

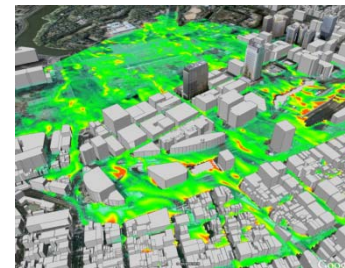
IFAC ROCOND2012, Aalborg, Denmark
June 21, 2012

Robustness in Networked Dynamical Systems

Shinji HARA
(The University of Tokyo, Japan)

Future Direction in Control

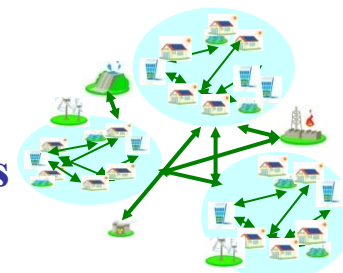
Realization of High Quality Products
→ Solving Social Problems such as
Energy, Environments, and Medicine



Meteorological Phenomena



Bio-systems



Power NWs



Transportation

Multiple Functions

Hybrid Control

High Performance

Robust Control

Linear motor car

Automation

Modern Control

Engine control

Stabilization

Classical Control

Robotics
Aerospace

Mechatronics

Steel process

Chemical process

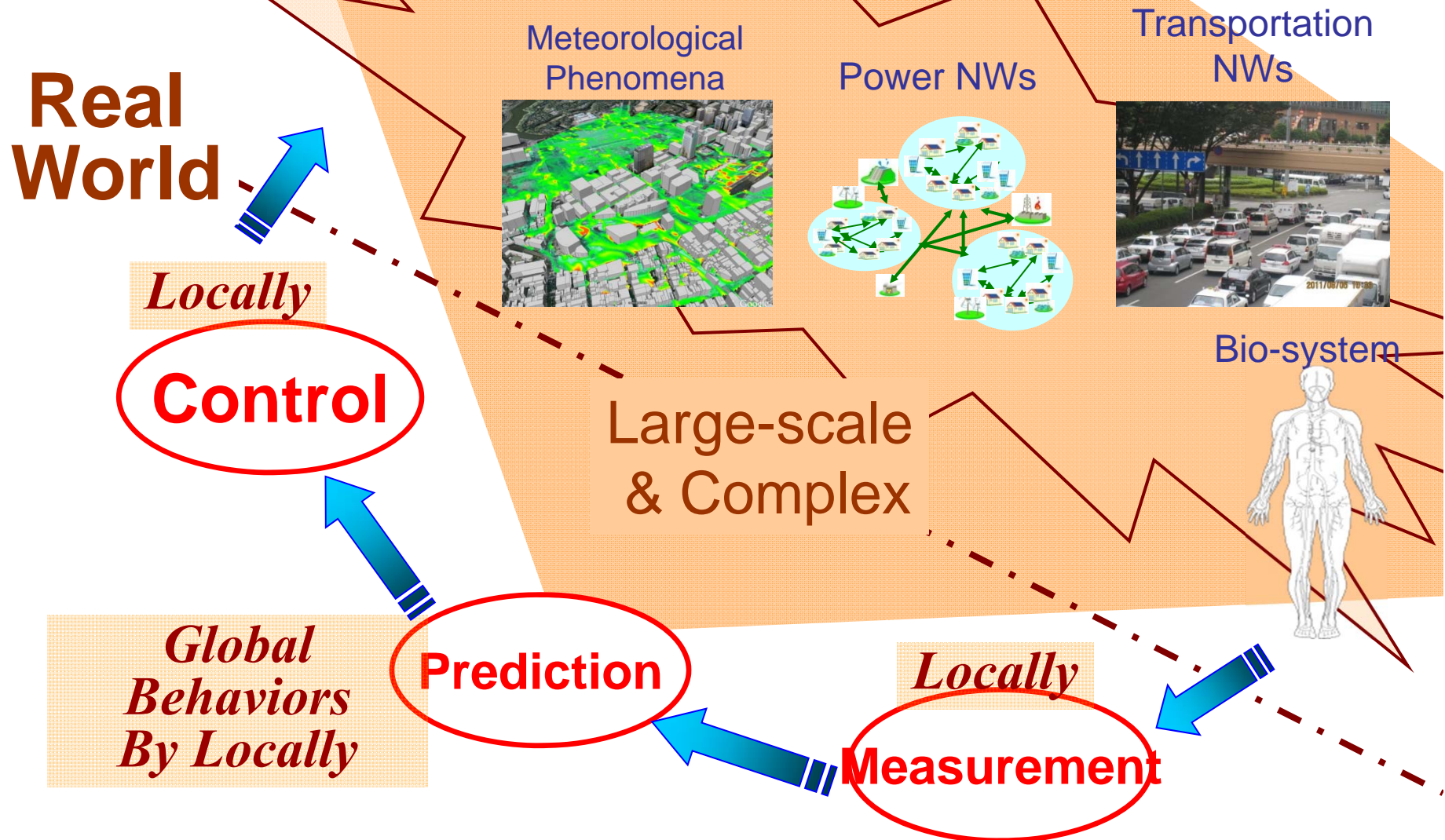
Glocal Control

Watt



Idea of “Glocal Control” ?

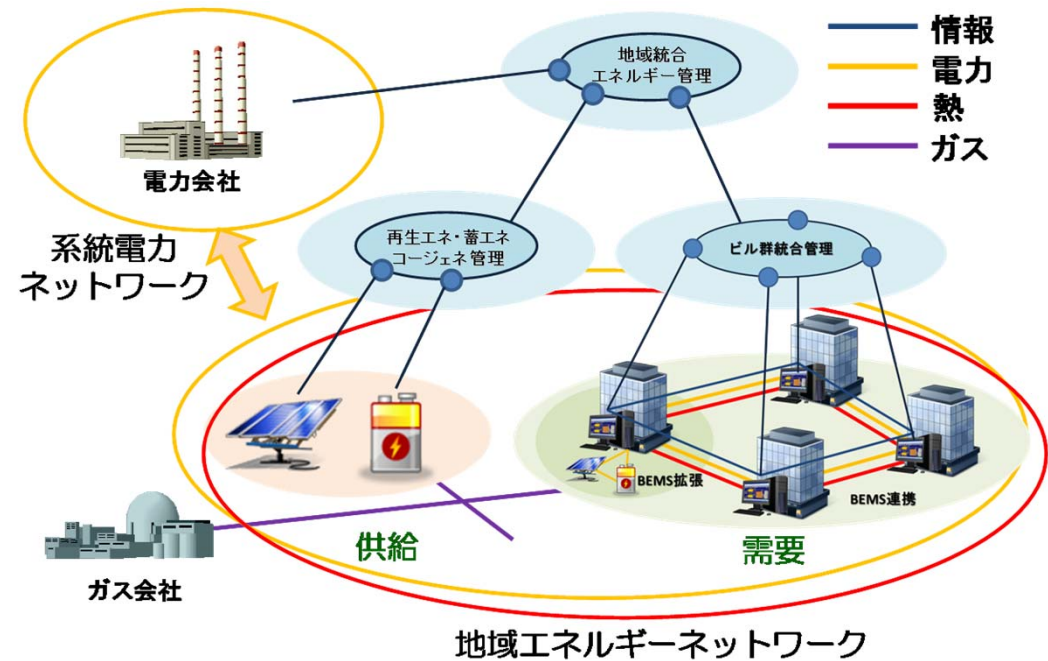
Glocal Control : to achieve desired global behavior by only **local** actions of measurement and control



Smart Energy NW and Energy Saving

Smart Energy Network

Electric power network
+ Gas energy network



Hierarchical Air Conditioning System

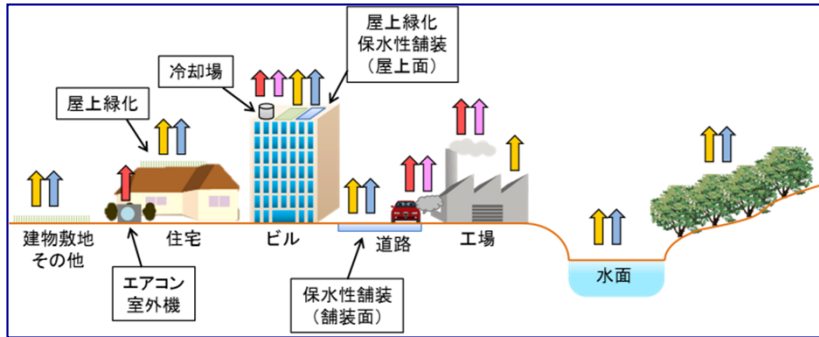
Area: Group of buildings

Building: Set of floors

Floor: Set of rooms

Urban Heat Island Problem

Local Actions of Measurement & Control

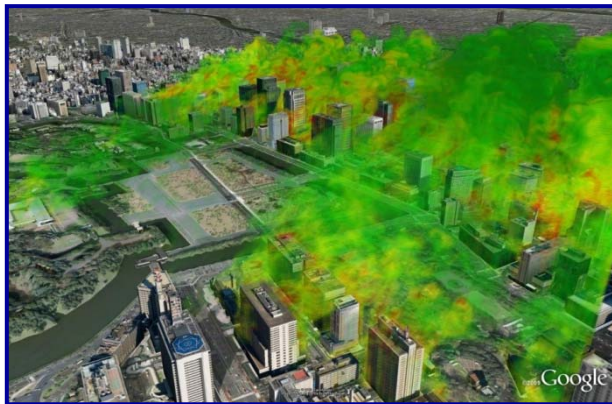


Scale of buildings and roads

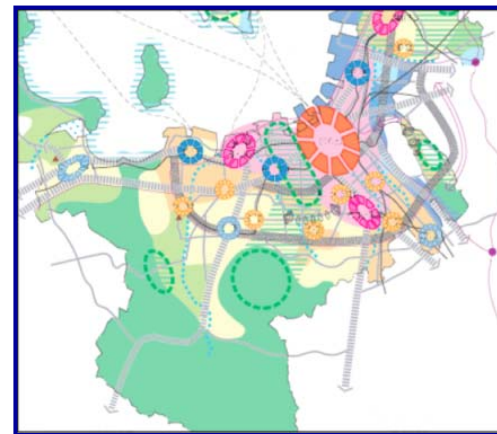
Realization of Global Desired Environment of a Whole City



Glocal Control

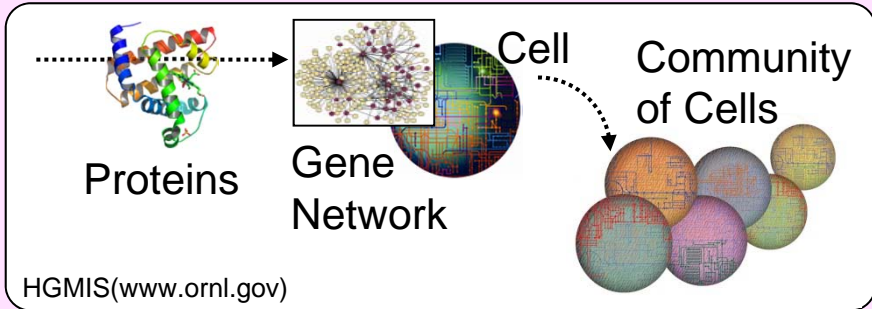


Scale of residential and business areas

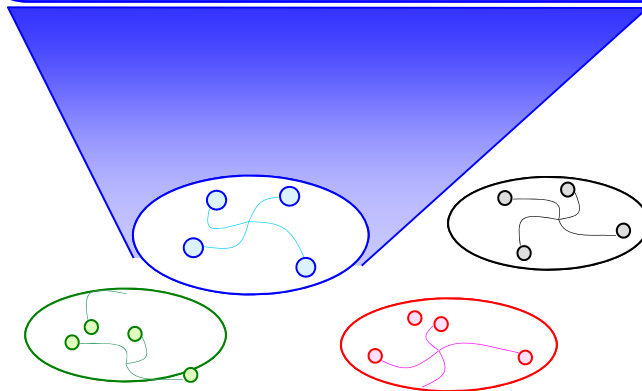
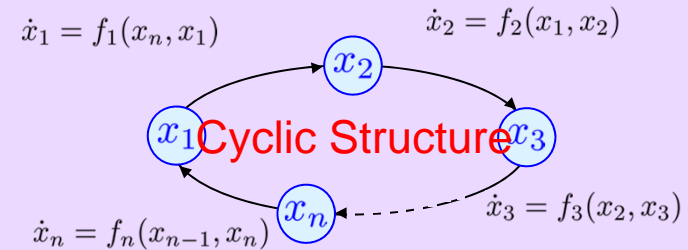
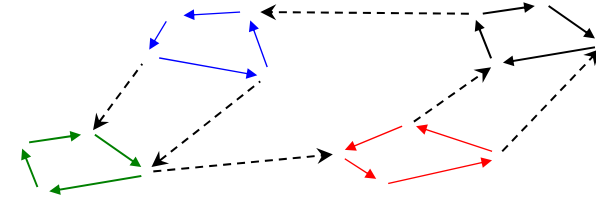
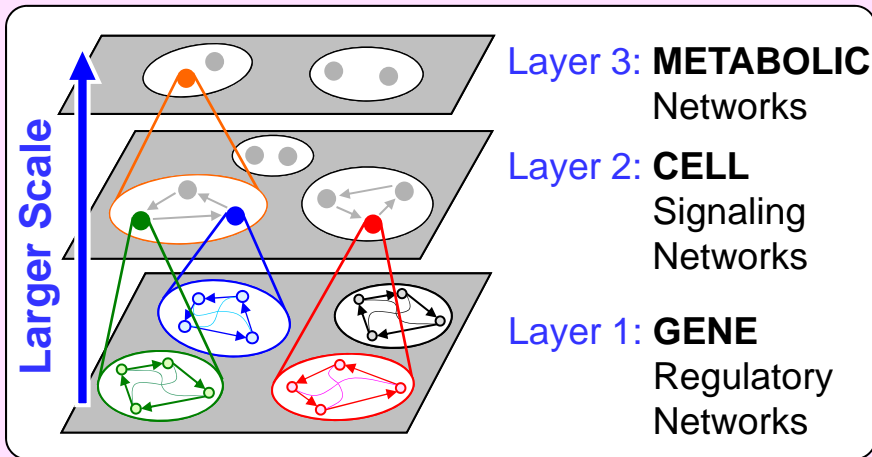


Scale of districts/towns

Hierarchical Bio-Network Systems



Hierarchical Bio-Network Systems



- : Subsystem 1 ○ : Subsystem 2
- : Subsystem 3 ○ : Subsystem 4

Framework for Glocal Control

**Realization of Global Functions
by Local Measurement and Control**

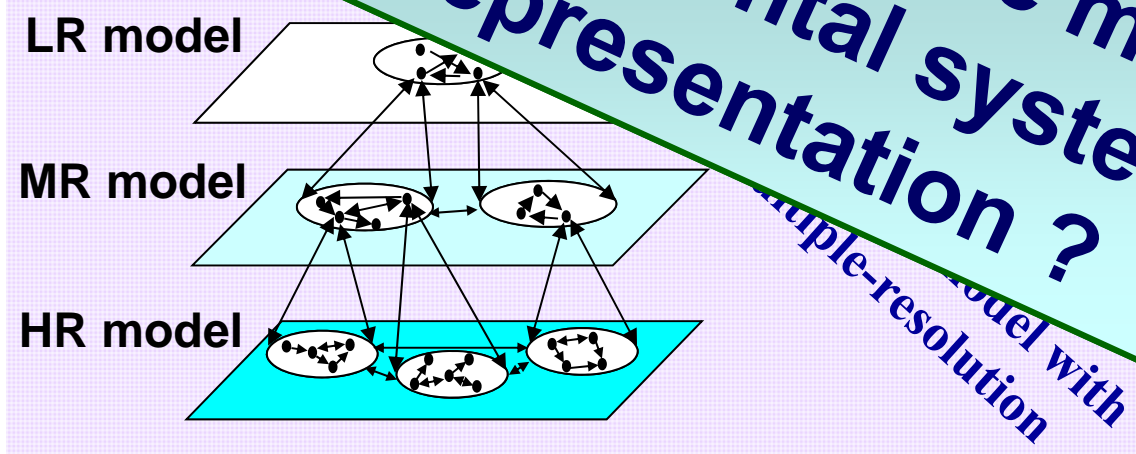
Real World



**Glocal Control
System**

**Hierarchical Dynamical Systems
with Multi-resolution**

**Q1: What is the most
fundamental system
representation ?**

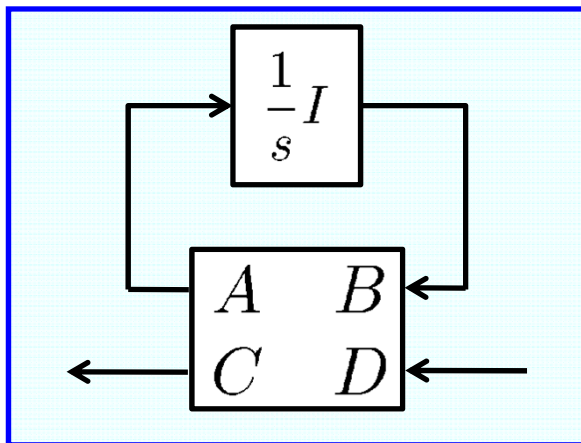


**Local
Measurement**

LTI System with Generalized Frequency Variable

A unified representation for homogeneous multi-agent dynamical systems

$$C(sI - A)^{-1}B + D$$



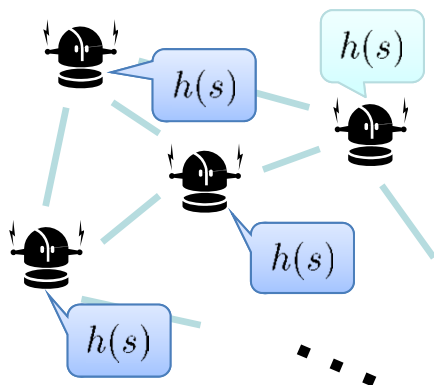
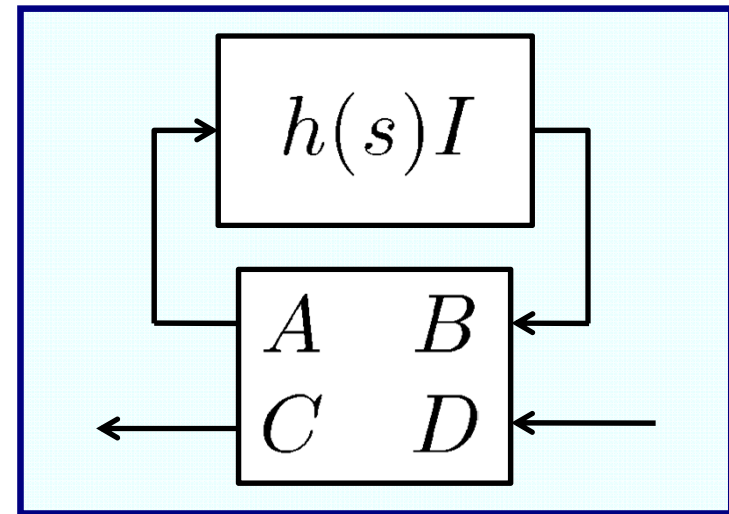
$$1/s \rightarrow h(s)$$



