

# Robust and Optimal Control, Spring 2015

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## C. $H_\infty$ Loop Shaping Design

C.1 Perturbations of Coprime Factors [SP05, Sec. 4.1.5]

C.2 Robust Stabilization [SP05, Sec. 9.4.1]

C.3 Loop Shaping Design Procedure [SP05, Sec. 9.4.2]

C.4 Design Example

## Reference:

[SP05] S. Skogestad and I. Postlethwaite,

*Multivariable Feedback Control; Analysis and Design,*  
Second Edition, Wiley, 2005.



# Robust Stabilization

K. Glover and D. McFarlane, “Robust stabilization of normalized coprime factor plant descriptions with  $H_\infty$  bounded uncertainty,”  
IEEE TAC, **34-8**, 821-830, 1989

D. McFarlane K. Glover

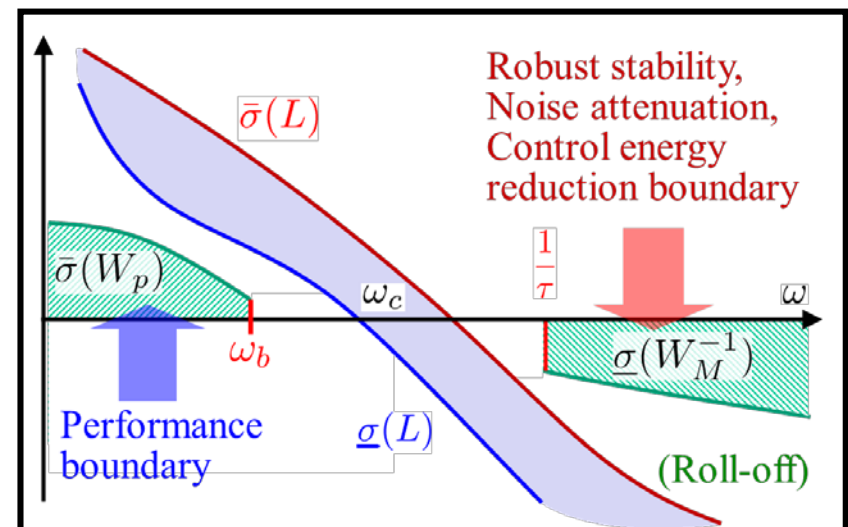
## Mixed Sensitivity Problem

$$\text{find } K(s) \text{ s.t. } \left\| \begin{array}{l} W_P(s)S(s) \\ W_M(s)T(s) \end{array} \right\|_\infty < 1$$

## Remark

Pole/Zero Cancellation

Loop Shaping (NS, NP+RS)





# Perturbations to Coprime Factors

## Coprime Factorization of Transfer Functions

$p \times m$  Transfer Function Matrix

$$G = \left[ \begin{array}{c|c} A & B \\ \hline C & 0 \end{array} \right] = \underset{\text{(Left)}}{M_l^{-1}} N_l = \underset{\text{(Right)}}{N_r} M_r^{-1}$$

$(A, B)$  Controllable       $(A, C)$  Observable

$$M_l, N_l, M_r, N_r \in \mathcal{H}_\infty$$

$$\text{rank} \begin{bmatrix} N_l & M_l \end{bmatrix} = p, \quad \forall \text{Re}(s) \geq 0 \quad \text{rank} \begin{bmatrix} N_r \\ M_r \end{bmatrix} = m, \quad \forall \text{Re}(s) \geq 0$$

[ There are no “common zeros” in  $N$  and  $M$  in the right half plane ]

[Ex.]  $G(s) = N(s)M^{-1}(s) = \frac{1}{s}$

➔  $N(s) = \frac{1}{s+1}, \quad M^{-1}(s) = \frac{s+1}{s}$

# Normalized Coprime Factorization

(NLCF: Normalized Left-Coprime Factorization)

$$M_l(j\omega)M_l(j\omega)^* + N_l(j\omega)N_l(j\omega)^* = I, \quad \forall \omega$$
$$\left[ M(s)^* = M^T(-s) \right]$$

Note: Given any coprime factorization of  $G = M_l^{-1}N_l$  then  
for  $R$  (need the poles and zeros of  $R$  to be in the LHP)

$$G = (RM_l)^{-1}(RN_l)$$

$$(RM_l)(M_l^*R^*) + (RN_l)(N_l^*R^*) = R(M_lM_l^* + N_lN_l^*)R^*$$

[Ex.]  $G(s) = \frac{N(s)}{M(s)} = \frac{1}{s} \quad \rightarrow \quad N(s) = \frac{1}{s+1}, \quad M(s) = \frac{s}{s+1}$

$$MM^* + NN^* = \frac{j\omega}{j\omega + 1} \cdot \frac{-j\omega}{-j\omega + 1} + \frac{1}{j\omega + 1} \cdot \frac{1}{-j\omega + 1}$$
$$= \frac{\omega^2}{1 + \omega^2} + \frac{1}{1 + \omega^2} = 1$$

# Coprime Factor Uncertainty [SP05, pp. 304, 365]

## Nominal Plant Model [SP05, p. 122]

### (Left Coprime Factorization)

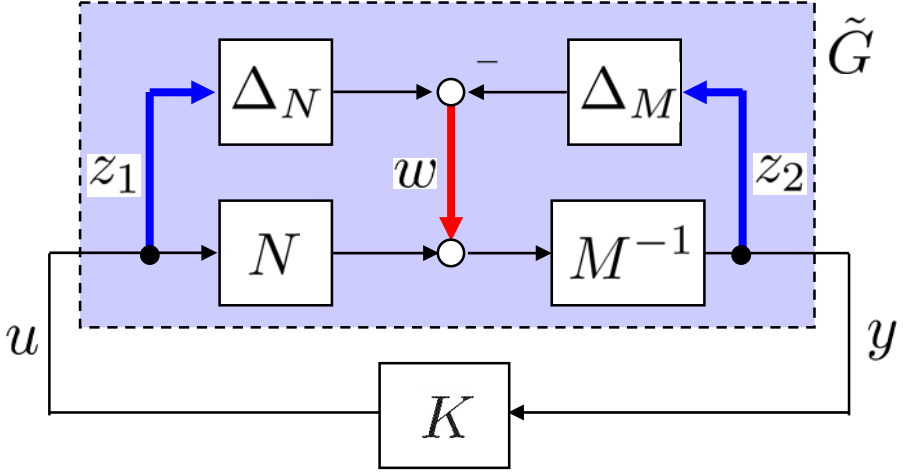
$$G = M^{-1}N$$

### The set of Plant Models

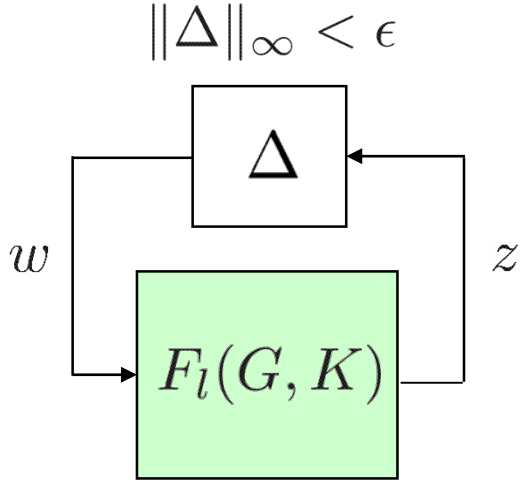
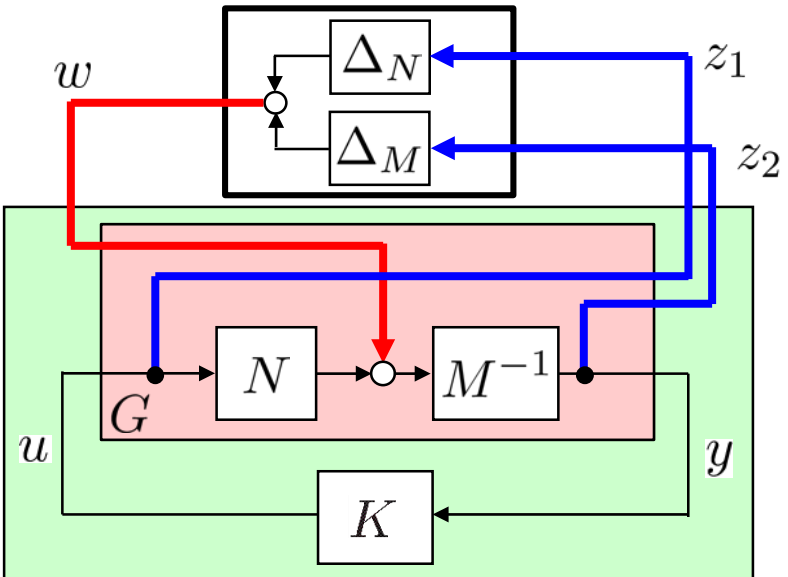
$$\tilde{G} = (M + \Delta_M)^{-1}(N + \Delta_N)$$

$$\| [\Delta_N \quad \Delta_M] \|_\infty < \epsilon$$

$$M, N, \Delta_M, \Delta_N \in \mathcal{RH}_\infty$$



### Full-matrix



# Coprime Factor Uncertainty

Nominal Plant Model [SP05, p. 122]

