



Design of Cyber-Physical Systems Using Passivity/Dissipativity

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Outline

- **A Brief Introduction to Cyber-Physical Systems (CPS).
What are they and why are they important and interesting?**
- ***Studying CPS using Passivity and Dissipativity.***
- ***Background on Passivity and Passivity Indices.
Preserving passivity when interconnecting systems.***
- ***Passivation using Passivity Indices. Performance.***
- ***Application to Automotive and Human-Operator Applications.***
- ***System Approximations and Passivity/Dissipativity.***
- ***Switched/Hybrid Systems, Networked, Discrete Event Systems.***
- ***Systems with Symmetries and approximate Symmetries.***

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Cyber-Physical Systems (CPS)

-As computers become ever-faster and communication bandwidth ever-cheaper, computing and communication capabilities will be embedded in all types of objects and structures in the physical environment.

-***Cyber-physical systems (CPS)*** are physical, biological and engineered systems whose ***operations are monitored, coordinated, controlled and integrated by a computing and communication core.***

-This intimate coupling between the cyber and physical will be manifested from the nano-world to large-scale wide-area systems of systems. And at multiple time-scales.

-Applications with enormous societal impact and economic benefit will be created. ***Cyber-physical systems will transform how we interact with the physical world just like the Internet transformed how we interact with one another.***

-***We should care about CPS because our lives depend on them***

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Technological and Economic Drivers

- The decreasing cost of computation, networking, and sensing.
- A variety of social and economic forces will require us to use national infrastructures more efficiently.
- Environmental pressures will mandate the rapid introduction of technologies to improve energy efficiency and reduce pollution.
- As the national population ages, we will need to make more efficient use of our health care systems, ranging from facilities to medical data and information.

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PCAST Report



Leadership Under Challenge:
Information Technology R&D in a Competitive World
An Assessment of the Federal Networking and Information Technology
R&D Program
President's Council of Advisors on Science and Technology
August 2007


New Directions in Networking and Information Technology (NIT)

Recommendation: No 1 Funding Priority:
NIT Systems Connected with the Physical World

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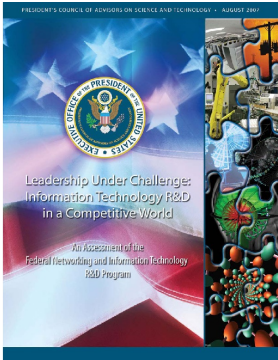
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PCAST Report



Chapter 4 in Report – Technical Priorities for
NIT R&D

1. NIT Systems Connected with the Physical World
2. Software
3. Data, Data Stores, and Data Streams
4. Networking
5. High End Computing
6. Cyber Security and Information Assurance
7. Human-Computer Interaction
8. NIT and the Social Sciences



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CPS Characteristics

What cyber physical systems have as defining characteristics:

- Cyber capability (i.e. networking and computational capability) in every physical component
- They are networked at multiple and extreme scales
- They are complex at multiple temporal and spatial scales.
- They are dynamically reorganizing and reconfiguring
- Control loops are closed at each spatial and temporal scale. Maybe human in the loop.
- Operation needs to be dependable and certifiable in certain cases
- Computation/information processing and physical processes are so **tightly integrated** that it is not possible to identify whether behavioral attributes are the result of computations (computer programs), physical laws, or both working together.

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Passivity, Symmetry and CPS

-Heterogeneity causes major challenges. Also dynamic changes (verification, security implications). In addition network uncertainties, time-varying delays, data rate limitations, packet losses.

-How do we guarantee desirable properties in a network of heterogeneous systems which may change dynamically, and expand or contract? Passivity inequalities. Comparison with Lyapunov stability.

-Can we start with a system and grow it in particular ways to preserve its properties? Symmetry.

-We impose passivity constraints on the components (also symmetry), and the design can accommodate heterogeneity and network effects. Approach also useful in human-interaction.

**- NSF CPS Large Project: "Science of Integration of CPS"
(with Vanderbilt, Maryland, GM R&D).**


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Background on Passivity



Definition of Passivity in Continuous-time

- Consider a continuous-time nonlinear dynamical system

$$\begin{aligned} \dot{x} &= f(x, u) \\ y &= h(x, u). \end{aligned}$$


- This system is *passive* if there exists a continuous storage function $V(x) \geq 0$ (for all x) such that

$$\int_{t_1}^{t_2} u^T(t)y(t)dt + V(x(t_1)) \geq V(x(t_2))$$

for all $t_2 \geq t_1$ and input $u(t) \in U$.

- When $V(x)$ is continuously differentiable, it can be written as:

$$u^T(t)y(t) \geq \dot{V}(x(t))$$



Alternative Definition of Passivity

- Passivity is inherently an input-output property.



- It is independent of internal representation. An alternative definition is that for all inputs $u(t) \in U$ and times T , there exists a β to satisfy the following inequality

$$\int_0^T u^T(t)y(t)dt \geq -\beta.$$

- When the system has zero initial conditions the inequality reduces to

$$\int_0^T u^T(t)y(t)dt \geq 0.$$

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Examples of Passive Systems

- G_1 is a passive system if the pole is **negative** ($a \geq 0$) but not passive for $a < 0$

$$G_1(s) = \frac{1}{s+a}$$

- G_2 is not passive for any a because of the negative gain

$$G_2(s) = \frac{-1}{s+a}$$

- Even a stable, **minimum** phase system can be non-passive if the phase shift is too large (G_3)

$$G_3(s) = \frac{s+5}{s^2+2s+2}$$

- G_3 would have been passive if the zero were closer to the origin (G_4)

$$G_4(s) = \frac{s+0.5}{s^2+2s+2}$$

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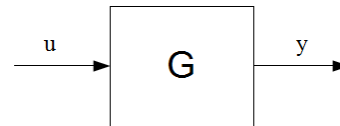


Passivity in Discrete-time

- Passivity can also be defined for discrete-time systems. Consider a nonlinear discrete time system

$$x(k+1) = f(x(k), u(k))$$

$$y(k) = h(x(k), u(k)).$$



- This system is *passive* if there exists a continuous storage function $V(x) \geq 0$ such that

$$\sum_{k=k_1}^{k_2} u^T(k)y(k) + V(x(k_1)) \geq V(x(k_2))$$

for all k_1, k_2 and all inputs $u(k) \in U$.

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Extended Definitions of Passivity

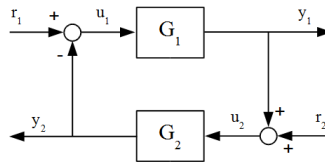
Passive	$u^T y \geq \dot{V}(x)$
Lossless	$u^T y = \dot{V}(x)$
Strictly Passive	$u^T y \geq \dot{V}(x) + \psi(x)$
Strictly Output Passive	$u^T y \geq \dot{V}(x) + \epsilon y^T y$
Strictly Input Passive	$u^T y \geq \dot{V}(x) + \delta u^T u$

- Note that $V(x)$ and $\psi(x)$ are positive definite and continuously differentiable. The constants ϵ and δ are positive. These equations hold for all times, inputs, and states.

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Interconnections of Passive Systems

- One of the strengths of passivity is when systems are interconnected. Passive systems are stable and passivity is preserved in many practical interconnections.
- For example, the negative feedback interconnection of two passive systems is passive.

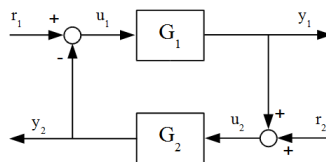


- If $u_1 \rightarrow y_1$ and $u_2 \rightarrow y_2$ are passive then the mapping $r = \begin{bmatrix} r_1 \\ r_2 \end{bmatrix} \rightarrow y = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$ is passive
- Note: the other internal mappings ($u_1 \rightarrow y_2$ and $u_2 \rightarrow y_1$) will be stable but may not be passive

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Interconnections of Stable Systems

- Compared with passive systems, the feedback interconnection of two stable systems is not always stable



- One notable special case is the small gain theorem where if G_1 and G_2 are finite-gain L_2 stable with gains γ_1 and γ_2 then the interconnection is stable if $\gamma_1 \gamma_2 < 1$.
- Both Passivity theory and the small gain theorem are special cases of larger frameworks including the conic systems theory and the passivity index theory.

