2$$

$$\zeta_2 = 2 \quad \text{potential drop between } v_2-v_3$$

$$\zeta_3 = 3 \quad \text{potential drop between } v_1-v_3$$

network equations

$$\bar{E}^T y = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ -2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

$$\bar{\mu} = \Psi(\zeta) = \begin{bmatrix} \frac{1}{2} \\ 2 \\ 3 \end{bmatrix}$$

$$\bar{E}\bar{\mu} = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \\ 0 & -1 & -1 \end{bmatrix} \begin{bmatrix} \frac{1}{2} \\ 2 \\ \star \end{bmatrix} \neq \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathbf{G} = 2.25 > 1.5$$

### Proposition: Minimum of Cocontent

Consider a strictly positive network system  $(\mathcal{G}, \Sigma, \Pi)$  with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E}_>)$ , and all edge functions are monotonically increasing. Identify  $p, q \in \mathcal{V}$  as two terminals of interest, and the tension between nodes  $p$  and  $q$  is specified as  $\zeta_{pq}$ . If the equivalent edge function between nodes  $p$  and  $q$  exists, then the minimum cocontent of the network system  $(\mathcal{G}, \Sigma, \Pi)$  is the cocontent of the equivalent edge function between nodes  $p$  and  $q$ , denoted as  $\min \mathbf{G}|_{\zeta_{pq}} = G_{pq}|_{\zeta_{pq}}$ .

Establishes connection between  
equivalent edge functions and cocontent

# CONVERGENCE ANALYSIS OF NONLINEAR INTEGRATORS

Nonlinear integrator dynamics:

$$\Sigma_i : \dot{x}_i(t) = \gamma_i(u_i(t)), \quad y_i(t) = x_i(t), \quad i \in \mathcal{V}.$$

where  $\gamma_i(\cdot)$  satisfies  $u_i \cdot \gamma_i(u_i) \geq 0$ , and equality holds if and only if  $u_i = 0$ .

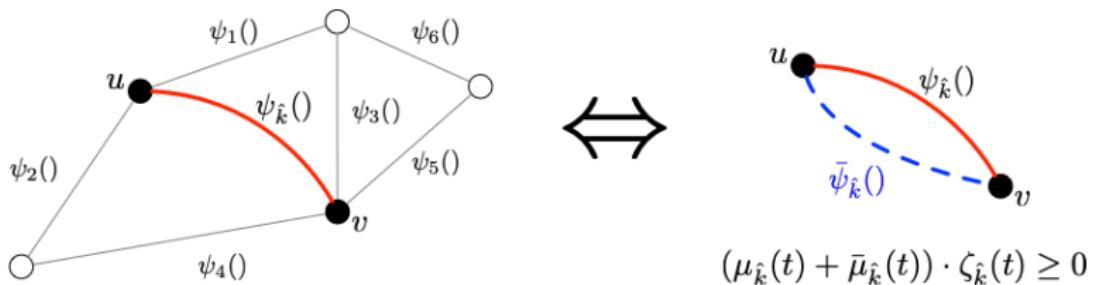
## Theorem

Consider a signed network system  $(\mathcal{G}, \Sigma, \Pi)$  of nonlinear integrators with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ . Suppose there is only one non-strictly positive edge  $\hat{k}$  in  $\mathcal{E}$ , with edge function  $\mu_{\hat{k}}(t) = \psi_{\hat{k}}(\zeta_{\hat{k}}(t))$ , and  $\psi_{\hat{k}}(0) = 0$ . Furthermore,  $\forall k \in \mathcal{E}_{>}$ ,  $\psi_k(\cdot)$  is monotonically increasing. Identify  $p, q \in \mathcal{V}$ , which are connected by edge  $\hat{k}$ , as the two terminals of the strictly positive subnetwork system  $(\mathcal{G}_{>}, \Sigma, \bar{\Pi})$ . If the equivalent edge function  $\bar{\mu}_{pq}(t) = \bar{\psi}_{pq}(\zeta_{\hat{k}}(t))$  between  $p$  and  $q$  in  $(\mathcal{G}_{>}, \Sigma, \bar{\Pi})$  exists, and

$$(\mu_{\hat{k}}(t) + \bar{\mu}_{pq}(t)) \cdot \zeta_{\hat{k}}(t) \geq 0$$

holds for any  $\zeta_{\hat{k}}(t) \in \mathbb{R}$ , then  $\lim_{t \rightarrow \infty} \mathbf{u}(t) = \mathbf{0}$ , and  $\lim_{t \rightarrow \infty} \mu_{\hat{k}}(t) + \bar{\mu}_{pq}(t) = 0$ .