(Left Coprime Factorization)

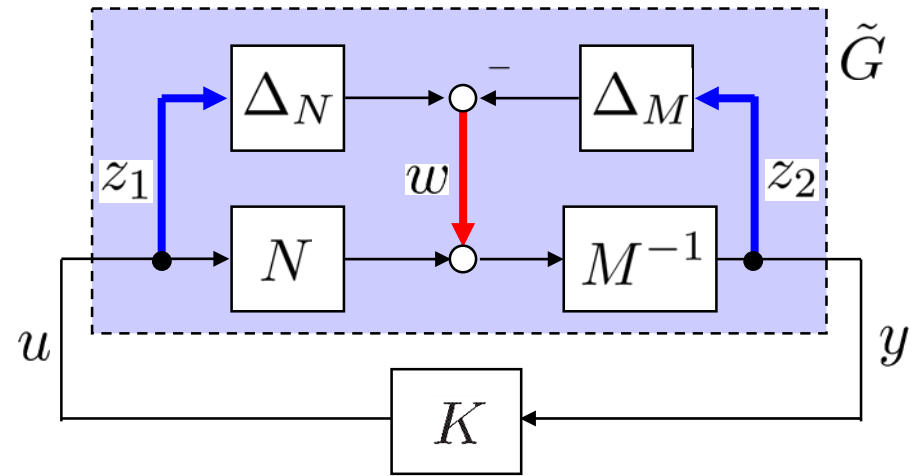
$$G = M^{-1}N$$

The set of Plant Models

$$\tilde{G} = (M + \Delta_M)^{-1}(N + \Delta_N)$$

$$\| \begin{bmatrix} \Delta_N & \Delta_M \end{bmatrix} \|_{\infty} < \epsilon$$

$$M, N, \Delta_M, \Delta_N \in \mathcal{RH}_{\infty}$$



[Ex.]  $G(s) = \frac{1}{s} \rightarrow N(s) = \frac{1}{s+1}, M(s) = \frac{s}{s+1}$

$$\tilde{G}(s) = \frac{N + \Delta_N}{M + \Delta_M} = \frac{1 + \Delta_N(s+1)}{s + \Delta_M(s+1)} \quad \text{with } |\Delta_N|^2 + |\Delta_M|^2 < \epsilon^2$$

If  $\Delta_M$  is real constant, then pole is moved to  $-\frac{\Delta_M}{1 + \Delta_M}$ . Hence poles move across  $s = j\omega$  with small  $|\Delta_M|$  but very large  $|G_{\Delta} - G_o|$  changes.

**Note:** Coprime factor perturbations are not unique.

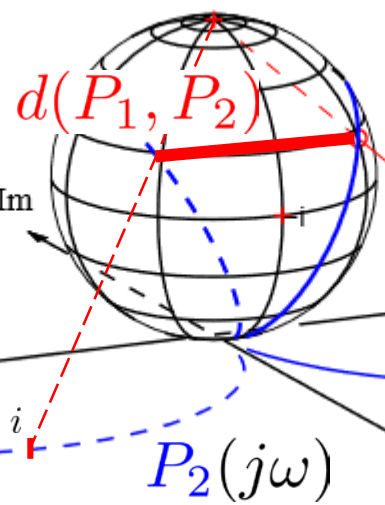
# Vinnicombe Metric ( $\nu$ -gap Metric) [ZD97, Chap.17] [AM09, pp. 349-352]



$k = 100$

$$P_1(s) = \frac{100}{s + 1}$$

$$P_2(s) = \frac{100}{(s - 1)}$$



$$d(P_1, P_2) = \frac{2k}{(1 + k^2)} = \frac{200}{10001}$$

## Vinnicombe metric ( $\nu$ -gap Metric)

$\delta_v(P_1, P_2) = d(P_1, P_2) \in [0, 1]$  if  $(P_1, P_2) \in \mathcal{C}$  G. Vinnicombe

A distance measure that is appropriate for closed loop systems

$$d(P_1, P_2) = \sup_{\omega} \frac{|P_1(j\omega) - P_2(j\omega)|}{\sqrt{(1 + |P_1(j\omega)|^2)(1 + |P_2(j\omega)|^2)}} \in [0, 1]$$

[AP09, Ex 12.2]  $\delta_v(P_1, P_2) = 0.98$

[AP09, Ex 12.3]  $\delta_v(P_1, P_2) = 0.02$

# Robust Stability Condition [SP05, pp. 305, 366]

## Plant Model

$$G = M^{-1}N$$

## The set of Plant Models

$$\tilde{G} = (M + \Delta_M)^{-1}(N + \Delta_N)$$

$$\left\| \begin{bmatrix} \Delta_N & \Delta_M \end{bmatrix} \right\|_\infty < \epsilon$$

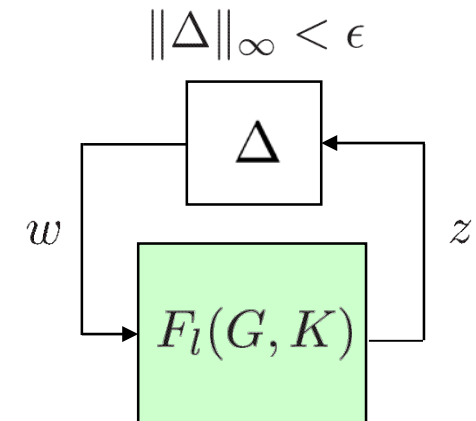
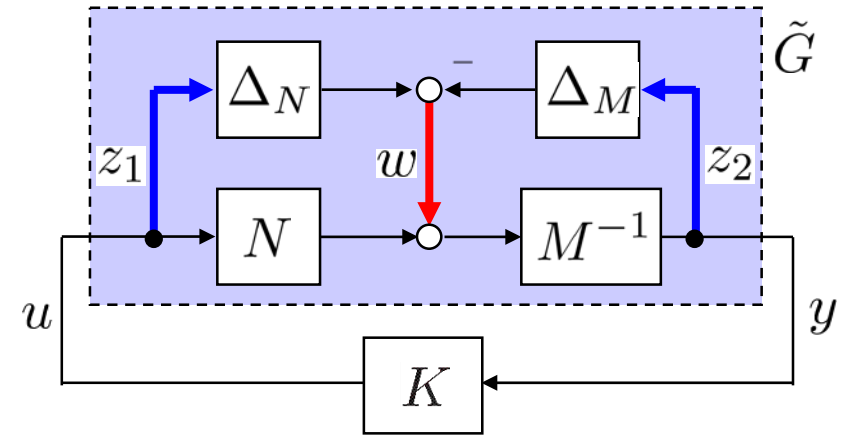
## Closed loop system (LFT)

$$\begin{aligned} \|F_l(G, K)\|_\infty &= \left\| \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} M^{-1} \right\|_\infty \\ &= \left\| \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} M^{-1} \begin{bmatrix} M & N \end{bmatrix} \right\|_\infty \\ &= \left\| \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} I & G \end{bmatrix} \right\|_\infty \end{aligned}$$

## Robust Stability Condition

$$\gamma = \left\| \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} I & G \end{bmatrix} \right\|_\infty \leq \frac{1}{\epsilon}$$

( $\because$  Small Gain Theorem)





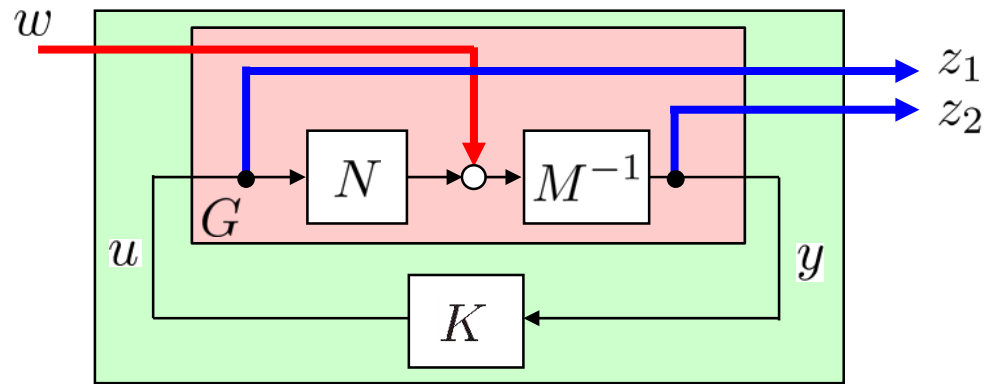
# $H_\infty$ Control Synthesis

## Nominal Plant Model

$$G = M_l^{-1} N_l = N_r M_r^{-1}$$

## Double Bezout Equation

$$\begin{bmatrix} V_l & -U_l \\ -N_l & M_l \end{bmatrix} \begin{bmatrix} M_r & U_r \\ N_r & V_r \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & I \end{bmatrix}$$