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Stability of Passive Systems

- Strictly passive systems ($\psi(x) > 0$) are asymptotically stable
- Output strictly passive systems ($\delta > 0$) are L_2 stable
- The following results hold in feedback
 - Two passive systems \rightarrow passive and stable loop
 - Passive system and a strictly passive system \rightarrow asymptotically stable loop
 - Two output strictly passive systems $\rightarrow L_2$ stable loop
 - Two input strictly passive systems ($\epsilon > 0$) $\rightarrow L_2$ stable loop

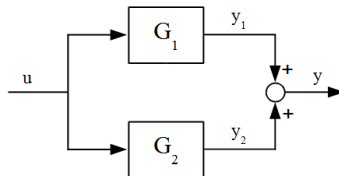
$$u^T y \geq \dot{V}(x) + \epsilon u^T u$$

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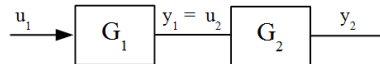


Other Interconnections

- The parallel interconnection of two passive systems is still passive



- However, this isn't true for the series connection of two systems



- For example, the series connection of any two systems that have 90° of phase shift have a combined phase shift of 180°

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Dissipativity, conic systems, and passivity indices

[McCourt and Antsaklis, ISIS-2009-009]
 [Kottenstette, McCourt, Xia and Antsaklis, 2014 Automatica]

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Definition of Dissipativity (CT)

- This concept generalizes passivity to allow for an arbitrary energy supply rate $\omega(u,y)$.
- A system is *dissipative* with respect to supply rate $\omega(u,y)$ if there exists a continuous storage function $V(x) \geq 0$ such that

$$\int_{t_1}^{t_2} \omega(u,y) dt \geq V(x(t_2)) - V(x(t_1))$$

for all t_1, t_2 and the input $u(t) \in U$.

- A special case of dissipativity is the QSR definition where the energy supply rate takes the following form:

$$\omega(u,y) = y^T Q y + 2y^T S u + u^T R u.$$

- QSR dissipative systems are L_2 stable when $Q < 0$

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Definition of Dissipativity (DT)

- The concept of dissipativity applies to discrete time systems for an arbitrary supply rate $\omega(u, y)$.
- A system is *dissipative* with respect to supply rate $\omega(u, y)$ if there exists a continuous storage function $V(x) \geq 0$ such that

$$\sum_{k=k_1}^{k_2} \omega(u, y) \geq V(x(k_2)) - V(x(k_1))$$

for all k_1, k_2 and the input $u(k) \in U$.

- A special case of dissipativity is the QSR definition where the energy supply rate takes the following form:

$$\omega(u, y) = y^T Q y + 2y^T S u + u^T R u.$$

- Dissipative DT systems are stable when $Q < 0$

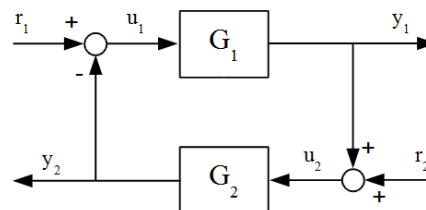
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Stability using dissipativity

- Dissipative systems may not be passive or stable

- Stability of the feedback interconnection of two dissipative systems can be assessed



- If G_1 is dissipative with (Q_1, S_1, R_1) and G_2 is dissipative with (Q_2, S_2, R_2) , the feedback interconnection of the two systems is stable if the following LMI is satisfied

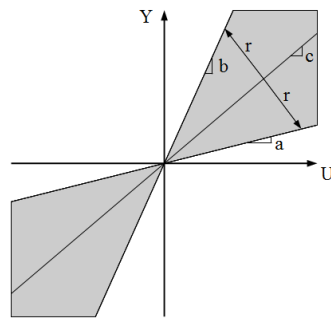
$$\begin{bmatrix} Q_1 + R_2 & S_1 - S_2^T \\ S_1^T - S_2 & Q_2 + R_1 \end{bmatrix} \leq 0$$

- This can also be seen as a control design tool. Say (Q_1, S_1, R_1) are known then stabilizing (Q_2, S_2, R_2) can be found. A controller can then be designed from (Q_2, S_2, R_2) to stabilize

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Conic Systems

- A conic system is one whose input-output behavior is constrained to lie in a cone of the $U \times Y$ inner product space



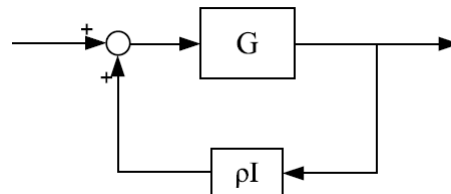
- A system is conic if the following dissipative inequality holds for all $t_2 \geq t_1$

$$\int_{t_1}^{t_2} \left[\left(1 + \frac{a}{b}\right) u^T y - a u^T u - \frac{1}{b} y^T y \right] dt \geq V(x(t_2)) - V(x(t_1))$$

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Output Feedback Passivity Index

The output feedback passivity index (OFP) is the largest gain that can be put in positive feedback with a system such that the interconnected system is passive.



Equivalent to the following dissipative inequality holding for G

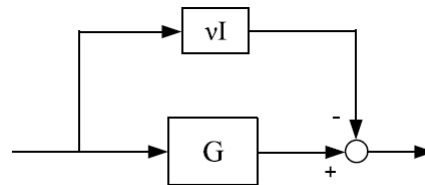
$$\int_{t_1}^{t_2} u^T y dt \geq V(x(t_2)) - V(x(t_1)) + \rho \int_{t_1}^{t_2} y^T y dt$$

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Input Feed-Forward Passivity Index

The input feed-forward passivity index (IFP) is the largest gain that can be put in a negative parallel interconnection with a system such that the interconnected system is passive.



Equivalent to the following dissipative inequality holding for G

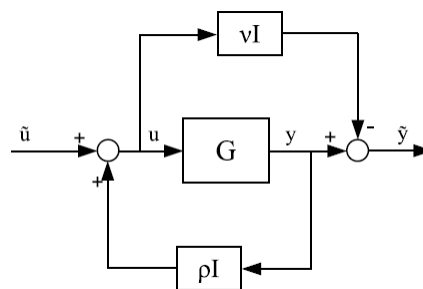
$$\int_{t_1}^{t_2} u^T y dt \geq V(x(t_2)) - V(x(t_1)) + \nu \int_{t_1}^{t_2} u^T u dt$$

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Simultaneous Indices

When applying both indices the physical interpretation as in the block diagram



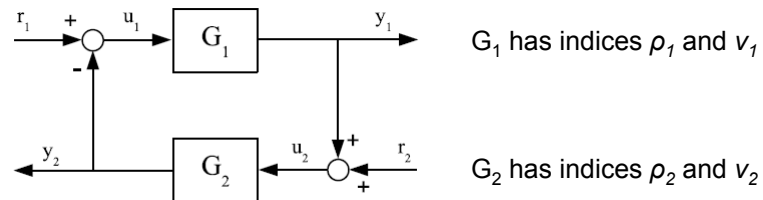
Equivalent to the following dissipative inequality holding for G

$$(1 + \rho\nu) \int_{t_1}^{t_2} u^T y dt \geq V(x(t_2)) - V(x(t_1)) + \rho \int_{t_1}^{t_2} y^T y dt + \nu \int_{t_1}^{t_2} u^T u dt$$

