Devolving System



Evolving Systems: An Introduction

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References



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- Balas, M. J. and Frost, S. A., "Evolving Systems: Inheriting Stability with Evolving Controllers", Proceedings 47th Israel Annual Conference on Aerospace Sciences, Tel-Aviv, Israel, 2007.
- Frost, S. A. and Balas, M. J., "Stability Inheritance and Contact Dynamics of Flexible Structure Evolving Systems", Proceedings 17th IFAC Symposium on Automatic Control in Aerospace, Toulouse, France, 2007.
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- Frost, S. A and Balas, M. J., "Adaptive Key Component Control and Inheritance of Almost Strict Passivity in Evolving Systems", AAS Landis Markley Symposium, Cambridge, MD, 2008
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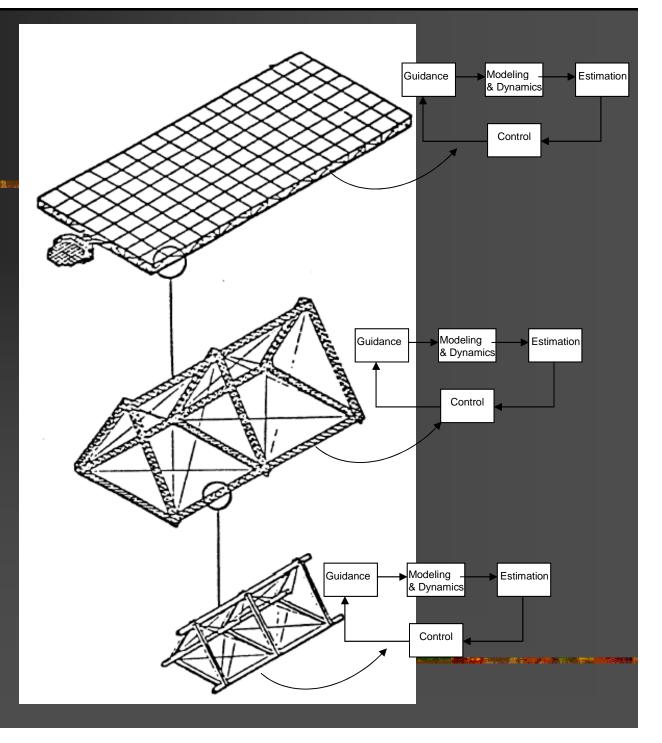
What It Is?

Evolving Systems= Autonomously

Assembled Active Structures

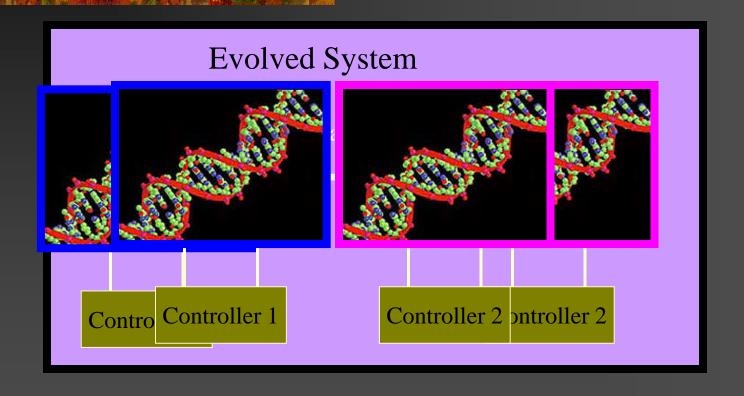
Or Self-Assembling
Structures,
which Aspire to a
Higher Purpose;
Cannot be attained
by Components Alone





Susan's Slide

How It Works



The Process of Evolving Systems

Active Component 1

Active Component 2

Active Component 3

Evolved System

Active Component 1

Active Component 2

Active Component 3

Mated Components

It's not theories about stars; it's the actual stars that count."
...... Freeman Dyson



Evolving Systems Applications

Autonomous Assembly in Space



International Space
Station after
9 December 2006
Mission

Evolving Systems Applications

- Autonomous Rendezvous and Docking
- Servicing and System Upgrades



DARPA's Orbital Express

(ASTRO Servicing Satellite pictured on left)

Stability is Essential During the Entire Evolution Process

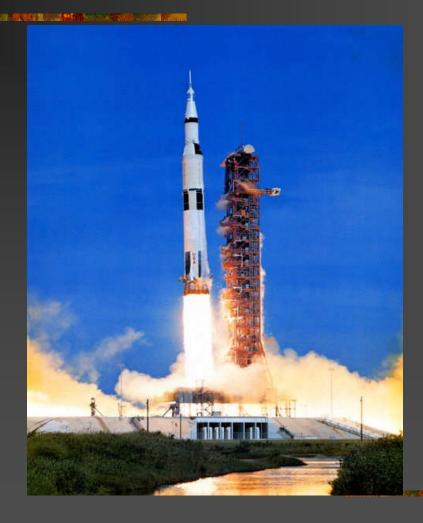


Orion Crew Exploration Vehicle Docking with the ISS

Launch Vehicles: Devolving

Systems

NASA-MSFC



Ares-Orion

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Constellations and Formations of Spacecraft (NASA-JPL)

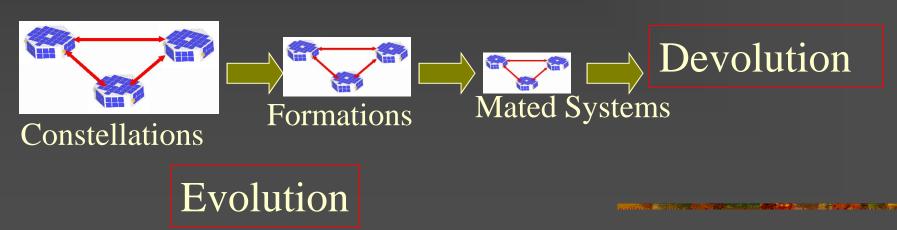


Separated Spacecraft Interferometers

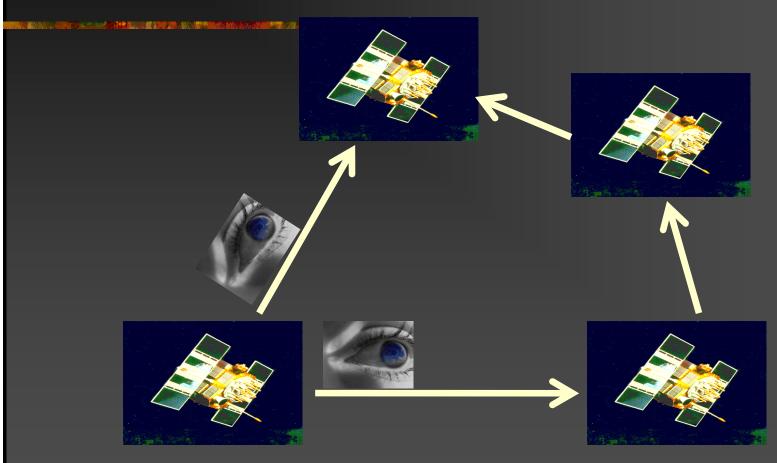
Global Persistent Surveillance: Constellations-Formations-Evolving Systems (NRO/DARPA)

The next step in network-centric warfare will be the creation of networked sensing suites that tailor their observations to the adversary's rate of activity. These various sensors will concentrate on observing changes rather than on observing scenery ...

Signal Magazine

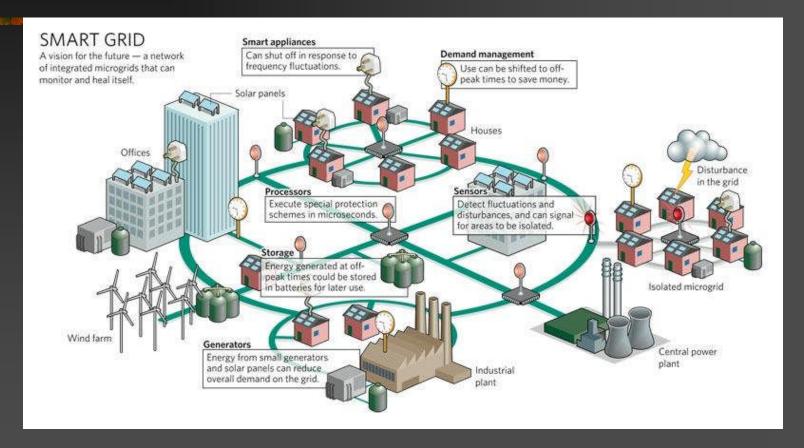


Evolving Spacecraft Formations



Joining Spacecraft

Smart Grids: Virtual Interconnecting Forces



"It is surprising how quickly we replace a human operator with an algorithm and call it SMART"

Affine Nonlinear Systems

$$\begin{cases} \dot{x} = A(x) + B(x)u \\ y = C(x) \end{cases}$$

With smooth vector fields: (A(x), B(x), C(x)) defined on a neighborhood a smooth manifold, or \Re^N