# CONVERGENCE ANALYSIS OF NONLINEAR INTEGRATORS



parallel edges must be positive!

# CONVERGENCE ANALYSIS OF NONLINEAR INTEGRATORS

## PROOF

Let  $V(t) := G_{\hat{k}}|_{\zeta_{\hat{k}}(t)} + \mathbf{G}_{>}|_{\zeta_{\hat{k}}(t)}$  be the cocontent of the network system  $(\mathcal{G}, \Sigma, \Pi)$ , where  $G_{\hat{k}}|_{\zeta_{\hat{k}}(t)}$  is the cocontent of edge  $\hat{k}$ , and  $\mathbf{G}_{>}|_{\zeta_{\hat{k}}(t)}$  is the cocontent of the subnetwork system  $(\mathcal{G}_{>}, \Sigma, \bar{\Pi})$  for a fixed value of  $\zeta_{\hat{k}}(t)$ . Then

$$V(t) \stackrel{(a)}{\geq} G_{\hat{k}}|_{\zeta_{\hat{k}}(t)} + G_{pq}|_{\zeta_{\hat{k}}(t)} \stackrel{(b)}{\geq} 0,$$

where (b) follows from  $(\mu_{\hat{k}}(t) + \bar{\mu}_{pq}(t)) \cdot \zeta_{\hat{k}}(t) \geq 0$ .

As  $\dot{G}_k|_{\zeta(t)}(t) = \mu_k(t)\dot{\zeta}_k(t)$ ,  $\forall k \in \mathcal{E}$ , then

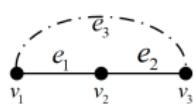
$$\dot{V}(t) = \boldsymbol{\mu}(t)^T \dot{\zeta}(t) = -\mathbf{u}(t)^T \dot{\mathbf{y}}(t) = -\sum_{i=1}^{|\mathcal{V}|} u_i(t) \cdot \gamma_i(u_i(t)) \leq 0.$$

From LaSalle's invariance principle,  $\lim_{t \rightarrow \infty} \mathbf{u}(t) = \mathbf{0}$ , and  $\lim_{t \rightarrow \infty} \mu_{\hat{k}}(t) + \bar{\mu}_{pq}(t) = 0$  follows the definition of equivalent edge function.

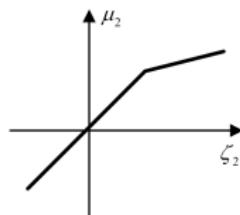
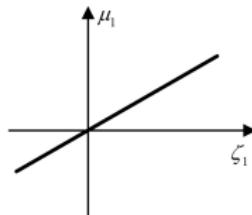
## Main Point

Cocontent function works as a Lyapunov Function

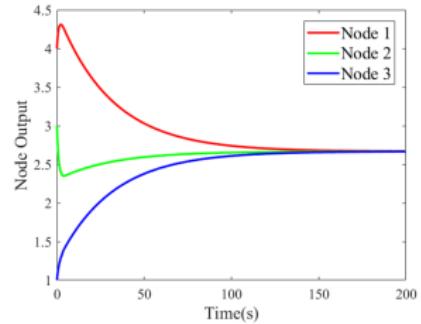
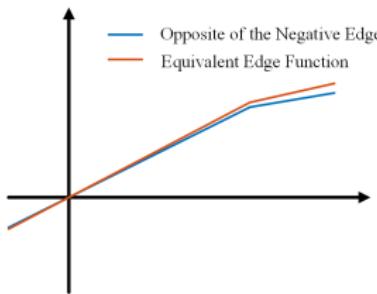
# CONVERGENCE ANALYSIS OF NONLINEAR INTEGRATORS