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Stability

We can assess the stability of an interconnection using the indices for G_1 and G_2



The interconnection is L_2 stable if the following matrix is positive definite

$$\begin{bmatrix} (\rho_1 + \nu_2)I & \frac{1}{2}(\rho_2\nu_2 - \rho_1\nu_1)I \\ \frac{1}{2}(\rho_2\nu_2 - \rho_1\nu_1)I & (\rho_2 + \nu_1)I \end{bmatrix} > 0$$

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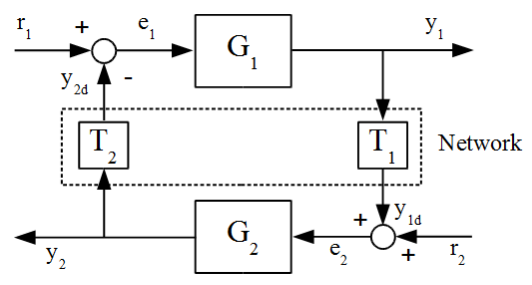
Networked Passive Systems

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Networked Systems

- Motivating Problem: The feedback interconnection of two passive systems is passive and stable. However, when the two are interconnected over a delayed network the result is not passive so stability is no longer guaranteed. How do we recover stability?



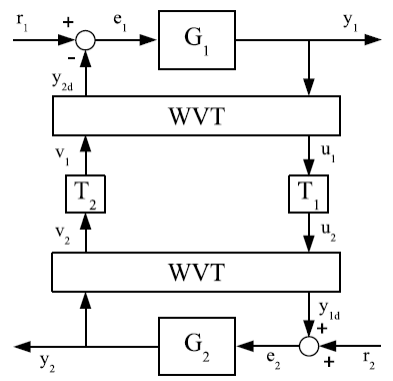
The systems G_1 and G_2 are interconnected over a network with time delays T_1 and T_2

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Stability of Networked Passive Systems

- One solution to interconnecting passive systems over a delayed network is to add an interface between the systems and the network
- The wave variable transformation forces the interconnection to guarantee stability for arbitrarily large time delays
- The WVT is defined below



$$\begin{bmatrix} u_1 \\ v_1 \end{bmatrix} = \frac{1}{\sqrt{2b}} \begin{bmatrix} bI & I \\ bI & -I \end{bmatrix} \begin{bmatrix} y_1 \\ y_{2d} \end{bmatrix}$$

$$\begin{bmatrix} u_2 \\ v_2 \end{bmatrix} = \frac{1}{\sqrt{2b}} \begin{bmatrix} bI & I \\ bI & -I \end{bmatrix} \begin{bmatrix} y_{1d} \\ y_2 \end{bmatrix}$$

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Passivity and Dissipativity for Switched Systems

[McCourt and Antsaklis, 2012 ACC]
[McCourt and Antsaklis, 2010 CDC]
[McCourt and Antsaklis, 2010 ACC]



Passivity for Switched Systems

- The notion of passivity has been defined for switched systems

$$\dot{x} = f_{\sigma}(x, u)$$

$$y = h_{\sigma}(x, u)$$

A *switched system* is *passive* if it meets the following conditions

- Each subsystem i is passive when active:

$$\int_{t_1}^{t_2} u^T y dt \geq V_i(x(t_2)) - V_i(x(t_1))$$

- Each subsystem i is dissipative w.r.t. ω_j^i when inactive:

$$\int_{t_1}^{t_2} \omega_j^i(u, y, x, t) dt \geq V_j(x(t_2)) - V_j(x(t_1))$$

- There exists an input u so that the cross supply rates (ω_j^i) are integrable on the infinite time interval.

[McCourt & Antsaklis 2010 ACC, 2010 CDC]





Dissipativity for Switched Systems

Definition of Dissipativity in Discrete-time

A discrete-time switched system is dissipative if for each subsystem i there exists a positive function V_i such that the following conditions hold

1. Each subsystem i is dissipative while it is active with respect to $\omega_i(u, y)$:

$$\omega_i(u(t), y(t)) \geq V_i(x(t+1)) - V_i(x(t))$$

2. Each subsystem is dissipative when inactive with respect to $\omega_{ij}(u, y, x, t)$ for each active subsystem j :

$$\omega_{ij}(u(t), y(t), x(t), t(t)) \geq V_i(x(t+1)) - V_i(x(t))$$

Stability for Dissipative Discrete-time Systems

Theorem. Dissipative switched systems are stable when $\omega_i < 0$ for all i and there exists an infinitely summable function $\phi(t)$ so that the inactive energy is bounded,

$$\phi(t) \geq \omega_{ij}(u, y, x, t).$$

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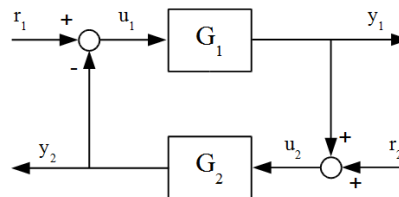


QSR Dissipativity for Switched Systems

- QSR dissipativity uses a quadratic supply rate to capture energy

$$\omega_i(u, y) = \begin{bmatrix} y \\ u \end{bmatrix}^T \begin{bmatrix} Q_i & S_i \\ S_i^T & R_i \end{bmatrix} \begin{bmatrix} y \\ u \end{bmatrix}$$

- Stability of switched systems can be assessed using Q_i



- Dissipativity of the feedback interconnection of two switched systems can be assessed with Q_i, S_i, R_i of both systems
- Large scale systems can be analyzed or designed using QSR dissipativity to ensure that the entire system is stable
- When dealing with passive switched systems ($Q_i = 0, S_i = 1/2 I, R_i = 0$), any sequential combination of systems in feedback or parallel can be shown to be passive and stable

[McCourt & Antsaklis 2012 ACC]

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Dissipativity for Switched Systems

Definition of Dissipativity in Discrete-time

A discrete-time switched system is dissipative if for each subsystem i there exists a positive function V_i such that the following conditions hold

1. Each subsystem i is dissipative while it is active with respect to $\omega_i(u, y)$:

$$\omega_i(u(t), y(t)) \geq V_i(x(t+1)) - V_i(x(t))$$

2. Each subsystem is dissipative when inactive with respect to $\omega_{ij}(u, y, x, t)$ for each active subsystem j :

$$\omega_{ij}(u(t), y(t), x(t), t) \geq V_i(x(t+1)) - V_i(x(t))$$

Stability for Dissipative Discrete-time Systems

Theorem. Dissipative switched systems are stable when $\omega_i < 0$ for all i and there exists an infinitely summable function $\phi(t)$ so that the inactive energy is bounded,

$$\phi(t) \geq \omega_{ij}(u, y, x, t).$$

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Passivity and Dissipativity in Networked Switched Systems

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Computational and Experimental Methods for Passivity and Dissipativity Determination

[McCourt and Antsaklis, 2014 ACC]
[McCourt and Antsaklis, ISIS-2013-008]
[Wu, McCourt and Antsaklis, ISIS-2013-002]

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Showing Passivity and Dissipativity

- Passivity and dissipativity are powerful properties for analysis and synthesis of dynamical systems.
- Requires finding a positive storage function V and an appropriate ω in the case of dissipativity.

$$\int_{t_1}^{t_2} \omega(u(t), y(t)) dt + V(x(t_1)) \geq V(x(t_2))$$

- In a switched system with m subsystems, dissipativity requires finding m storage functions and $\sim m^2$ dissipative rates
- In the worst case, this is a search similar to finding a Lyapunov function
- In many practical cases, this can be automated so that a program can generate an energy storage function (for LTI systems this is done using LMIs)

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LMI Methods – Passivity

- There are computational methods to find storage functions. For LTI passive systems, can always assume there exists a quadratic storage function

$$V(x) = \frac{1}{2} x^T P x \quad \text{where } P = P^T > 0.$$

- For continuous-time system this leads to the following LMI

$$\begin{bmatrix} A^T P + PA & PB - C^T \\ B^T P - C & -D - D^T \end{bmatrix} \leq 0$$