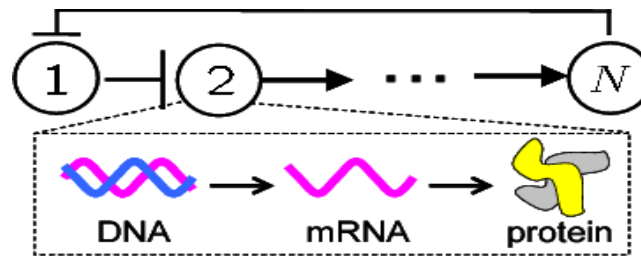
$$\Phi(s) = 1/h(s)$$

Generalized
Freq. Variable

$$C(\phi(s)I - A)^{-1}B + D$$



Group Robot



Gene Reg. Networks

**Dynamics
+
Information
Structure**

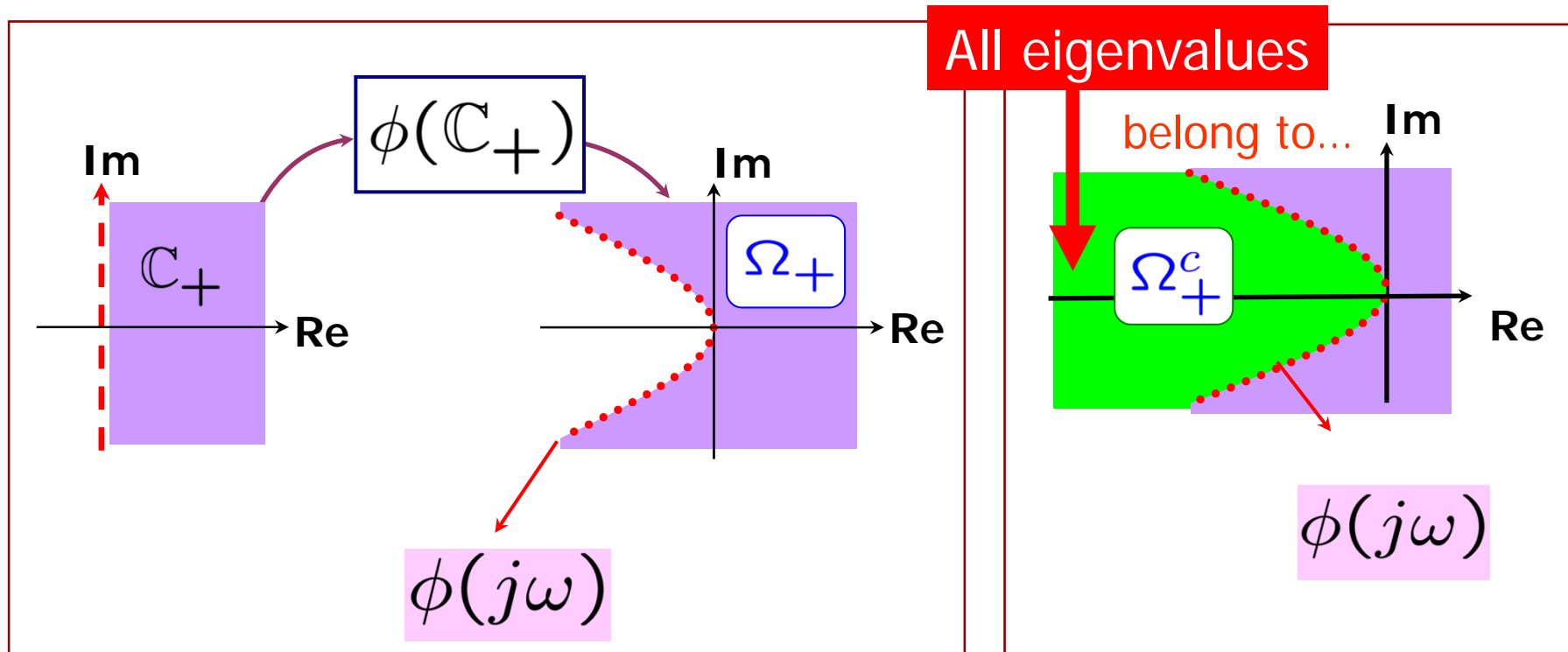
OUTLINE

- 1. Stability Analysis: Review**
- 2. D-Stability Analysis**
- 3. Robust Stability Analysis**
- 4. Application to Gene Regulatory Networks**
- 5. Nonlinear Stability Analysis**
- 6. Concluding Remarks**

Stability Region for LTISwGFV

(Hara et al. IEEE CDC, 2007)

- ❖ Define: Domains $\Omega_+ := \phi(\mathbb{C}_+)$, $\Omega_+^c := \mathbb{C} \setminus \Omega_+$



Q2A: *How to characterize the region ?*

Q2B: *How to check the condition ?*

Stability Tests for LTISwGFV

Graphical	Algebraic	Numeric (LMI)
Nyquist – type	Hurwitz – type	Lyapunov – type
Fax & Murray (2004) Hara et al. (2007)	Tanaka, Hara, Iwasaki (2009)	Tanaka, Hara, Iwasaki (2009)
$h(s)$ and $\sigma(A)$	$h(s)$ and $\sigma(A)$	$h(s)$ and A

Characteristic
Polynomial

Hurwitz test for
complex
coefficients

Generalized
Lyapunov Ineq.

$$p(\lambda, s) := d(s) - \lambda n(s) \quad \lambda \in \sigma(A) \quad (\text{complex})$$

Stability Conditions

(Tanaka et al., ASCC, 2009)

Given $h(s) = n(s)/d(s)$, A $\mathcal{H}_A(s)$ is stable



$\sigma(A) \subset \Lambda := \{ \lambda \in \mathbb{C} \mid d(s) - \lambda n(s) \text{ is Hurwitz stable} \}$

Key
lemma

Algebraic condition

$$\sigma(A) \subset \bigcap_{k=1}^{\nu} \Sigma_k$$

$$\Sigma_k := \{ \lambda \in \mathbb{C} \mid l_k(\lambda)^* \Phi_k l_k(\lambda) > 0 \}$$

$$(k = 1, 2, \dots, \nu)$$

Extended
Routh-Hurwitz
Criterion [Frank,1946]

Generalized Lyapunov inequality

LMI feasibility problem

$$X_k = X_k^T > 0 \text{ s.t. } L_k(A)^T (\Phi_k \otimes X_k) L_k(A) > 0$$

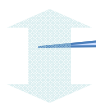
for each $k = 1, 2, \dots, \nu$

$$l_\ell(\lambda) := \begin{bmatrix} 1 \\ \lambda \\ \vdots \\ \lambda^\ell \end{bmatrix}, \quad L_\ell(A) := \begin{bmatrix} I \\ A \\ \vdots \\ A^\ell \end{bmatrix}$$

Numerical Example : 2nd order (1/2)

Given $h(s) = \frac{2s + 1}{s^2 + s + 1}$, $A \in \mathbb{R}^{n \times n}$

$\sigma(A) \subset \Lambda := \{ \lambda \in \mathbb{C} \mid (s^2 + s + 1) - \lambda(2s + 1) \text{ is Hurwitz stable} \}$

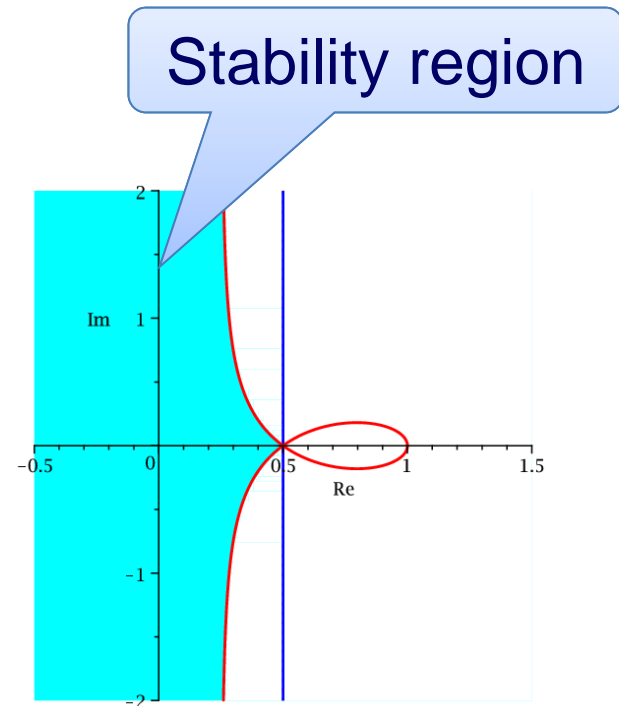
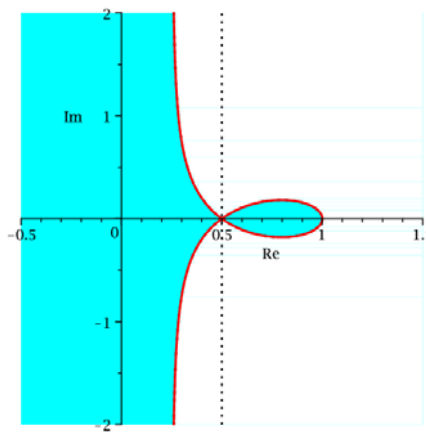
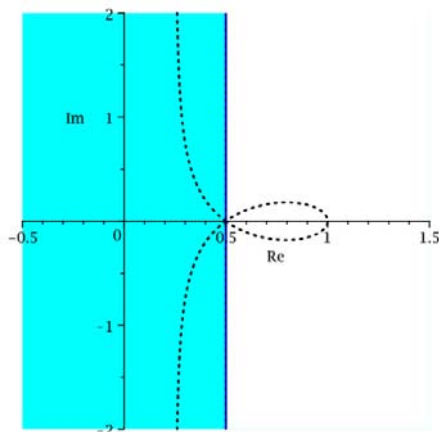


Extended Routh-Hurwitz Criterion (Frank, 1948)

$\sigma(A) \subset \Sigma := \{ \lambda \in \mathbb{C} \mid \Delta_1 > 0 \text{ and } \Delta_2 > 0 \}$

$$\Delta_1 = \begin{bmatrix} 1 \\ \lambda \end{bmatrix}^* \begin{bmatrix} 1 & -1 \\ -1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ \lambda \end{bmatrix} > 0$$

$$\Delta_2 = \frac{1}{4} \begin{bmatrix} 1 \\ \lambda \\ \lambda^2 \end{bmatrix}^* \begin{bmatrix} 4 & -10 & 7 \\ -10 & 18 & -8 \\ 7 & -8 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ \lambda \\ \lambda^2 \end{bmatrix} > 0$$



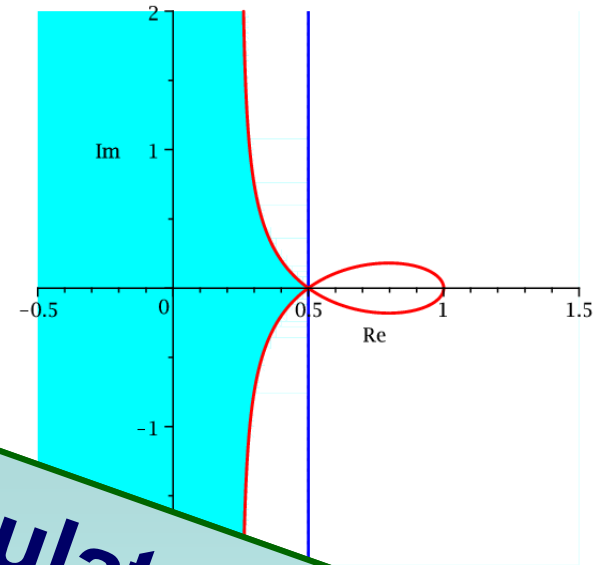
Numerical Example : 2nd order (2/2)

Given $h(s) = \frac{2s + 1}{s^2 + s + 1}$, $A \in \mathbb{R}^{n \times n}$

$\sigma(A) \subset \{ \lambda \in \mathbb{C} \mid \Delta_1 > 0 \text{ and } \Delta_2 > 0 \}$

$\Delta_1 =$

$\Delta_2 = \frac{1}{4} \begin{bmatrix} \lambda & \lambda^2 \\ \lambda^2 & \lambda \end{bmatrix}$



Q3: How can we calculate Φ_k ?
Is there any systematic way for doing it ?

Generalized Lyapunov

$$X_1 = X_1^T > 0 \quad \text{s.t.} \quad \begin{bmatrix} I \\ A \end{bmatrix}^T \left(\begin{bmatrix} 1 & -1 \\ -1 & 0 \end{bmatrix} \otimes X_1 \right) \begin{bmatrix} I \\ A \end{bmatrix} > 0$$

$$X_2 = X_2^T > 0 \quad \text{s.t.} \quad \begin{bmatrix} I \\ A \\ A^2 \end{bmatrix}^T \left(\begin{bmatrix} 4 & -10 & 7 \\ -10 & 18 & -8 \\ 7 & -8 & 0 \end{bmatrix} \otimes X_2 \right) \begin{bmatrix} I \\ A \\ A^2 \end{bmatrix} > 0$$

Algorithm

Given Data: all coefficients of numerator and denominator of $h(s)$

Algorithm h2Phi($h(s)$):

Input : $h(s) = \frac{b_1 s^{\nu-1} + \dots + b_\nu}{s^\nu + a_1 s^{\nu-1} + \dots + a_\nu}$

Output : ℓ_k and Φ_k

```

1.  $p_0 \leftarrow 1, q_0 \leftarrow 0$ 
   for  $i \leftarrow 1$  until  $2\nu - 1$  do
     if  $i \leq \nu$  then
        $p_i \leftarrow a_i - b_i x, q_i \leftarrow -b_i y$ 
     else
        $p_i \leftarrow 0, q_i \leftarrow 0$ 
2.  $\Delta_1 \leftarrow p_1$ 
   for  $k \leftarrow 2$  until  $2\nu$  do
      $M \leftarrow O^{(2k-1) \times (2k-1)}$ 
     for  $i \leftarrow 1$  until  $k - (i-2)/2$  do
       if  $i$  is even
         for  $j \leftarrow 1$  until  $\nu$  do
            $\Delta_k \leftarrow \Delta_k((\lambda + \bar{\lambda})/2, (\lambda - \bar{\lambda})/2j)$ 
         until  $\nu$  do
            $\Delta_k \leftarrow \Delta_k((\lambda + \bar{\lambda})/2, (\lambda - \bar{\lambda})/2j)$ 
            $\ell_k \leftarrow \text{maximum of the degree of } \lambda \text{ in } \Delta'_k(\lambda, \bar{\lambda}) - 1$ 
         until  $\ell_k - 1$  do
            $\Phi_k(m+1, l+1) \leftarrow \Delta_k(m+1, l+1)$ 
            $\Phi_k(m+1, l+1) \leftarrow \text{the coefficient of } \lambda^m \bar{\lambda}^l \text{ in } \Delta'_k(\lambda, \bar{\lambda})$ 
5. return  $\ell_k$  and  $\Phi_k$ .

