

A PASSIVITY ANALYSIS FOR NONLINEAR CONSENSUS ON DIGRAPHS

2025 IAAC CONTROL CONFERENCE

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MULTI-AGENT NETWORKS



Applications:

- ▶ Formation flying
- ▶ Power grid
- ▶ Automated transportation networks...

MULTI-AGENT NETWORKS



Applications:

- ▶ Formation flying
- ▶ Power grid
- ▶ Automated transportation networks...
- ▶ Fundamental problem: **Output consensus**

Multi-agent networks: A group of SISO agents Σ_i interact over a graph \mathcal{G} with SISO edge controllers Π_k :

$$\Sigma_i : \begin{cases} \dot{x}_i = f_i(x_i, u_i) \\ y_i = h_i(x_i, u_i) \end{cases}, i \in [1, n] \quad \Pi_k : \begin{cases} \dot{\eta}_k = \phi_k(\eta_k, \zeta_k) \\ \mu_k = \psi_k(\eta_k, \zeta_k) \end{cases}, k \in [1, m]$$

MULTI-AGENT NETWORKS AND CONSENSUS

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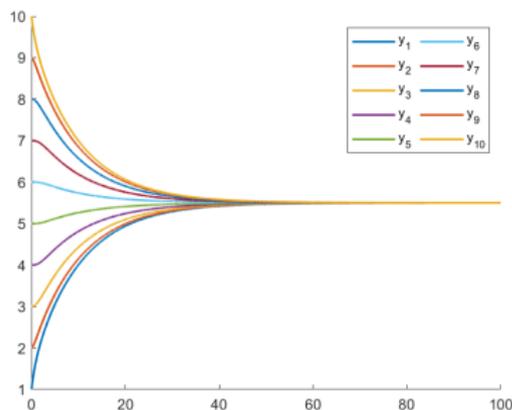
Output consensus problem:

Design distributed Π_k s, such that

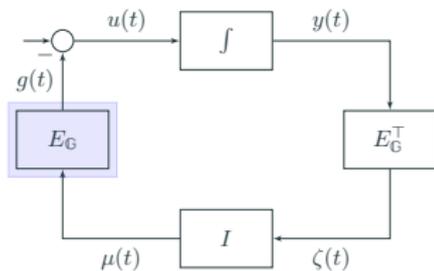
$$\lim_{t \rightarrow \infty} (y_i(t) - y_j(t)) = 0, \forall i, j$$

$$\Leftrightarrow \lim_{t \rightarrow \infty} y(t) \in \mathcal{S}$$

where $\mathcal{S} = \text{span}(\mathbf{1})$ denotes the **agreement space**.



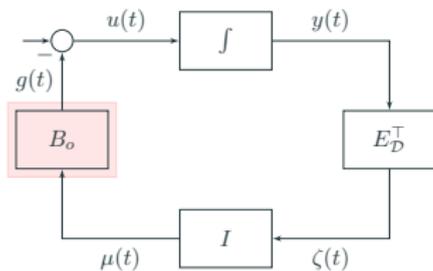
Graph Topologies Matter!



Undirected Networks

$$\dot{x}(t) = -E_G E_G^\top x(t)$$

$$y(t) = x(t)$$

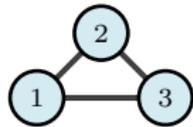
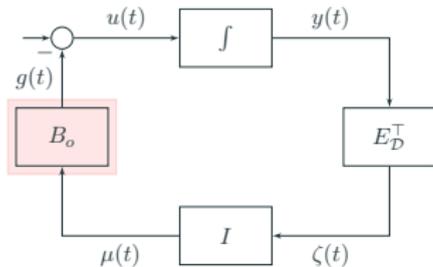
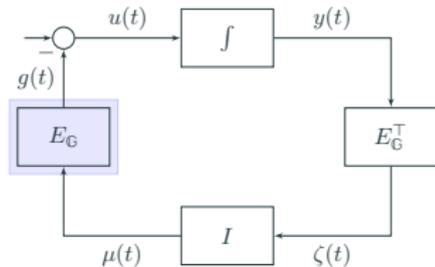


Directed Networks

$$\dot{x}(t) = -B_o E_D^\top x(t)$$

$$y(t) = x(t)$$

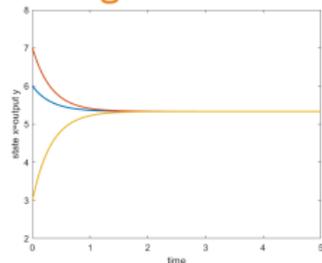
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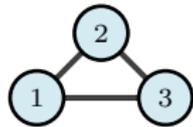
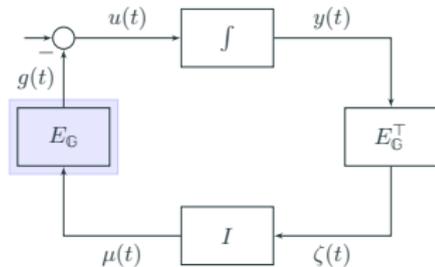
Undirected \mathbb{G}

$$\lim_{t \rightarrow \infty} y(t) = \frac{1}{n} \mathbf{1}_n \mathbf{1}_n^\top x(0)$$

Average consensus



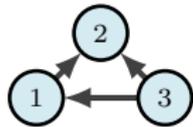
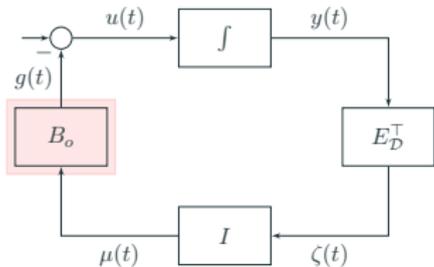
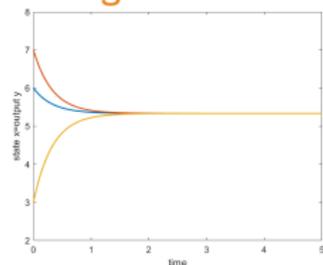
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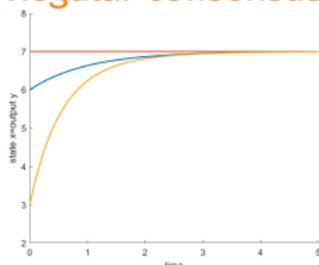
Average consensus