- In discrete-time the LMI becomes the following

$$\begin{bmatrix} A^T P A - P & A^T P B - C^T \\ B^T P A - C & B^T P B - D - D^T \end{bmatrix} \leq 0$$

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LMI Methods – QSR Dissipativity

- The same can be done to demonstrate that an LTI system is QSR dissipative. Once again, a quadratic storage function is used

$$V(x) = \frac{1}{2} x^T P x \quad \text{where } P = P^T > 0.$$

- For continuous-time system this leads to the following LMI

$$\begin{bmatrix} A^T P + PA - C^T Q C & PB - C^T Q D - C^T S \\ B^T P - D^T Q C - S^T C & -D^T Q D - S^T D - D^T S - R \end{bmatrix} \leq 0$$

- In discrete-time the LMI becomes the following

$$\begin{bmatrix} A^T P A - P - C^T Q C & A^T P B - C^T Q D - C^T S \\ B^T P A - D^T Q C - S^T C & B^T P B - D^T Q D - S^T D - D^T S - R \end{bmatrix} \leq 0$$

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LMI Example

- Consider the following linear system

$$\dot{x} = \begin{bmatrix} -1 & 1 \\ -8 & -1.5 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$$

$$y = [1 \quad 1]x + 0.5u$$

- The LMI solver in MATLAB determines the system to be passive with storage function

$$V(x) = \frac{1}{2} x^T P x \quad \text{with} \quad P = \begin{bmatrix} 12.9 & 1 \\ 1 & 1 \end{bmatrix}$$

- It can also be shown to be QSR dissipative (with respect to $Q = -.5$, $S = .5$, $R = 0$) with storage function

$$V(x) = \frac{1}{2} x^T P x \quad \text{with} \quad P = \begin{bmatrix} 4.75 & 0.5 \\ 0.5 & 0.5 \end{bmatrix}$$

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SOS Methods

- For nonlinear systems, there isn't a general method to find storage functions
- For polynomial nonlinear systems, there are sum of squares (SOS) methods
- Any problem that can be expressed as a search for a positive polynomial function with polynomial constraints can be solved using SOS optimization

- Consider a system

$$\begin{aligned} x(t+1) &= f(x(t)) + g(x(t))u(t) \\ y(t) &= h(x(t)) \end{aligned}$$

where f , g , and h are polynomials.

- Need to search for a polynomial storage function $V(x)$ to show dissipativity for a given Q , S , and R

[McCourt and Antsaklis, 2014 ACC]

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SOS Methods – Dissipativity

- The form of the storage function is based on the order of the desired storage function and a vector of lower order monomials

$$z = [x_1 \ x_2 \ x_1^2 \ x_1x_2 \ x_2^2 \ x_1^3 \ \dots]^T$$
- The search is for the coefficients of the monomials to make up the polynomial storage function

$$V(x) = z^T L z$$
- This can be formulated as a sum of squares optimization problem to find the elements of $L = \{\ell_{ij}\}$
- The optimization problem is to minimize a user chosen linear combination of the entries of L subject to $V(x) = z^T L z$ being polynomial and the dissipative inequality holding for the given system
- This opens up a class of nonlinear systems that can be identified as passive or dissipative in an automated way

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SOS Example

- Consider the following unstable (non-passive) nonlinear system

$$\dot{x} = \begin{bmatrix} -1.2x_1 + .65x_2 \\ .2x_2 - x_1^3 + u \end{bmatrix}$$

$$y = x_2$$
- This system isn't passive but it is QSR dissipative (with respect to $Q=0.4$, $S=0.5$, $R=-0.9$) with storage function

$$V(x) = 0.49x_1^2 + 0.91x_2^2 + 0.69x_2^4$$
- This system can be feedback stabilized with another system with (for example) $Q=0$, $S=0.5$, $R=1$. A simple example would be

$$\dot{x} = u$$

$$y = x + u$$

(continued)

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Experimental Methods

- The passivity levels (indices) may also be determined experimentally if input/output simulations or physical experimental settings are available.
- Restricted to classes of inputs.
- If certain system parameters are adjustable the passivity indices may be optimized using non-derivative based optimization;
- Performance can also be adjusted.

[Wu, McCourt and Antsaklis, ISIS-2013-002]



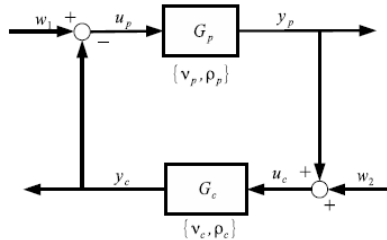
Feedback Passivation Using Passivity Levels

[Zhu, Xia and Antsaklis, 2014 ACC]



Problem Formulation

- Assume that the passivity levels for subsystems are known,



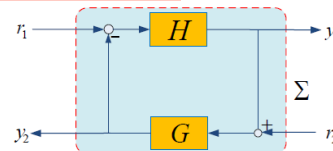
- Two problems are considered:
 - Passivity analysis – to determine the passivity levels of the interconnected system;
 - Passivation synthesis – to render a non-passive system passive (called passivation) through feedback.

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Existing Results

- Feedback Interconnections of IF-OFP systems:

- If system H and G are passive, then system Σ is passive.
- If system H and G are output strictly passive (OSP), then system Σ is OSP;
- If system H is IF-OFP(ϵ_1, δ_1) and system G is IF-OFP(ϵ_2, δ_2), where $\epsilon_1 + \delta_2 > 0$, $\epsilon_2 + \delta_1 > 0$, then system Σ is finite gain stable (FGS). \square

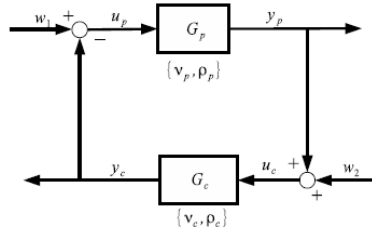


- Restrictions:
 - either assume passivity of subsystems ;
 - or focus on stability of the interconnected system.

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Main Results (1)

- Assume that the passivity levels for subsystems are known,



- Theorem: the passivity levels for the interconnected system satisfy:

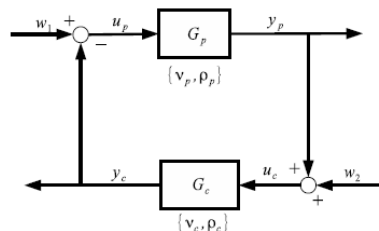
$$\begin{cases} \epsilon < \min \{ \nu_p, \nu_c \} \\ \delta \leq \min \left\{ \rho_c - \frac{\epsilon \nu_p}{\nu_p - \epsilon}, \rho_p - \frac{\epsilon \nu_c}{\nu_c - \epsilon} \right\} \end{cases}$$

- Comment: in general, the precise values of the passivity levels are difficult to obtain (especially for nonlinear systems).

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Main Results (2)

- Assume that the passivity levels for subsystems are known,



$$\begin{cases} w_2 = 0 \\ \nu_p \geq 0 \\ \rho_c \geq 0 \\ \nu_p + \rho_c > 0. \end{cases}$$

- Theorem: the passivity levels for the interconnected system satisfy:

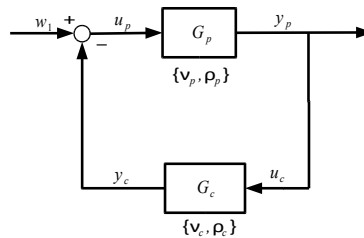
$$\begin{cases} \epsilon \leq \frac{\nu_p \rho_c}{\nu_p + \rho_c} \\ \delta \leq \nu_c + \rho_p \end{cases}$$

- Note that when $w_2 = 0$, then the passivity of the system that maps w_1 to y_p can be guaranteed even if one subsystem is non-passive.