```

Systematic methods for stability analysis Hurwitz-type & LMI

Result:
 $\Phi_k \quad (k = 1, 2, \dots, \nu)$

Numerical Example : 4th order

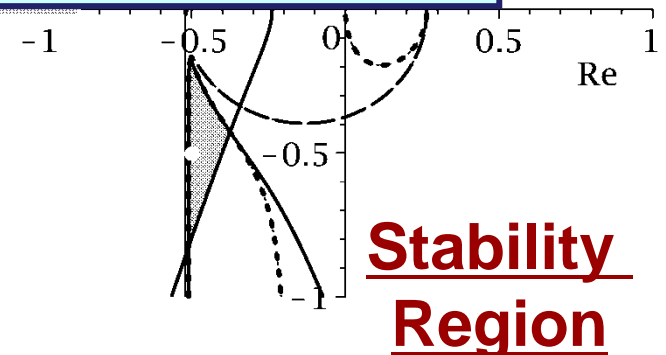
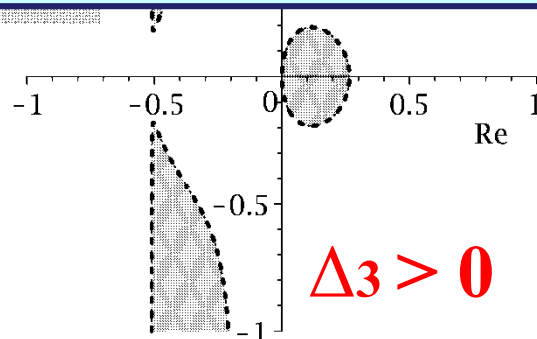
$$h(s) = \frac{100(s+2)\left(\frac{19}{10}s^2 - \frac{1}{500000}s + \frac{21}{10}\right)}{(s-1)^2(s+1)(s+100)}$$

Unstable
Non-minimum phase

Summary 1: LTI System with GFV

- One of nice classes for NDSs
- Stability and robust stability conditions
- Many application fields

synchronization of coupled oscillators
bio-systems



Further Fundamental Questions

Robustness Issues in
Networked Dynamical Systems
(LTI systems with GFV) ?

Q4 : *D-Stability Analysis* ?

Stability Margins

Q5 : *Robust Stability Analysis* ?

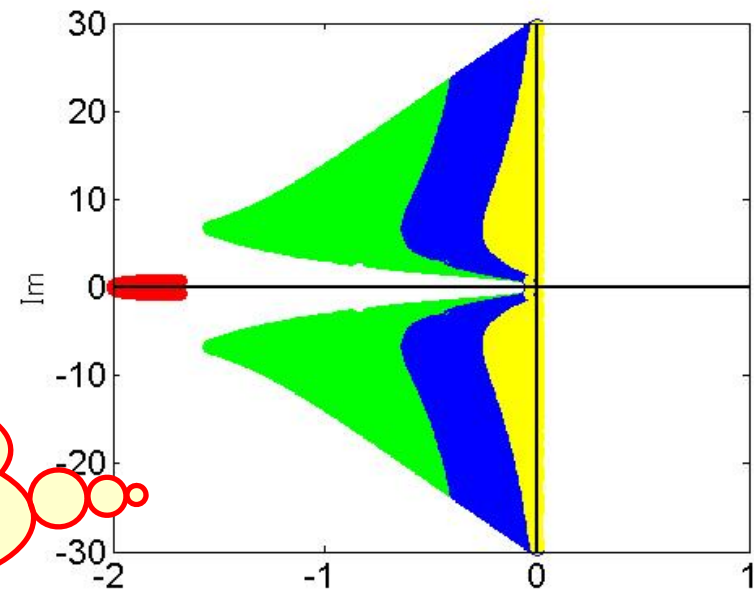
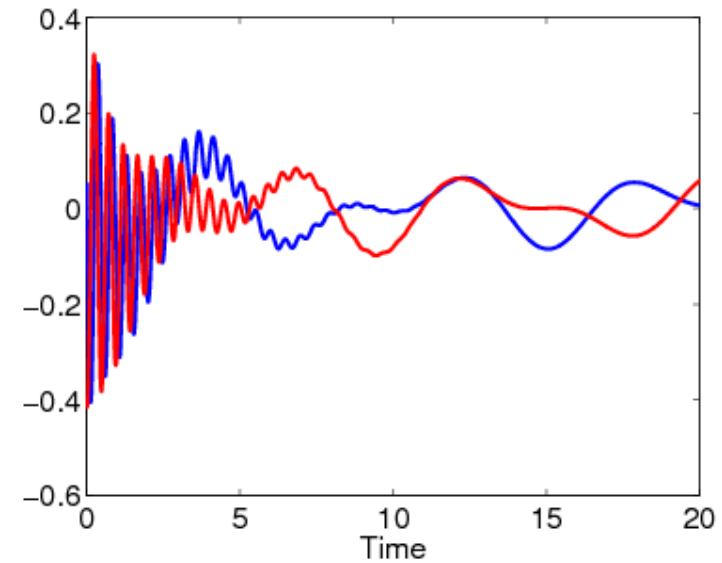
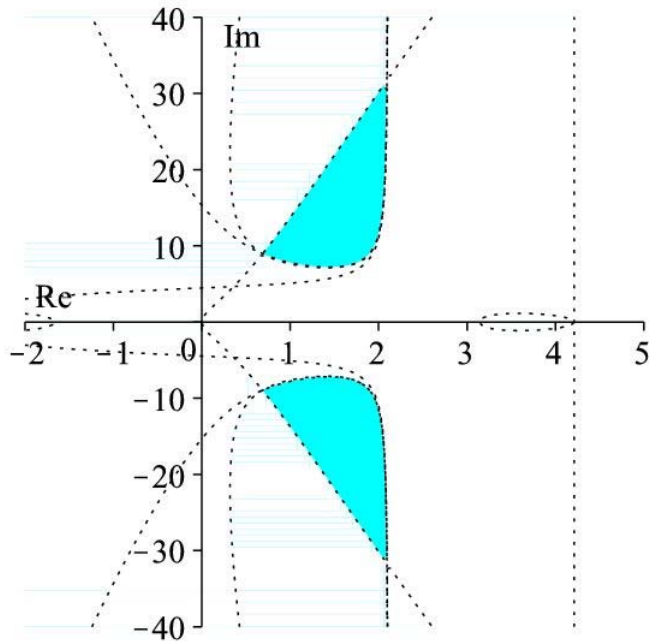
Homogeneous → Heterogeneous

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Why D-Stability Analysis ?

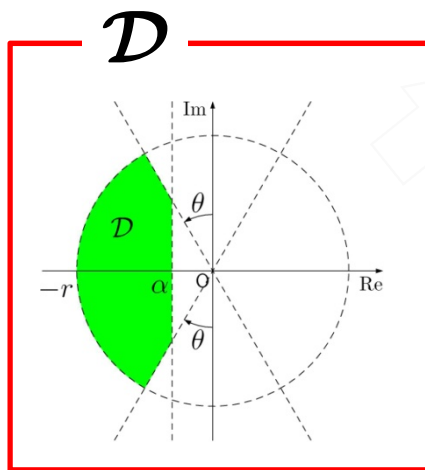
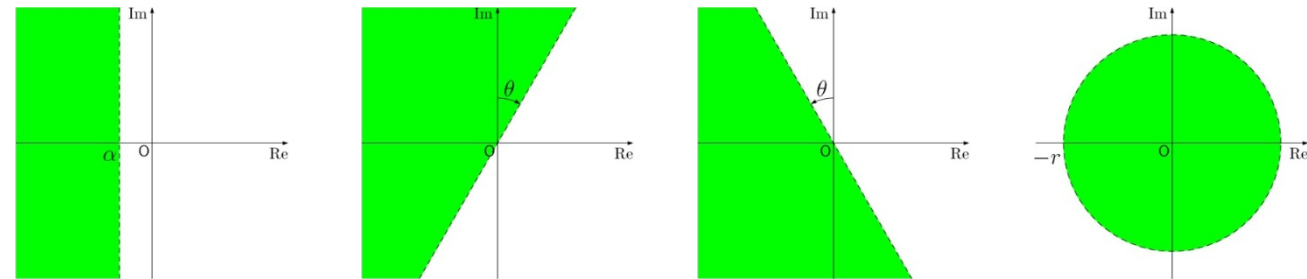
$$h(s) = \frac{(\frac{1}{2}s + 1)(\frac{19}{10}s^2 - \frac{1}{500}s + \frac{21}{10})}{s(s - 2)(s + 1)(s + 5)}$$



We can not assign all the closed-loop poles as you want.

Closed-loop Poles

Unified Approach to D-Stability Analysis



Unified Stability Analysis for
Disks and Half Planes

Derive the stability conditions

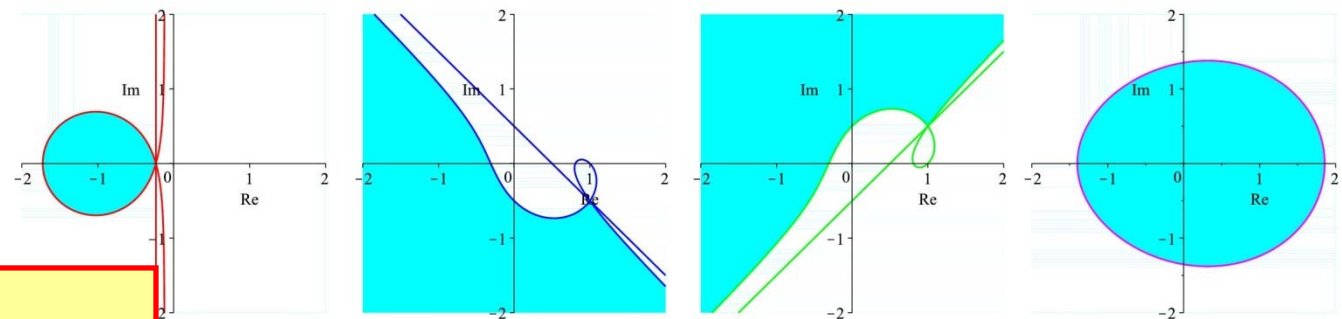
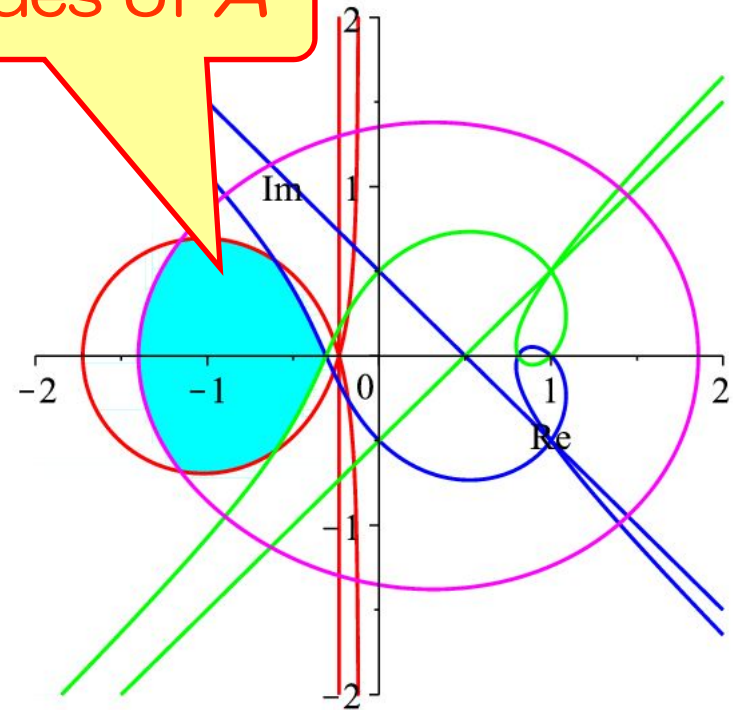
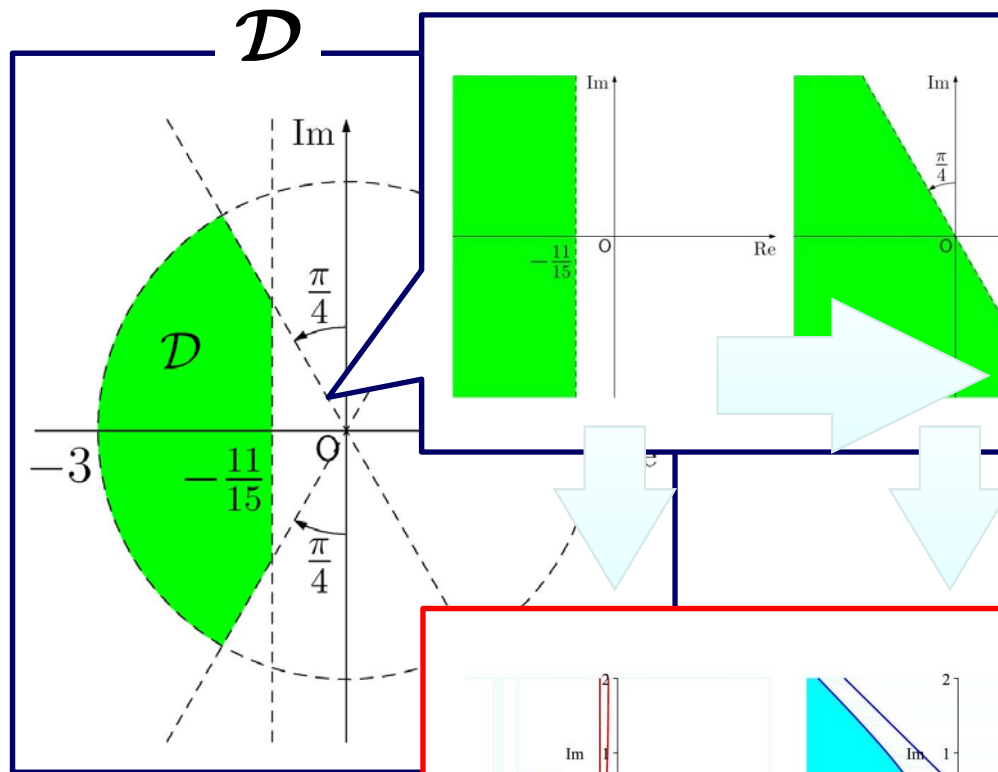
Intersection

D-stability Condition

A Numerical Example

$$h(s) = \frac{2s + 1}{s^2 + s + 1}$$

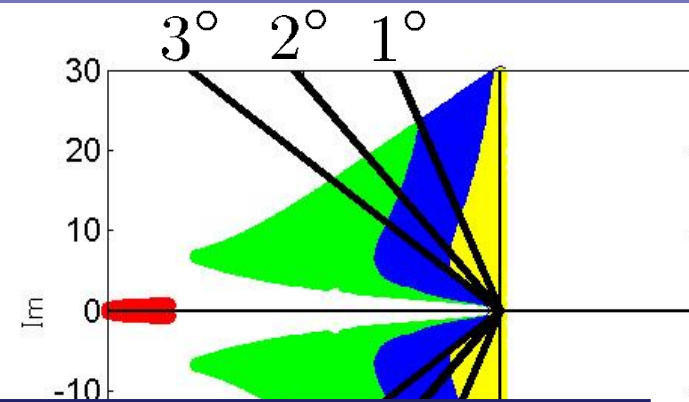
Eigenvalues of A



Stability Regions

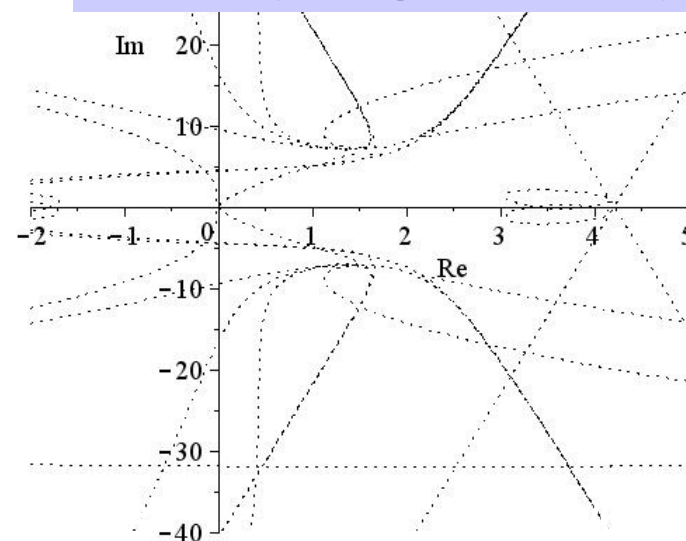
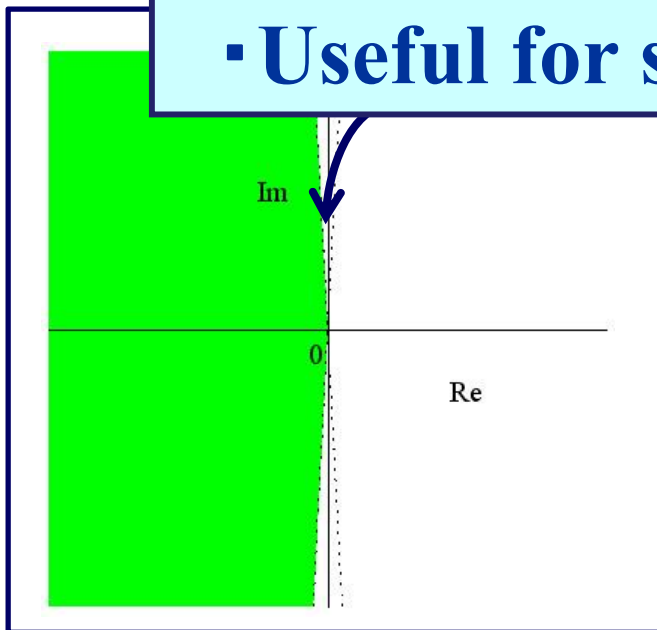
Motivating Example

$$h(s) = \frac{(\frac{1}{2}s + 1)(\frac{19}{10}s^2 - \frac{1}{500}s + \frac{21}{10})}{s(s - 2)(s + 1)(s + 5)}$$