## All Stabilizing Controllers (Youla Parameterization)

$$\begin{aligned} K &= (V_l + Q N_l)^{-1} (U_l + Q M_l) \\ &= (U_r + M_r Q)(V_r + N_r Q)^{-1} \quad \text{for } Q \in \mathcal{H}_\infty \end{aligned}$$

## Closed-loop Transfer Function

$$\begin{aligned} F_l(G, K) &= \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} I & G \end{bmatrix} \\ &= \begin{bmatrix} U_r \\ V_r \end{bmatrix} \begin{bmatrix} M_l & N_l \end{bmatrix} + \begin{bmatrix} M_r \\ N_r \end{bmatrix} Q \begin{bmatrix} M_l & N_l \end{bmatrix} \end{aligned}$$

## $H_\infty$ Sub-optimal Control Problem

Given  $\gamma > \gamma_{\min}$ , find all stabilizing controllers  $K$  such that

$$\left\| \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} I & G \end{bmatrix} \right\|_\infty \leq \gamma = \frac{1}{\epsilon}$$

### Solution

Minimum Value of  $H_\infty$ -norm (Maximum Stability Margin)

$$\gamma_{\min}^2 = \epsilon_{\max}^{-2} = 1 + \lambda_{\max}(XZ)$$

Maximum eigenvalues of the matrix  $XZ$

Sub-optimal Solution ( $H_\infty$  controller)

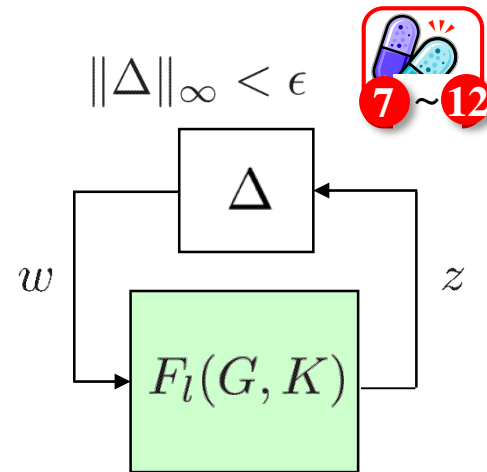
$$K = (K_{11}\Phi + K_{12})(K_{21}\Phi + K_{22})^{-1}$$

$\Phi$  : a transfer function satisfying  $\|\Phi\|_\infty < 1$

$$\gamma^{-1} = \epsilon_{\max} = \sqrt{1 - \|\tilde{N}_S \tilde{M}_S\|_H^2} \leq 1$$

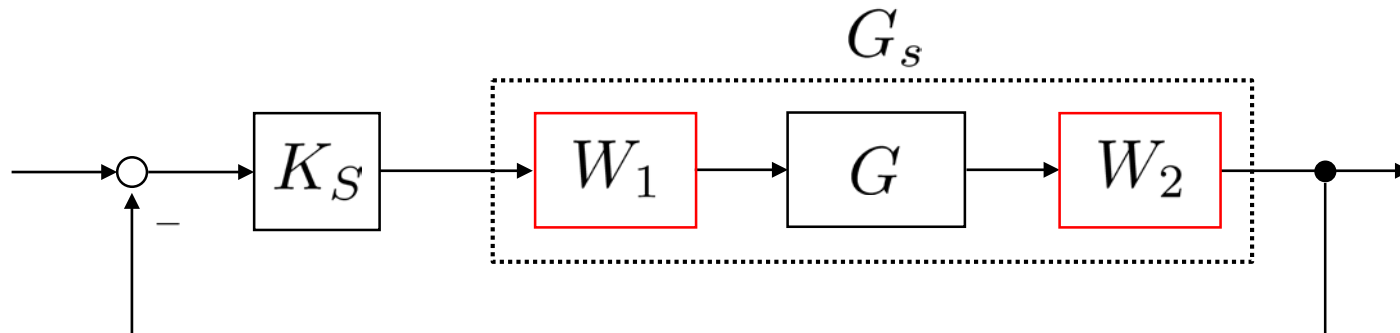
Central controller ( $\Phi = 0$ )

$$K = \left[ \begin{array}{c|c} \frac{A + BF + \gamma^2 W_1^{-T} Z C^T (C + DF)}{B^T X} & \frac{\gamma^2 W_1^{-T} X C^T}{-D^T} \end{array} \right]$$



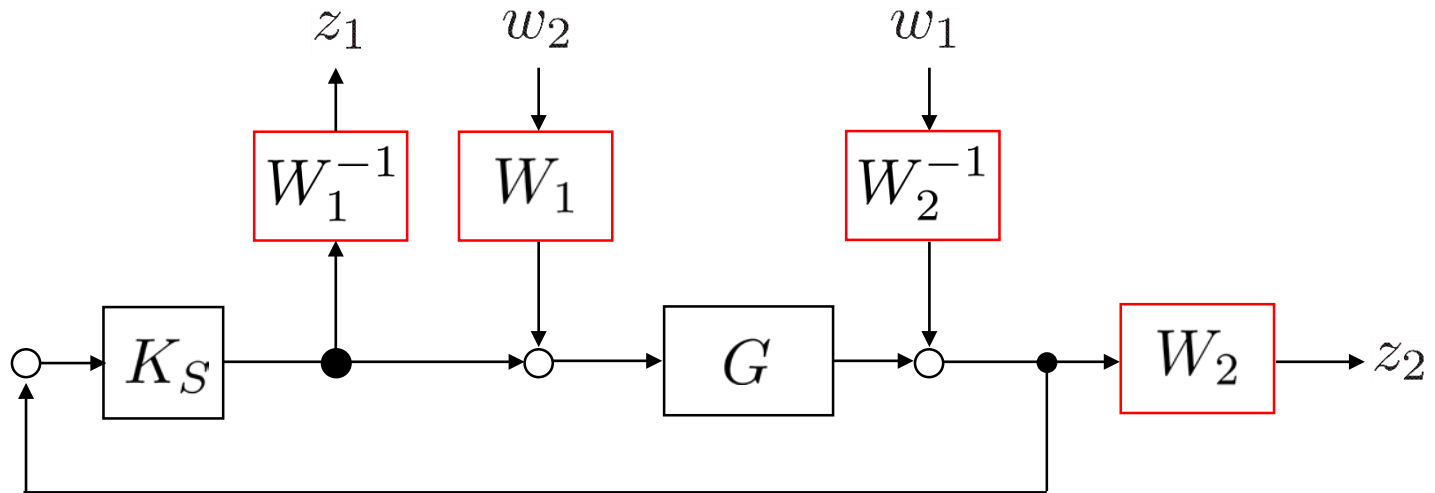
# Loop Shaping Design Procedure (LSDP)

D. McFarlane and K. Glover, "A loop Shaping Design Procedure Using  $H_\infty$  Synthesis," IEEE TAC, **37**-6, 759-769, 1992



The shaped plant and controller

# Loop Shaping Design

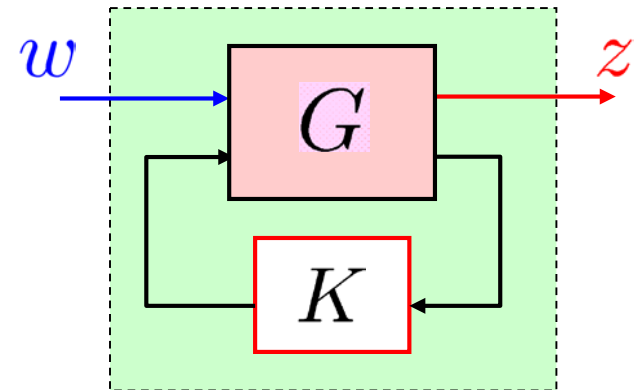


$$F_l(G, K)$$

$$= \begin{bmatrix} W_1^{-1}K \\ W_2 \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} W_2^{-1} & GW_1 \end{bmatrix}$$

$$= \begin{bmatrix} K_S \\ I \end{bmatrix} (I - G_S K_S)^{-1} \begin{bmatrix} I & G_S \end{bmatrix}$$

$$G_S = W_2 G W_1, \quad K = W_1 K_S W_2$$



$$z = F_l(G, K)w$$

# Loop Shaping Design Procedure (LSDP) [SP05, p. 368]

## STEP 1: Loop Shaping $G_s = W_2GW_1$

To shape the plant  $G$  using shaping functions  $W_1$  and  $W_2$

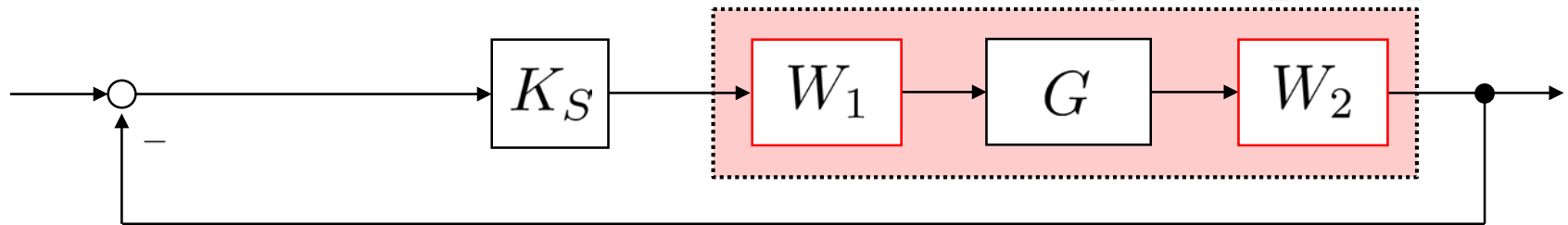
$W_1, W_2$  are chosen to satisfy keeping unstable pole of the model  $G$