Evolving Systems: General

i th Component

$$\begin{cases} \dot{x}_i = A_i(x_i) + B_i(x_i)u_i^c \\ y_i = C_i(x_i) \end{cases}$$

$$\Rightarrow \begin{cases} \dot{x} = C_i(x_i) \\ \dot{x} = A(x) + B(x)u \\ y = C(x) \end{cases}$$
 Evolved System with $x \equiv \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$

where
$$\dot{x}_i = A_i(x_i) + B_i(x_i)u_i + \sum_{j=1}^{L} \varepsilon_{ij} \underbrace{A_{ij}(x_i, x_j, u_j)}_{Interconnections}; \begin{cases} 0 \le \varepsilon_{ij} \le 1 \\ \varepsilon_{ij} = \varepsilon_{ji} \end{cases}$$

Local Controller : $\begin{cases} u_i^c = h_i(z_i) + u_i \\ \dot{z}_i = l_i(z_i, y_i, u_i) \end{cases}$

Evolving Systems: 2 Components

$$0 \le \varepsilon_{12} = \varepsilon_{21} \equiv \varepsilon \le 1$$

Component 1

$$\begin{cases} \dot{x}_1 = A_1(x_1) + B_1(x_1)u_1 + \varepsilon A_{12}(x_1, x_2, u_2) \\ y_1 = C_1(x_1) \end{cases}$$

Component 2

$$\begin{cases} \dot{x}_1 = A_1(x_1) + B_1(x_1)u_1 + \varepsilon A_{12}(x_1, x_2, u_2) \\ y_1 = C_1(x_1) \end{cases} \begin{cases} \dot{x}_2 = A_2(x_2) + B_2(x_2)u_2 + \varepsilon A_{21}(x_2, x_1, u_1) \\ y_2 = C_2(x_2) \end{cases}$$

$$\Rightarrow$$
 Evolved System $\{$

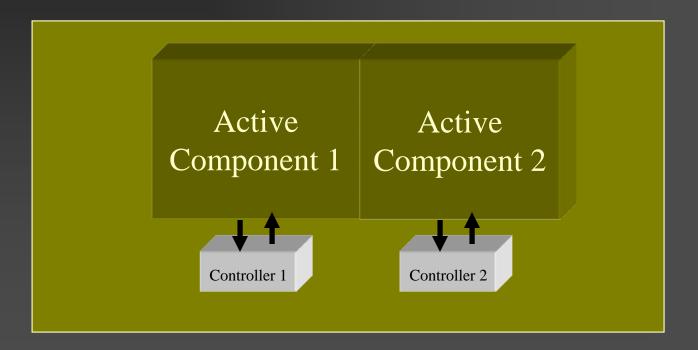
$$\begin{vmatrix} \dot{x} = \begin{bmatrix} A_1(x_1) \\ A_2(x_2) \end{bmatrix} + \begin{bmatrix} B_1(x_1) & 0 \\ 0 & B_2(x_2) \end{bmatrix} u + \varepsilon \begin{bmatrix} A_{12}(x_1, x_2, u_2) \\ A_{21}(x_2, x_1, u_1) \end{bmatrix}$$

$$y = \begin{bmatrix} C_1(x_1) \\ C_2(x_2) \end{bmatrix} \equiv C(x)$$

$$\begin{cases} \varepsilon = 0 \text{ unconnecte d} \\ \varepsilon = 1 \text{ fully connected} \end{cases}$$

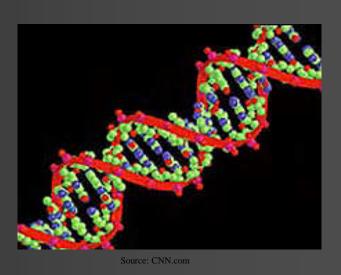
Evolving Systems: Evolution=Homotopy

Evolved System



$$0 \le \varepsilon \le 1$$

Genetics of Evolving Systems: Inheritance of Component Traits



- Controllability/Observability
- Stability
- Dissipativity
- Optimality
- Robustness
- Disturbance Rejection/Signal Tracking

How To Analyze Evolving Systems

- Admittance/Impedance
- Perturbation Methods
- ☐Graph Theory
- ☐ Differential Geometry- Lie Groups
- □Other stuff

Impedance and Admittance of Components

$$v_1 \equiv \dot{x}_1 = \dot{x}_2$$

$$x_1 = x_2$$

Component 1

Component 2

 $f_2 = -f_1$

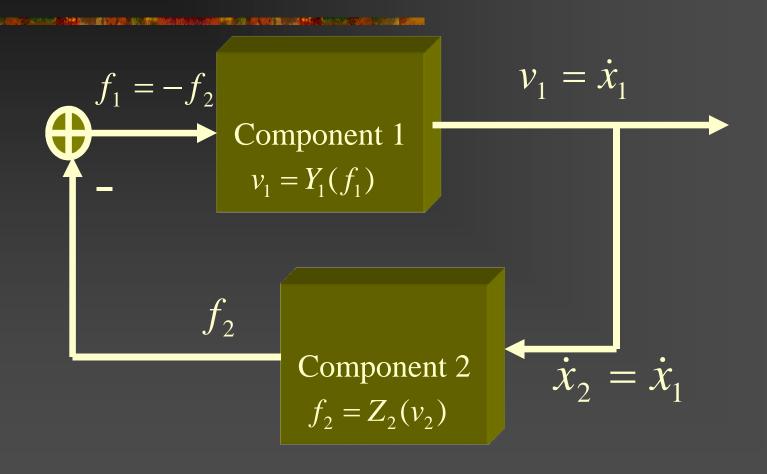
$$Z \equiv \text{Impedance Operator}: F = Z(V)$$

$$Y \equiv \text{Admittance Operator}: V = Y(F)$$

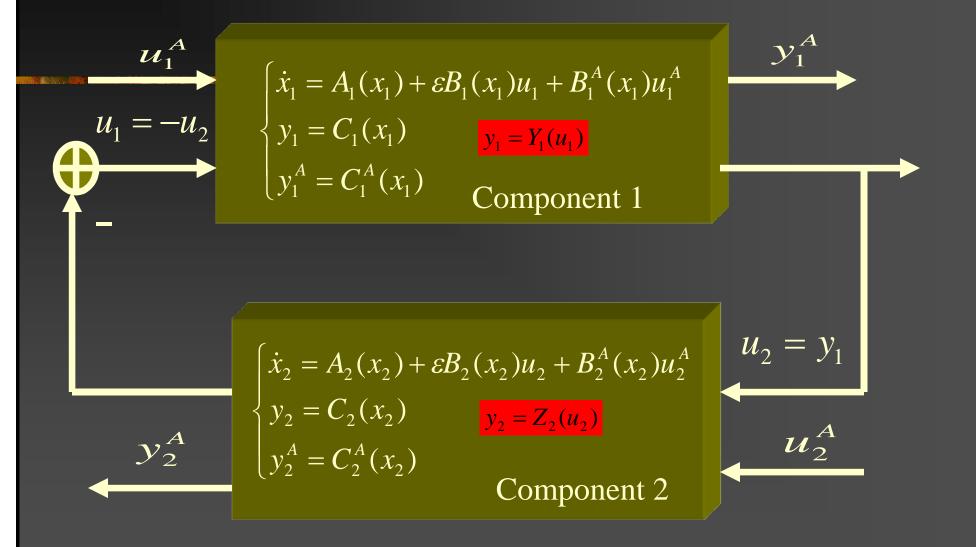
$$Z \equiv Y^{-1}, Y \equiv Z^{-1}$$

Linear Case: C. M. Harris and A. G. Piersol, Shock and Vibration Handbook, McGraw-Hill,

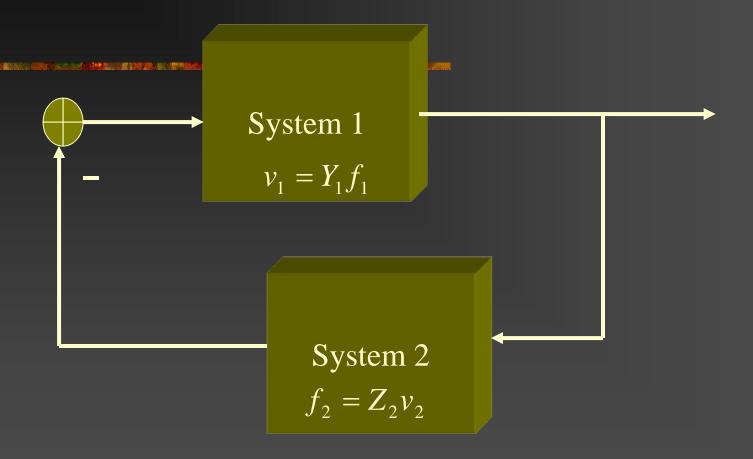
Fully Evolved System



Nonlinear State Space Version: Admittance-Impedance



Characterize LINEAR Evolved System Stability



Q: Given System 1, can we find all System 2 so that connected system is stable?