Summary 2 : D-Stability Condition

- Complicated but straightforward
- Useful for some control performances



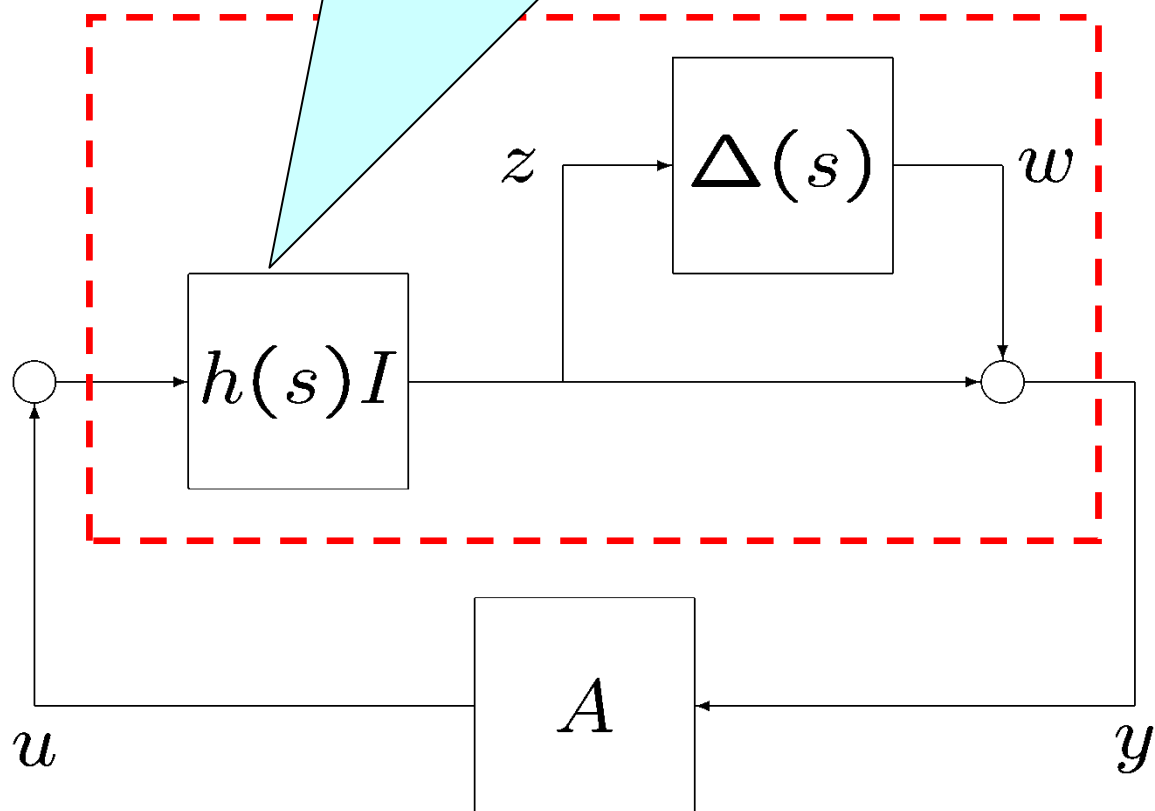
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Multiplicative Perturbations

$$\tilde{H}(s) = (I + \Delta(s)) \cdot h(s)$$

Nominal : Homogeneous



Three Classes of Perturbations

Multiplicative Perturbation:

$$\tilde{H}(s) = (I + \Delta(s)) \cdot h(s)$$

Three Classes:

Full perturbation:

$$\Delta_\gamma := \{ \Delta(s) \in \Delta_p \mid \|\Delta\|_\infty \leq 1/\gamma \}$$

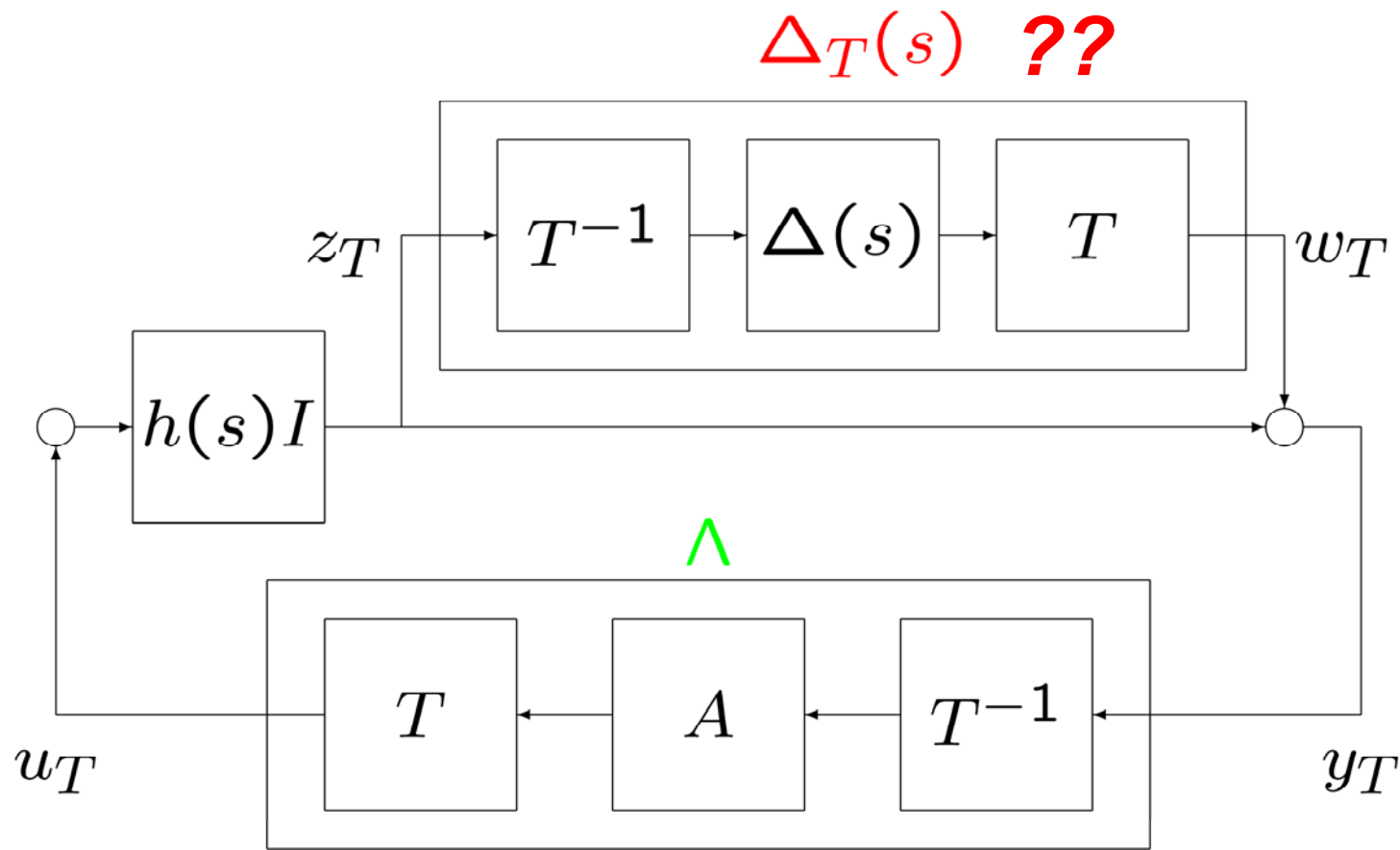
Heterogeneous:

$$\Delta_{d\gamma} := \{ \Delta(s) \in \Delta_\gamma \mid \Delta(s) : \text{diagonal} \}$$

Homogeneous:

$$\Delta_{I\gamma} := \{ \Delta(s) \in \Delta_\gamma \mid \Delta(s) = \delta(s)I \}$$

Basic Idea

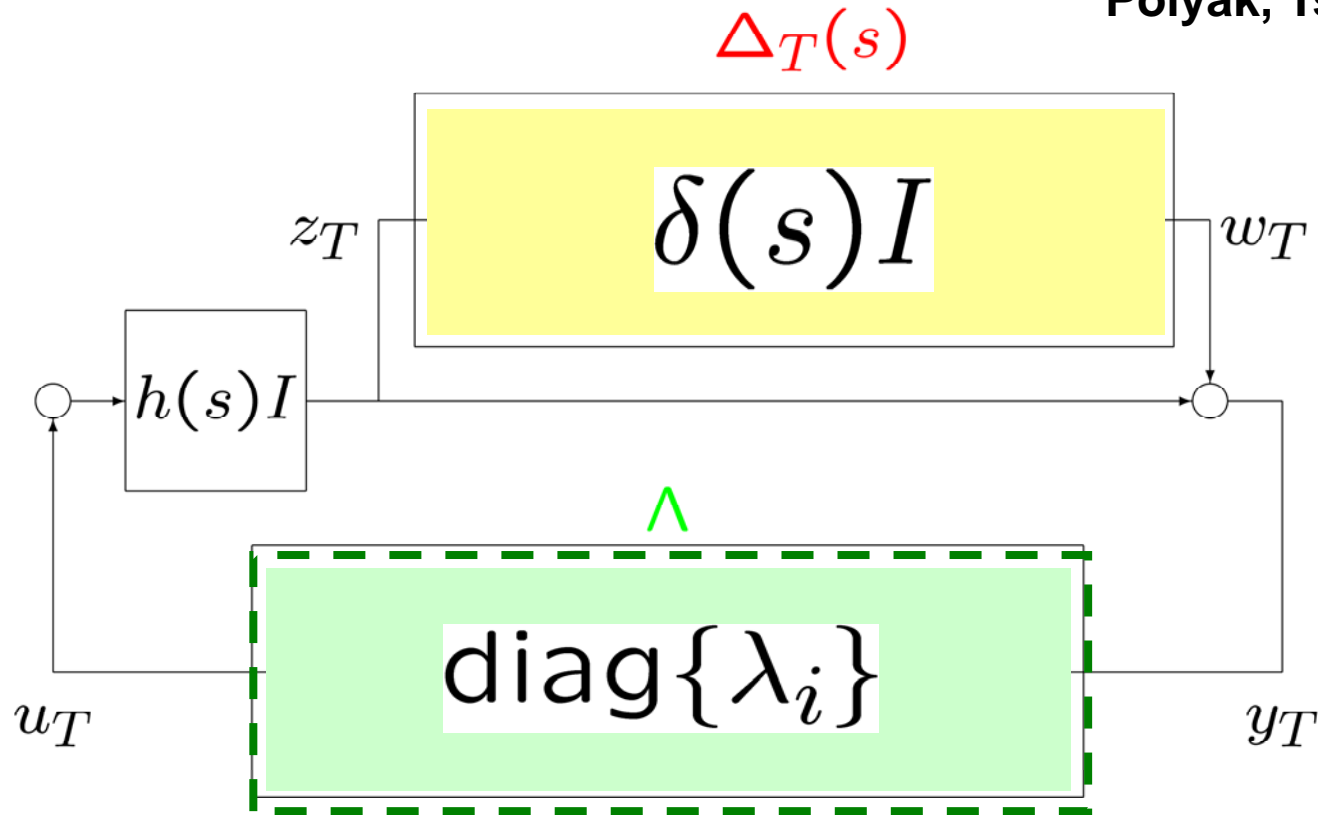


$$\Lambda := TAT^{-1} = \text{diag}\{\lambda_i\}$$

A: diagonalizable

Homogeneous Perturbations

Polyak, Tsytkin (1996)

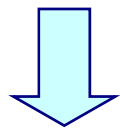


$$w_T \rightarrow z_T : \text{diag}\left\{\frac{\lambda_i h(s)}{1 - \lambda_i h(s)}\right\}$$

Complementary Sensitivity function ($h(s), \lambda_i$)

Robust Stability Condition for Homogeneous Perturbations

$$\tilde{H}(s) = (1 + \delta(s)) \cdot h(s)I$$



Small Gain Criterion

Theorem: The following three conditions are equivalent.

(i) The system is robustly stable for $\Delta_{I\gamma}$.

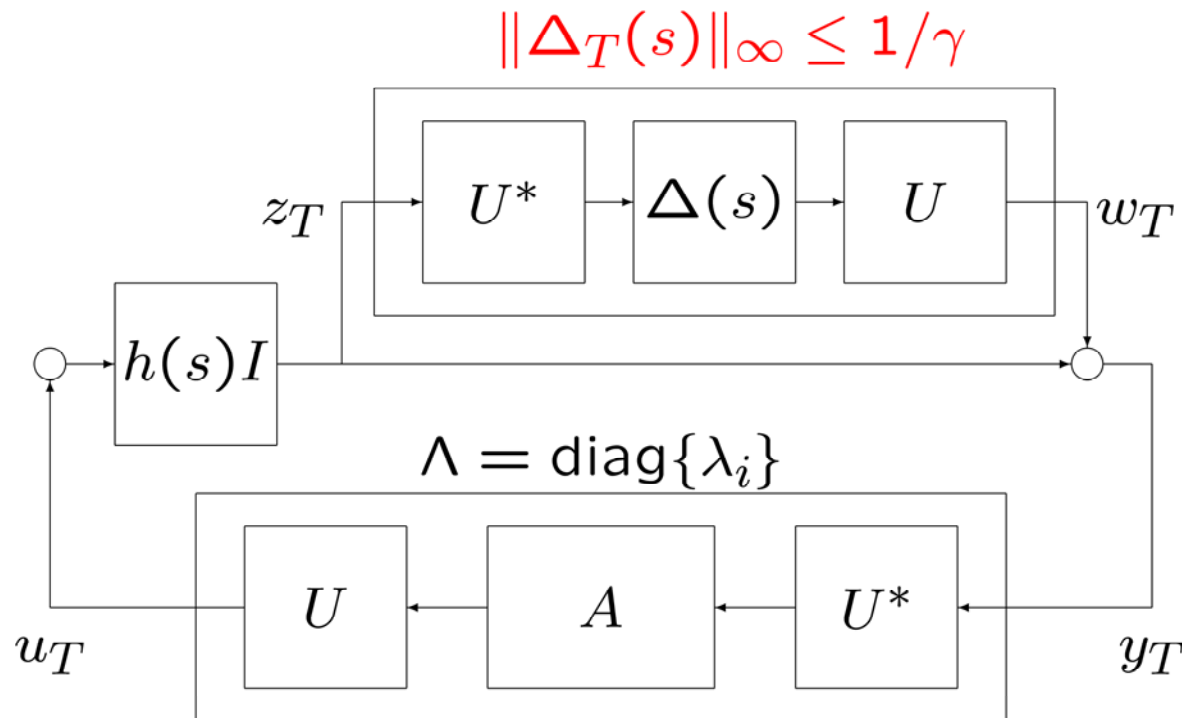
$$(ii) \left\| \frac{\lambda h}{1 - \lambda h} \right\|_{\infty} < \gamma, \quad \forall \lambda \in \sigma(A)$$

$$(iii) \left| \frac{\lambda}{\phi - \lambda} \right| < \gamma, \quad \forall \lambda \in \sigma(A),$$

$$\forall \phi \in \Phi := \{1/h(j\omega) \mid \omega \in \mathbb{R}\}.$$

A : Normal ($T = U$: Unitary Matrix)

$A \in \mathbb{R}^{n \times n}$ is normal, i.e., $A^T A = A A^T$.