## STEP 2: Robust Stabilization

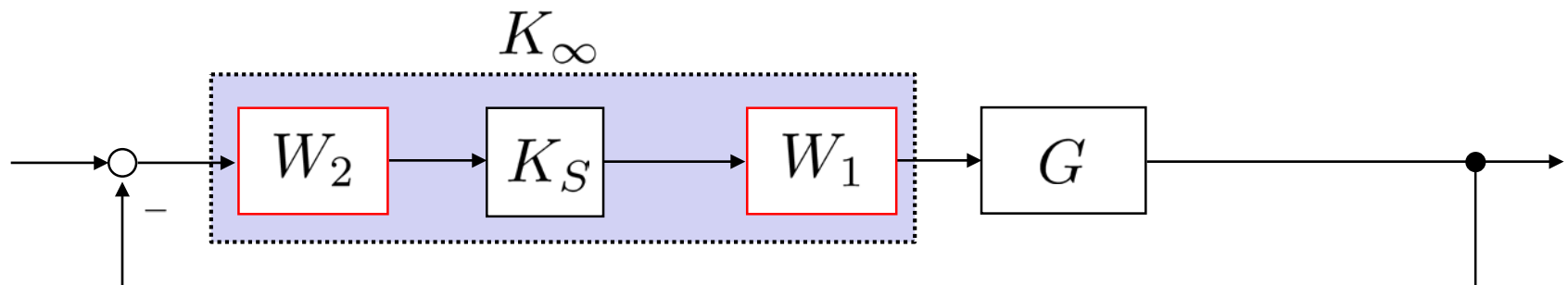
Find  $K_S$  such that  $\left\| \begin{bmatrix} K_S \\ I \end{bmatrix} (I - G_S K_S)^{-1} \begin{bmatrix} I & G_S \end{bmatrix} \right\|_{\infty} \leq \gamma$

(maximize  $b_0(G_S, K_S)$ )

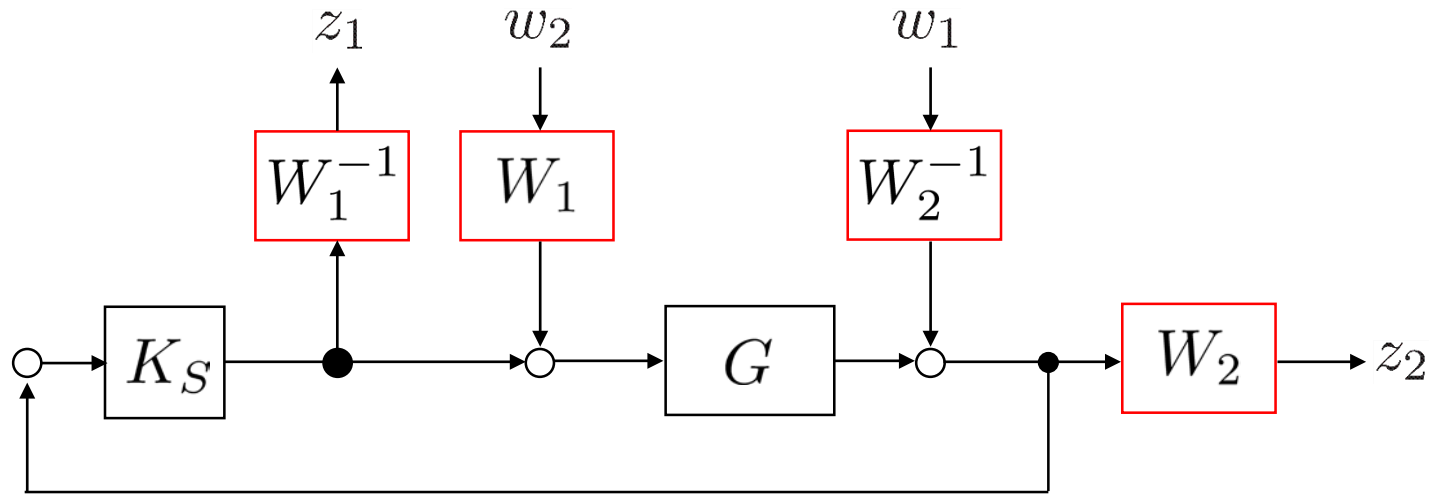
**Target Loop**  $G_s = W_2GW_1$



## STEP 3: $H_{\infty}$ Controller $K_{\infty} = W_1K_SW_2$



# LSDP: Decision Method of Weights $W_1, W_2$



$$G_s = W_2 G W_1 = \tilde{M}_s^{-1} \tilde{N}_s$$

$$\bar{\sigma}(\tilde{N}_s) = \left( \frac{\bar{\sigma}^2(G_s)}{1 + \bar{\sigma}^2(G_s)} \right)^{1/2} \quad \bar{\sigma}(\tilde{M}_s) = \left( \frac{1}{1 + \underline{\sigma}^2(G_s)} \right)^{1/2}$$

$$\left\{ \begin{array}{l} \bar{\sigma}(K(I - GK)^{-1}) \leq \gamma \bar{\sigma}(\tilde{M}_s) \bar{\sigma}(W_1) \bar{\sigma}(W_2) \\ \bar{\sigma}((I - GK)^{-1}) \leq \gamma \bar{\sigma}(\tilde{M}_s) c(W_2) \\ \bar{\sigma}(K(I - GK)^{-1}G) \leq \gamma \bar{\sigma}(\tilde{N}_s) c(W_1) \\ \bar{\sigma}((I - GK)^{-1}G) \leq \frac{\gamma \bar{\sigma}(\tilde{N}_s)}{\underline{\sigma}(W_1) \underline{\sigma}(W_2)} \end{array} \right.$$