Directed \mathcal{D}

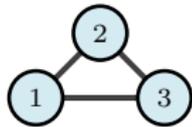
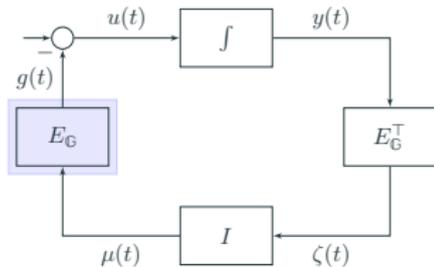
$$\lim_{t \rightarrow \infty} y(t) = (q_1^\top x(0)) \mathbf{1}_n$$

Regular consensus



GRAPHS AND LINEAR CONSENSUS PROTOCOLS

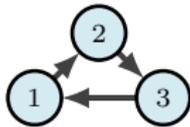
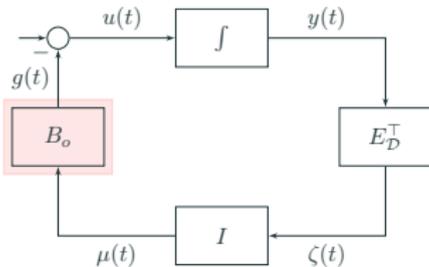
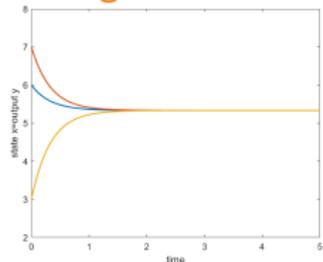
Graph Topologies Matter!



Undirected \mathbb{G}

$$\lim_{t \rightarrow \infty} y(t) = \frac{1}{n} \mathbf{1}_n \mathbf{1}_n^{\top} x(0)$$

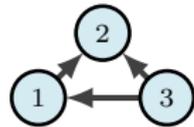
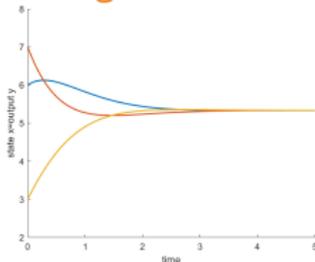
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Balanced \mathcal{D}_b

$$\lim_{t \rightarrow \infty} y(t) = \frac{1}{n} \mathbf{1}_n \mathbf{1}_n^{\top} x(0)$$

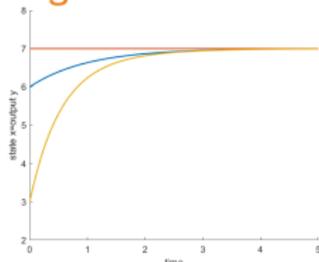
Average consensus



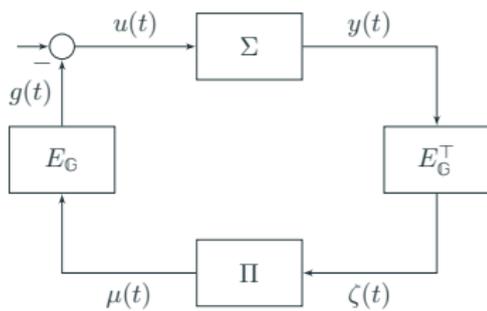
Directed \mathcal{D}

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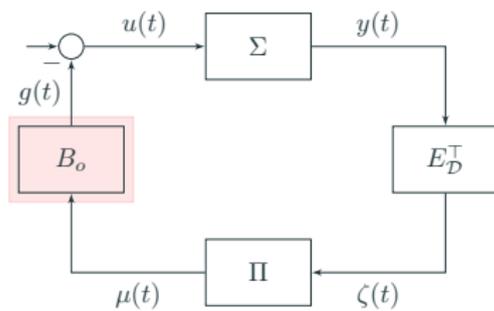
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NETWORK SYSTEMS AND PASSIVITY

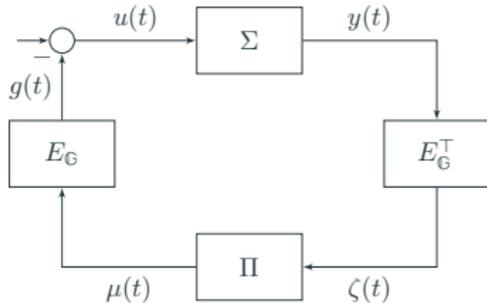


Undirected $(\Sigma, \Pi, \mathcal{G})_E$



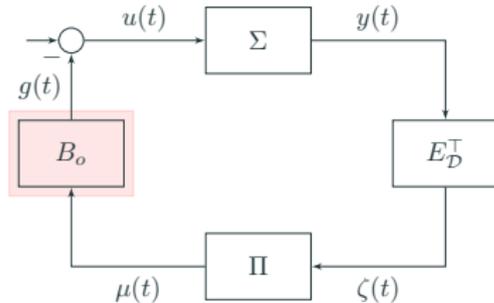
Directed $(\Sigma, \Pi, \mathcal{D})_{B_o}$

NETWORK SYSTEMS AND PASSIVITY



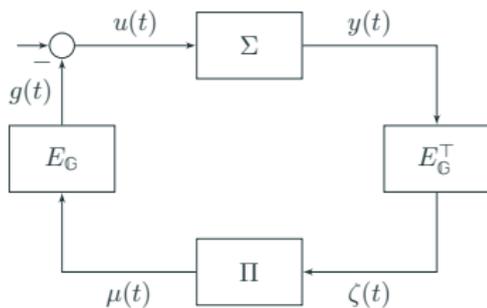
Undirected $(\Sigma, \Pi, \mathcal{G})_E$