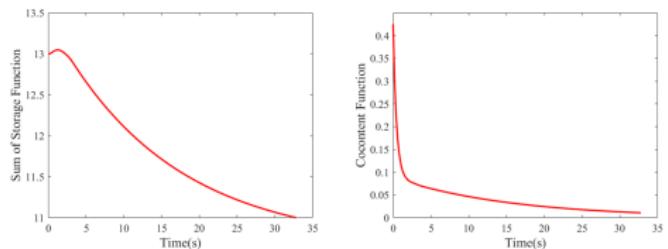
$$\bar{E} = \begin{bmatrix} 1 & 0 & 1 \\ -1 & 1 & 0 \\ 0 & -1 & -1 \end{bmatrix}$$



$$\mu_1 = 0.5\zeta_1 \quad \mu_2 = \begin{cases} \zeta_2, & \text{when } \zeta_2 < 1; \\ 0.2\zeta_2 + 0.8, & \text{when } \zeta_2 \geq 1. \end{cases}$$



Consensus!



Derivative of storage function  
is indefinite!

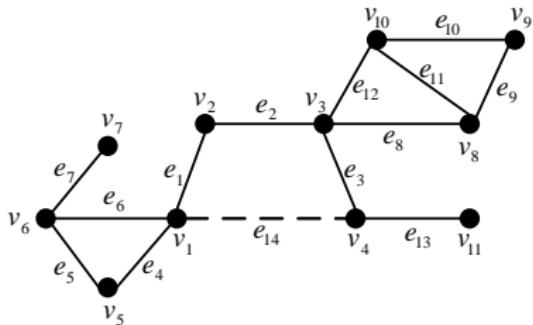
## Corollary: Agreement

Consider a signed network system  $(\mathcal{G}, \Sigma, \Pi)$  of nonlinear integrators with connected graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ . Suppose there is only one non-strictly positive edge  $\hat{k}$  in  $\mathcal{E}$ , with edge function  $\mu_{\hat{k}}(t) = \psi_{\hat{k}}(\zeta_{\hat{k}}(t))$ , and  $\psi_{\hat{k}}(0) = 0$ . Furthermore,  $\forall k \in \mathcal{E}_>$ ,  $\psi_k(\cdot)$  is monotonically increasing. Identify  $p, q \in \mathcal{V}$ , which are connected by edge  $\hat{k}$ , as the two terminals of the strictly positive subnetwork system  $(\mathcal{G}_>, \Sigma, \bar{\Pi})$ . If the equivalent edge function  $\bar{\mu}_{pq}(t) = \bar{\psi}_{pq}(\zeta_{\hat{k}}(t))$  between  $p$  and  $q$  in  $(\mathcal{G}_>, \Sigma, \bar{\Pi})$  exists, and

$$(\mu_{\hat{k}}(t) + \bar{\mu}_{pq}(t)) \cdot \zeta_{\hat{k}}(t) \geq 0$$

holds for any  $\zeta_{\hat{k}}(t) \in \mathbb{R}$ , and  $(\mu_{\hat{k}}(t) + \bar{\mu}_{pq}(t)) \cdot \zeta_{\hat{k}}(t) = 0$  if and only if  $\zeta_{\hat{k}}(t) = 0$ , then  $\lim_{t \rightarrow \infty} \zeta(t) = \mathbf{0}$ , and  $\lim_{t \rightarrow \infty} \mathbf{y}(t) = \beta \mathbf{1}$ ,  $\beta \in \mathbb{R}$ .