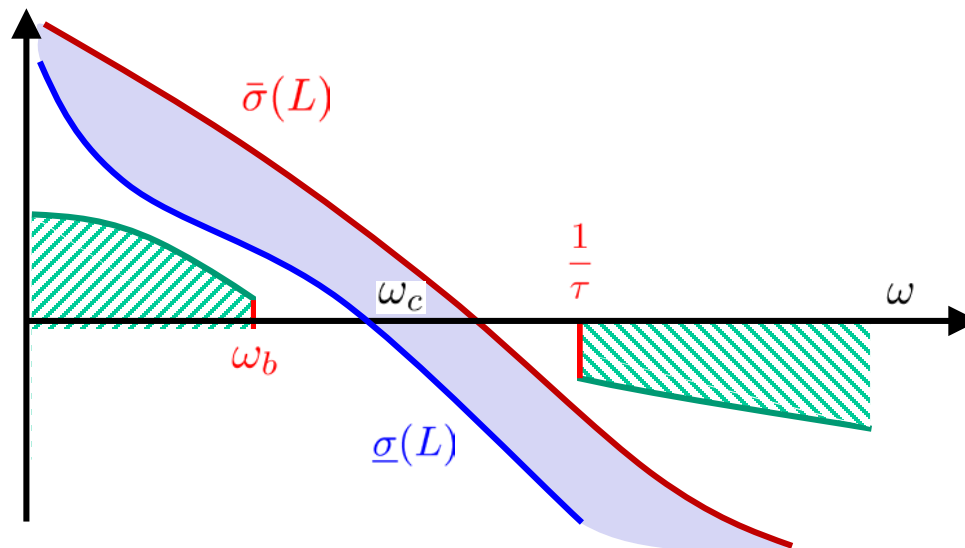
# LSDP in MIMO Systems

For Low Frequencies  $\underline{\sigma}(W_2GW_1) \gg 1$

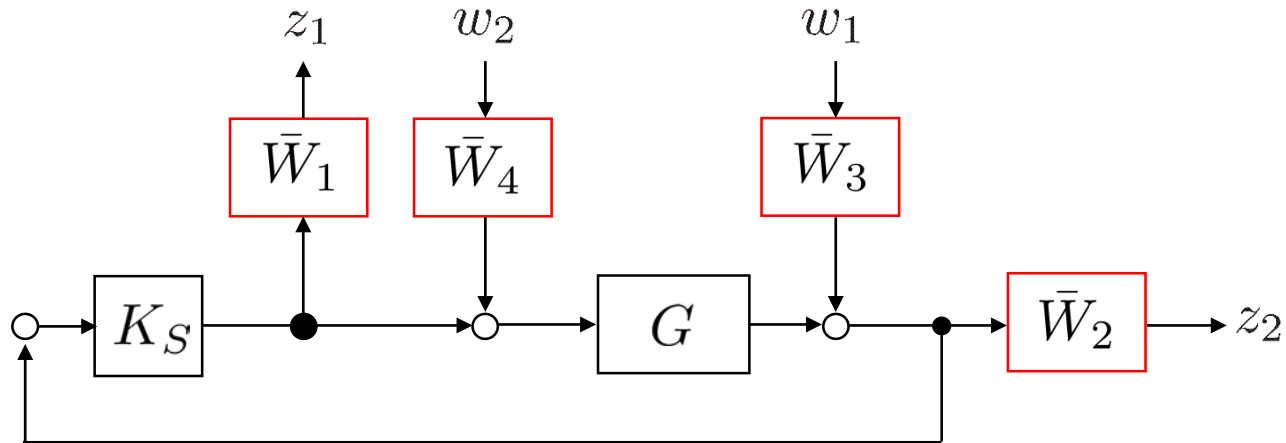
$$\Rightarrow \begin{cases} \bar{\sigma}((I - GK)^{-1}) \leq \frac{\gamma}{\underline{\sigma}(G)\underline{\sigma}(W_1)\underline{\sigma}(W_2)} \\ \bar{\sigma}((I - GK)^{-1}G) \leq \frac{\gamma}{\underline{\sigma}(W_1)\underline{\sigma}(W_2)} \end{cases}$$

For High Frequencies  $\bar{\sigma}(W_2GW_1) \ll 1$

$$\Rightarrow \begin{cases} \bar{\sigma}(K(I - GK)^{-1}) \leq \gamma\bar{\sigma}(W_1)\bar{\sigma}(W_2) \\ \bar{\sigma}(K(I - GK)^{-1}G) \leq \gamma\bar{\sigma}(G)\bar{\sigma}(W_1)\bar{\sigma}(W_2) \end{cases}$$

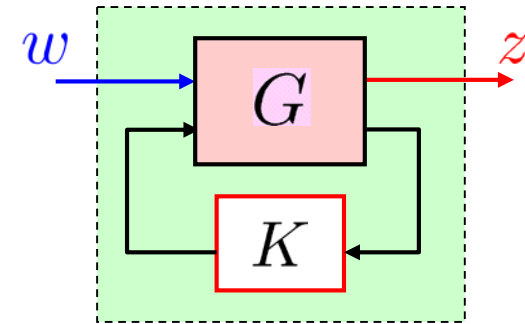


# Generalized Weighted Formulation



$$F_l(G, K) = \begin{bmatrix} \bar{W}_1 K \\ \bar{W}_2 \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} \bar{W}_3 & G\bar{W}_4 \end{bmatrix} \quad z = F_l(G, K)w$$

$$= \begin{bmatrix} K_S \\ I \end{bmatrix} (I - G_S K_S)^{-1} \begin{bmatrix} I & G_S \end{bmatrix}$$



$$G_S = \bar{W}_2 G \bar{W}_4, \quad K = \bar{W}_1^{-1} K_S \bar{W}_3^{-1}$$

$$\bar{W}_1 = W_1^{-1}, \quad \bar{W}_2 = W_2, \quad \bar{W}_3 = W_2^{-1}, \quad \bar{W}_4 = W_1$$

$\bar{W}_1 = 0, \bar{W}_4 = 0$  : minimum sensitivity problem

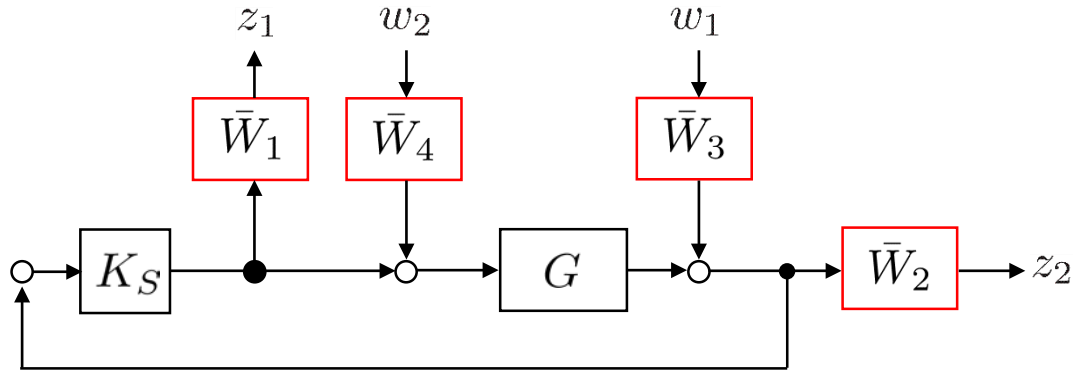
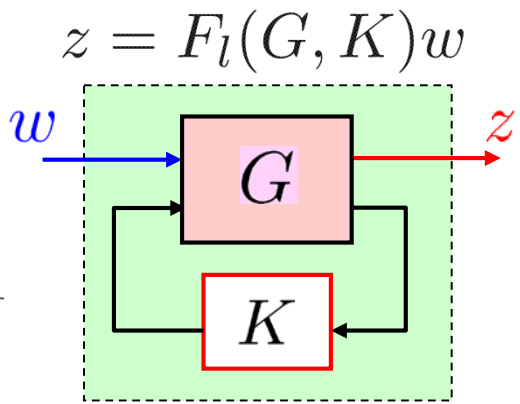
$\bar{W}_2 = 0, \bar{W}_4 = 0$  : robust control problem for additive uncertainty

$\bar{W}_4 = 0$  : mixed sensitivity problem

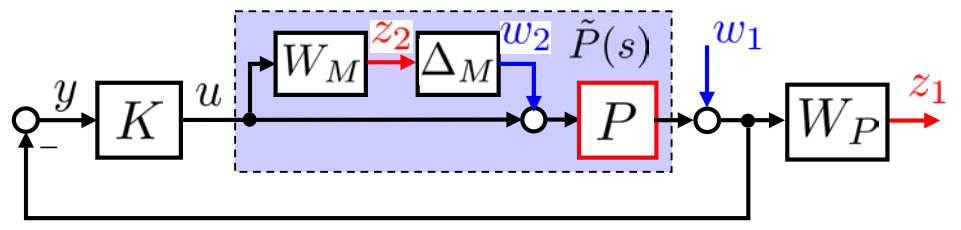


# Robust Performance

$$\begin{aligned}
 & \|F_l(G, K)\|_\infty \\
 &= \left\| \begin{bmatrix} K_s \\ I \end{bmatrix} (I - G_s K_s)^{-1} \begin{bmatrix} I & G_s \end{bmatrix} \right\|_\infty \\
 &= \left\| \begin{bmatrix} K_s S_s & K_s S_s G_s \\ S_s & S_s G_s \end{bmatrix} \right\|_\infty \quad S_s = (I - G_s K_s)^{-1}
 \end{aligned}$$



$$\left[ \begin{array}{l} \text{RP: Robust Performance} \\ F_l(G, K) \\ = \begin{bmatrix} -W_M T_I & -W_M K S_o \\ W_P S_o P & W_P S_o \end{bmatrix} \end{array} \right]$$



## Advantage of LSDP [SP05, p. 372]

- (1) LSDP is relatively easy to use, based on classical loop-shaping ideas
- (2) There exists a closed formula for the  $H_\infty$  optimal cost  $\gamma_{\min}$ , which in turn corresponds to a maximum stability margin  $\epsilon_{\max} = \gamma_{\min}^{-1}$
- (3) No  $\gamma$ -iteration is required in the solution  $\gamma_{opt}$ .  
[ **hinfsyn/dksyn** :  $\gamma$ -iteration ]
- (4) In case a process has a pole on the imaginary axis, LSDP does not require the additional operation to solve the problem.  