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Passivation

- If one of the system is non-passive, say e.g. $\rho_p < 0$, how to passivate the system (from w_1 to y_p using passivity levels?



- The controller need to satisfy the following conditions:

$$\rho_c \geq 0 \quad \nu_c > -\rho_p$$

- The closed-loop system has passivity levels:

$$\epsilon \leq \frac{\nu_p \rho_c}{\nu_p + \rho_c} \quad \delta \leq \nu_c + \rho_p$$

- Comment: can design controller based on desired ϵ and δ .

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Example (1) – Linear Systems

- The plant G_p is linear system with $\rho_p = -0.2$ and $\nu_p = 0$.

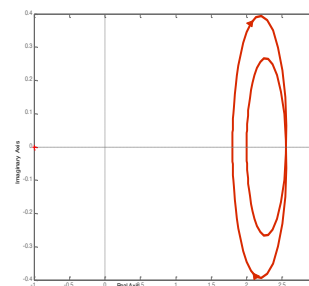
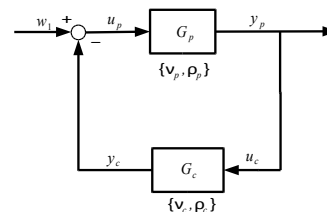
$$G_p = \frac{s + 0.5}{s - 0.1}$$

- The controller G_c is a linear controller with $\rho_c = 0$ and $\nu_c = 1$.

$$G_c = \frac{s + 4}{s + 2}$$

- The closed loop system has the passivity levels $\rho = 0.8$ and $\nu = 0$.
- The closed-loop transfer function is OSP with passivity index 1.8.

$$\frac{s^2 + 2.5s + 1}{2s^2 + 6.4s + 1.8}$$



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Example (2) – Nonlinear Systems

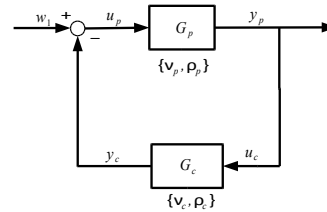
- The plant G_p is a nonlinear system with $\rho_p = -0.5$ and $\nu_p = 1.5$.

$$G_p \begin{cases} \dot{x}_1 = & x_2 \\ \dot{x}_2 = & -0.5x_1^3 + 0.5x_2 + 2u_p \\ y_p = & x_2 + u_p \end{cases}$$

- The controller G_c is a linear controller with $\rho_c = 0$ and $\nu_c = 1$.

$$G_c = \frac{s+4}{s+2}$$

- The closed-loop system has the passivity levels $\rho = 0.5$ and $\nu = 0$.



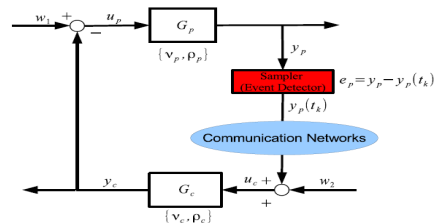
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Conclusions

- We considered passivity analysis and passivation problems for feedback interconnected IF-OPF systems.
- Passivity levels were characterized for feedback systems.
- Passivation conditions were provided to obtain required passivity levels in design of nonlinear systems.
- Results can be extended to event-triggered feedback systems.

[Zhu, Xia and Antsaklis, 2014 IFAC]



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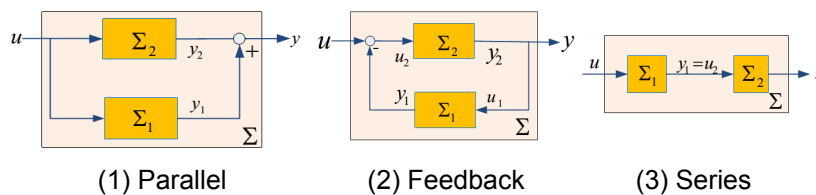
Generalized Passivation of Systems with Application to Systems with Input/Output Delays

[Xia, Antsaklis and Gupta, 2014 CDC]
[Xia, Antsaklis and Gupta, ISIS-2014-002]



Passivation

- Passivation methods refer to methods that can render a non-passive system passive, e.g. parallel, feedback and series interconnections



- Feedback passivation cannot passivate systems that are non-minimum phase or have relative degree larger than one, e.g.

$$\frac{s-1}{s+1}$$

$$\frac{1}{s^2 + s + 1}$$

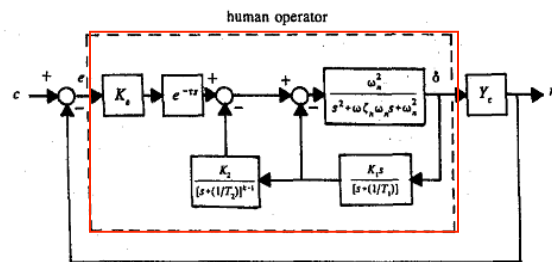


Motivation

- Consider linear systems with input-output delay (e.g. chemical systems, human operators etc.)

$$G(s) = G_0(s)e^{-\tau s}$$

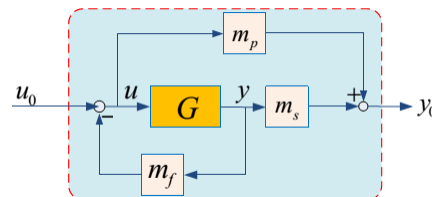
- G_0 is a SISO, stable, proper, rational transfer function;
- $\tau > 0$ denotes the transport delay;
- an example for linear human operator model
- such systems cannot be passivated through feedback alone.



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Problem Setup

- The following passivation method can be viewed as a combination of parallel, feedback and series interconnections



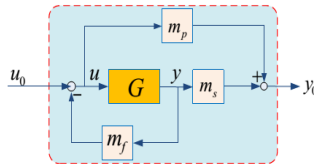
$$\begin{bmatrix} u_0 \\ y_0 \end{bmatrix} = M \begin{bmatrix} u \\ y \end{bmatrix} \quad \text{and} \quad M \triangleq \begin{bmatrix} 1 & m_f \\ m_p & m_s \end{bmatrix}$$

- How to select the passivation parameters so that system is passive? In addition, can they improve system $\Sigma_0 : u_0 \rightarrow y_0$ performance?

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Passivation Results

- **Theorem:** Let system G be finite-gain stable with gain γ . Consider the passivation method as shown in the following figure.



If the passivation are chosen such that

$$1 > m_f \gamma > 0, \quad m_f m_p > m_s > 0,$$

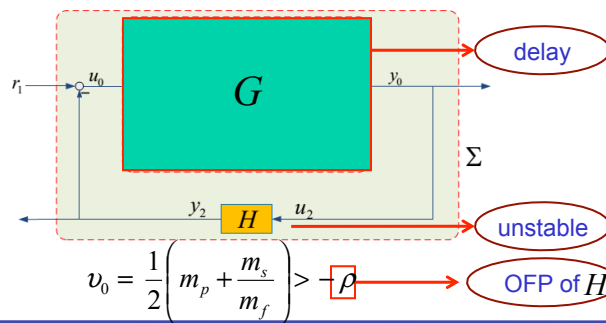
then the passivated system $\Sigma_0 : u_0 \rightarrow y_0$ is ISP with IFP level

$$v_0 = \frac{1}{2} \left(m_p + \frac{m_s}{m_f} \right) > 0.$$

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Passivation Results contd

- In the passivity theorem, to guarantee passivity or stability of the closed-loop system, both plant and controller need to be passive, or at least one of them has to be input/output strictly passive.
- If such conditions do not hold (e.g. both systems are non-passive), then the passivation method can be used to guarantee passivity.