## SIMULATION RESULTS



- The original network consists of 11 nodes and 13 edges, and  $e_{14}$  is introduced by an attacker.
- The original 13 edges are **strictly positive**, with edge functions described by

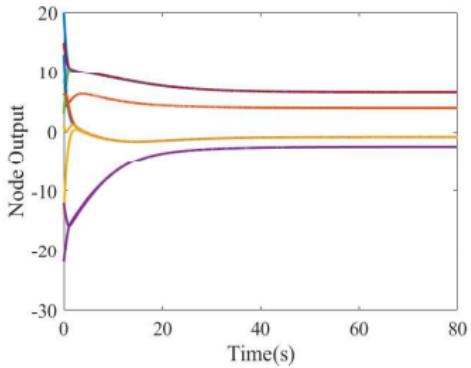
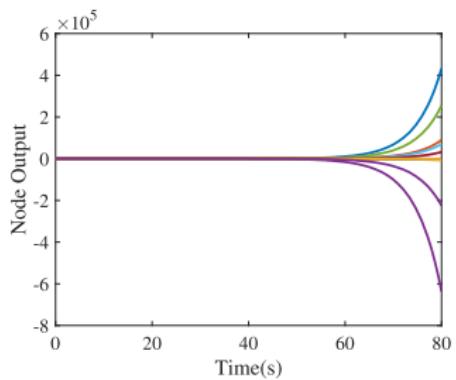
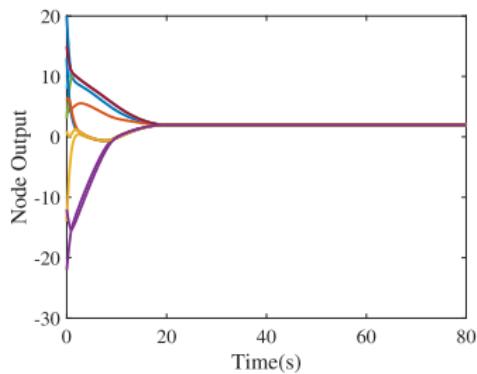
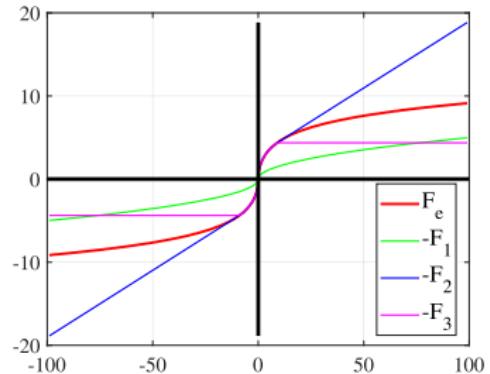
$$\mu_k(t) = w_k \cdot \text{sign}(\zeta_k(t)) \cdot |\zeta_k(t)|^{\alpha_k},$$

and  $\mathbf{w} = (3, 2, 4, 1, 2, 1, 3, 2, 2, 1, 1, 1, 2)^T$ ,

$\boldsymbol{\alpha} = (0.4, 0.5, 0.2, 0.8, 0.4, 0.4, 0.5, 0.5, 0.5, 0.6, 0.8, 0.2, 0.5)^T$ .

- **Only one cycle** contains  $e_{14}$ , i.e., the cycle consisting of nodes  $v_1, v_2, v_3$  and  $v_4$ , and edges  $e_1, e_2, e_3$  and  $e_{14}$ .
- We consider three **strictly negative** candidate functions for  $e_{14}$ .

# SIMULATION RESULTS



## CONCLUSION

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- We generalize the definition of signed linear networks to graphs with **nonlinear functions** on the edges.
- Convergence analysis for positive and strictly positive networks. Spanning subgraphs of strictly positive edges are required for consensus.
- For networks comprised of nonlinear integrator agents, we show a connection to notions from electrical circuit theory and the equivalent circuit model to derive convergence results for networks with non-positive edges. We also propose an algorithm for constructing equivalent edge functions.

Future research directions include:

- Convergence analysis of general MEIP nodes in a signed network of more than one non-strictly positive edges.
- Real-world applications.

## References

### Signed Linear Networks

D. Zelazo and M. Bürger, *On the definiteness of the weighted Laplacian and its connection to effective resistance*, IEEE Conference on Decision and Control, 2014.

D. Zelazo and M. Bürger, *On the robustness of uncertain consensus networks*, IEEE Control of Network Systems, 4(2):170–178, 2017.

### Signed Nonlinear Networks

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# THANK YOU!

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