[	<b>hinfsyn</b> : <b>Assumptions</b> (A1) $(A, B_2)$ is stabilizable and $(C_2, A)$ is detectable (A2) $(A, B_1)$ is controllable and $(C_1, A)$ is observable <b>Full rank on the imaginary axis</b>	]
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- (5) Except for special systems, ones with all-pass factors, there are no pole-zero cancellations between the plant and controller
- (6) LSDP can permit a wider error of the model

# MATLAB Command loopsyn

$H_\infty$  optimal controller synthesis for LTI plant

```
[K, Cl, Gam, info] = loopsyn( G, Gd, RANGE )
```

Input argument

**G** (Generalized) LTI Plant

**Gd** Desired Loop-shape (LTI model)

(option) **RANGE** Desired frequency range for loop-shaping  
 $\{\omega_{\min}, \omega_{\max}\}$ ,  $10\omega_{\min} < \omega_{\max}$  (Default)  $\{0, \infty\}$

Output argument

**K** LTI Controller

**Cl** LTI Closed-loop system

**Gam** Loop-shaping accuracy  $\gamma \geq 1$ .  $\gamma = 1$  : perfect fit.

**Info** Information of output result

(option) Info.W,  $W$  satisfying  $\sigma(G_d(j\omega)) \approx \sigma(G(j\omega)W(j\omega))$ ,  $\forall \omega$

Info.Gs,  $G_s = GW$

Info.Ks,  $K_s = WK$

Info.range  $\{\omega_{\min}, \omega_{\max}\}$

# MATLAB Command ncfsyn

Loop shaping design using Glover-McFarlane method

```
[K, Cl, Gam, info] = ncfsyn( G, W1, W2, 'ref' )
```

Input argument

**G** (Generalized) LTI Plant

**W1, W2** Weights

**(option) 'ref'** Compute normalized coprime factor loop-shaping controller

Output argument

**K** LTI Controller

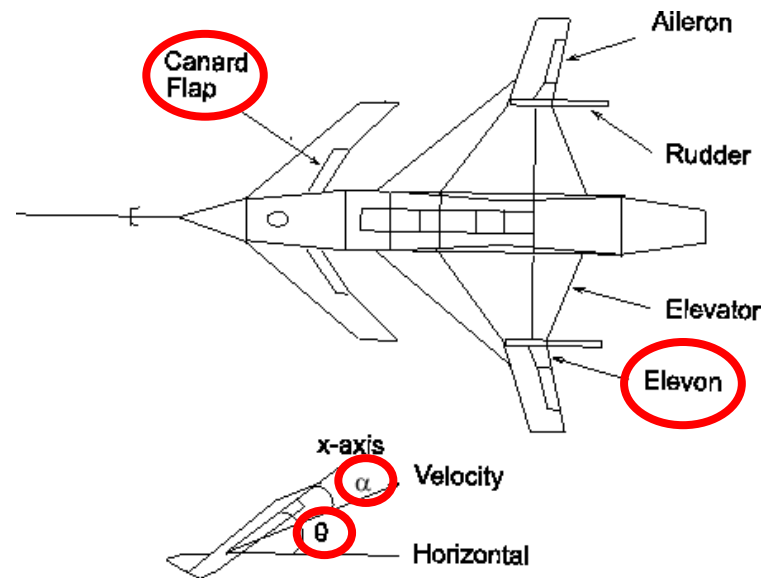
**Cl** LTI Closed-loop system

**Gam** Loop-shaping accuracy  $\gamma \geq 1$  .  $\gamma = 1$  : perfect fit.

**Info** Information of output result



# The 2-by-2 NASA HIMAT aircraft model (Loop Shaping of HIMAT Pitch Axis Controller)



$$x^T = \begin{pmatrix} \dot{\alpha} & \alpha & \dot{\theta} & \theta & x_e & x_c \end{pmatrix}$$

$$u^T = \begin{pmatrix} \delta_e & \delta_c \end{pmatrix} \quad y^T = \begin{pmatrix} \alpha & \theta \end{pmatrix}$$

$\alpha$  : angle of attack  
 $\theta$  : attitude angle  
 $\delta_e$  : elevon actuator  
 $\delta_c$  : canard actuator

# HIMAT: Nominal Plant Model

## State Space Form (Matrix Representation)

$$G = \left[ \begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$$

$$x^T = \left( \dot{\alpha} \quad \alpha \quad \dot{\theta} \quad \theta \quad x_e \quad x_c \right)$$

$$u^T = \left( \delta_e \quad \delta_c \right) \quad y^T = \left( \alpha \quad \theta \right)$$

$$A = \begin{pmatrix} -0.023 & -36.62 & -18.90 & -32.09 & 3.251 & -0.763 \\ 0.000 & -1.900 & -0.983 & -0.000 & -0.171 & -0.005 \\ 0.012 & 11.72 & -2.632 & 0.000 & -31.60 & 22.40 \\ 0 & 0 & 1.000 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -30.00 & 0 \\ 0 & 0 & 0 & 0 & 0 & -30.00 \end{pmatrix} \quad B = \begin{pmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ 30 & 0 \\ 0 & 30 \end{pmatrix}$$

$$C = \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{pmatrix} \quad D = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

### MATLAB Command

```
% NASA HiMAT model G(s)
ag = [-2.2567e-02 -3.6617e+01 -1.8897e+01 -3.2090e+01 3.2509e+00 -7.6257e-01;
      9.2572e-05 -1.8997e+00 9.8312e-01 -7.2562e-04 -1.7080e-01 -4.9652e-03;
      1.2338e-02 1.1720e+01 -2.6316e+00 8.7582e-04 -3.1604e+01 2.2396e+01;
      0 0 1.0000e+00 0 0 0;
      0 0 0 0 -3.0000e+01 0;
      0 0 0 0 0 -3.0000e+01];
bg = [0 0; 0 0; 0 0; 0 0; 30 0; 0 30];
cg = [0 1 0 0 0 0; 0 0 0 1 0 0];
dg = [0 0; 0 0];
G=ss(ag,bg,cg,dg);
```

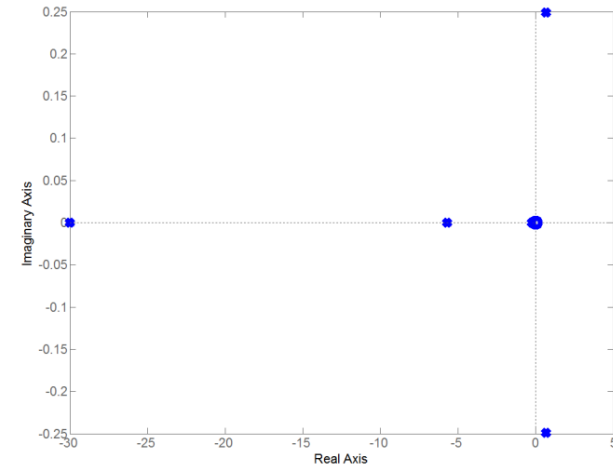
# HIMAT: Nominal Plant Model

Controllability ○

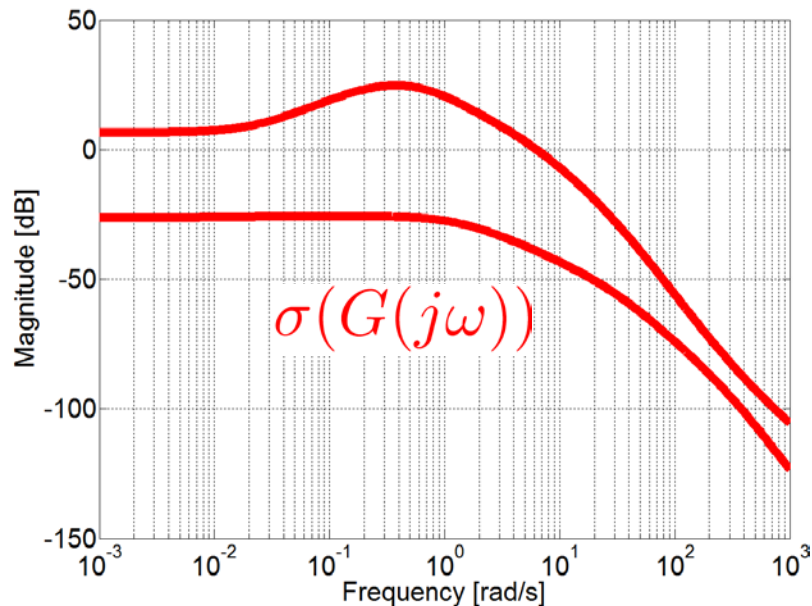
Observability ○

Poles (Stability) -5.6757, -0.2578, -30, -30,  
 $0.6898 \pm 0.2488i$   
Unstable

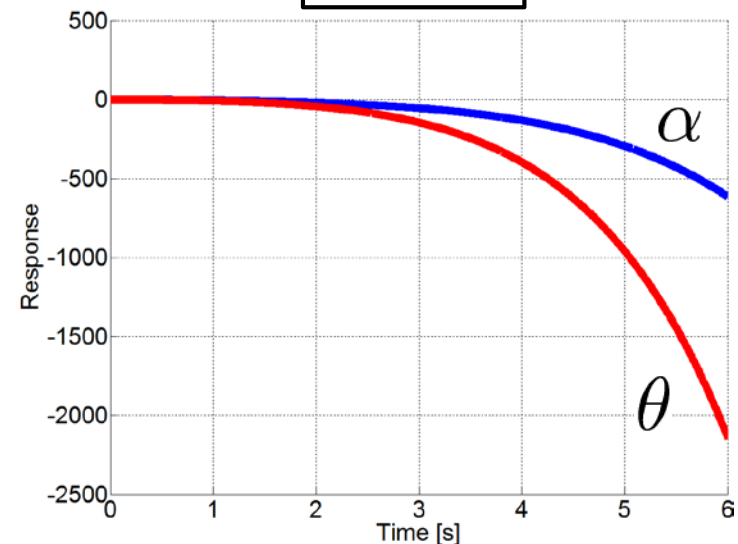
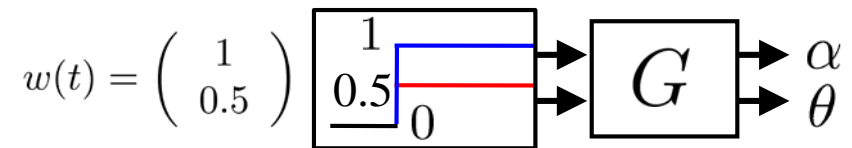
Zeros -0.0210