- * Symmetric
- * Skew - Symmetric
- * Circulant

Sufficiency: small gain condition

Necessity: worst case $\Delta(s) = \delta(s)I$

Robust Stability Condition for Full Perturbations

Hara, Tanaka, Iwasaki (ACC2010)

Assumption

$A \in \mathbb{R}^{n \times n}$ is normal, i.e., $A^T A = A A^T$.

Theorem: The following three conditions are equivalent.

(i) The system is robustly stable for Δ_γ .

$$(ii) \quad \left\| \frac{\lambda h}{1 - \lambda h} \right\|_\infty < \gamma, \quad \forall \lambda \in \sigma(A)$$

$$(iii) \quad \left| \frac{\lambda}{\phi - \lambda} \right| < \gamma, \quad \forall \lambda \in \sigma(A),$$

$$\forall \phi \in \Phi := \{1/h(j\omega) \mid \omega \in \mathbb{R}\}.$$

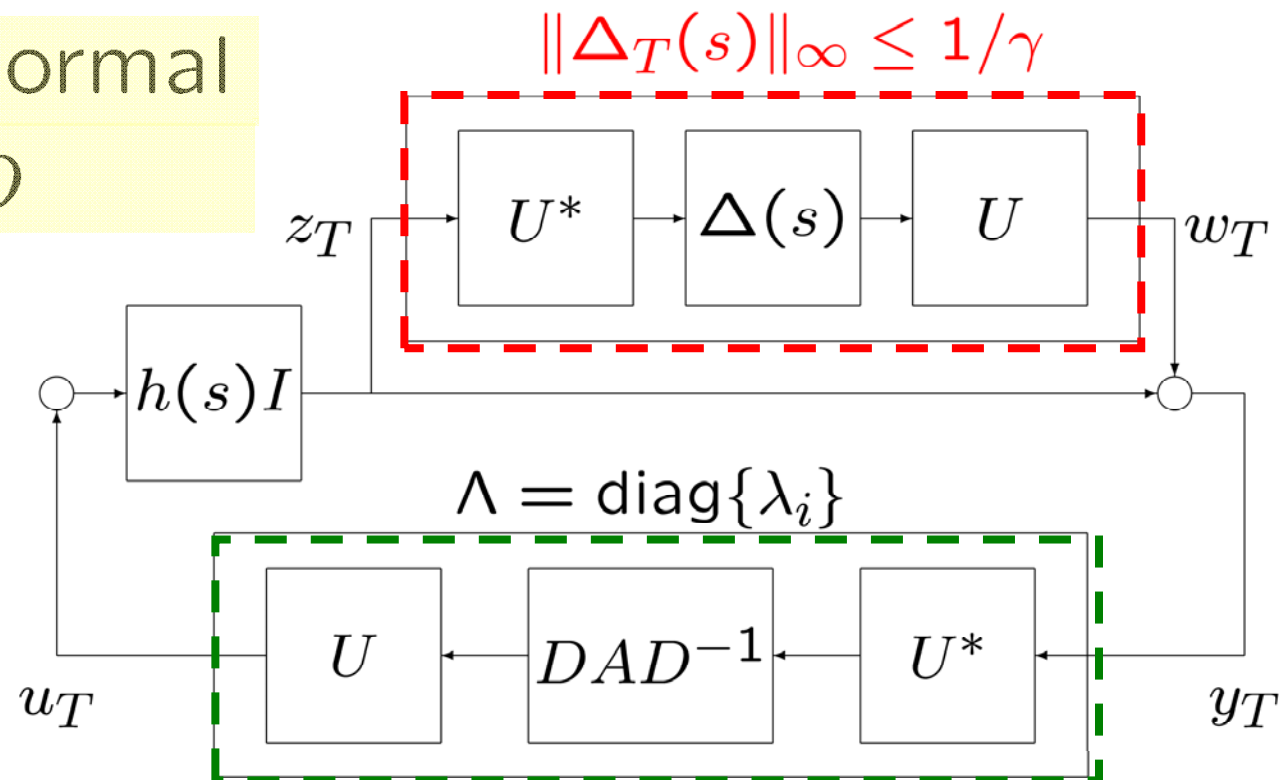
Heterogeneous Perturbations

$$\Delta(s) = \text{diag}\{\delta_i(s)\}$$

$$\forall D = \text{diag}\{d_i\} > 0 \text{ s.t. } D\Delta(s)D^{-1} = \Delta(s)$$

DAD^{-1} is normal

$$T = UD$$



Robust Stability Condition for Heterogeneous Perturbations

Assumption

$\exists D$: diagonal s.t. DAD^{-1} is normal

Theorem: The following three conditions are equivalent.

(i) The system is robustly stable for $\Delta_{d\gamma}$.


$$(ii) \quad \left\| \frac{\lambda h}{1 - \lambda h} \right\|_{\infty} < \gamma, \quad \forall \lambda \in \sigma(A)$$

$$(iii) \quad \left| \frac{\lambda}{\phi - \lambda} \right| < \gamma, \quad \forall \lambda \in \sigma(A),$$

$$\forall \phi \in \Phi := \{1/h(j\omega) \mid \omega \in \mathbb{R}\}.$$

Coprime Factor Perturbations (1/2)

$$G(s) := \begin{bmatrix} A \\ I \end{bmatrix} (I - h(s)A)^{-1} \begin{bmatrix} h(s)I & I \end{bmatrix}$$

 $A = U^* \Lambda U$

$$G(s) = \begin{bmatrix} U^* & 0 \\ 0 & I \end{bmatrix} \begin{bmatrix} \Lambda \\ I \end{bmatrix} (I - h(s)\Lambda)^{-1} \begin{bmatrix} h(s)I & I \end{bmatrix} \begin{bmatrix} U & 0 \\ 0 & I \end{bmatrix}$$

$$\|G\|_\infty < \gamma \Leftrightarrow \left\| \begin{bmatrix} \lambda \\ 1 \end{bmatrix} (1 - h\lambda)^{-1} \begin{bmatrix} h & 1 \end{bmatrix} \right\|_\infty < \gamma, \\ \forall \lambda \in \sigma(A)$$

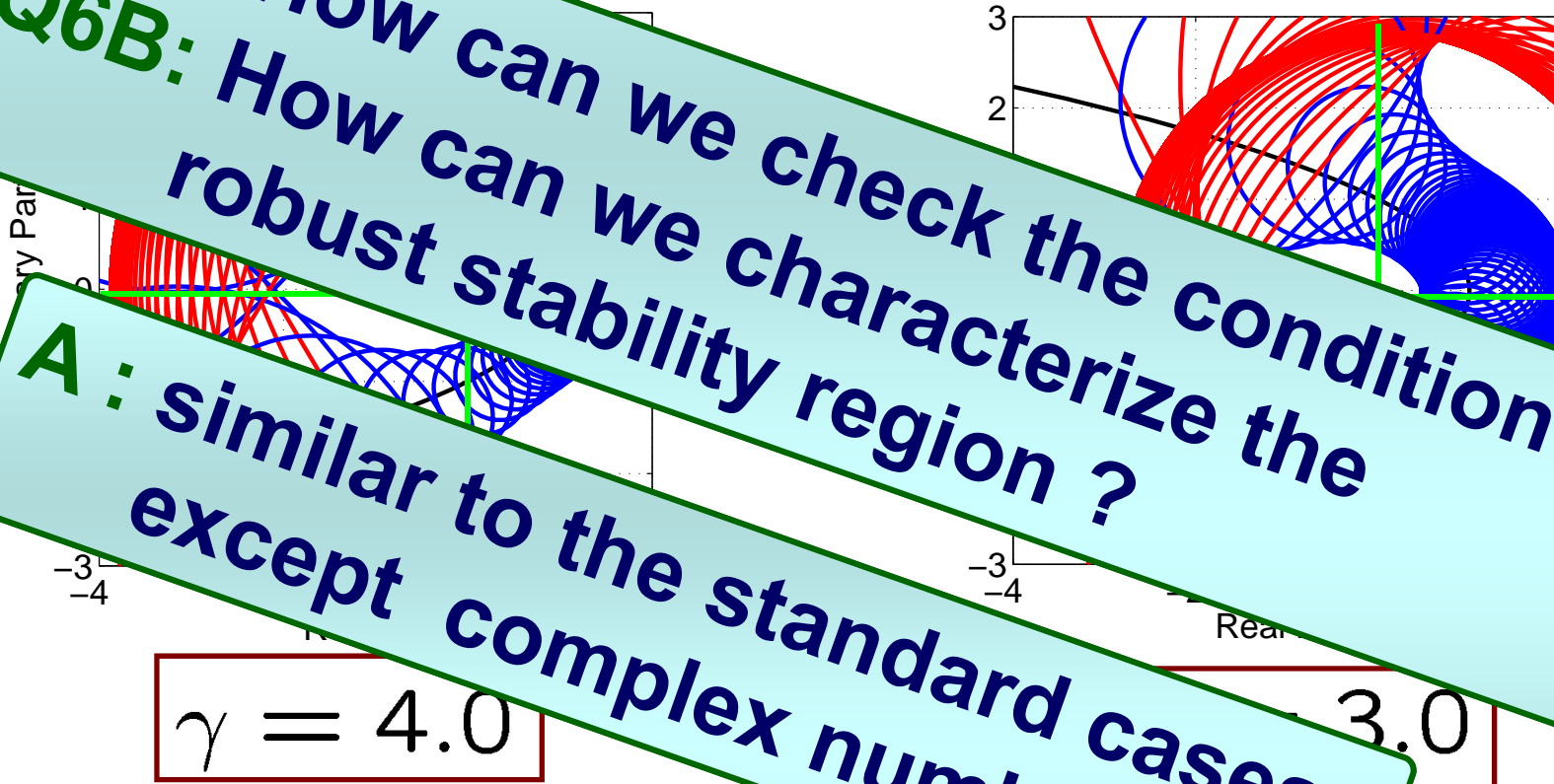
Coprime Factor Perturbations (2/2)

$$h(s) = \frac{1}{s^2 + s + 1}$$

Q6A: How can we check the condition ?
Q6B: How can we characterize the robust stability region ?

A : similar to the standard cases except complex number λ

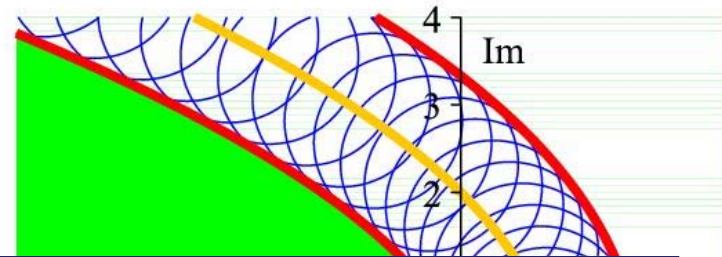
Outside of Blue Circle & Inside of Red Circle



Analytic Expressions by QE

$$h(s) = 1/(s^2 + 2s + 1)$$

$$\lambda = x + jy$$



Summary 3 : Robust Stability Conditions

- Class of $A \Leftrightarrow$ Class of Δ
- Small gain (H_∞ -norm) condition is necessary & sufficient
- D -scaling technique is quite useful
- Some applications

bio-systems

$$+ (8x^3 + 6x^2 - 30x - 1)y^2$$

$$+ (16x^3 - 24x^2 - 15x - 2)x > 0$$

$$\begin{bmatrix} 0 & -\frac{1}{4} & 1 & \frac{21}{8} & -\frac{51}{4} & -\frac{7}{2} \\ \frac{1}{16} & -1 & \frac{27}{4} & -\frac{51}{4} & -8 & -1 \\ 0 & -\frac{1}{16} & \frac{3}{4} & -\frac{7}{2} & -1 & 0 \end{bmatrix}$$

OUTLINE

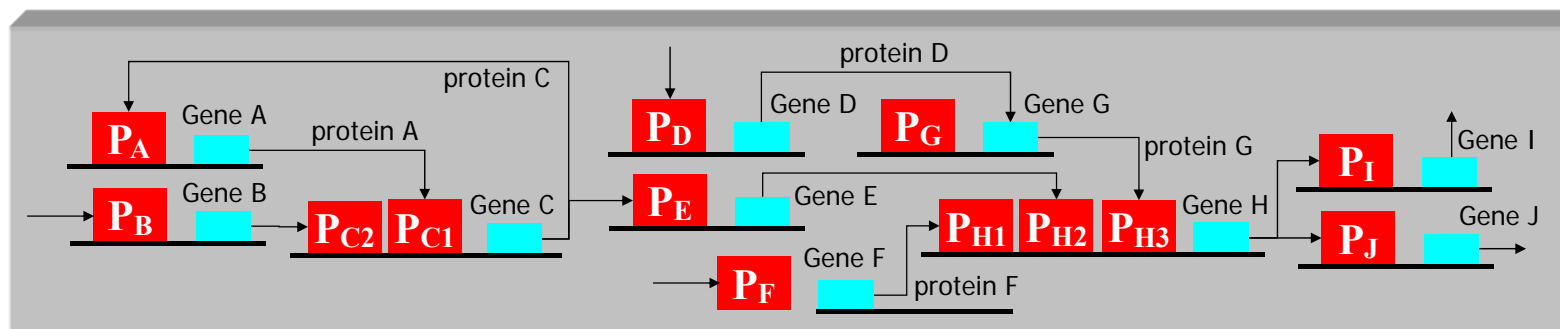
1. Stability Analysis: Review
2. D-Stability Analysis
3. Robust Stability Analysis
- 4. Application to Gene Regulatory Networks**
5. Nonlinear Stability Analysis
6. Conclusion

An Application : Biological rhythms

Motivation



- **Biological rhythms**
 - 24h-cycle, heart beat, sleep cycle etc.
 - caused by periodic oscillations of protein concentrations in *Gene Regulatory Networks*
- **Medical and engineering applications**
 - Artificially engineered biological oscillators (e.g.) Repressilator [Elowitz & Leibler, *Nature*, 2000]

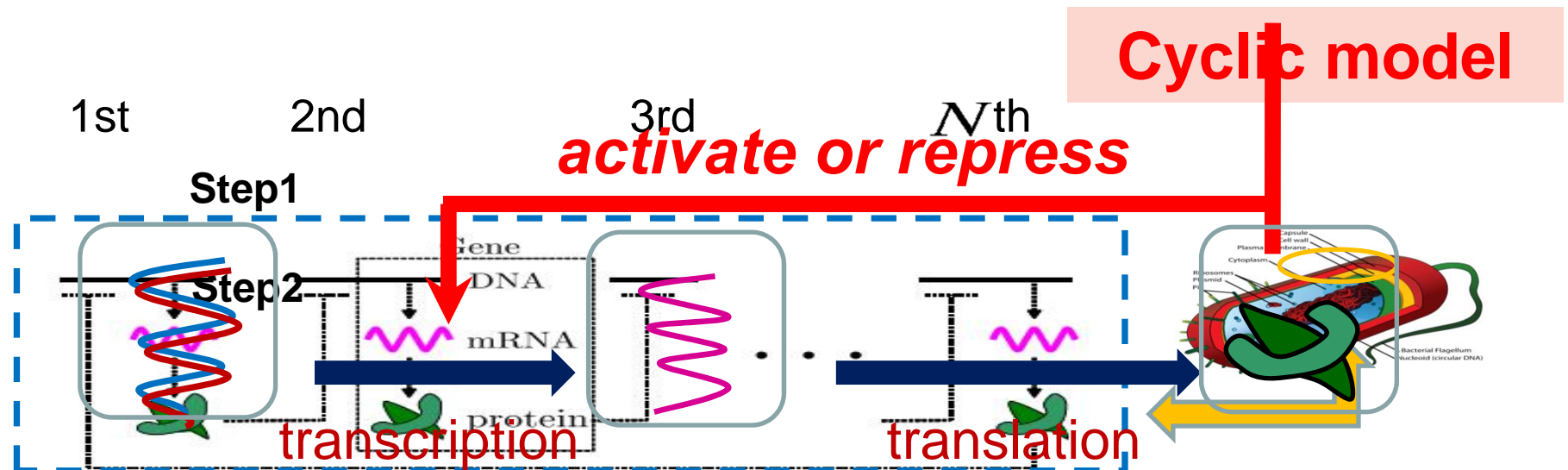
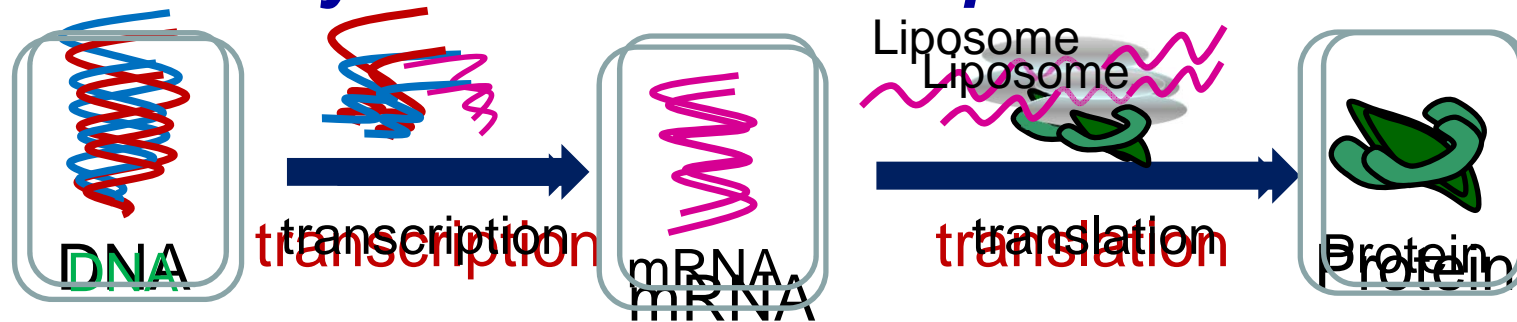