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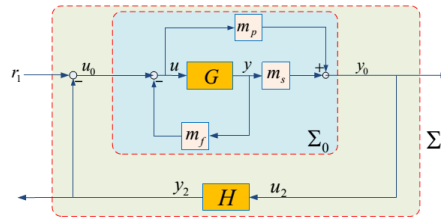
Feedback Interconnections

➤ **Theorem:** Consider the feedback configuration in the following figure, where r_1 can be seen as the reference to the controller G .

Assume that the plant H has an OFP level $\rho < 0$. If the passivation parameters (m_p, m_s, m_f) are chosen such that

$$v_0 = \frac{1}{2} \left(m_p + \frac{m_s}{m_f} \right) > -\rho, \quad 1 > m_f \gamma > 0, \quad m_f m_p > m_s > 0,$$

then the closed-loop system is output strictly passive. Furthermore, the system is finite-gain stable with gain no larger than $\frac{1}{\rho + v_0}$.



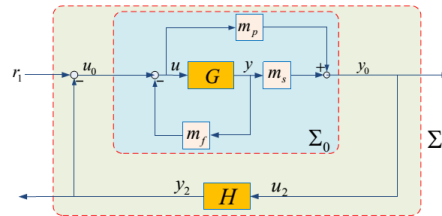
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Performance Optimization

➤ The passivation parameters can be selected to optimize system performance, such as minimizing the tracking error.

tracking error

Minimize: $J = \int_0^T \|r_1 - y_2\|^2 dt$
 w.r.t: m_p, m_s, m_f
 subject to: passivation conditions



- If the dynamics for the plant H are known, one can use gradient-based optimization methods.
- If the dynamics for the plant is unknown, unreliable or expensive, one can use non-derivative optimization methods.

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Performance Optimization

- A simulation-based control design (co-simulation) setup
- Optimization algorithm: the method of Hooke and Jeeves

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Adaptive Cruise Control Design

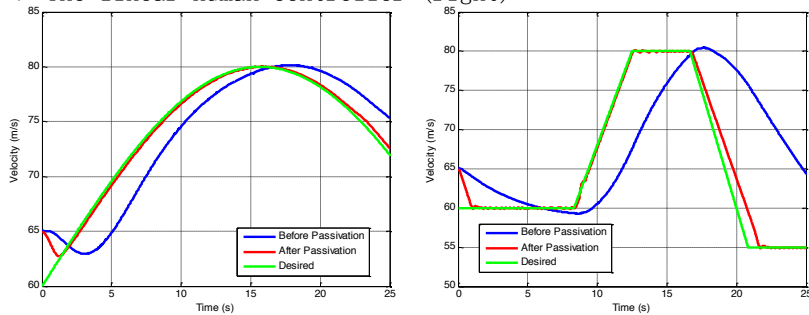
- Adaptive cruise control design for automotive systems
 - two control modes: speed control and spacing control
 - a typical value for the time delay: 0.5 seconds

a_{des} : desired acceleration v_h : host car velocity
 σ_{des} : desired throttle a_h : host car acceleration
 $P_{mc_{des}}$: desired master cylinder pressure

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Adaptive Cruise Control Design contd

- Two controllers were considered for minimizing the tracking error
 - PI controller with delay (left)
 - The linear human controller (right)

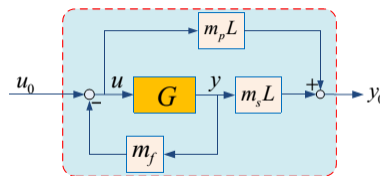


- The passivation parameters can greatly improve system performance in addition to guaranteeing passivity.

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Passivation Using Transfer Functions

- Motivation: braking frequently (left)
- Solution: use low pass filter for passivation (right) $L = \frac{0.08s + 1}{s + 1}$



⟨#⟩



Summary of Contributions

- The passivation method represents a combination of series, parallel and feedback interconnections
- Such a general passivation method works for any finite gain stable (linear or nonlinear) systems, e.g. system with input-output delays
- The passivation parameters can guarantee not only passivity but also desired performance
- Validated through simulations in CarSim and Simulink
- On-going and future work may include:
 - Use of non-positive passivation parameters and transfer functions for passivation
 - Extensions to passivation of switched systems
 - Analytic relation between performance and passivity indices

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Passivity/Dissipativity for Hybrid/Switched and DES Systems

- Switched Systems. Definitions and compositionality results. [McCourt, Dissertation, Apr. 2013]
- Hybrid Systems. Definitions.
- DES abstractions of continuous systems. Define granularity to preserve passivity. [Sajja, Gupta and Antsaklis, ISIS-2014-005]
- Approximations. An alternative method to define passive DES and Hybrid systems.

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Passivity and QSR-dissipativity Analysis of a System and its Approximations

[Xia, Antsaklis, Gupta and Zhu, 2014 TAC, submitted]
 [Xia, Antsaklis and Gupta, 2013 ACC]
 [Xia, Antsaklis and Gupta, ISIS-2012-007]

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Problem Statement

- Problem Setup (input-output mapping):

$$u \rightarrow \Sigma_2 \rightarrow y_2 = y_1 + \Delta y$$

$$u \rightarrow \Sigma_1 \rightarrow y_1$$

- Error constraint (for the "worst" case):

$$\langle \Delta y, \Delta y \rangle_T \leq \gamma^2 \langle u, u \rangle_T + \varepsilon, \quad \forall u \in U, T \geq 0.$$

- For two linear systems given by G_1 and G_2 , γ is an upper bound on the H-infinity norm of the difference between two transfer functions $\Delta G = G_1 - G_2$, i.e. if $\|\Delta G\|_{H_\infty} \leq \gamma$, then the error constraint holds.

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General Result

➤ Assume that the error γ is 'small'.

$$u \rightarrow \Sigma_2 \rightarrow y_2 = y_1 + \Delta y$$

$$u \rightarrow \Sigma_1 \rightarrow y_1$$

$$\langle \Delta y, \Delta y \rangle_T \leq \gamma^2 \langle u, u \rangle_T + \varepsilon$$

Approximate Model Σ_2

↓ γ

System of Interest Σ_1

➤ when an approximate model Σ_2 has an excess of passivity:
 $\rho > 0$ or $v > 0$

➤ when an approximate model Σ_2 is not necessarily passive:
 $\rho < 0$ and $v < 0$

Then certain passivity levels for system Σ_1 can be guaranteed.

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Very Strictly Passive Systems

➤ Recall the definition for VSP systems:

$$\dot{V} \leq u^T y - v u^T u - \rho y^T y, \quad (\rho > 0, v > 0)$$

➤ **Theorem:** Consider the two systems as shown in the following figure. Suppose that the error constraint is satisfied, i.e.

$$\langle \Delta y, \Delta y \rangle_T \leq \gamma^2 \langle u, u \rangle_T + \varepsilon,$$

and system Σ_2 is VSP for (ρ, v) . Assume that $\gamma < \rho$ and $\gamma < v$. If the following condition is satisfied,

$$\gamma^2 - \left(\rho - \frac{2}{\rho}\right)\gamma + v^2 - 2 \geq 0,$$

then system Σ_1 is VSP for $(\rho - \gamma, v - \gamma)$.

$$u \rightarrow \Sigma_2 \rightarrow y_2 = y_1 + \Delta y$$

$$u \rightarrow \Sigma_1 \rightarrow y_1$$

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(Q,S,R)-dissipative Systems

- Consider a special case for non-passive systems:

$$\dot{V} \leq u^T y - v u^T u - \rho y^T y, \quad (\rho < 0, v < 0)$$

- **Theorem:** Consider the two systems as shown in the following figure. Suppose that the error constraint is satisfied, i.e.

$$\langle \Delta y, \Delta y \rangle_T \leq \gamma^2 \langle u, u \rangle_T + \varepsilon,$$

and system Σ_2 is has passivity levels $(\rho_2 < 0, v_2 < 0)$. Then Σ_1 has passivity levels (ρ_1, v_1) , if we can find $\xi > 0$ such that

$$\rho_2 - \rho_1 - \xi \rho_1^2 \geq 0,$$

$$u \rightarrow \Sigma_2 \rightarrow y_2 = y_1 + \Delta y$$

$$\frac{\gamma^2}{\xi} + \rho_1 \leq v_2 - v_1.$$

$$u \rightarrow \Sigma_1 \rightarrow y_1$$

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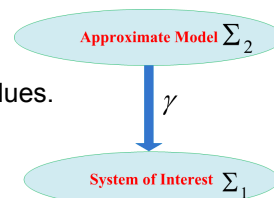
Particular Approximation Methods

- Linearization around an equilibrium point:
 - γ is determined by the radius of the ball around (0,0).