## Frequency Response

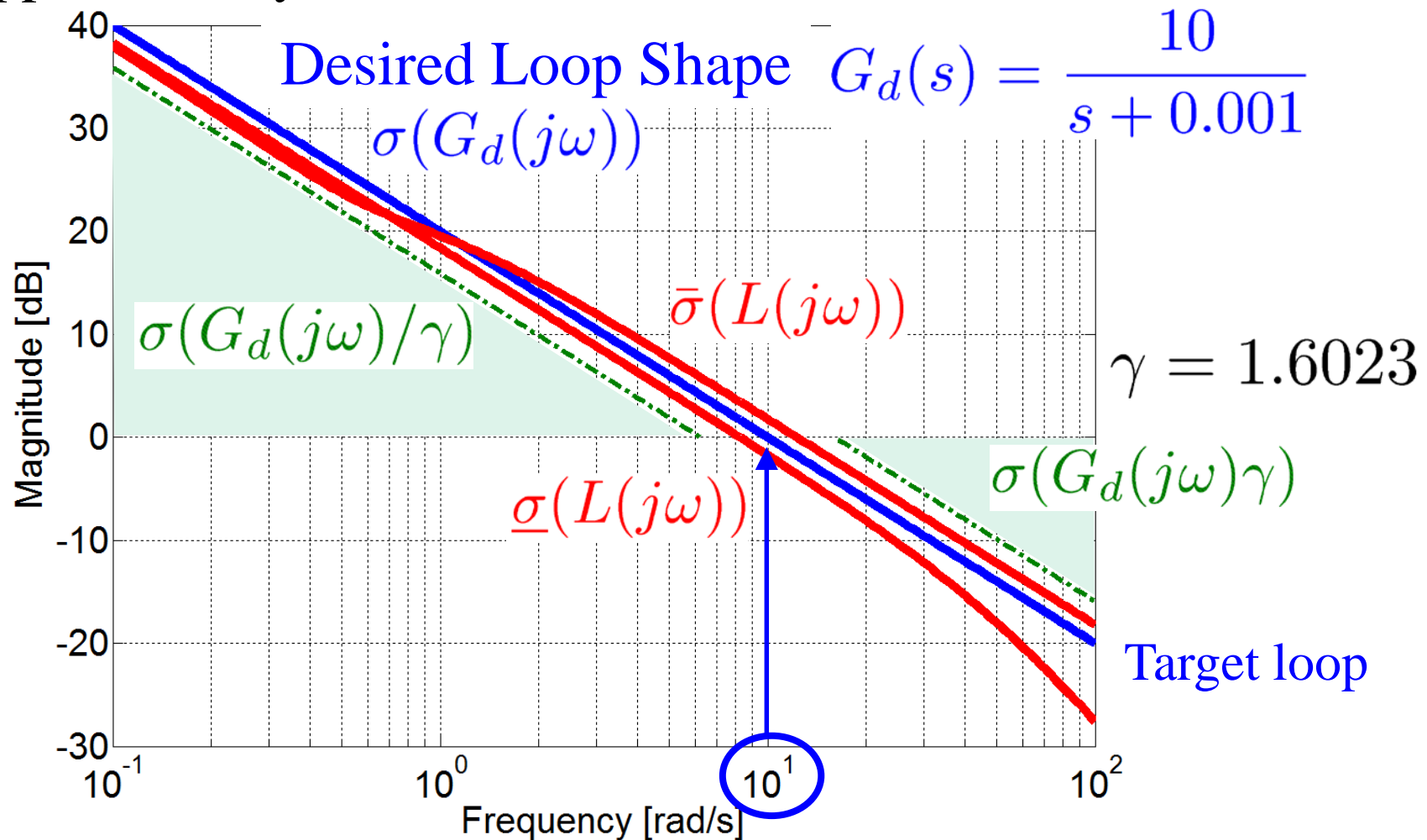


## Step Response for Nominal Plant Model



# HIMAT: Specifications and Open-loop

Approximately a bandwidth of 10 rad/s

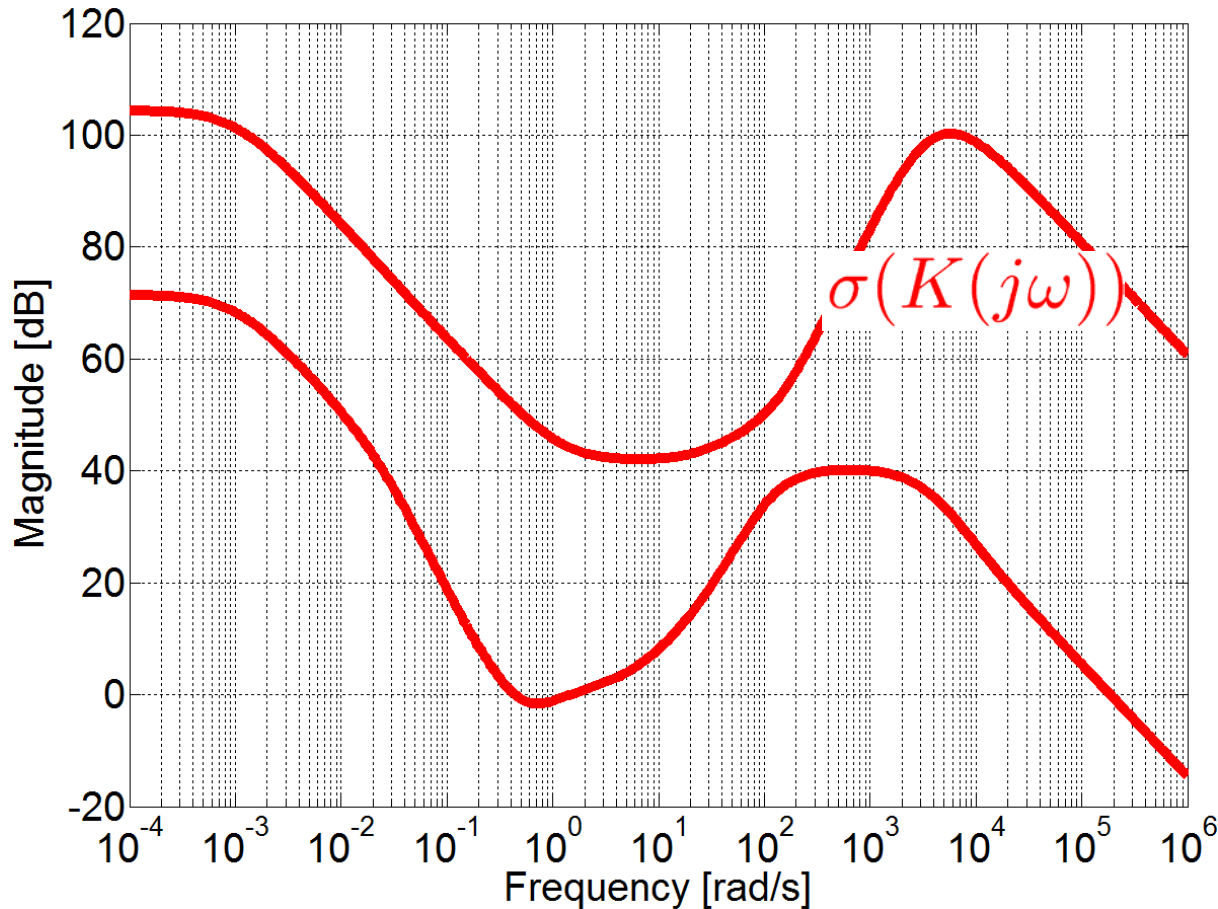


## MATLAB Command

```
Gd= tf( 10, [1 .001] );  
[K,CL,GAM,INFO]=loopsyn(G,Gd);  
sigma(Gd, 'b', G*K, 'r', Gd/GAM, 'g:', Gd*GAM, 'g:', {.1,100})
```