Gene Regulatory Network Systems

- **Biological rhythms:** 24h-cycle, heart beat
periodic oscillations of protein concentration:
in *Gene Regulatory Networks*

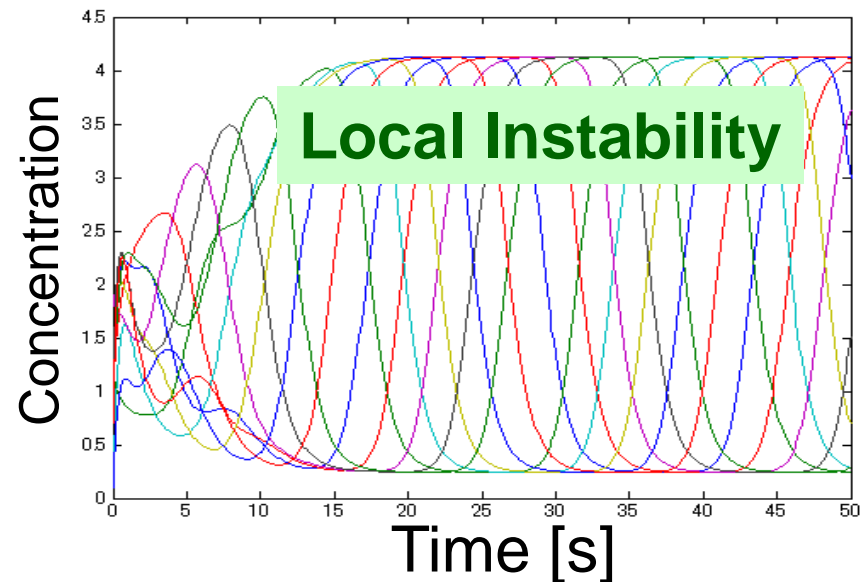
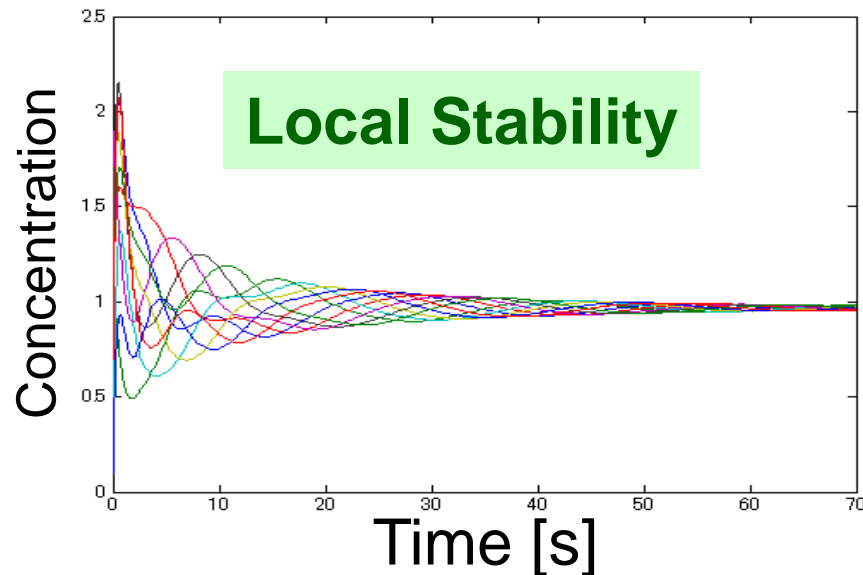


- **Protein synthesis : transcription & translation**



Convergence or Oscillations ?

- **Numerical simulations**
 - Changing chemical parameters



Q7: What are the conditions for convergence and the existence of oscillations ?

Nonlinear Analysis

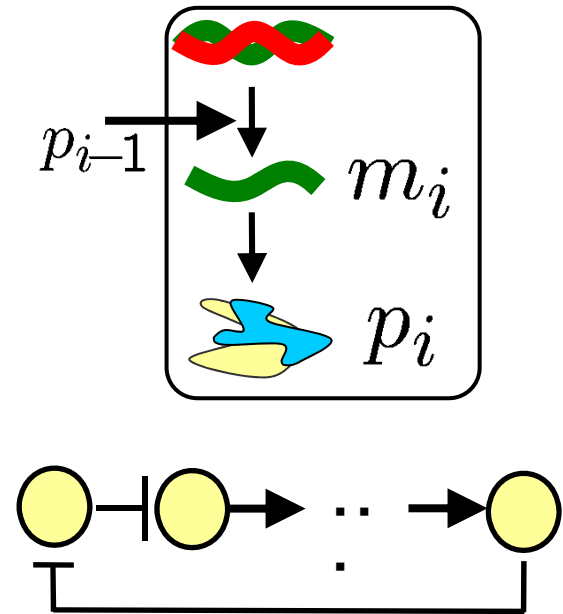
Gene Regulatory Network Model

gene model ($i = 1, \dots, N$)

$$\frac{d}{dt} \begin{bmatrix} r_i \\ p_i \end{bmatrix} = \begin{bmatrix} -a_i & 0 \\ c_i & -b_i \end{bmatrix} \begin{bmatrix} r_i \\ p_i \end{bmatrix} + \begin{bmatrix} \beta_i \\ 0 \end{bmatrix} f_i(p_{i-1})$$

$a_i, b_i > 0$: Degradation rates
(1/Time constants)

$c_i, \beta_i > 0$: Production rates



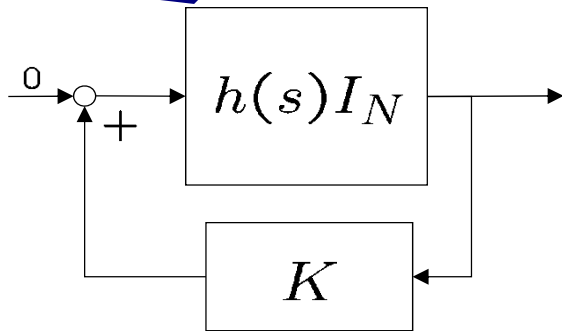
$f_i(p_{i-1})$: Hill function

$$f_i(p_{i-1}) := \begin{cases} \frac{p_{i-1}^\nu}{1 + p_{i-1}^\nu} & \text{(Mono. increasing for activation)} \\ \frac{1}{1 + p_{i-1}^\nu} & \text{(Mono. decreasing for repression)} \end{cases}$$

Linearized Gene Network Model

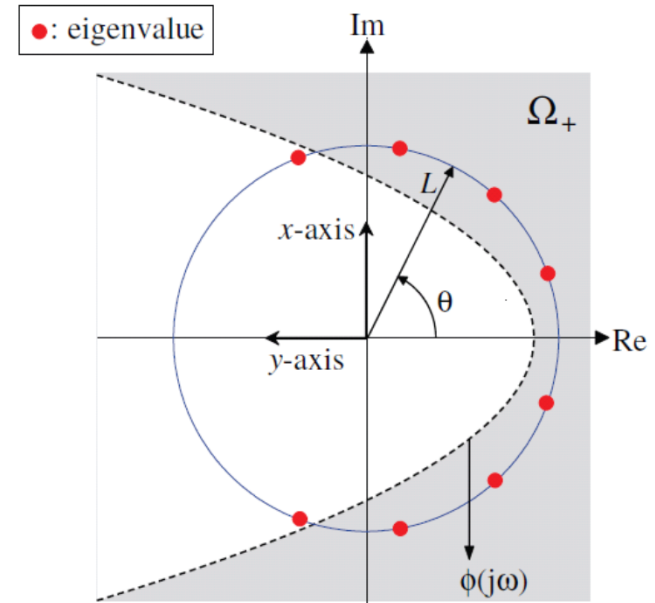
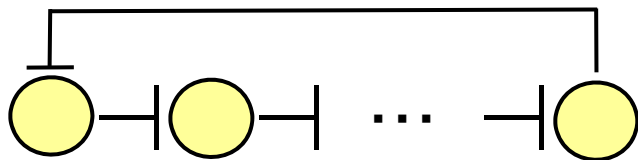
Each gene's dynamics

$$h(s) := \frac{1}{(T_a s + 1)(T_b s + 1)}$$

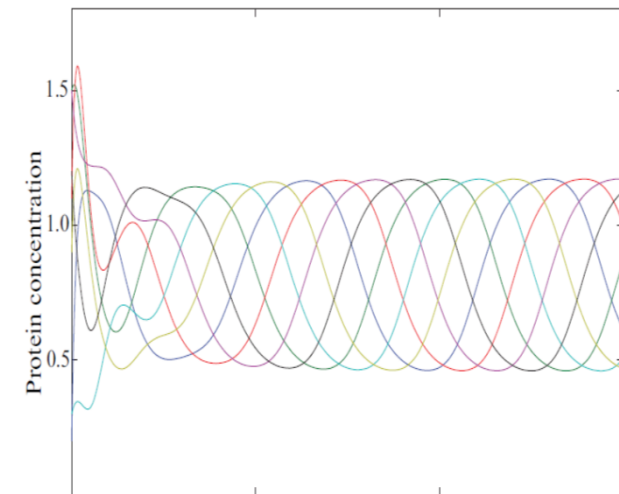


$$\begin{bmatrix} 0 & 0 & 0 & \dots & R_1^2 f_1'(p_N^*) \\ R_2^2 f_2'(p_1^*) & 0 & 0 & \dots & 0 \\ 0 & R_3^2 f_3'(p_2^*) & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & R_N^2 f_N'(p_{N-1}^*) & 0 \end{bmatrix}$$

Interaction structure



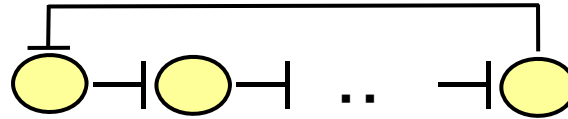
Protein concentration



Time

Analytic Criteria

- Assumptions:



- All interactions are repressive

$$R = \frac{\sqrt{c\beta}}{\sqrt{ab}}$$

$$\frac{T_a T_b}{(T_b)/2}$$

Theorem [Hori et al., 09]

The cyclic GP

a

when

Time-delay case
Non-homogeneous case
Robustness analysis
Experiments with RIKEN

$$\frac{(\pi/N) - \cos(\pi/N)}{1 - \cos(\pi/N)}$$

$$\frac{1 + \cos(\frac{\pi}{N})}{(N, Q) + \nu Q(1 + \cos(\frac{\pi}{N}))}$$

$$\frac{1}{1 + \cos(\frac{\pi}{N})}$$

- Four initial parameters (N, ν, R, Q) which determine the existence of periodic oscillations

- This coincides with [H. E. Samad et al., 05] $N=3, Q=1$

Robust Stability Condition

$$h(s) = \frac{1}{(T_a s + 1)(T_b s + 1)}$$

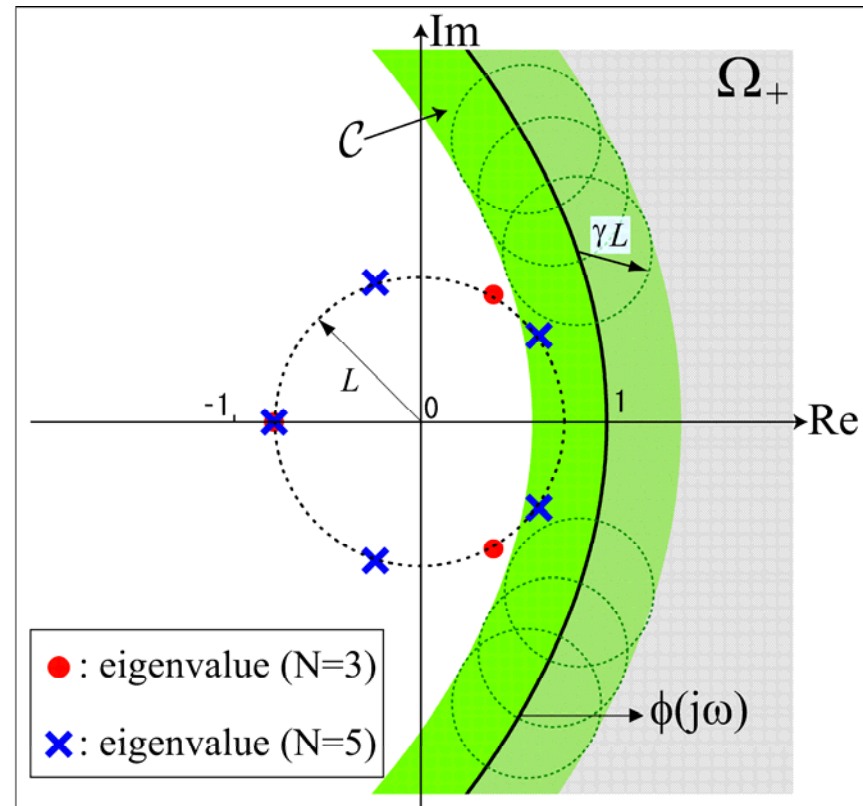
$$A = R^2 \begin{bmatrix} 0 & 0 & 0 & \cdots & \kappa_1 \\ \kappa_2 & 0 & 0 & \cdots & 0 \\ 0 & \kappa_3 & 0 & \cdots & 0 \\ \vdots & \vdots & \cdots & \cdots & \vdots \\ 0 & 0 & \cdots & \kappa_N & 0 \end{bmatrix}$$

$$\left\| \frac{h}{1 - \lambda h} \right\|_{\infty} < \gamma, \quad \forall \lambda \in \sigma(A)$$

$\exists D$: diagonal s.t.
 DAD^{-1} is normal

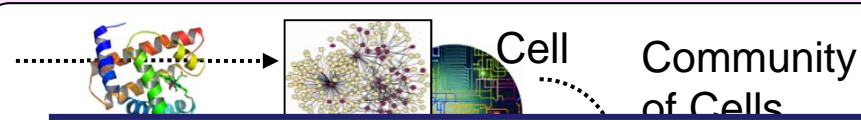
$$Q := \frac{\sqrt{T_a T_b}}{(T_a + T_b)/2} \quad R := \frac{\sqrt{c\beta}}{\sqrt{ab}}$$

$$L := \prod_{k=1}^N \left| \frac{df_i}{dp} \right|_{p^*}^{\frac{1}{N}}$$



More Robust as N, R^2, Q, L decrease.