► **Symmetric operator** $E_G \Pi E_G^\top$



Directed $(\Sigma, \Pi, \mathcal{D})_{B_o}$

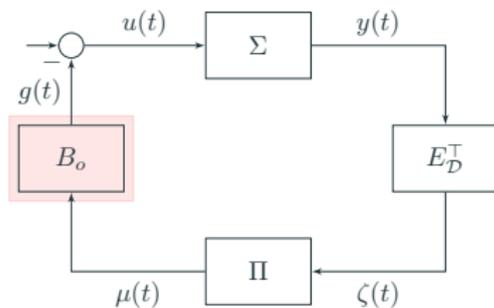
NETWORK SYSTEMS AND PASSIVITY



Undirected $(\Sigma, \Pi, \mathcal{G})_E$

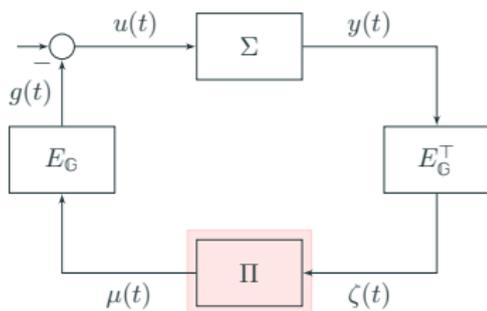
► **Symmetric operator** $E_G \Pi E_G^\top$

► **Passivity Analysis** ✓

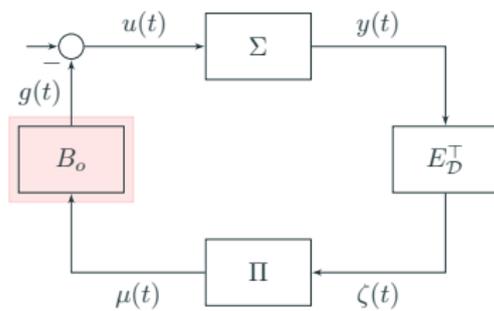


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NETWORK SYSTEMS AND PASSIVITY



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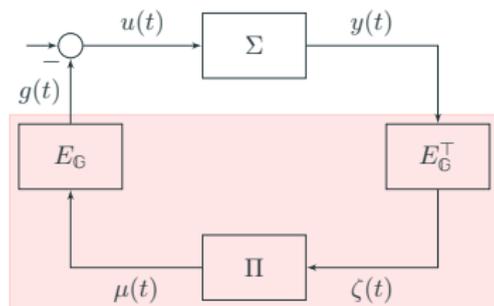
Directed $(\Sigma, \Pi, \mathcal{D})_{B_o}$

► **Symmetric operator** $E_{\mathcal{G}}\Pi E_{\mathcal{G}}^{\top}$

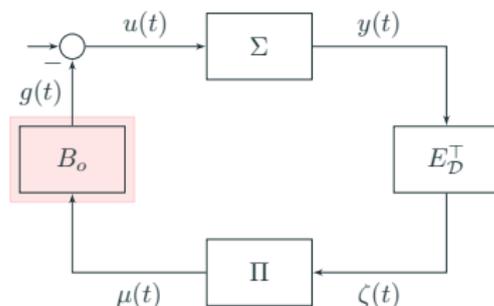
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Passive Π $\mu^{\top}(t)\zeta(t) \geq \dot{V}(\eta(t))$

NETWORK SYSTEMS AND PASSIVITY



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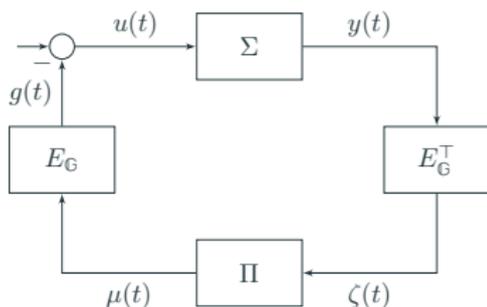
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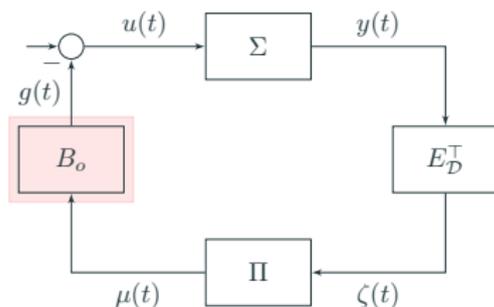
⇒ **Passive** $E_{\mathcal{G}}\Pi E_{\mathcal{G}}^{\top}$

$g^{\top}(t)y(t) = \mu^{\top}(t)E_{\mathcal{G}}^{\top}y(t) = \mu^{\top}(t)\zeta(t)$

NETWORK SYSTEMS AND PASSIVITY



Undirected $(\Sigma, \Pi, \mathcal{G})_E$



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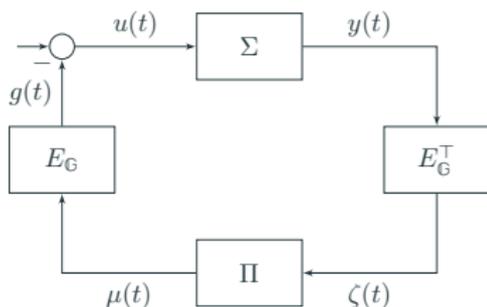
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- A decoupled analysis
- Convergence, stability