Use Youla Parametrization for nonproper Impedances/Admittances.

Youla or Q Parameterization

Component 1

 $Y_1 u = ND^{-1}u$ where N,D are coprime and stable

Component 2

 $Z_2 = (Y - QN)^{-1}(X + QD)$ Let X, Y be coprime and stable, and NX + DY = I, and Q be any stable rational function.



All controllers that make the closed-loop stable.

M. Vidyasagar, Control System Synthesis: A Factorization Approach, MIT Press,1985

Dissipativity: "Higher" Form of Stability

u y System

Energy Stored ≤ Energy Supplied e.g. Springs and Masses
Inductors and Capacitors
Roommates/Spouses

Definition of Dissipativity

$$\begin{cases} \dot{x} = A(x) + B(x)u \\ y = C(x) \end{cases}$$

Energy Storage Function :
$$\begin{cases} V(x) > 0; \forall x \neq 0; \\ V(0) = 0 \end{cases} \Rightarrow d'$$

System is Dissipativ e when

$$\begin{cases} L_{A(x)}V \equiv \nabla V A(x) \le 0 \\ L_{B(x)}V \equiv \nabla V B(x) = C^{T}(x) \end{cases}$$

$$\Rightarrow \underline{dV(x(t))/dt} \equiv \nabla V \dot{x}(t) = \nabla V[A(x) + B(x)u]$$
energy storage rate
$$\leq 0 + C^{T}(x)u = y^{T}(t)u(t) \equiv \langle y, u \rangle$$
External
Power
or

$$\underbrace{V(x(t))}_{EnergyStored} \leq \underbrace{V(x(0))}_{Initial} + \int_{0}^{t} \langle y, u \rangle d\tau$$
External
EnergyStored
EnergyStored

Strict Dissipativity

Energy Storage Function :
$$\begin{cases} V(x) > 0; \forall x \neq 0; \\ V(0) = 0 \end{cases}$$

System is Strictly Dissipativ e when

$$\begin{cases} L_{A(x)}V \equiv \nabla V A(x) \le -S(x) \\ L_{B(x)}V \equiv \nabla V B(x) = C^{T}(x) \end{cases}$$

$$\Rightarrow \underbrace{dV(x(t))/dt}_{\text{energy storage rate}} \equiv \nabla V \dot{x}(t) = \nabla V [A(x) + B(x)u]$$

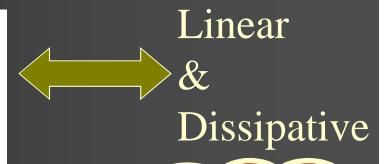
$$\leq -S(x) + C^{T}(x)u = \langle y, u \rangle - S(x)$$

Internal
Power
Dissipated

Positive Realness (PR)

Given
$$(A, B, C)$$

and $T(s) \equiv C(sI - A)^{-1}B + D$:
There is $P > 0$ so that
$$K - Y \begin{cases} A^T P + PA = -L^T L \equiv Q \leq 0 \\ PB = C^T \end{cases}$$



A Frequency
Domain
Condition

$$\operatorname{Re} T(j\omega) \equiv (T(j\omega) + T^*(j\omega))/2 \ge 0 \text{ for all } \omega$$

This is often taken as the definition of Positive Real(PR)

For Linear Systems: PR= Dissipative= Passive

What does **Strictly** Positive Real (SPR) mean?

Answer: Lotsa Things, Not all equivalent!! (See J.Wen,AC-33,1988)

Here is ONE Definition of SPR:

There is $\mu > 0$ so that

Re
$$T(j\omega - \mu) \equiv (T(j\omega - \mu) + T^*(j\omega - \mu))/2$$
 is PR

Relation to K - Y:

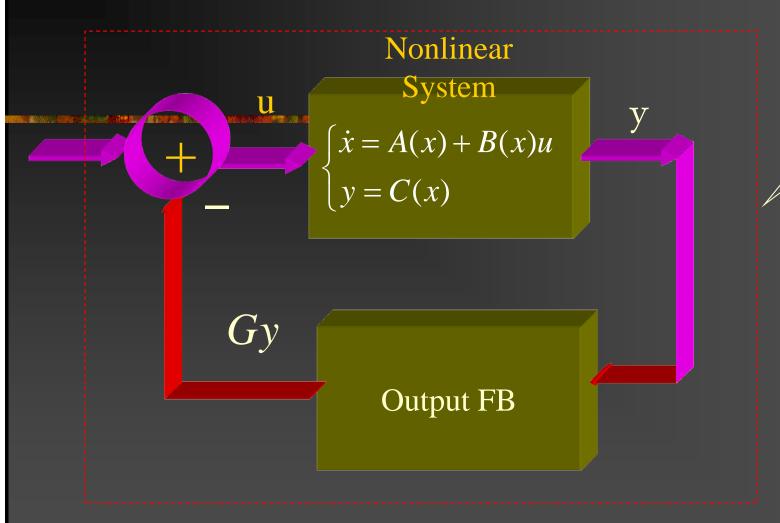
Just Change A to $A + \mu I$ in K - Y

$$\begin{cases} A^T P + PA = -L^T L - 2\mu P \equiv -Q < 0 \\ PB = C^T \end{cases}$$

Linear &
Strictly
Dissipative

For Linear Systems: SPR=Strictly Dissipative=Strictly Passive

Almost Strictly Dissipative (ASD)



Strictly
Dissipative

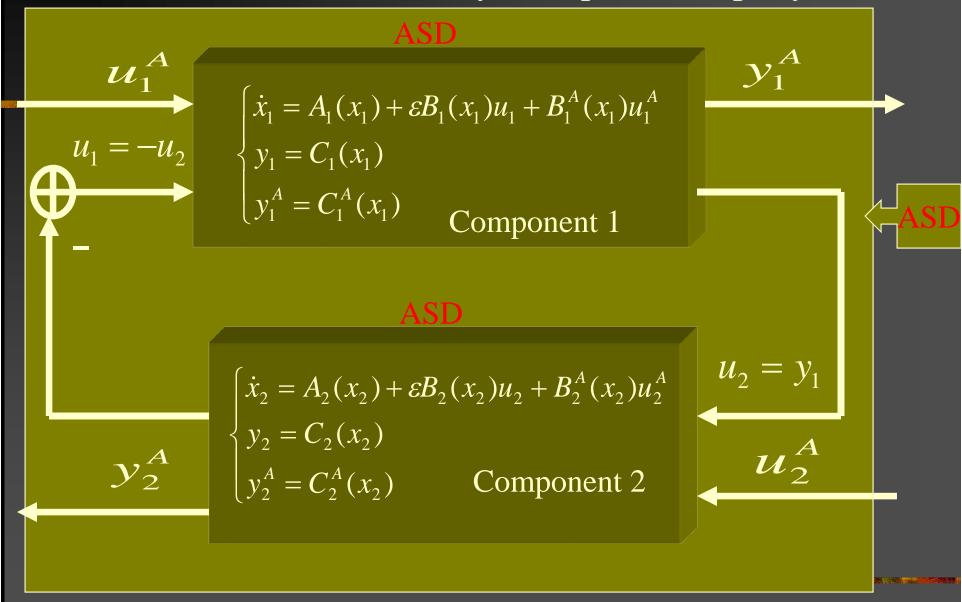
 $\therefore (A_C(x) \equiv A(x) + B(x)GC(x), B(x), C(x))$ is ASD