Hierarchical Bio-Network Systems



Summary 4 : Applications to GRNSs

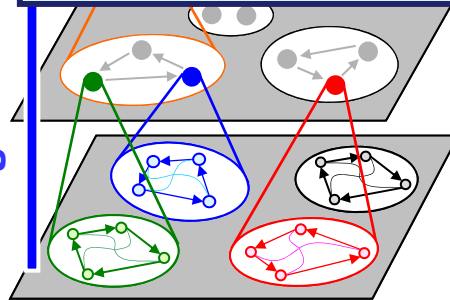
- We can apply our theoretical results to cyclic GRNSs
- We can get new biological insights from the view point of control
- We may treat the cell level behavior

HGM

(2)

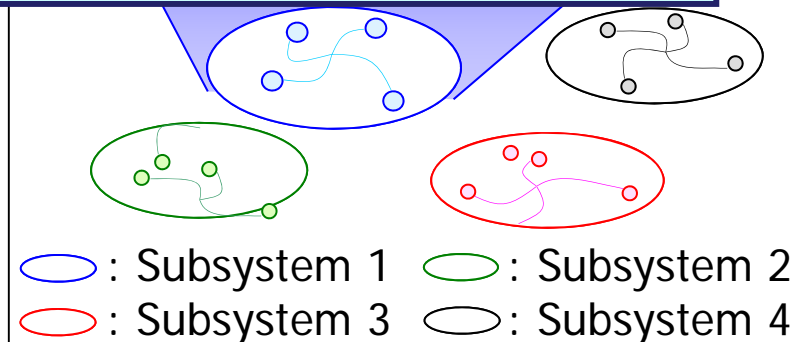
(x_2, x_3)

Larger Scale



Layer 2: **CELL**
Signaling
Networks

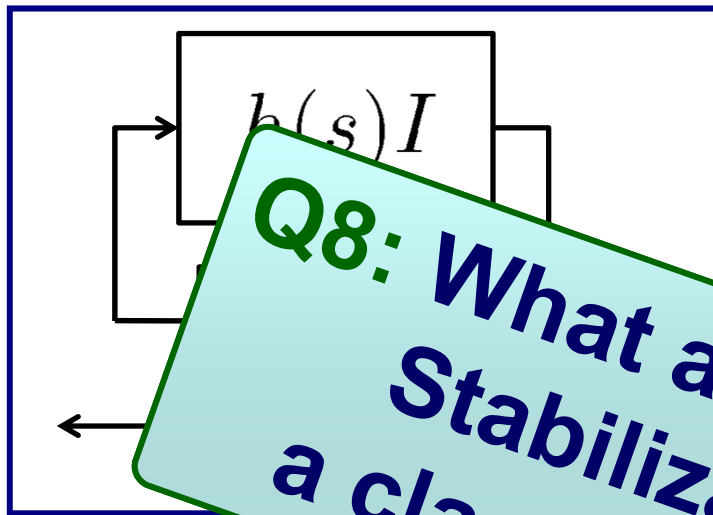
Layer 1: **GENE**
Regulatory
Networks



OUTLINE

1. Stability Analysis: Review
2. D-Stability Analysis
3. Robust Stability Analysis
4. Application to Gene Regulatory Networks
- 5. Nonlinear Stability Analysis**
6. Concluding Remarks

Linear \rightarrow Nonlinear



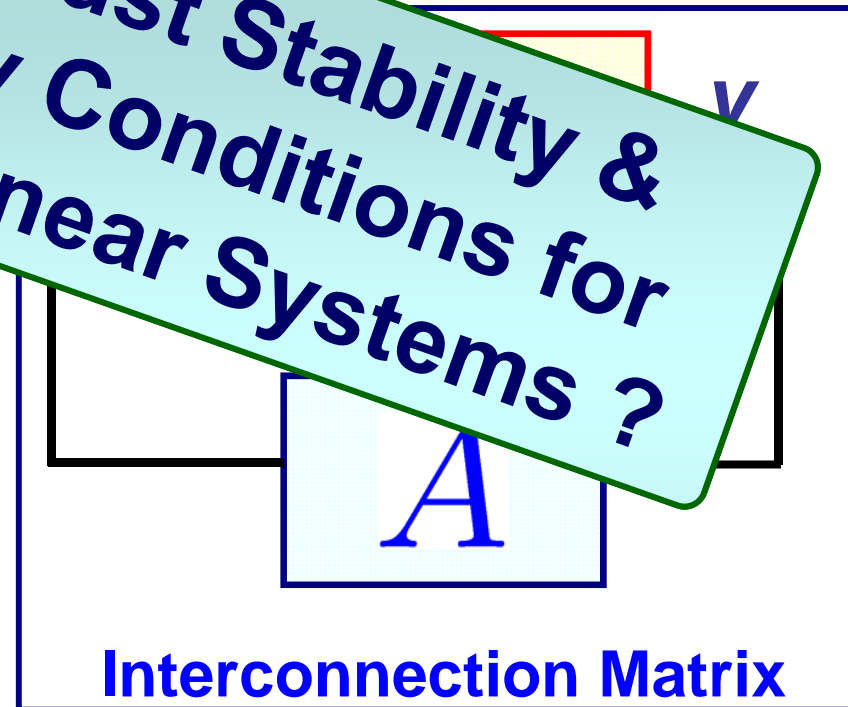
Homogeneous
 \rightarrow Heterogeneous
Robust Stability Analysis

$$\Sigma_i : h_i(s)$$

Q8: What are Robust Stability & Stabilizability Conditions for a class of Nonlinear Systems ?

Linear \rightarrow Nonlinear
Nonlinear Analysis

$$\Sigma_i : \mathcal{N}_i$$



(Q, S, R) Dissipativity

Definition

A system is called (Q, S, R) -*dissipative* if there exists a positive definite function $V(x)$ called *storage function*, such that for all $x \in \mathcal{X}$

$$V(x(T)) - V(x(0)) \leq \int_0^T w(u(t), y(t)) dt$$

holds for all inputs $u \in \mathcal{U}$ and all finite $T \geq 0$, where $w(u, y)$ is **quadratic supply rate** given by

$$w(u, y) = y^T Q y + 2y^T S u + u^T R u$$

with $R = R^T \in \mathbb{R}^{m \times m}$, $S \in \mathbb{R}^{p \times m}$, $Q = Q^T \in \mathbb{R}^{p \times p}$.

Stability for Dissipative Agents

(Hirsch, Hara: IFAC2008)

Agent Dynamics — SISO (Q, S, R) -dissipative

$$\begin{aligned}\dot{x}_i &= f_i(x_i) + g_i(x_i)u_i \\ y_i &= h_i(x_i)\end{aligned}$$

$$\begin{aligned}Q &= \text{diag}\{Q_i\} \leq 0, \\ S &= \text{diag}\{S_i\}, \\ R &= \text{diag}\{R_i\} \geq 0.\end{aligned}$$

Theorem (LMI)

$$V := \sum_{i=1}^N d_i \cdot V_i$$

If \exists a diagonal matrix $D > 0$ such that

$$A^T D R A + D S A + A^T S^T D + D Q < 0$$

holds, then the networked system is asymptotically stable.

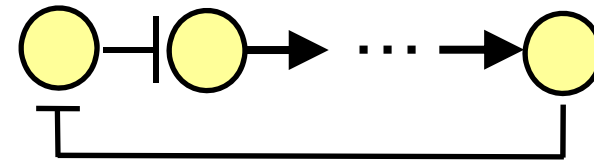
If $R = 0$ and $S > 0$, then

$A + S^{-1}Q/2$: diagonally stable

Stability Condition for GRNs

Cyclic Structure with Negative Feedback

$$A = \begin{bmatrix} 0 & 0 & 0 & \cdots & -1 \\ 1 & 0 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & 0 \\ \vdots & \vdots & \cdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 & 0 \end{bmatrix}$$



$$= A + S^{-1}Q/2 = \begin{bmatrix} -1/\gamma_1 & 0 & 0 & \cdots & -1 \\ 1 & -1/\gamma_2 & 0 & \cdots & 0 \\ 0 & 1 & -1/\gamma_3 & \cdots & 0 \\ \vdots & \vdots & \cdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 & -1/\gamma_N \end{bmatrix}$$

$$\text{diag}\{1/\gamma_i\} \begin{bmatrix} -1 & 0 & 0 & \cdots & -\gamma_1 \\ \gamma_2 & -1 & 0 & \cdots & 0 \\ 0 & \gamma_3 & -1 & \cdots & 0 \\ \vdots & \vdots & \cdots & \ddots & \vdots \\ 0 & 0 & \cdots & \gamma_N & -1 \end{bmatrix}$$

Secant Criterion

$$\gamma_1 \cdots \gamma_N < \sec(\pi/N)^N$$



Diagonally Stable

(Arcak & Sontag. Automatica, 2006)

Stabilization

Theorem (LMI)

If there exist a solution (X, Y) represented by

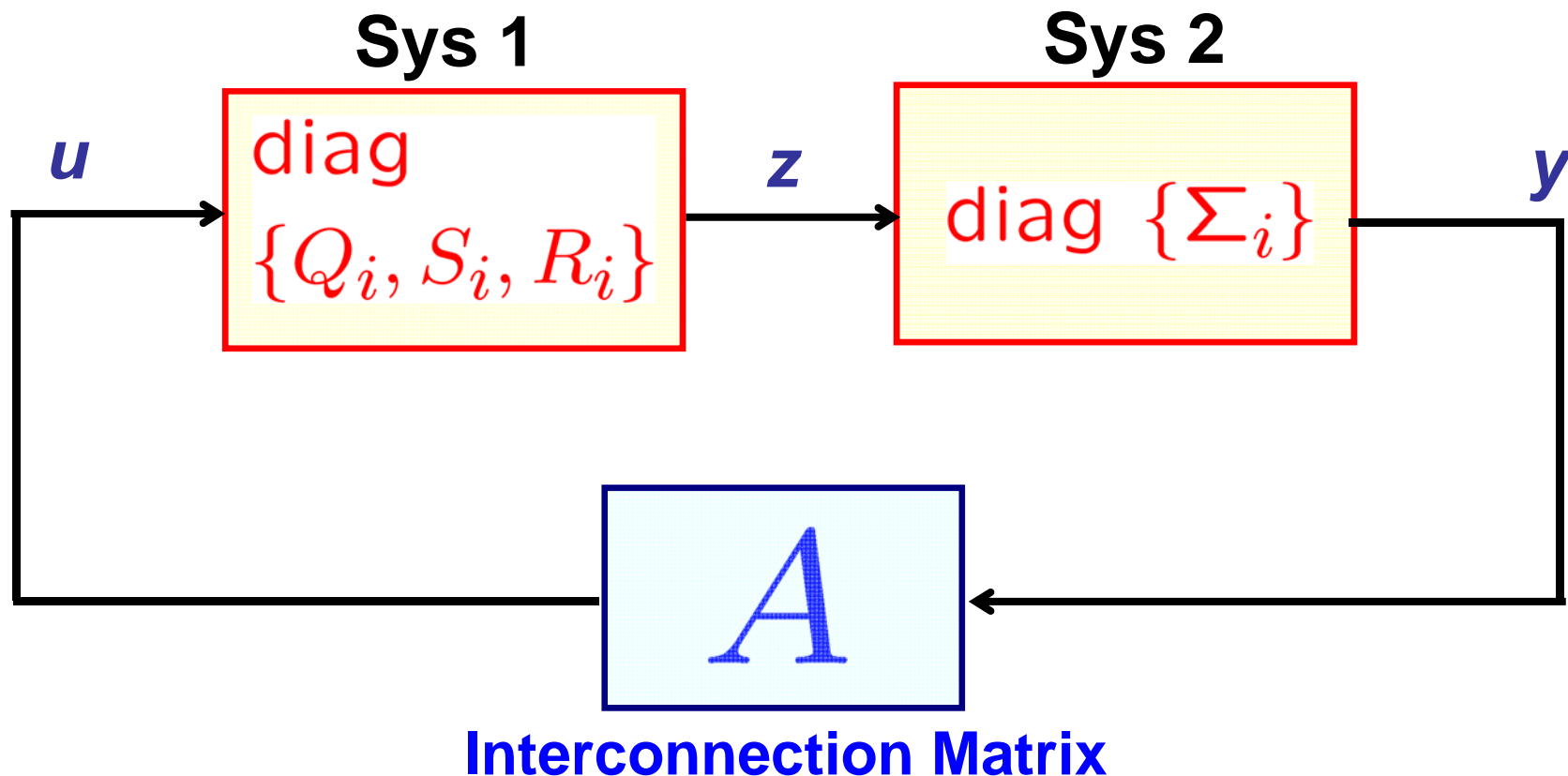
$$Y > 0.$$

Q9: Dissipative (Passive) \rightarrow
beyond Dissipative (Passive) ?
Can we apply the results to
non-dissipative systems ?

Note: X preserves the network structure of A ,
since Y is diagonal.

MADS with Cascaded Dissipative Systems

A class of multi-agent dynamical systems based on dissipative properties



Two Cascaded Dissipative Systems

Sys 2

$$\text{diag} \{Q_i, S_i, R_i\} \Leftrightarrow A = \begin{bmatrix} 0 & I_N \\ A_2 & 0 \end{bmatrix} \in \mathbb{R}^{2N \times 2N}$$



$$\begin{aligned} Q &= \text{diag}\{Q_1, Q_2\}, \\ S &= \text{diag}\{S_1, S_2\}, \\ R &= \text{diag}\{R_1, R_2\} \end{aligned}$$

Stability Condition (LMI)

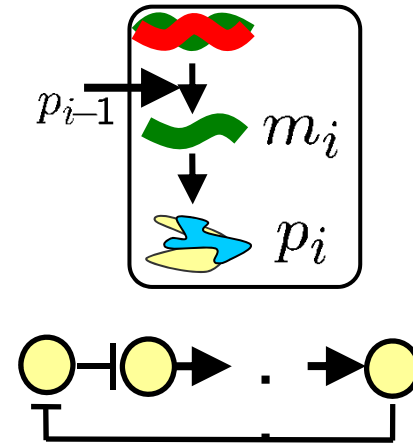
$$\begin{bmatrix} D_1 Q_1 & D_1 S_1 + A_2^T S_2 D_2 & 0 & A_2^T R_2 D_2 \\ D_1 S_1 + D_2 S_2 A_2 & D_2 Q_2 & D_1 R_1 & 0 \\ 0 & D_1 R_1 & -D_1 R_1 & 0 \\ D_2 R_2 A_2 & 0 & 0 & -D_2 R_2 \end{bmatrix} < 0$$

$D_1 > 0, D_2 > 0$: diagonal

Gene Regulatory Network

$$Q = \text{diag}\{Q_1, Q_2\},$$

$$S = \text{diag}\{S_1, S_2\}, R = 0$$



Stability Condition (LMI)

$$\begin{bmatrix} D_1 Q_1 & D_1 S_1 + A_2^T D_2 S_2 \\ S_1 D_1 + S_2 D_2 A_2 & D_2 Q_2 \end{bmatrix} < 0$$

Stabilization Condition (LMI)

$$\begin{bmatrix} Y_1 Q_1 & Y_1 S_1 + X_2^T S_2 \\ S_1 Y_1 + S_2 X_2 & Y_2 Q_2 \end{bmatrix} < 0 \quad A_2 = X_2 Y_2^{-1}$$

Dissipative + Integrator

Sys 2

Theorem

1 . J

Suppose $R = 0$ and $Q < 0$.

Summary 5 : Dissipative Networked Systems

- LMI stability & stabilizability conditions
- Stability and robust stability conditions
- D -scaling technique is also useful
- Dissipative \rightarrow Non-dissipative

D : positive diagonal s.t.

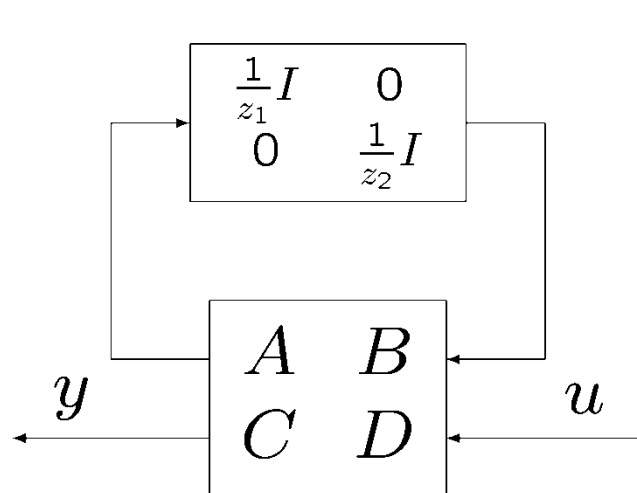
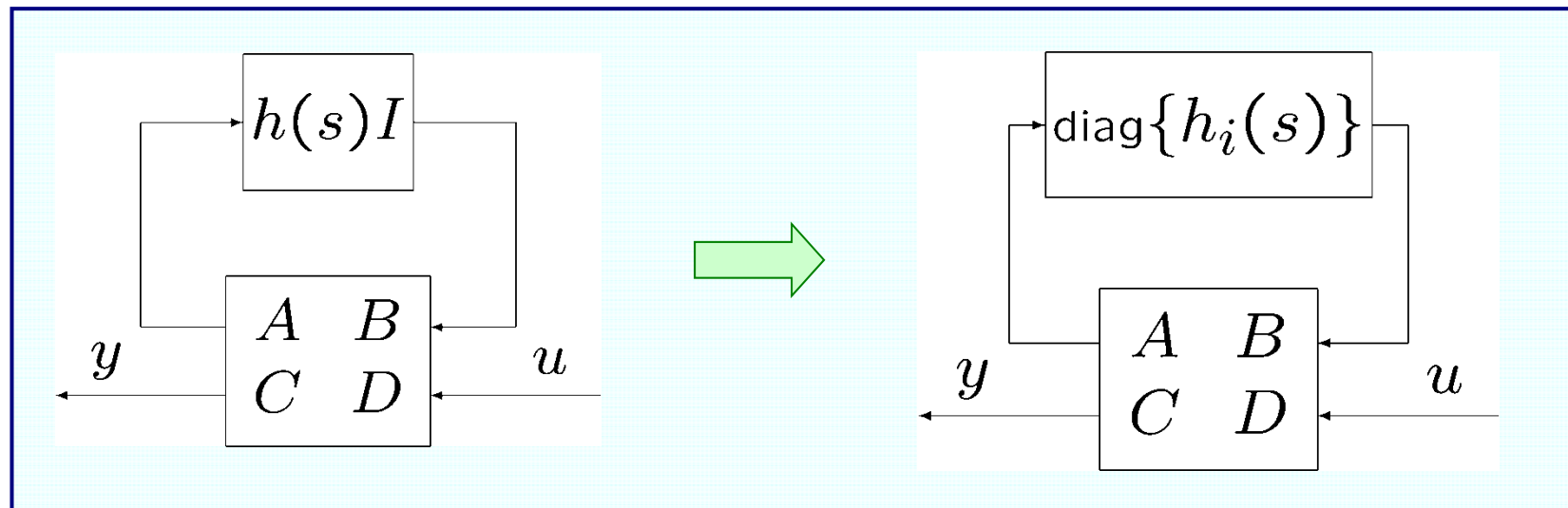
$A_{2D} := DA_2D^{-1}$ is normal. Then,

A_2 is stable $\Leftrightarrow A_2$ is diagonally stable .