NETWORK SYSTEMS AND PASSIVITY



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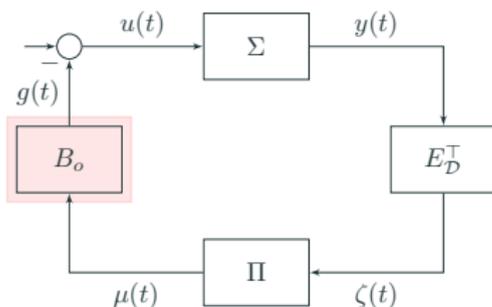
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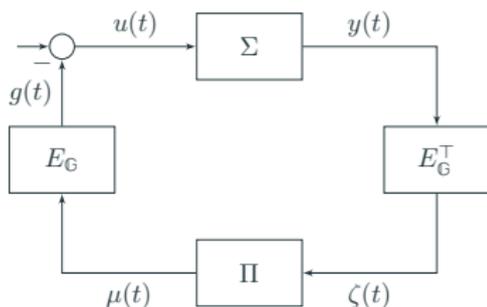
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Directed $(\Sigma, \Pi, \mathcal{D})_{B_o}$

► **Asymmetric operator** $B_o \Pi E_D^\top$

NETWORK SYSTEMS AND PASSIVITY



Undirected $(\Sigma, \Pi, \mathcal{G})_E$

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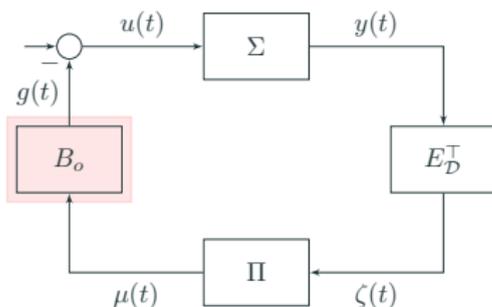
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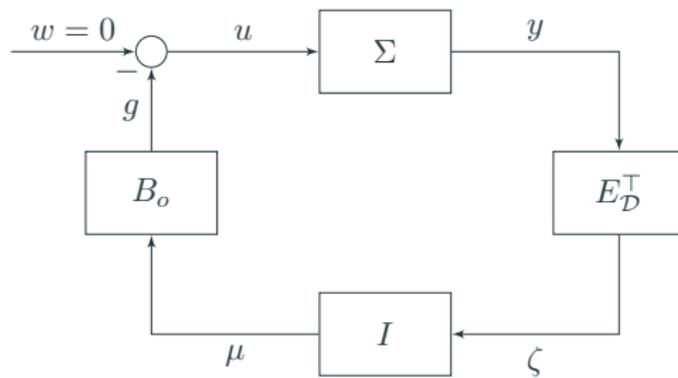


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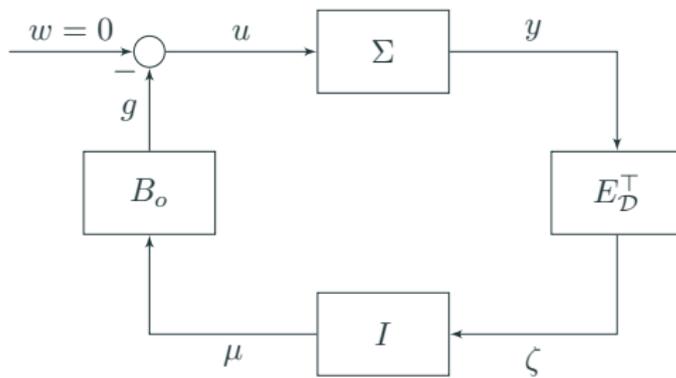
► **Passivity Analysis?**

CONTRIBUTION 1: PASSIVITY ANALYSIS FOR $(\Sigma, \Pi, \mathcal{D})_{B_o}$



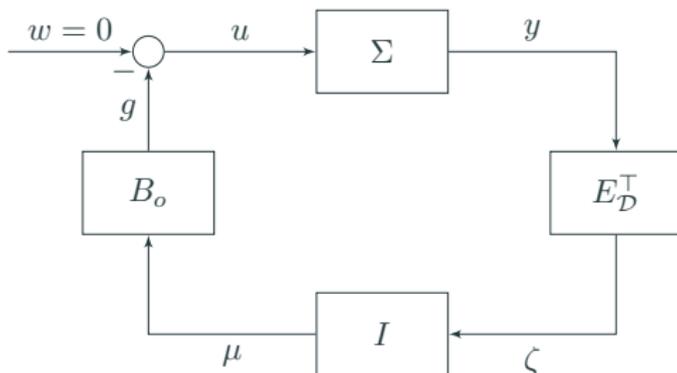
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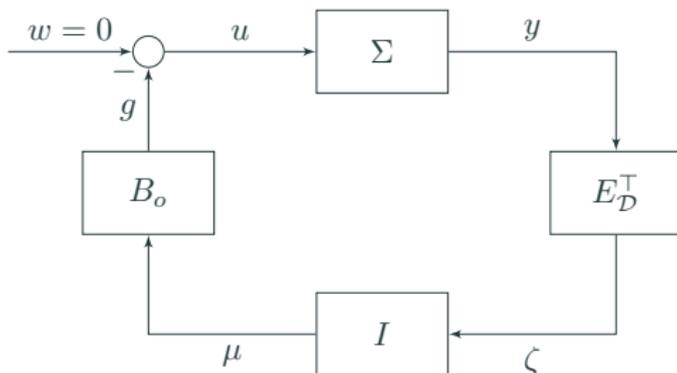
- ▶ General agents Σ , Linear controllers $\Pi = I : \mu = \zeta$
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CONTRIBUTION 1: PASSIVITY ANALYSIS FOR $(\Sigma, \Pi, \mathcal{D})_{B_o}$