For LTI Systems: ASD=ASPR

Linear ASPR via Non-Orthogonal Projections

```
Balas&Fuentes:
1) (A,B,C) Almost SPR if and only if CB positive definite and
              open-loop transfer function P(s) \equiv C(sI - A)^{-1}B
                    is minimum phase
                           (i.e. all transmission zeros stable)
 2) Almost PR if and only if
        CB positive definite,
        open-loop transfer function is weakly minimum phase
        (i.e. can have some marginally stable transmission zeros),
        and
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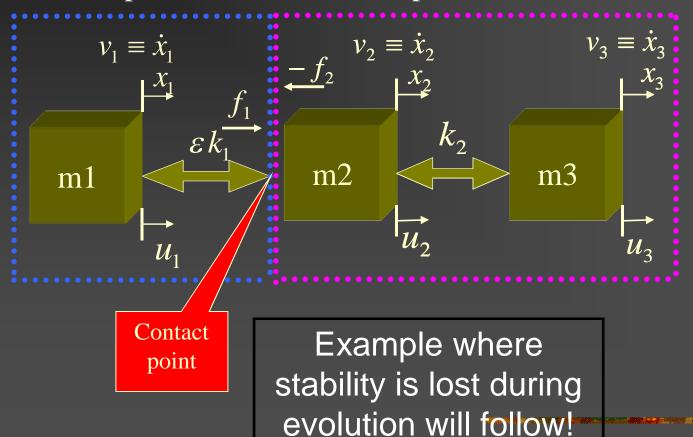
Inheritance of Almost Strictly Dissipative Property



Two Component Flexible Structure Evolving System



Component 2



System That Does Not Inherit **Stability**

Component 1:

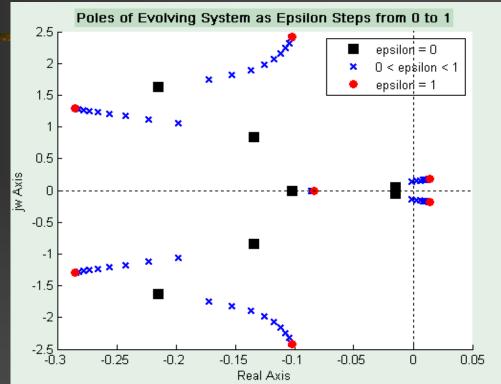
$$\begin{cases} m_1 \ddot{q}_1 = u_1 - \varepsilon k_{12} (q_1 - q_2) \\ y_1 = [q_1, \dot{q}_1]^T \end{cases}$$

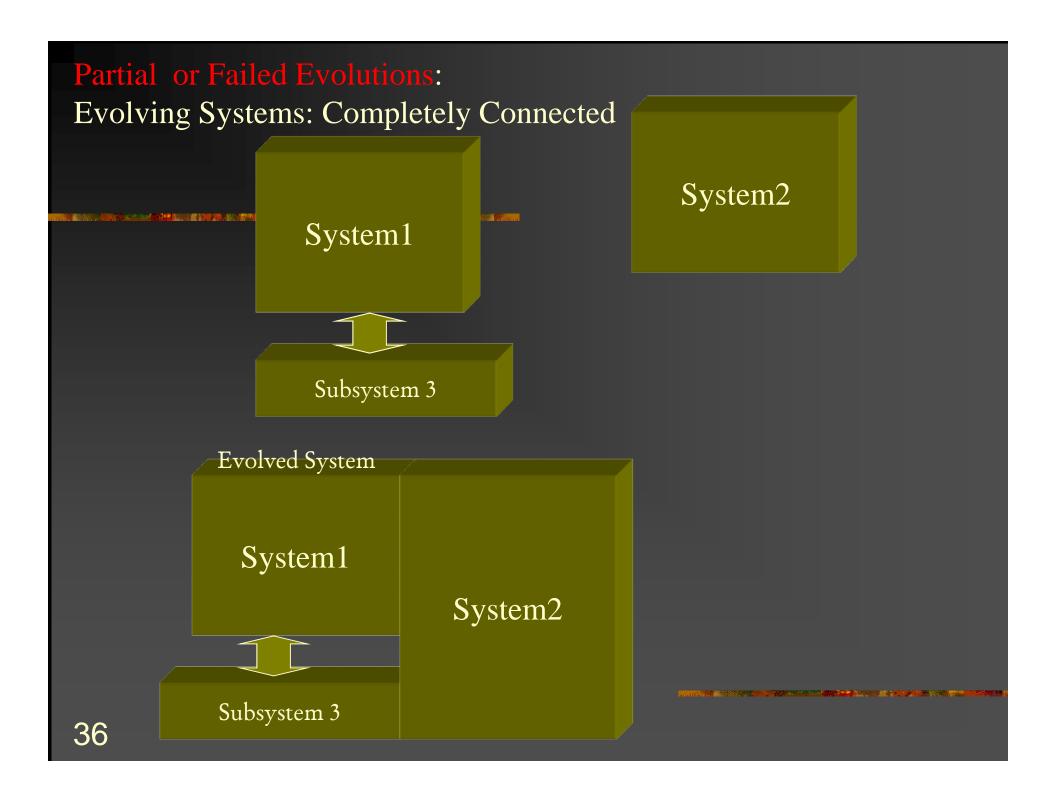
$$\begin{cases} m_2 \ddot{q}_2 = u_2 - k_{23} (q_2 - q_3) - \varepsilon k_{12} (q_2 - q_1) \\ m_3 \ddot{q}_3 = u_3 - k_{23} (q_3 - q_2) \end{cases}$$

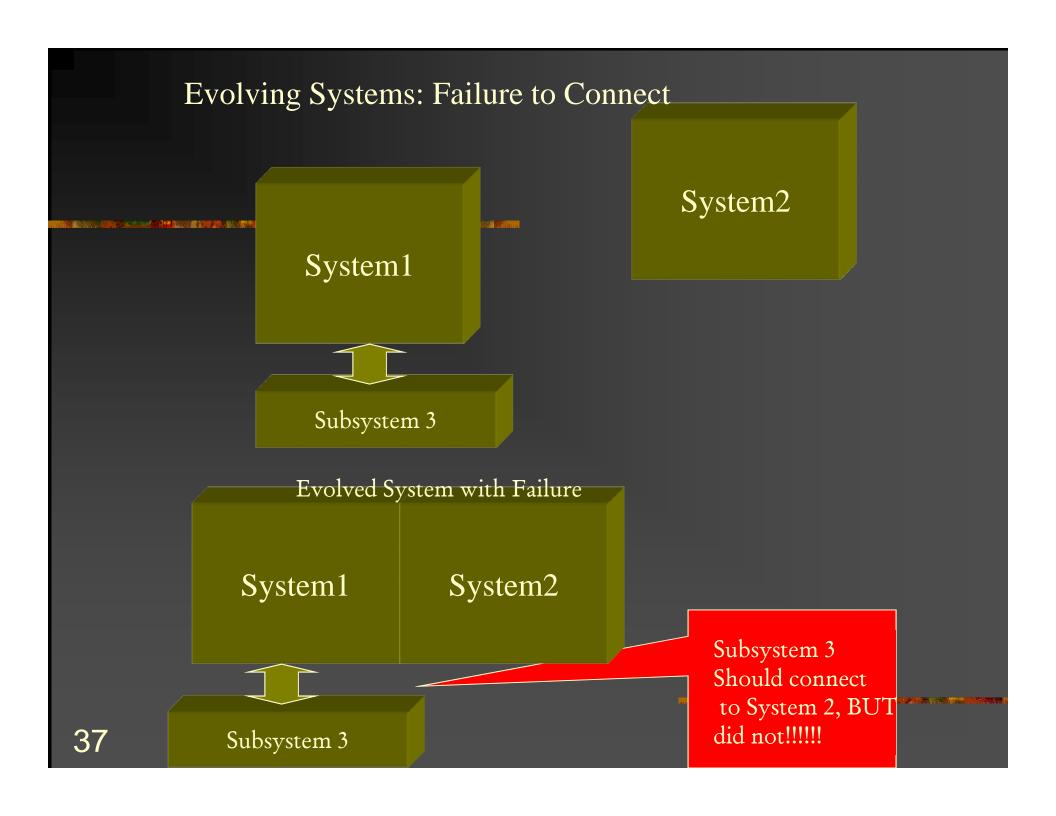
$$\begin{cases} y_2 = [q_2, \dot{q}_2]^T \\ y_3 = [q_3, \dot{q}_3]^T \end{cases}$$
(0.2003)

35

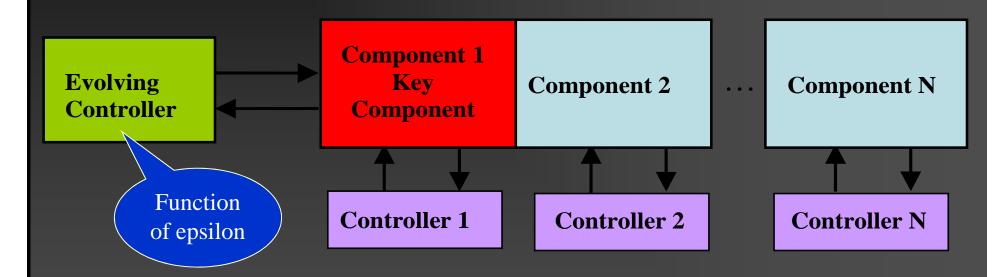
Component Controllers:
$$\begin{cases} u_1 = -(0.9s + 0.1)q_1 \\ u_2 = -\left(\frac{0.1}{s} + 0.2s + 0.5\right)q_2 \\ u_3 = -(0.6s + 1)q_3 \end{cases}$$







Key Component Evolving Controller

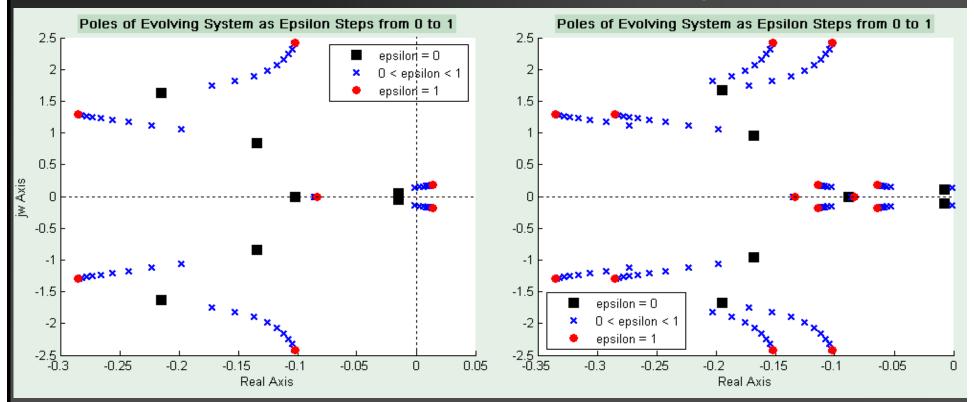


Local Controllers 1 ... N and their components' input-output ports remain unchanged

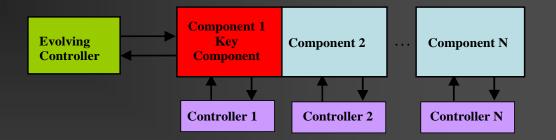
Closed Loop Poles of Example

No Evolving Controllers

System with Key Component Evolving Controller

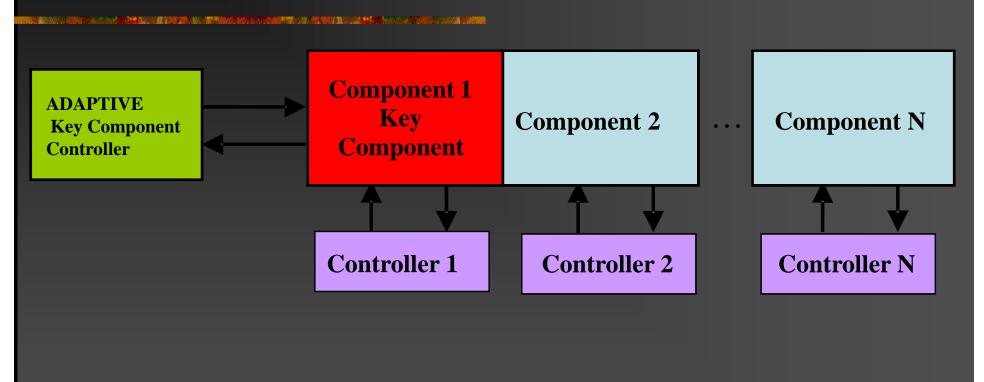


Benefits of Key Component Evolving Controllers

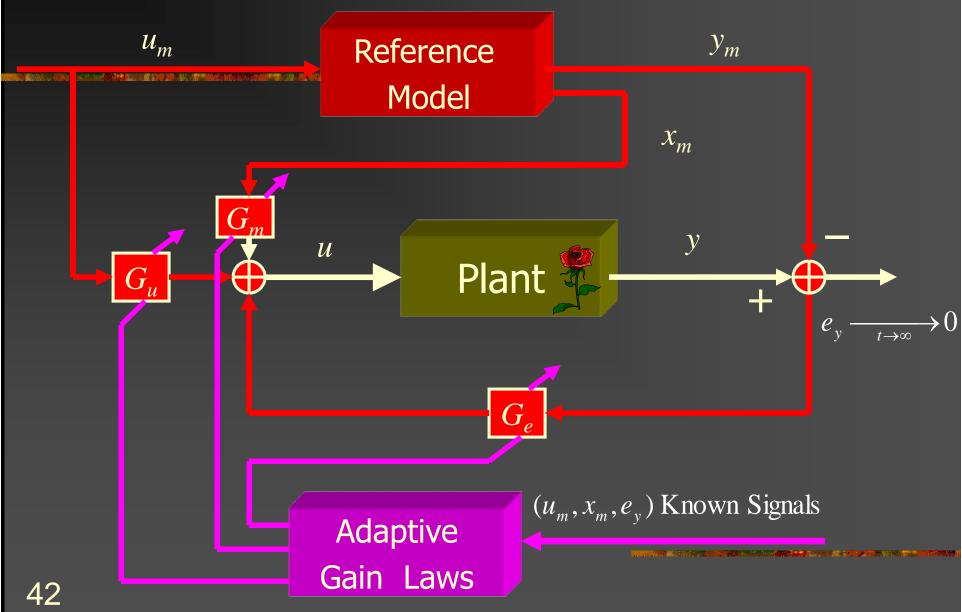


- Maintains stability during Evolution
- Interchangeability of non-key components
- Cost savings & risk mitigation
- No clear design methodology

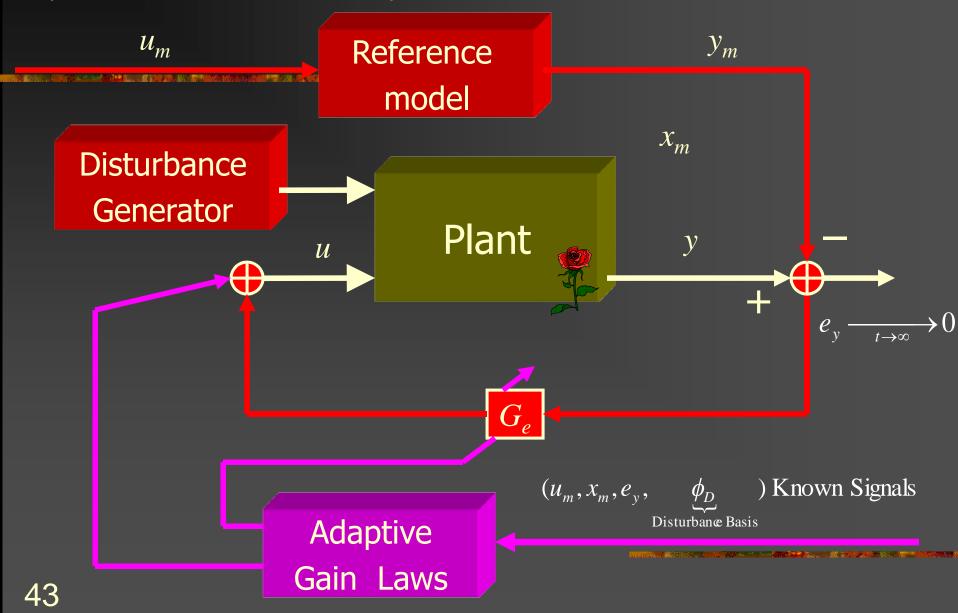
ADAPTIVE Key Component Evolving Controller



Direct Adaptive Model Following Control (Wen-Balas 1989)



Direct Adaptive Persistent Disturbance Rejection (Fuentes-Balas 2000)



Adaptive Key Component Control



_____AS

$$u_1 = -u_2$$

 u_1^A

$$\begin{cases} \dot{x}_1 = A_1(x_1) + \varepsilon B_1(x_1) u_1 + B_1^A(x_1) u_1^A \\ y_1 = C_1(x_1) \\ y_1^A = C_1^A(x_1) \end{cases}$$
We we Connected

Key Component 1

Strictly Dissipative

$$\begin{cases} \dot{x}_2 = A_2(x_2) + \varepsilon B_2(x_2) u_2 \\ y_2 = C_2(x_2) \end{cases}$$