- Model reduction of linear systems:
 - γ is determined by the Hankel singular values.

- Sampled-data systems:
 - γ is determined by the sampling period.

- Quantization (e.g. logarithmic quantizers):
 - γ is determined by the quantizer parameters.



$$\langle \Delta y, \Delta y \rangle_T \leq \gamma^2 \langle u, u \rangle_T + \varepsilon$$

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Model Reduction

- Algorithm: truncated balanced realization (TBR)
- Error bound is given by Hankel singular values:

$$\|G_1 - G_2\|_{H_\infty} \leq 2 \sum_{i=r+1}^n \sigma_i \quad \longrightarrow \quad \langle \Delta y, \Delta y \rangle_T \leq \gamma^2 \langle u, u \rangle_T + \varepsilon$$

- **Corollary:** Let G_1 be a stable LTI system with order n . Let G_2 be a reduced order model of G_1 with order r , obtained using the TBR procedure. Define

$$\gamma = 2 \sum_{i=r+1}^n \sigma_i, \quad (\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n),$$

where σ_i is the i th Hankel singular values of system G_1 . Assume system G_1 is VSP for (ρ, v) . If $\gamma < \rho$ and $\gamma < v$, and the condition

$$\gamma^2 - \left(\rho - \frac{2}{\rho}\right)\gamma + v^2 - 2 \geq 0$$

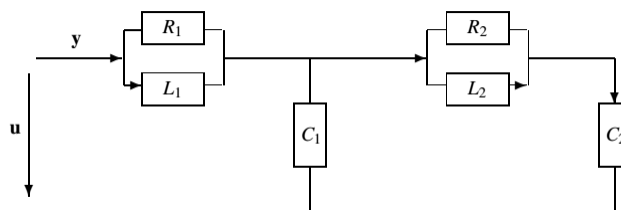
is satisfied, then system G_2 is VSP for $(\rho - \gamma, v - \gamma)$.

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Numerical Example

- Example: an RLC circuit



- The model is given by a 4-th order transfer function

$$G_1 = \frac{s^4 + 1.833s^3 + 1.25s^2 + 0.25s}{s^4 + 3.5s^3 + 2.417s^2 + 0.8333s + 0.1667}$$

$$(R_1 = 1, R_2 = 2, C_1 = 0.5, C_2 = 1, L_1 = 3, L_2 = 4)$$

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Numerical Example contd

- Model reduction using TBR algorithms, we obtain a 2nd order approximate model:

$$G_2 = \frac{s^2 + 1.399s - 0.08816}{s^2 + 3.069s + 0.2669}$$

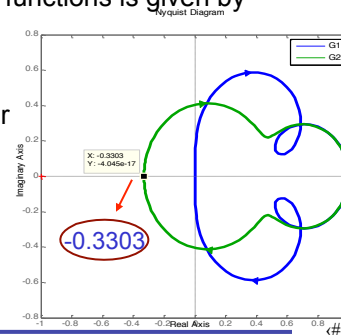
- The difference between the two transfer functions is given by

$$\gamma = 0.3303$$

- According to our results, the IFP level for system G_2 is less than

$$v_2 < v_1 - \gamma = -\gamma.$$

- Verified through Nyquist plots:



Summary of Contributions


- Passivity properties of a system can be obtained by analyzing its approximations.
- Robustness properties of passivity and dissipativity with respect to modeling uncertainties, etc
- Particular approximation methods ranging from linearization, model reduction, sampling and quantization

$$u \rightarrow \Sigma_2 \rightarrow y_2 = y_1 + \Delta y$$

$$u \rightarrow \Sigma_1 \rightarrow y_1$$

- Future Work may include:
 - Control design using approximate models
 - Extension to hybrid dynamical systems, discrete-event systems

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Symmetry and Dissipativity

[Wu, Ghanbari and Antsaklis, 2014 TAC, submitted]
[Wu and Antsaklis, 2011 MED]
[Wu and Antsaklis, 2010 MED]

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Symmetry in Systems

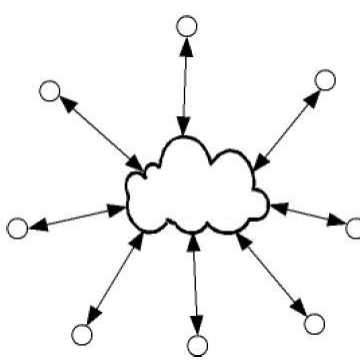
- Symmetry: A basic feature of shapes and graphs, indicating some degree of repetition or regularity
 - Symmetry in characterizations of information structure
 - Identical dynamics of subsystems
 - Invariance under group transformation e.g. rotational symmetry
- Why Symmetry?
 - Decompose into lower dimensional systems with better understanding of system properties such as stability and controllability
 - Construct symmetric large-scale systems without reducing performance if certain properties of low dimensional systems hold

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Simple Examples

Star-shaped Symmetry



$$u = u_e - \tilde{H}y$$

$$\tilde{H} = \begin{bmatrix} H & b & \dots & b \\ c & h & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ c & 0 & \dots & h \end{bmatrix}$$

$$u_0 = u_{e0} - Hy_0 - by_1 - \dots - by_m$$

$$u_1 = u_{e1} - cy_0 - hy_1$$

$$\vdots$$

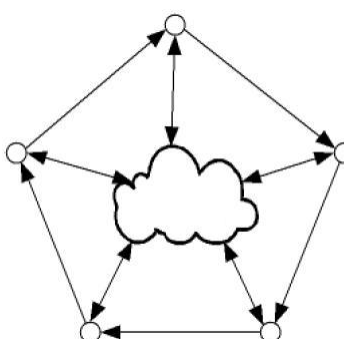
$$u_m = u_{em} - cy_0 - hy_m$$

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Simple Examples

Cyclic Symmetry



$$u = u_e - \tilde{H}y$$

$$\tilde{H} = \begin{bmatrix} H & b & \dots & b \\ c & & & \\ \vdots & & \tilde{h} & \\ c & & & \end{bmatrix}$$

$$\tilde{h} = circ([v_1, v_2, \dots, v_m])$$

$$u_0 = u_{e0} - Hy_0 - by_1 - \dots - by_m$$

$$u_1 = u_{e1} - cy_0 - v_1y_1 - v_2y_2 - \dots - v_my_m$$

$$\vdots$$

$$u_m = u_{em} - cy_0 - v_2y_1 - v_3y_2 - \dots - v_1y_m$$

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Main Results (1)

Theorem (Star-shaped Symmetry)

Consider a (Q, S, R) – dissipative system extended by m star-shaped symmetric (q, s, r) – dissipative subsystems. The whole system is finite gain input-out stable if

$$m < \min\left(\frac{\underline{\sigma}(\hat{Q})}{\underline{\sigma}(c^T r c + \beta(\hat{q} - b^T R b)^{-1} \beta^T)}, \frac{\hat{q}}{b^T R b}\right)$$

where

$$\begin{aligned} \hat{Q} &= -H^T R H + S H + H^T S^T - Q > 0 \\ \hat{q} &= -h^T r h + s h + h^T s^T - q > 0 \\ \beta &= S b + c^T s^T - H^T R b - c^T r h \end{aligned} \quad \tilde{H} = \begin{bmatrix} H & b & \cdots & b \\ c & h & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ c & 0 & \cdots & h \end{bmatrix}$$