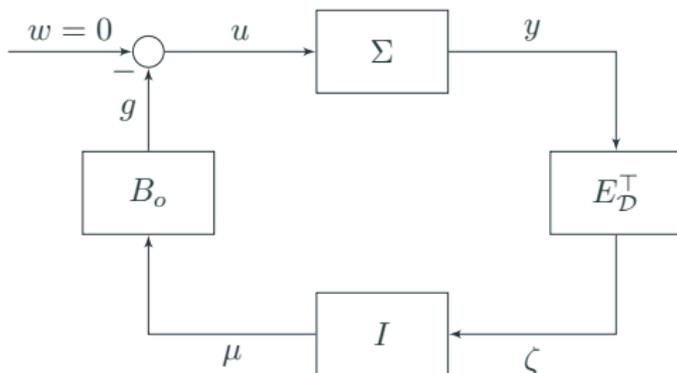
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CONTRIBUTION 1: PASSIVITY ANALYSIS FOR $(\Sigma, \Pi, \mathcal{D})_{B_o}$



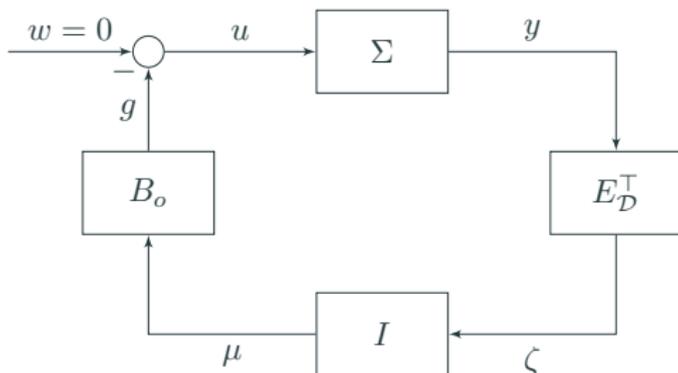
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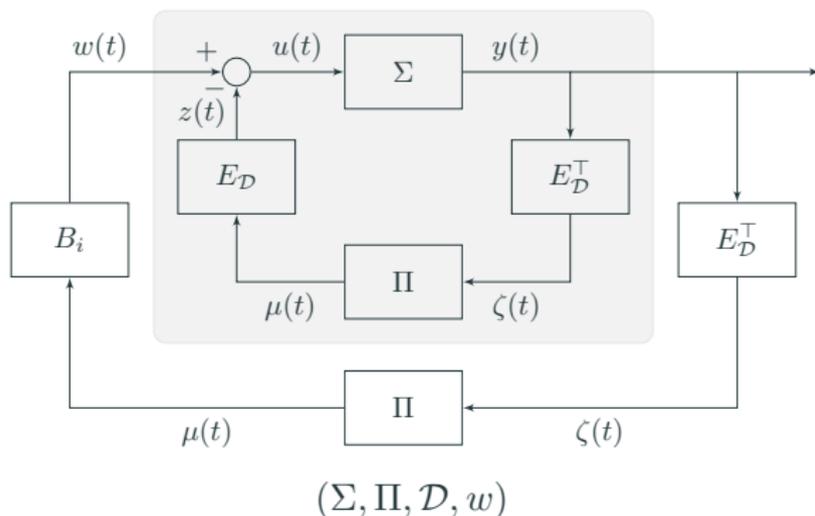
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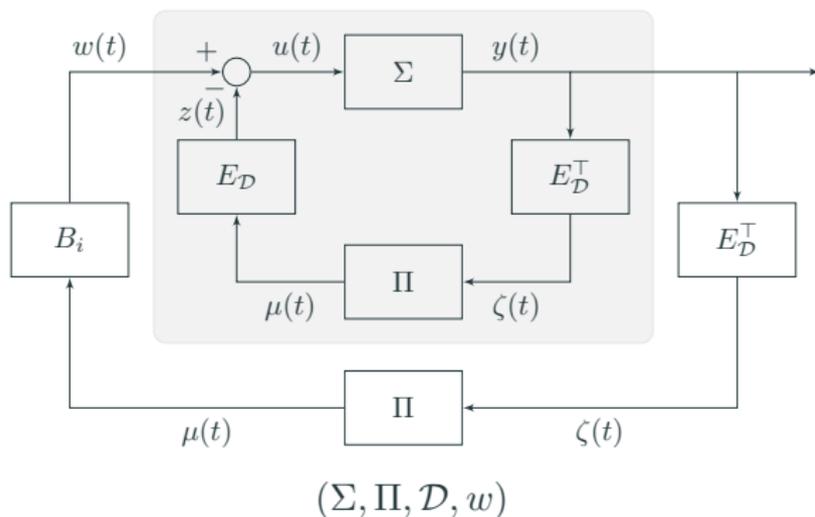
- ▶ General agents Σ , Linear controllers $\Pi = I : \mu = \zeta$
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 - **Balanced** \mathcal{D} : Passive $\xrightarrow{\text{passive } \Sigma}$ passivity analysis \checkmark
 - **General** \mathcal{D} : Not Passive
- ▶ For **general** controller dynamics Π : the operator $B_o \Pi E_{\mathcal{D}}^{\top}$ **may not** be passive.

CONTRIBUTION 1: A GENERAL APPROACH FOR DIRECTED COUPLING



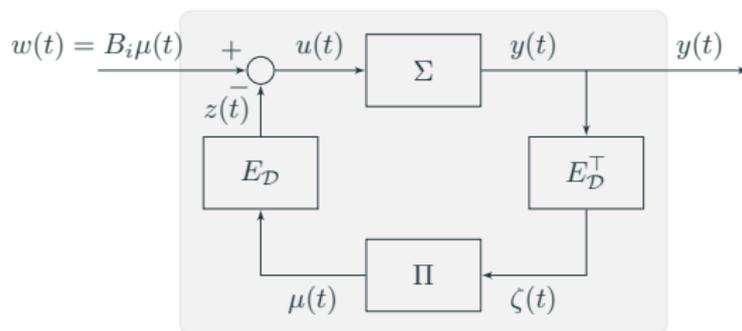
- ▶ Loop decomposition of $(\Sigma, \Pi, \mathcal{D})_{B_o}$: $E_{\mathcal{D}} = B_o + B_i$: **equivalence**
- ▶ First branch ($y \rightarrow z$): $E_{\mathcal{D}}\Pi E_{\mathcal{D}}^{\top}$ is passive, given passive Π .
- ▶ Second branch ($y \rightarrow w$): **external input** with directed information