 All Other Components



A Theorem <u>Is</u> Worth A Thousand Simulations

(despite its limitations)

Adaptive Key Component Controller Theorem

If Key Component 1 (u_1^A, y_1^A) is Almost Strictly Dissipative and Component 2 (u_2, y_2) is Strictly Dissipative, then

Adaptive
Key Component
Controller

$$\begin{cases} u_1^A = G_1 y_1^A \\ \dot{G}_1 = -y_1^A (y_1^A)^T \gamma_1; \gamma_1 > 0 \end{cases}$$

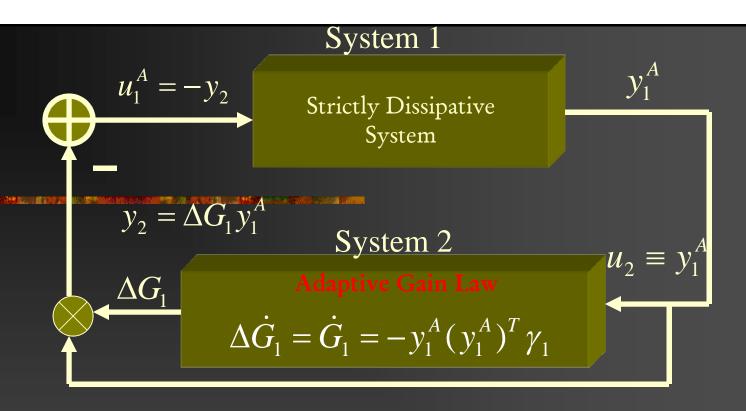
Produces $x_1, x_2 \xrightarrow[t \to \infty]{} 0$ and G_1 is bounded

throughout the entire evolution $0 \le \varepsilon \le 1$

Now Let's See Some Detailed Mathematical Proofs

No No Please, I'd Rather Be Eaten Alive by Radioactive Spiders





Let
$$V_2 \equiv \frac{1}{2} trace \Delta G \cdot \Delta G^T > 0$$

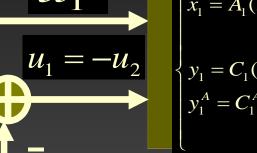
$$\Rightarrow \dot{V_2} = trace \Delta \dot{G_1} \gamma_1^{-1} \Delta G_1^T = trace(y_1^A (y_1^A)^T \gamma_1 \gamma_1^{-1} \Delta G^T)$$

$$= (\Delta G y_1^A)^T y_1^A = \langle y_2, u_2 \rangle \quad \therefore \text{System 2 is dissipativ e}$$

Adaptive Key Component Control Can Mitigate Persistent







$$\dot{x}_1 = A_1(x_1) + \varepsilon B_1(x_1)u_1 + B_1^A(x_1)u_1^A + \underbrace{\Gamma_1 u_D^1}_{Persistent Disturbani}$$

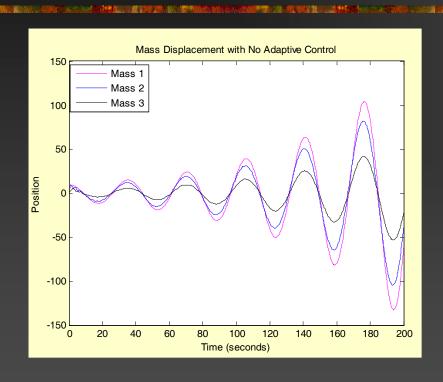
$$\begin{cases} y_1 = C_1(x_1) \\ y^A = C^A(x_1) \end{cases}$$

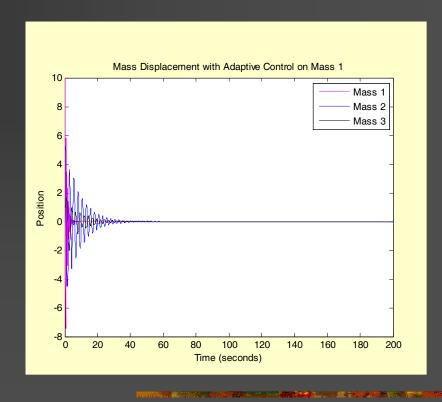
 $y_1^A = C_1^A(x_1)$ Key Component 1

$$\begin{cases} \dot{x}_2 = A_2(x_2) + \varepsilon B_2(x_2) u_2 \\ y_2 = C_2(x_2) \end{cases}$$

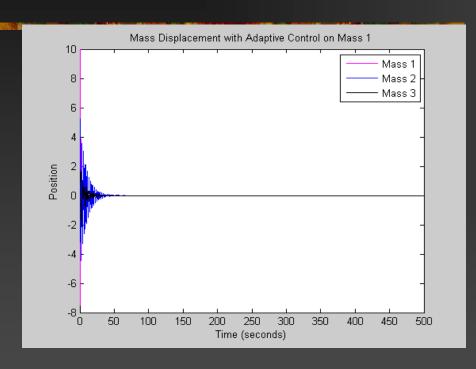
All Other Components

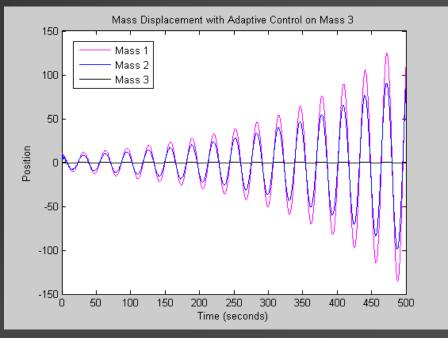
Adaptive Key Component Control of Nonlinear Plant





Adaptive Key Component Controller

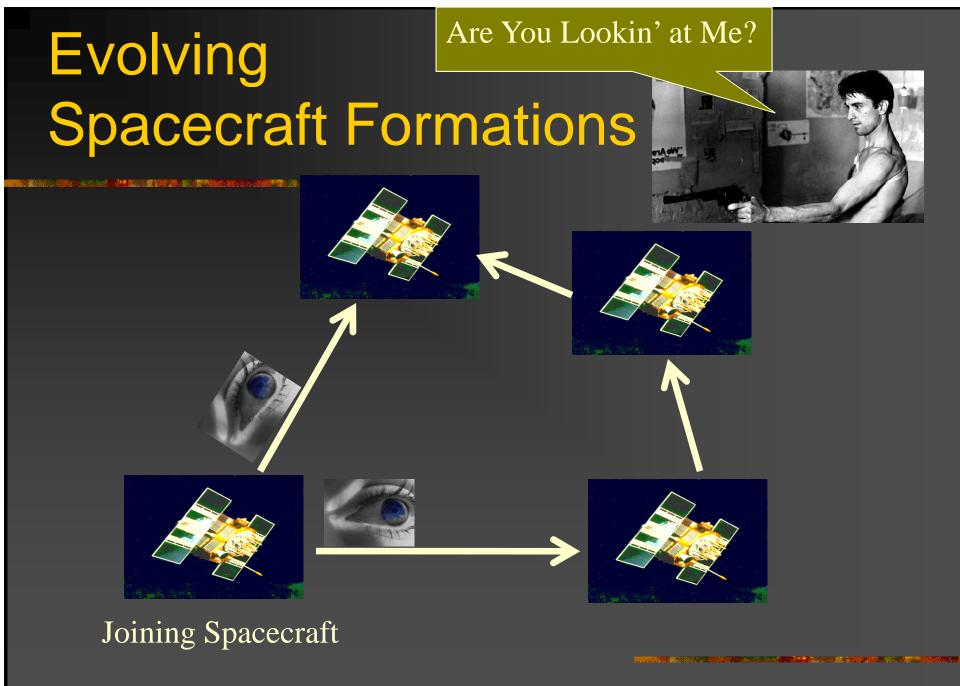




Component 1 is Key Component w/ I/O ports on mass 1

Component 2 is Key Component w/ I/O ports on mass 2

Nonminimum phase zeros: 0.0051466 + 0.20089i, 0.0051466 - 0.20089i



Control Based on Relative State

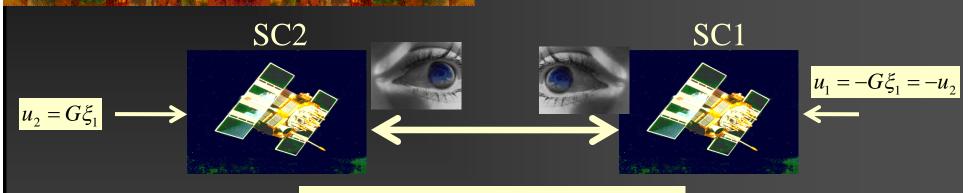


$$\left| \xi_1 \equiv x_2 - x_1 - r_1 \xrightarrow[t \to \infty]{} 0 \right|$$

$$\begin{aligned} u_2 &= G\xi_1 \\ &= \left[g_P \quad g_D \begin{bmatrix} q_2 - q_1 - r_P^1 \\ \dot{q}_2 - \dot{q}_1 - r_D^1 \end{bmatrix} \right] \\ &= \underbrace{g_P (q_2 - q_1 - r_P^1)}_{\text{preloaded spring}} + \underbrace{g_D (\dot{q}_2 - \dot{q}_1 - r_D^1)}_{\text{preloaded damper}} \end{aligned}$$

Uni-directional Newton's 3rdLaw

Reciprocity= Usual Newton's Laws



$$\left|\xi_1 \equiv x_2 - x_1 - r_1 \xrightarrow[t \to \infty]{} 0\right|$$

$$u_{2} = G\xi_{1} = -u_{1}$$