⊕



Main Results (2)

Theorem (Cyclic Symmetry)

Consider a (Q, S, R) – dissipative system extended by m cyclic symmetric (q, s, r) – dissipative subsystems. The whole system is finite gain input-out stable if

$$m < \min\left(\frac{\underline{\sigma}(\hat{Q})}{\underline{\sigma}(c^T r c + \beta_m \Lambda^{-1} \beta_m^T)}, \frac{-r \sigma(\tilde{h}) \overline{\sigma(\tilde{h})} + s(\sigma(\tilde{h}) + \overline{\sigma(\tilde{h})}) - q}{b^T R b}\right)$$

where

$$\begin{aligned} \tilde{h} &= \text{circ}([v_1, v_2, \dots, v_m]) \\ \sigma(\tilde{h}) &= \sum_{j=1}^m v_j \lambda_i^j = \sum_{j=1}^m v_j e^{\frac{2\pi i j}{m}} \end{aligned} \quad \tilde{H} = \begin{bmatrix} H & b & \cdots & b \\ c & & & \\ \vdots & & \tilde{h} & \\ c & & & \end{bmatrix}$$

⊕

Main Results (3)

(cont.)

$$\Lambda = -r\tilde{h}^T\tilde{h} + s(\tilde{h}^T + \tilde{h}) - q \otimes I_m - b^T Rb \otimes \text{circ}([1, 1, \dots, 1])$$

$$\beta = Sb + c^T s^T - H^T Rb - c^T r\tilde{h}$$

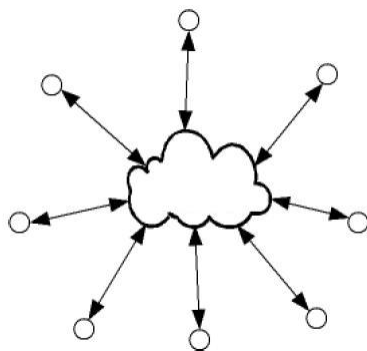
$$\beta_m = [\underbrace{\beta\beta\dots\beta}_m]$$

the spectral characterization of β should satisfy

$$\|\sigma(\tilde{h}) - \frac{s}{r}\| < \sqrt{\frac{s^2}{r^2} - \frac{q + mb^T Rb}{r}}$$

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Simple Examples



$$u = u_e - \tilde{H}y$$

$$\tilde{H} = \begin{bmatrix} 0.9 & -0.8 & -0.8 & \dots & -0.8 \\ -0.8 & 0.1 & 0 & \dots & 0 \\ -0.8 & 0 & 0.1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -0.8 & 0 & 0 & \dots & 0.1 \end{bmatrix}$$

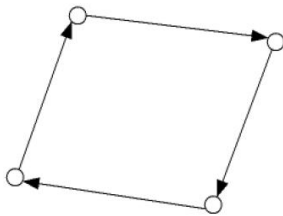
$$Q = q = -I, S = s = 0, R = r = \frac{1}{4}I$$

$$\Rightarrow m < \min(3.11, 6.25) = 3.11$$

Remark: $(-I, 0, \alpha^2 I)$ – dissipative systems corresponding to systems with gain less or equal to α (here $\alpha = \frac{1}{2}$)

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Simple Examples



$$u = u_e - \tilde{H}y \quad q = 0, s = \frac{1}{2}, r = 1$$

$$\tilde{H} = \tilde{h} = \begin{bmatrix} 0.1 & 0.2 & 0 & \dots & 0 \\ 0 & 0.1 & 0.2 & \dots & 0 \\ 0 & 0 & 0.1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0.2 & 0 & 0 & \dots & 0.1 \end{bmatrix}$$

The cyclic symmetric system is stable if $\|\sigma(\tilde{h}) - \frac{s}{r}\| = \|\sum_{j=0}^{m-1} v_j e^{\frac{2\pi i j}{m}}\| \leq 0.3 < 0.5 = \sqrt{\frac{s^2}{r^2} - \frac{q}{r}}$

The above stability condition is always satisfied. Also $m < \min(+\infty, +\infty)$

Thus the system can be extended with infinite numbers of subsystems without losing stability.

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Main Results (4)

Theorem (Star-shaped Symmetry for **Passive Systems**)

Consider a passive system extended by m star-shaped symmetric passive subsystems. The whole system is finite gain input-output stable if

where
$$m < \frac{\sigma(\hat{Q})}{\sigma(\beta \hat{Q}^{-1} \beta^T)}$$

$$\hat{Q} = \frac{H + H^T}{2} > 0$$

$$\hat{q} = \frac{h + h^T}{2} > 0$$

$$\beta = \frac{b + c^T}{2}$$

$$\tilde{H} = \begin{bmatrix} H & b & \dots & b \\ c & h & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ c & 0 & \dots & h \end{bmatrix}$$

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Main Results (5)

Theorem (Cyclic Symmetry for **Passive Systems**)

Consider a passive system extended by m cyclic symmetric passive subsystems. The whole system is finite gain input-output stable if

where
$$m < \frac{\sigma(\hat{Q})}{\sigma(\beta_m \Lambda^{-1} \beta_m^T)}$$

$$\hat{Q} = \frac{H + H^T}{2} > 0 \quad \Lambda = \frac{p/q \ p/s}{2}$$

$$\beta = \frac{b + c^T}{2} \quad \beta_m = [\beta \beta \dots \beta]$$

$$\sigma(\tilde{h}) = \sum_{j=1}^m v_j \lambda_i^j = \sum_{j=1}^m v_j e^{\frac{2\pi i j}{m}}$$

$$\tilde{H} = \begin{bmatrix} H & b & \dots & b \\ c & & & \\ \vdots & & \tilde{h} & \\ c & & & \end{bmatrix}$$

$$\tilde{h} = \text{circ}([v_1, v_2, \dots, v_m])$$

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Symmetry and Passivity/Dissipativity

- **Approximate symmetries. Robustness of results.**
- **The subsystem dynamics may not be identical as long as they satisfy the same q, s, r inequalities.**
- **Results still valid when the strengths of the interconnections (b, c) are not exactly the same.**

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Additional Passivity/Dissipativity Results

- Event-triggered control for networked systems using passivity. Event triggered control is used to reduce communication in networked control systems. Output Synchronization [Yu and Antsaklis 2011 CDC]
- Passivity of Systems in Series. A Passivity Measure Of Systems In Cascade Based On Passivity Indices [Yu and Antsaklis 2010 CDC]
- Compensating for Quantization [Zhu, Yu, McCourt, 2012 HSCC]
- Linearization. Preserving passivity. [Xia, Antsaklis, Gupta and McCourt, 2014 TAC, to appear]
- Model Predictive Control [Yu, Zhu, Xia, 2013 ECC]
- Shraavan Sajja, Vijay Gupta and Panos. J. Antsaklis, Passivity based Supervisory Control, [ISIS Technical Report ISIS-2014-005, July 2014. Revised March 2015].

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Concluding Remarks

- Main points
 - **New ways of thinking needed to deal effectively with the CPS problems. New ways to determine research directions.**
 - Passivity/Dissipativity and Symmetry are promising
 - Need deeper understanding of fundamentals that cut across disciplines.
 - CPS**, Distributed, Embedded, Networked Systems. Analog-digital, large scale, life cycles, safety critical, end to end high-confidence.
 - **Need to expand our horizons. Control Systems at the center.**
 - Collaborations with, build bridges to Computer Science, Networks, Biology, Physics. Also Sociology, Psychology...

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