CONTRIBUTION 1: A GENERAL APPROACH FOR DIRECTED COUPLING



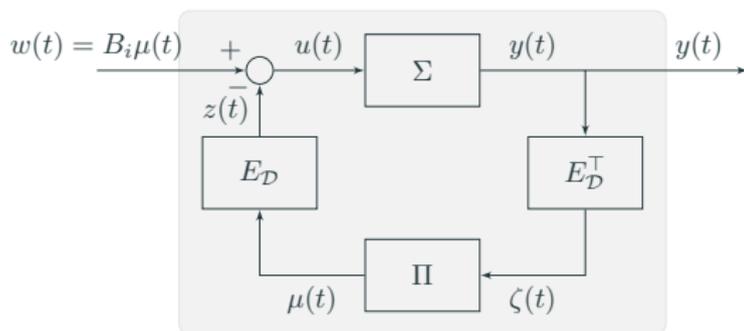
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- ▶ Second branch ($y \rightarrow w$): **external input** with directed information
- ▶ $(\Sigma, \Pi, \mathcal{D})_{B_o} \Leftrightarrow (\Sigma, \Pi, \mathcal{D}, w)$

OUTPUT CONSENSUS AND CONVERGENCE TO A SUBMANIFOLD



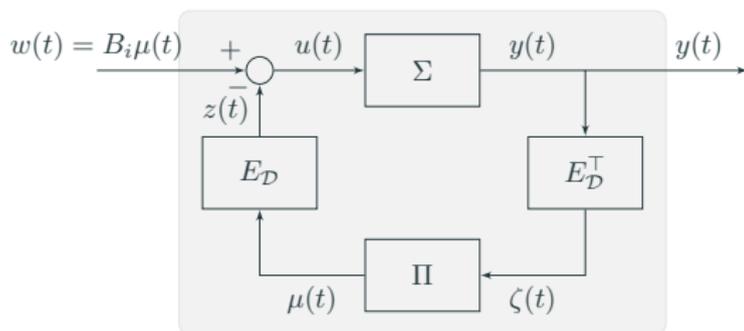
- Output consensus: $\lim_{t \rightarrow \infty} y(t) \in \text{span}(\mathbf{1}) = S$ **submanifold!**

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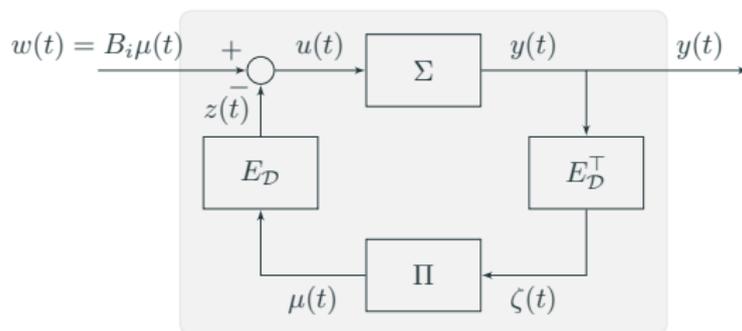
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- ▶ Converge to the agreement submanifold S : $\lim_{t \rightarrow \infty} \text{Proj}_{S^\perp}(y(t)) = 0$

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- ▶ **Connection to Passivity?**

OUTPUT CONSENSUS AND CONVERGENCE TO A SUBMANIFOLD



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- ▶ **Connection to Passivity?**

Passivity relations^[1] (point-wise)

$$u(t)^\top \text{Proj}_{S^\perp}(y(t)) \geq l \|u(t)\|^2 + e \|\text{Proj}_{S^\perp}(y(t))\|^2$$

$$z(t)^\top \text{Proj}_{S^\perp}(y(t)) \geq l \|z(t)\|^2 + e \|\text{Proj}_{S^\perp}(y(t))\|^2$$

CONTRIBUTION 2: PASSIVITY W.R.T. SUBMANIFOLD

$$\Lambda : \dot{x}(t) = f(x(t), u(t)), y(t) = h(x(t)), \quad f : (\mathbb{R}^n, \mathbb{R}^p) \rightarrow \mathbb{R}^n, \quad h : \mathbb{R}^n \rightarrow \mathbb{R}^p$$

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Recall **Classical Passivity**

- ▶ Storage Function $V : \mathbb{R}^n \rightarrow \mathbb{R}$
(1) $V(x) \geq 0$; (2) $V(0) = 0$
- ▶ Passivity indices $\exists \delta, \varepsilon \geq 0$

$$\textbf{Passive: } u^\top(t)y(t) \geq \dot{V}(x(t)) + \delta \|u(t)\|_2^2 + \varepsilon \|y(t)\|_2^2, \quad \forall t$$

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$$\textbf{Passive: } u^\top(t)y(t) \geq \dot{V}(x(t)) + \delta \|u(t)\|_2^2 + \varepsilon \|y(t)\|_2^2, \quad \forall t$$

Passivity w.r.t. Submanifold S [1]:

CONTRIBUTION 2: PASSIVITY W.R.T. SUBMANIFOLD

$$\Lambda : \dot{x}(t) = f(x(t), u(t)), y(t) = h(x(t)), \quad f : (\mathbb{R}^n, \mathbb{R}^p) \rightarrow \mathbb{R}^n, \quad h : \mathbb{R}^n \rightarrow \mathbb{R}^p$$

Recall **Classical Passivity**

- ▶ Storage Function $V : \mathbb{R}^n \rightarrow \mathbb{R}$
(1) $V(x) \geq 0$; (2) $V(0) = 0$
- ▶ Passivity indices $\exists \delta, \varepsilon \geq 0$