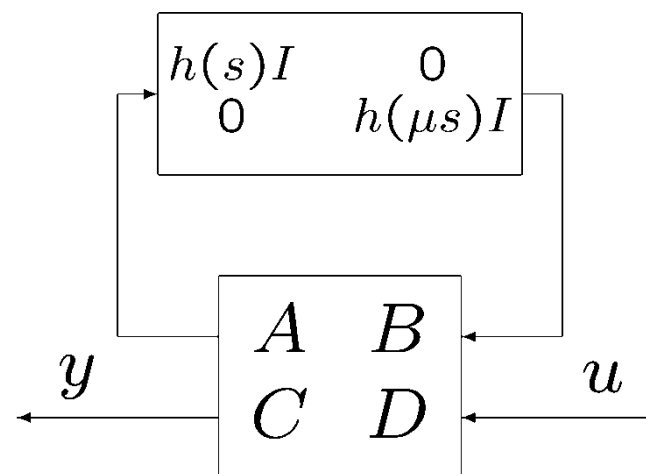
OUTLINE

1. Stability Analysis: Review
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- 6. Concluding Remarks**

New Framework for System Theory



2D System

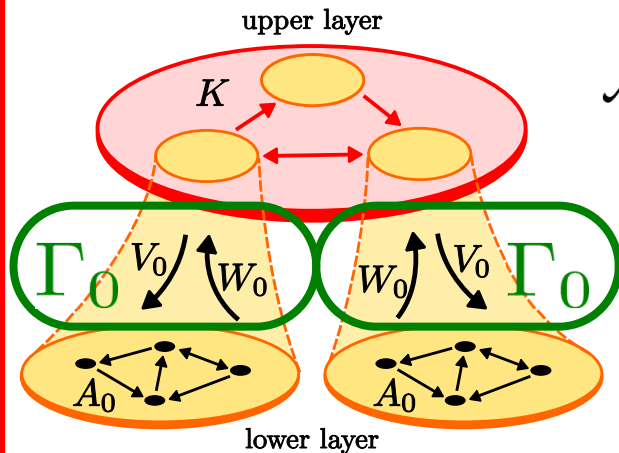


**Singular
Perturbed
System**

**Multi-
resolved
Systems**

Homogeneous vs Heterogeneous

Homogeneous

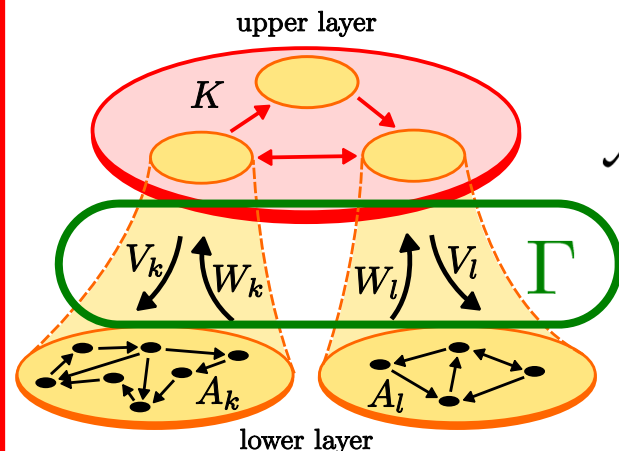


$$\mathcal{A} = \begin{bmatrix} A_0 & & \\ & \ddots & \\ & & A_0 \end{bmatrix} + \begin{bmatrix} k_{11}\Gamma_0 & \cdots & k_{1M}\Gamma_0 \\ \vdots & \ddots & \vdots \\ k_{M1}\Gamma_0 & \cdots & k_{MM}\Gamma_0 \end{bmatrix}$$

$$= I_M \otimes A_0 + K \otimes \Gamma_0 ; \Gamma_0 = V_0 W_0^\top$$

\mathcal{A} can be written using **Kronecker product** \otimes .

Heterogeneous



$$\mathcal{A} = \begin{bmatrix} A_1 & & \\ & \ddots & \\ & & A_M \end{bmatrix} + \begin{bmatrix} k_{11}\Gamma_{11} & \cdots & k_{1M}\Gamma_{1M} \\ \vdots & \ddots & \vdots \\ k_{M1}\Gamma_{M1} & \cdots & k_{MM}\Gamma_{MM} \end{bmatrix}$$

$$= \text{diag} \{ A_k \} + K \odot \Gamma$$

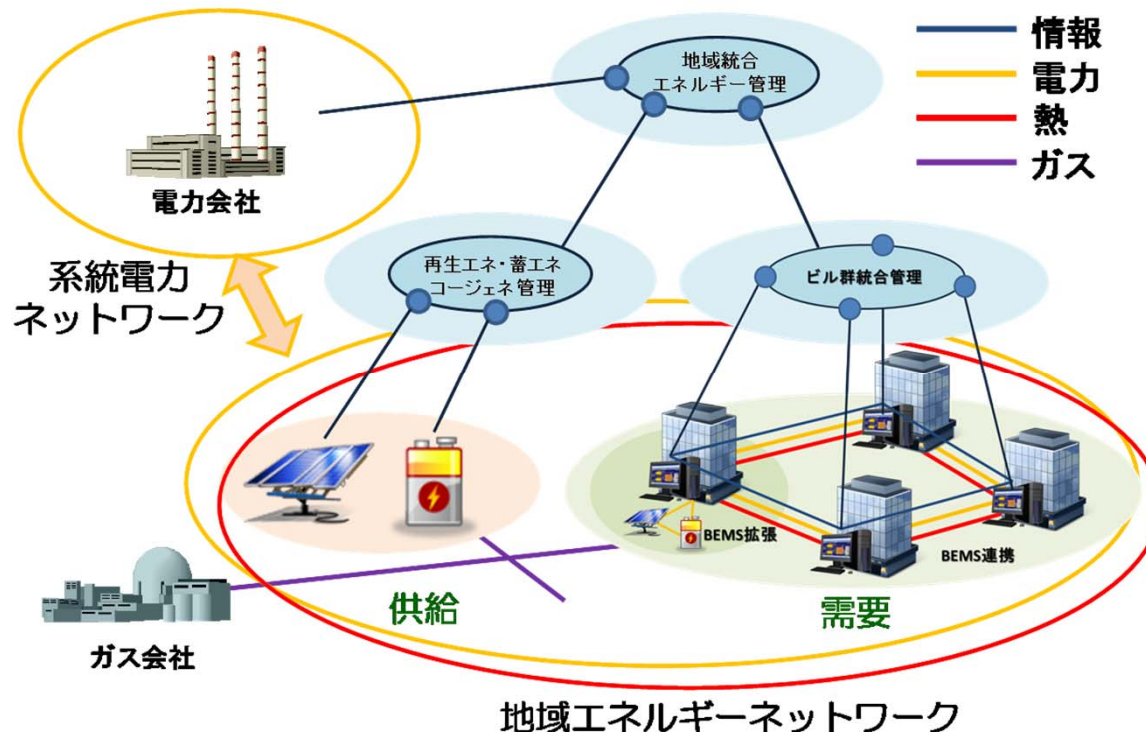
We need **Khatori-Rao product** \odot .

Smart Energy NW and Energy Saving

Smart Energy Network

Electric power network
+ Gas energy network

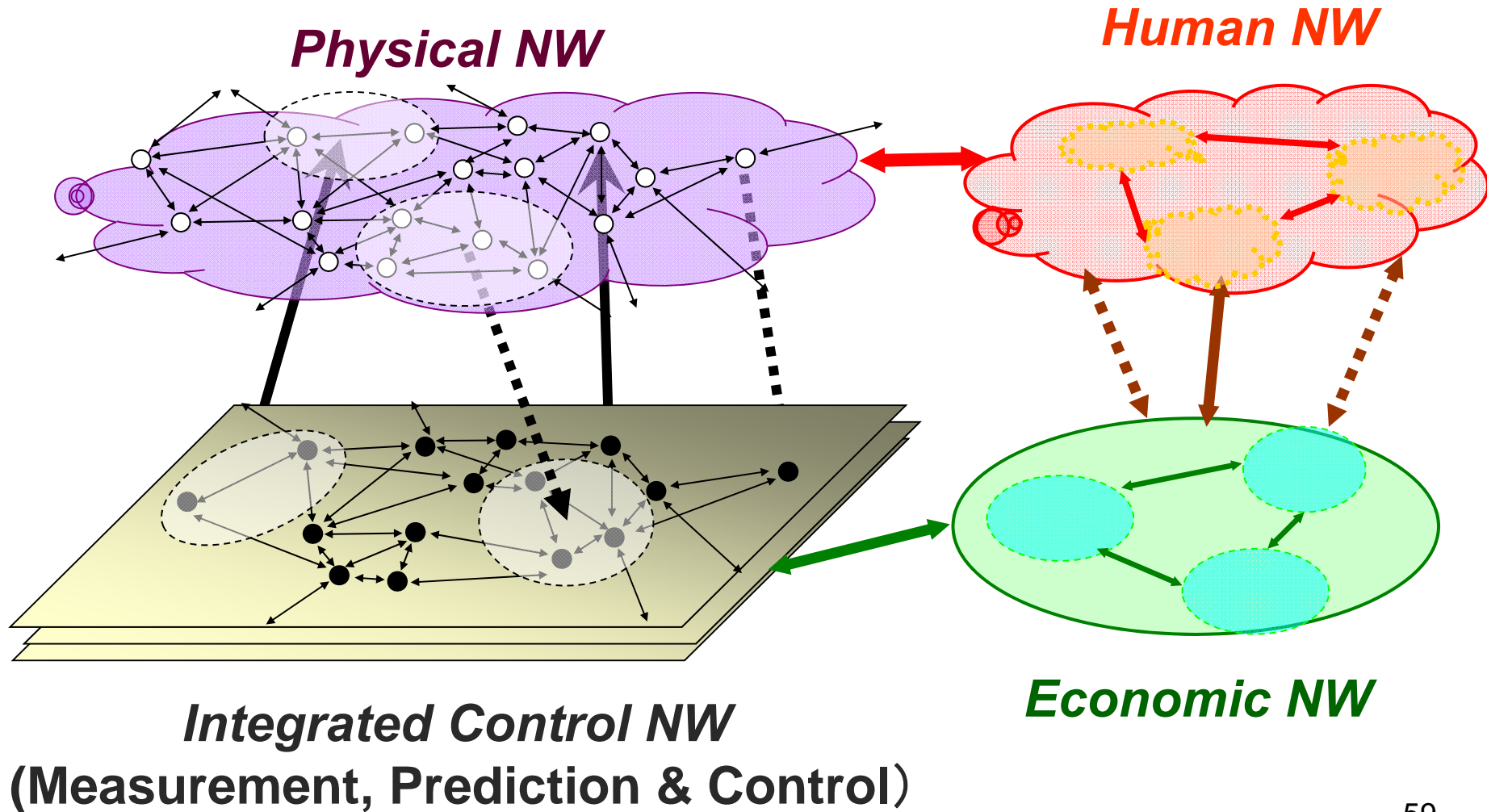
Multi-resolved Hierarchical Modeling
→ **Multi-resolved Prediction**
→ **Hierarchical Decentralized Control**



U Tokyo
Tokyo Gas
Fujitsu
Azbil

Harmony with Nature and Social System

Networked Hierarchical Cyber Physical System



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Hideaki Tanaka (U. Tokyo → Denso)

Masaaki Kanno (Niigata U.)

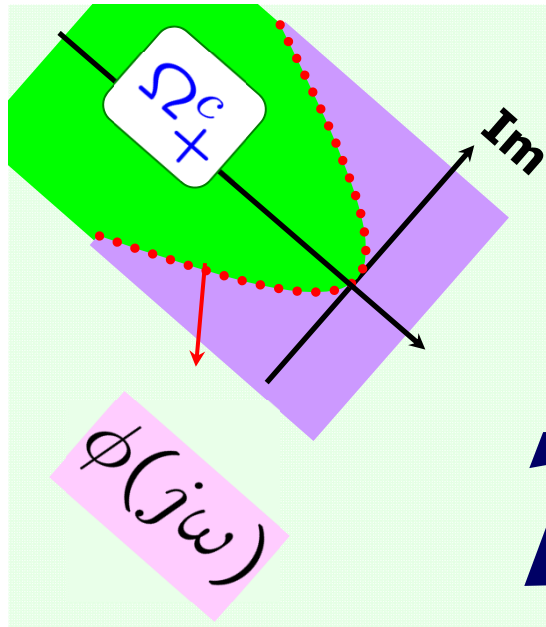
③ **Gene Regulatory Networks**

Yutaka Hori (U. Tokyo)

Tae-Hyoung Kim (Chung-Ang U.)

④ **Nonlinear Analysis**

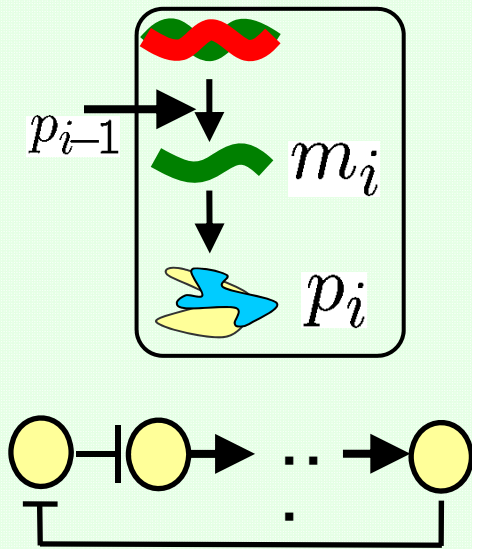
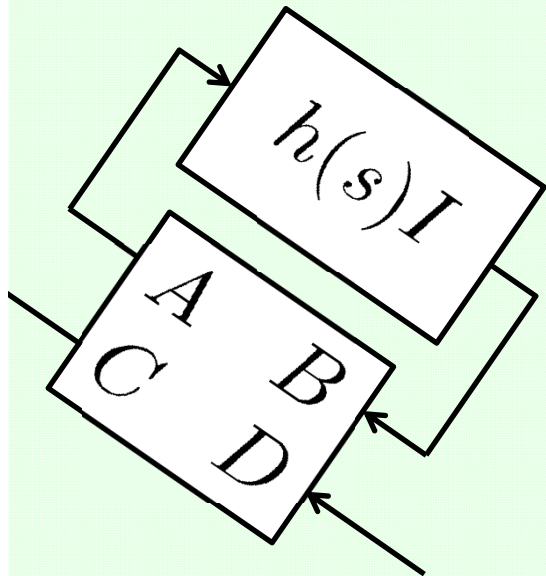
Sandra Hirsch (TU Munich)



(ii) $\left\| \frac{h}{1 - \lambda h} \right\|_{\infty} < \gamma, \forall \lambda \in \sigma(M)$

(iii) $\left| \frac{1}{\phi - \lambda} \right| < \gamma, \forall \lambda \in \sigma(M)$

***Thank you
very much !***



Framework for Glocal Control

**Realization of Global Functions
by Local Measurement and Control**

Real World

**Glocal Control
System**

**Hierarchical Dynamical Systems
with Multi-resolution**



**Local
Control**

**Local
Measurement**

**Global
Prediction**

*through
hierarchical model with
multiple-resolution*

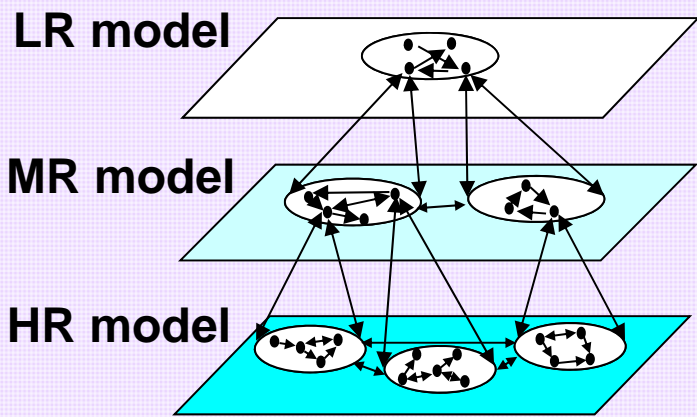


Image of Glocal Control System