$$= \underbrace{g_{P}(q_{2} - q_{1} - r_{P}^{1})}_{\text{preloaded spring}} + \underbrace{g_{D}(\dot{q}_{2} - \dot{q}_{1} - r_{D}^{1})}_{\text{preloaded damper}}$$

Spacecraft Dynamics

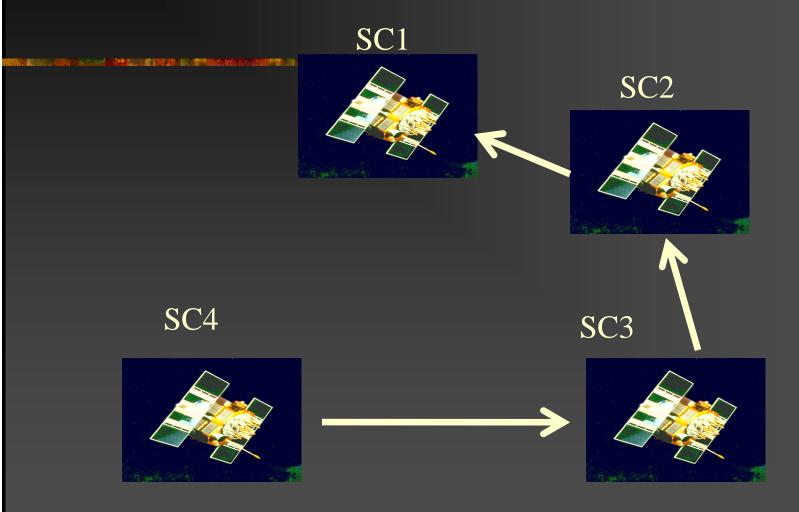
Identical Spacecraft

$$SC_k : \dot{x}_k = Ax_k + Bu_k + \Gamma u_k^D; k = 1,..., N$$

Disturbance Generator:
$$\begin{cases} u_k^D = \theta z_k^D \\ \dot{z}_k^D = F z_k^D \end{cases} \text{ or } z_k^D = L_k \phi_D$$

Double Integrator
$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = \begin{bmatrix} 1 & 0.1 \end{bmatrix}$$

Open Chain Formation



Relative Measurements

$$SC_k: y_k = C\xi_{k-1} \equiv C(x_k - x_{k-1} - r_{k-1})$$

Chained Output Feedback:
$$\begin{cases} u_k = Gy_k + G_D z_k^D = GC \xi_{k-1} + G_D z_k^D \\ u_1 = 0 \end{cases}$$

$$\Rightarrow \Delta u_k \equiv u_{k+1} - u_k = G(y_{k+1} - y_k) = GC(\xi_k - \xi_{k-1}) + G_D \Delta z_k^D$$

$$\therefore \dot{\xi}_{k} = A\xi_{k} + B\Delta u_{k} + \Gamma\Delta u_{k}^{D}$$

$$= (A + BGC)\xi_{k} - BGC\xi_{k-1} + (BG_{D} + \Gamma\theta)\Delta z_{k}^{D}; k \ge 2$$

$$= 0 \text{ or } R(\Gamma) \subset R(B)$$

&
$$\dot{\xi}_1 = (A + BGC)\xi_1$$

Open Chain Stability

 \overline{A}_C stable $\Leftrightarrow A + BGC$ stable

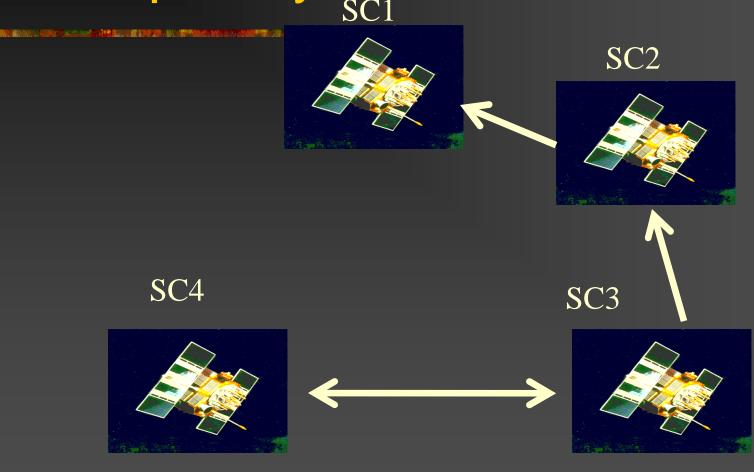


$$\Rightarrow \dot{\xi}_{k} = (A + BGC)\xi_{k} - BGC\xi_{k-1}$$

$$\text{Let } \xi \equiv \begin{bmatrix} \xi_{1} \\ \xi_{2} \\ \xi_{3} \end{bmatrix} \Rightarrow \dot{\xi} = \begin{bmatrix} A + BGC & 0 & 0 \\ -BGC & A + BGC & 0 \\ 0 & -BGC & A + BGC \end{bmatrix} \xi$$

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Open Chain Formation with Reciprocity



Stability Open Chain with Reciprocity

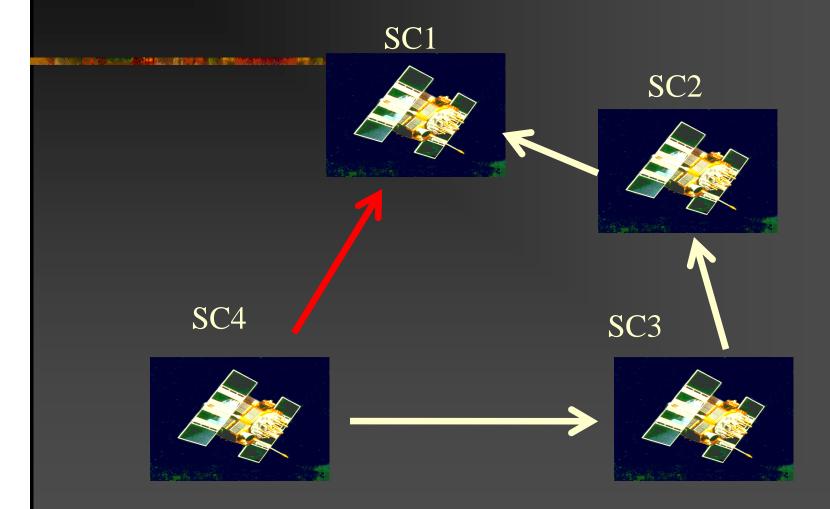
Output Feedback: $\begin{aligned} u_1 &= 0 \\ u_2 &= GC\xi_1 + G_D z_2^D \\ u_3 &= GC\xi_2 + G_D z_3^D + (-GC\xi_3) \\ u_4 &= GC\xi_3 + G_D z_4^D \end{aligned}$

Theorem: Open Chain with Reciprocity

Proof:
$$\begin{bmatrix} I & -\alpha I \\ 0 & I \end{bmatrix} \begin{bmatrix} A + 2BGC & -BGC \\ -BGC & A + BGC \end{bmatrix} \begin{bmatrix} I & \alpha I \\ 0 & I \end{bmatrix} = \begin{bmatrix} A + (2+\alpha)BGC & (\alpha^2 + \alpha - 1)BGC \\ -BGC & A + (1-\alpha)BGC \end{bmatrix} \#$$

Cor.
$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = \begin{bmatrix} 1 & 0.1 \end{bmatrix} \Rightarrow \overline{A}_C \text{ stable } \forall G = g > 0$$

Open Chain Formation

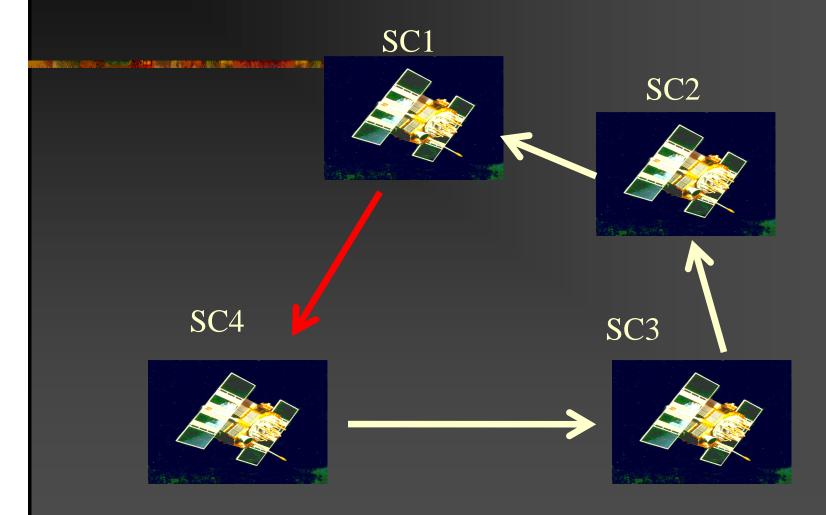


Theorem:

$$\overline{A}_{C} = \begin{bmatrix} A + BGC & 0 & 0 \\ -BGC & A + BGC & 0 \\ BGC & 0 & A + 2BGC \end{bmatrix}$$
 stable
$$\Leftrightarrow A + BGC \text{ and } A + 2BGC \text{ stable}$$

Cor.