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Passivity w.r.t. Submanifold S [1]:

- ▶ **Constrained** Storage Function $Q : \mathbb{R}^n \rightarrow \mathbb{R}$
(1) $Q(x) \geq 0$; (2) $Q(x) = 0, \forall h(x) \in S$
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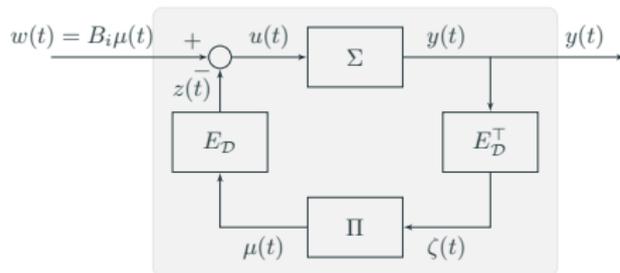
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$$\textbf{S-Passive: } u^\top(t) \text{Proj}_{S^\perp}(y(t)) \geq \dot{Q}(x(t)) + \varepsilon\|\text{Proj}_{S^\perp}(y(t))\|_2^2 + \delta\|u(t)\|_2^2, \quad \forall t$$

CONTRIBUTION 3: A PASSIVITY-BASED ANALYSIS

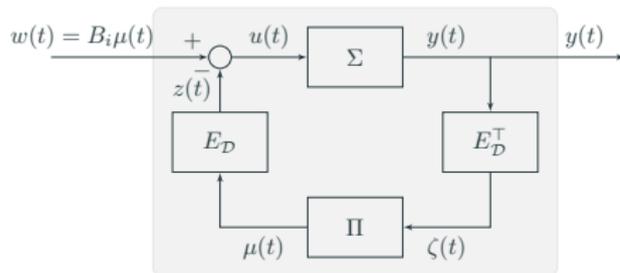


$(\Sigma^o, \Pi, \mathcal{D}, w)$

Integrator-like agents:

$$\Sigma_i^o : \begin{cases} \dot{x}_i(t) = u_i(t), \\ y_i(t) = h_i(x_i(t)), \end{cases} \quad i \in [1, n]$$

CONTRIBUTION 3: A PASSIVITY-BASED ANALYSIS



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Under what **passivity conditions** on Σ and Π does the **output** of the system converge to the agreement submanifold S ?

CONTRIBUTION 3: A PASSIVITY-BASED ANALYSIS

Directed networks $(\Sigma^o, \Pi, \mathcal{D}, w)$

Integrator-like agents $\Sigma_i^o : \dot{x}_i(t) = u_i(t), y_i(t) = h_i(x_i(t)), i \in [1, n]$

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Conditions

1. h_i s are continuously differentiable and monotone passive
2. h_i s have bounded derivatives, i.e., $\frac{\partial h_i(x)}{\partial x} \leq m$

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Passivity of agents Σ_i^o

$$\blacktriangleright u_i y_i \geq \dot{V}_i(x_i), V_i(x_i) = \int_0^{x_i} h(\sigma) d\sigma \quad [1]$$

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S-Passivity of Σ

$$\blacktriangleright \text{Constrained } Q(x) = \frac{1}{2} h^\top(x) (I - \frac{1}{|\mathcal{V}|} \mathbb{1}\mathbb{1}^\top) h(x)$$

$$\blacktriangleright M = \max(1, |1 - m|)$$

$$u^\top \text{Proj}_{S^\perp}(y) \geq \dot{Q}(x) - \frac{M}{2} \|u\|_2^2 - \frac{M}{2} \|\text{Proj}_{S^\perp}(y)\|_2^2$$

CONTRIBUTION 3: A PASSIVITY-BASED ANALYSIS

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Passivity of controllers Π_k

- ▶ $\zeta_k \mu_k \geq \dot{W}_k(\eta_k) + \alpha_k \mu_k^2 + \gamma_k \zeta_k^2, \alpha_k, \gamma_k > 0$
- ▶ $\alpha_k \gamma_k < 1/4, \alpha = \min(\alpha_k), \gamma = \min(\gamma_k)$

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Passivity of $E_{\mathcal{D}} \Pi E_{\mathcal{D}}^{\top}$

- ▶ $z^{\top} \text{Proj}_{S^{\perp}}(y) \geq \sum_{k=1}^p \dot{W}_k + \alpha \|\mu\|_2^2 + \gamma \lambda_2 \|\text{Proj}_{S^{\perp}}(y)\|_2^2$
- ▶ Passive $(z, \text{Proj}_{S^{\perp}}(y))$

CONTRIBUTION 3: A PASSIVITY-BASED ANALYSIS

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Integrator-like agents $\Sigma_i^o : \dot{x}_i(t) = u_i(t), y_i(t) = h_i(x_i(t)), i \in [1, n]$

Theorem

1. h_i are continuously differentiable and monotone passive
2. h_i have bounded derivatives, i.e., $\frac{\partial h_i(x)}{\partial x} \leq m$, $M = \max(1, |1 - m|)$
3. Controllers Π : $\zeta^\top \mu \geq \sum \dot{W}_k(\eta_k) + \alpha \|\mu\|_2^2 + \gamma \|\zeta\|_2^2$, $\alpha\gamma < \frac{1}{4}$

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4. $\alpha \geq \max(D_o) \frac{M}{2}$ and $\gamma\lambda_2 > \frac{M}{2}$

where $\max(D_o)$ and λ_2 denote the maximal out-degree and the algebraic connectivity of graph \mathcal{D} .

CONTRIBUTION 3: A PASSIVITY-BASED ANALYSIS

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Then, the network $(\Sigma^o, \Pi, \mathcal{D}, w)$ achieves **output consensus**.

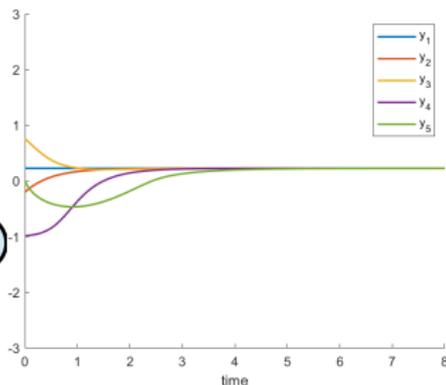
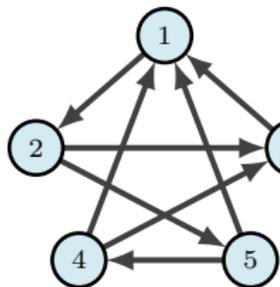
CASE STUDY: A HETEROGENEOUS NETWORK SYSTEM

► Systems

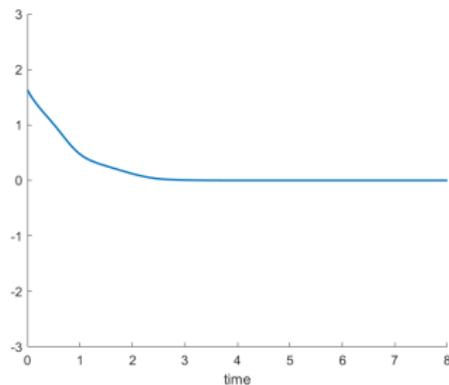
$$\Sigma^o : \dot{x}(t) = u(t), \quad y(t) = [x_1(t), x_2(t), \tanh(x_3(t)), \tanh(x_4(t)), \frac{x_5(t)}{1+|x_5(t)|}]^\top$$

► Parameters

- Constrained storage function: $Q(x) = \frac{1}{2}h^\top(x)(I - \frac{1}{|V|}\mathbb{1}\mathbb{1}^\top)h(x)$
- Algebraic connectivity: $\lambda_2 = 3$
- Maximal out-degree: $\max(D_o) = 2$



Outputs of agents



Evolution of $Q(x(t))$

CASE STUDY: A HETEROGENEOUS NETWORK SYSTEM

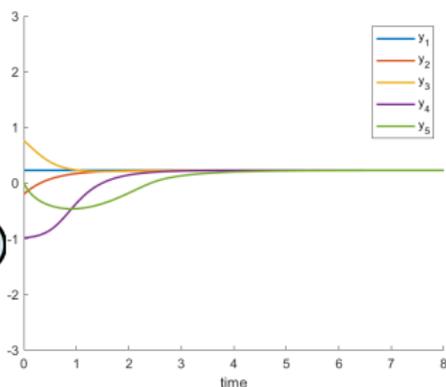
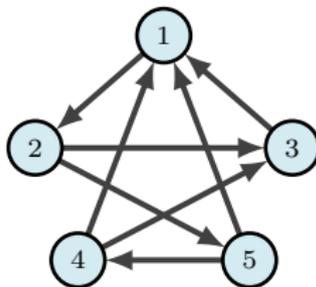
► Systems

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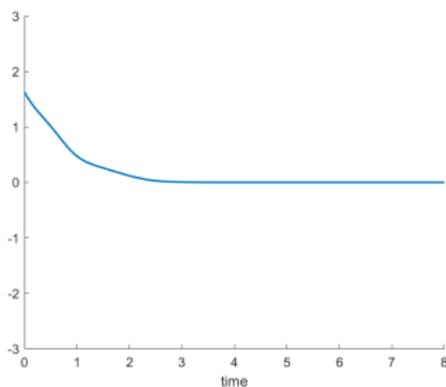
$$\Pi : \mu(t) = 2\zeta(t)$$

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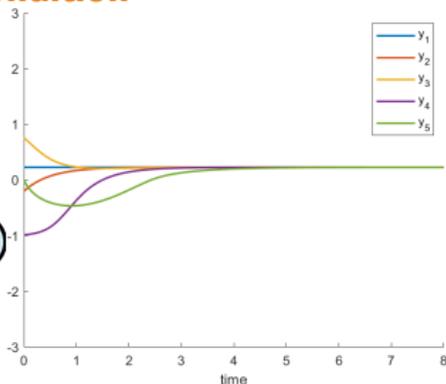
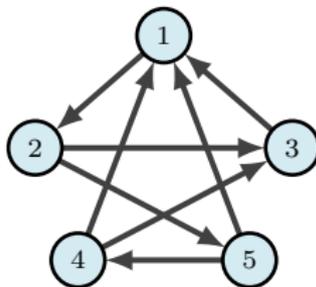
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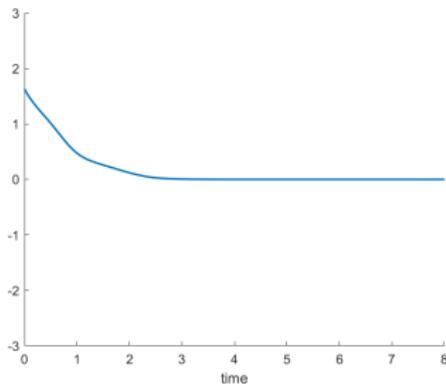
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► A sufficient condition



Outputs of agents



Evolution of $Q(x(t))$

Contributions:

- ▶ A general approach that enables a passivity analysis for the network systems with directed coupling.
- ▶ Constrained storage functions, Passivity w.r.t. submanifolds.
- ▶ A passivity-based analysis for integrator-like agents that interact over digraphs.

Future work:

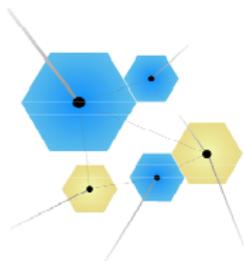
- ▶ complex dynamics, other passivity properties.
- ▶ A sufficient and necessary condition.

Thank-You!



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and Controls Lab