$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = \begin{bmatrix} 1 & 0.1 \end{bmatrix} \Rightarrow \overline{A}_C \text{ stable } \forall G = -g < 0$$

Closed Chain Formation



$$\overline{A}_{C}(\varepsilon) = \begin{bmatrix} A + (1+\varepsilon)BGC & \varepsilon BGC & \varepsilon BGC \\ -BGC & A + BGC & 0 \\ 0 & -BGC & A + BGC \end{bmatrix}$$

$$= \begin{bmatrix} A + BGC & 0 & 0 \\ -BGC & A + BGC & 0 \\ 0 & -BGC & A + BGC \end{bmatrix} + \varepsilon \begin{bmatrix} BGC & BGC & BGC \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

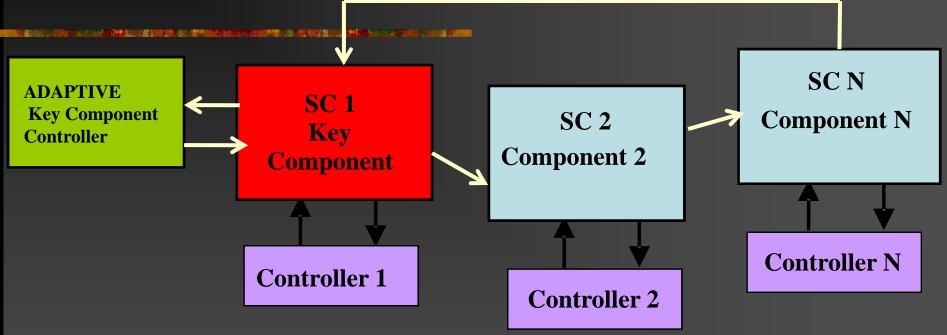
$$\overline{A}_{0} \text{ stable} \Leftrightarrow A + BGC \text{ and } A + 2BGC \text{ stable}$$

$$\Rightarrow \exists \varepsilon_{0} > 0 \Rightarrow, \forall 0 \le \varepsilon < \varepsilon_{0}, \ \overline{A}_{C}(\varepsilon) \text{ is stable}.$$

Cor.
$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = \begin{bmatrix} 1 & 0.1 \end{bmatrix}$$

 $\Rightarrow \overline{A}_C(\varepsilon)$ stable $\forall 0 \le \varepsilon < \varepsilon_0 = 0.02$ (theory $\sim 10^{-13}$), and unstable for $\varepsilon = 1$

ADAPTIVE Key Component Controller



Was Unstable, but Adaptive Key Component Controller Restores Stability

Example

Double Integrator:
$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, B = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, C = \begin{bmatrix} 1 & 0.1 \end{bmatrix}$$

$$C(s) \equiv -(10 + \frac{1}{s})$$

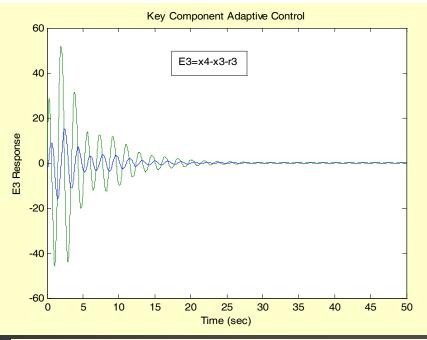
$$r_1 \equiv \begin{bmatrix} 3 \\ 0 \end{bmatrix}; r_2 \equiv \begin{bmatrix} 5 \\ 0 \end{bmatrix}; r_3 \equiv \begin{bmatrix} 8 \\ 0 \end{bmatrix}; r_4 \equiv \begin{bmatrix} 16 \\ 0 \end{bmatrix}$$

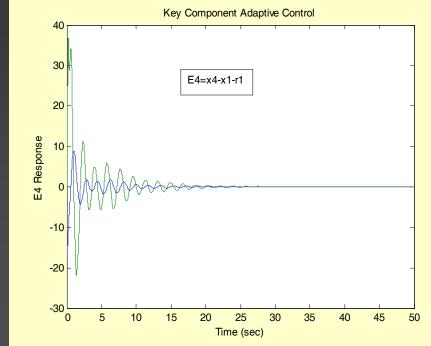
Key Component Adaptive Controller

$$u_{4} = \underbrace{GC\xi_{3}}_{Original} + \underbrace{G_{e}\widetilde{y}_{4} + G_{D}\phi_{D}}_{Adaptive}; \phi_{D} \equiv 1$$
Original Adaptive Control

Adaptive Gains $\begin{cases} \dot{G}_e = -(\tilde{y}_4)^2 \gamma_e; \gamma_e \equiv 100 \\ \dot{G}_D = -\tilde{y}_4 \phi_D \gamma_e; \gamma_e \equiv 1 \end{cases}$

Measured Output : $\tilde{y}_4 = Cx_4$





Conjecto-Theorem

If individual spacecraft dynamics (A, B, C) are ASPR,

ie CB > 0 and P(s) = $C(sI - A)^{-1}B$ is minimum phase, then

Key Component Adaptive Controller (on the joining spacecraft)

$$u_{N} = \underbrace{GC\xi_{N-1}}_{Original} + \underbrace{G_{e}\widetilde{y}_{N} + G_{D}\phi_{D}}_{Adaptive}; \phi_{D} \text{ bounded}$$

$$\underbrace{GC\xi_{N-1}}_{Original} + \underbrace{G_{e}\widetilde{y}_{N} + G_{D}\phi_{D}}_{Control}; \phi_{D} \text{ bounded}$$

Adaptive Gains
$$\begin{cases} \dot{G}_e = -(\tilde{y}_N)^2 \gamma_e; \gamma_e \equiv 100 \\ \dot{G}_D = -\tilde{y}_N \phi_D \gamma_e; \gamma_e \equiv 1 \end{cases}$$

Measured Output : $\tilde{y}_N = Cx_N$

produces $\xi_k = x_{k+1} - x_k - r_k \xrightarrow[t \to \infty]{} 0 \forall k$ (and rejects bounded disturbances) with bounded adaptive gains

Proof: WLOG use 4 SC

Let
$$\xi = \begin{bmatrix} \xi_1 \\ \xi_2 \\ \xi_3 \end{bmatrix} \Rightarrow \begin{cases} \dot{\xi} = \overline{A}_C(\varepsilon)\xi + \widetilde{B}u_4 \\ \widetilde{y}_4 = C_4x_4 = \widetilde{C}\xi \end{cases}$$

where $\overline{A}_C(\varepsilon) = \begin{bmatrix} A + (1+\varepsilon)BGC & \varepsilon BGC & \varepsilon BGC \\ -BGC & A + BGC & 0 \\ 0 & -BGC & A + BGC \end{bmatrix}$

$$= \begin{bmatrix} A & 0 & 0 \\ 0 & A & 0 \\ 0 & 0 & A \end{bmatrix} + \begin{bmatrix} B & 0 & 0 \\ 0 & B & 0 \\ 0 & 0 & B \end{bmatrix} \begin{bmatrix} (1+\varepsilon)G & \varepsilon G & \varepsilon G \\ -G & G & 0 \\ 0 & 0 & C \end{bmatrix} \begin{bmatrix} C & 0 & 0 \\ 0 & C & 0 \\ 0 & 0 & C \end{bmatrix}$$

$$\xrightarrow{\check{A}} \xrightarrow{\check{B}} \xrightarrow{\check{G}(\varepsilon)} \xrightarrow{\check{G}(\varepsilon)} \xrightarrow{\check{C}}$$

Find
$$G_* \ni u_4 = G_* \widetilde{y}_4$$

$$\Rightarrow \widetilde{A}_C(\varepsilon) = \overline{A}_C(\varepsilon) + \begin{bmatrix} 0 \\ 0 \\ B \end{bmatrix} G_* \begin{bmatrix} 0 & 0 & C \end{bmatrix}$$

$$= (\begin{bmatrix} A & 0 & 0 \\ 0 & A & 0 \\ 0 & 0 & A \end{bmatrix} + \begin{bmatrix} B & 0 & 0 \\ 0 & B & 0 \\ 0 & 0 & B \end{bmatrix} \begin{bmatrix} (1+\varepsilon)G & \varepsilon G & \varepsilon G \\ -G & G & 0 \\ 0 & 0 & G \end{bmatrix} \begin{bmatrix} C & 0 & 0 \\ 0 & C & 0 \\ 0 & 0 & C \end{bmatrix}$$

$$+ \begin{bmatrix} B & 0 & 0 \\ 0 & B & 0 \\ 0 & 0 & B \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} G_* \begin{bmatrix} 0 & 0 & I \end{bmatrix} \begin{bmatrix} C & 0 & 0 \\ 0 & C & 0 \\ 0 & 0 & C \end{bmatrix}$$

$$= \overline{A} + \overline{B}(\overline{G} + \overline{G}_*)\overline{C}$$

Clearly $\overline{CB} > 0 \Leftrightarrow CB > 0$

and $\overline{P}(s)$ minimum phase $\Leftrightarrow P(s)$ minimum phase.

$$: (A,B,C)ASPR \Rightarrow (\overline{A},\overline{B},\overline{C})ASPR \Rightarrow (\overline{A}_{C}(\varepsilon),\widetilde{B},\widetilde{C})ASPR \#$$

Future Formation Stuff

Nonlinear:
$$\begin{cases} \dot{x}_k = Ax_k + \underbrace{g(x_k)}_{\text{Lipschitz \& weak enough}} + Bu_k \\ y_k = Cx_k \end{cases}$$

Harder but Doable

$$\begin{cases} \dot{x}_k = A(x_k) + B(x_k)u_k \\ y_k = C(x_k) \end{cases}$$

•Maintaining a Formation Shape (Tracking); Immutability of Formation Shapes

•Swarms: George Hill's Eqs (or Johnny-Come-Lately: Clohessy-Wiltshire)

A Sortof Paradigm