# HIMAT: Controller



MATLAB Command

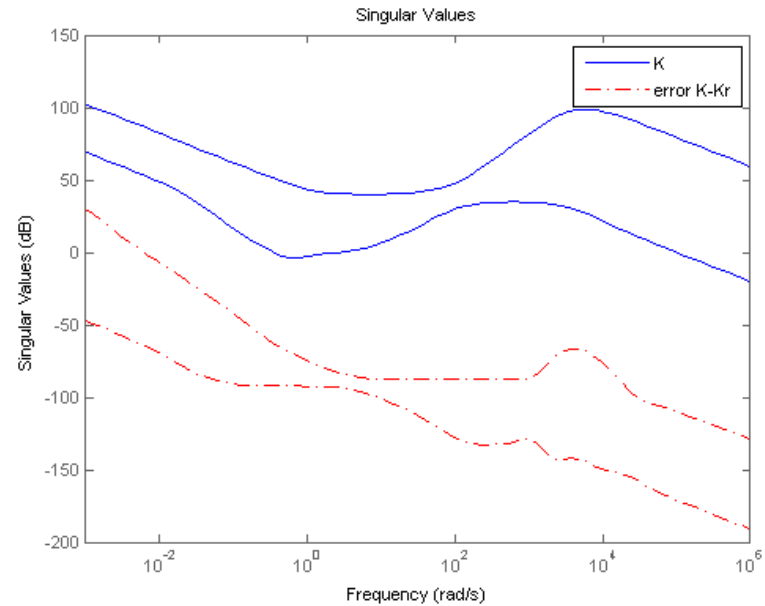
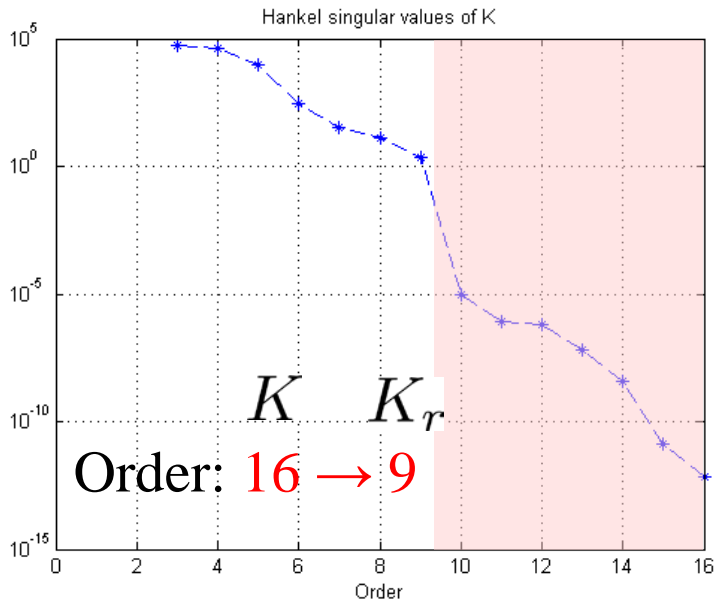
```
sigma(K);
```

Order: 16

Numerical problems or inaccuracies may be caused too high order

➔ Difficult to implement

# HIMAT: Controller Model Reduction



## Step responses

### MATLAB Command

```
size(K)
```

```
hsv = hankelsv(K);
```

```
semilogy(hsv,'*--'), grid
```

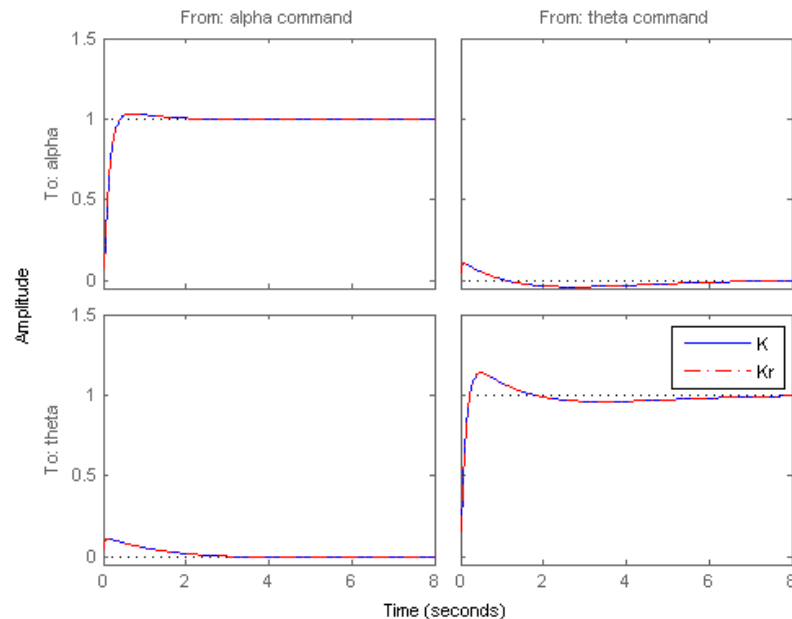
```
title('Hankel singular values of K'),
```

```
xlabel('Order')
```

```
Kr = reduce(K,9); order(Kr)
```

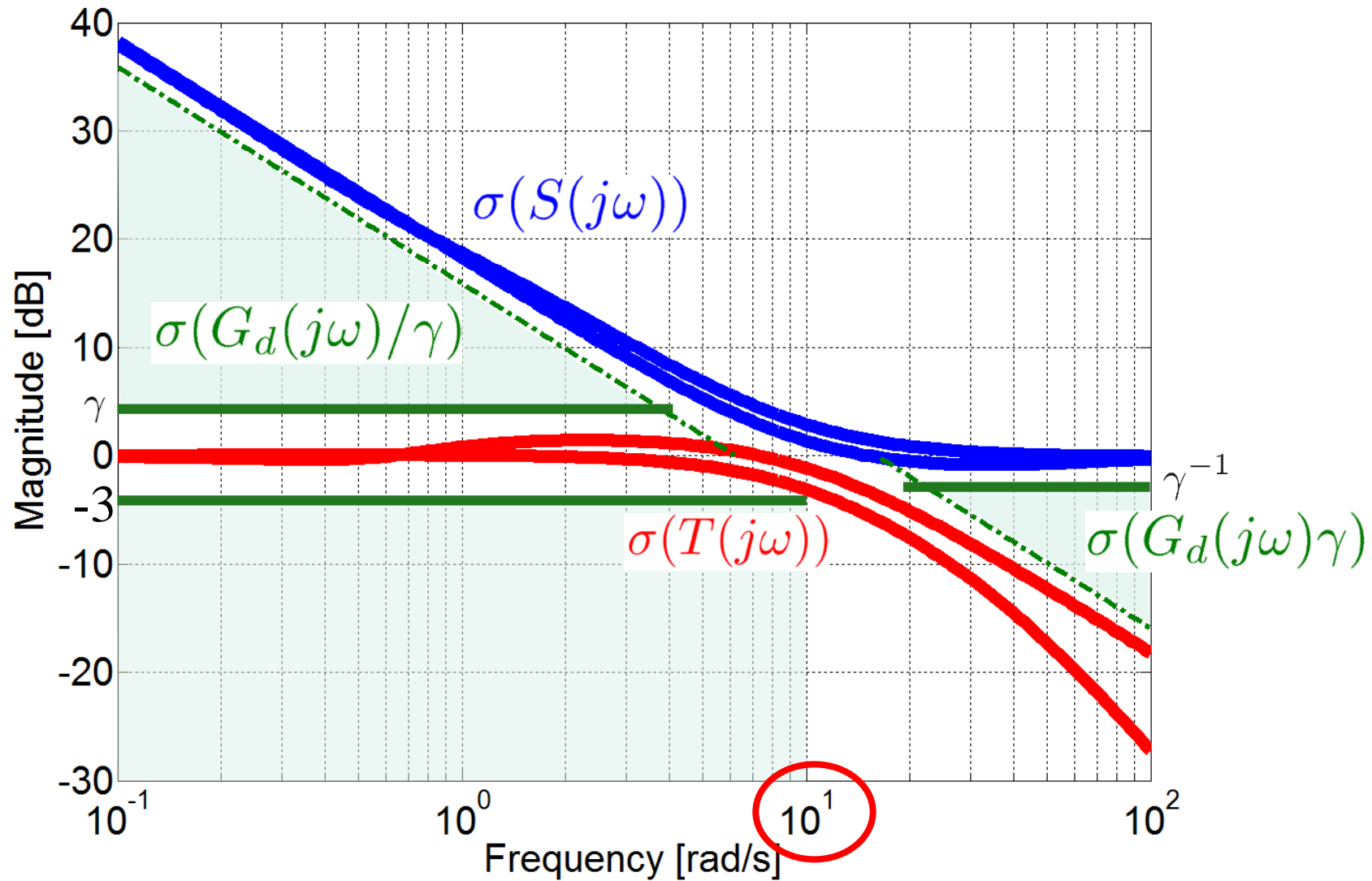
```
sigma(K,'b',K-Kr,'r-.')
```

```
legend('K','error K-Kr')
```



—  $K$   
 - -  $K_r$

# HIMAT: Sensitivity

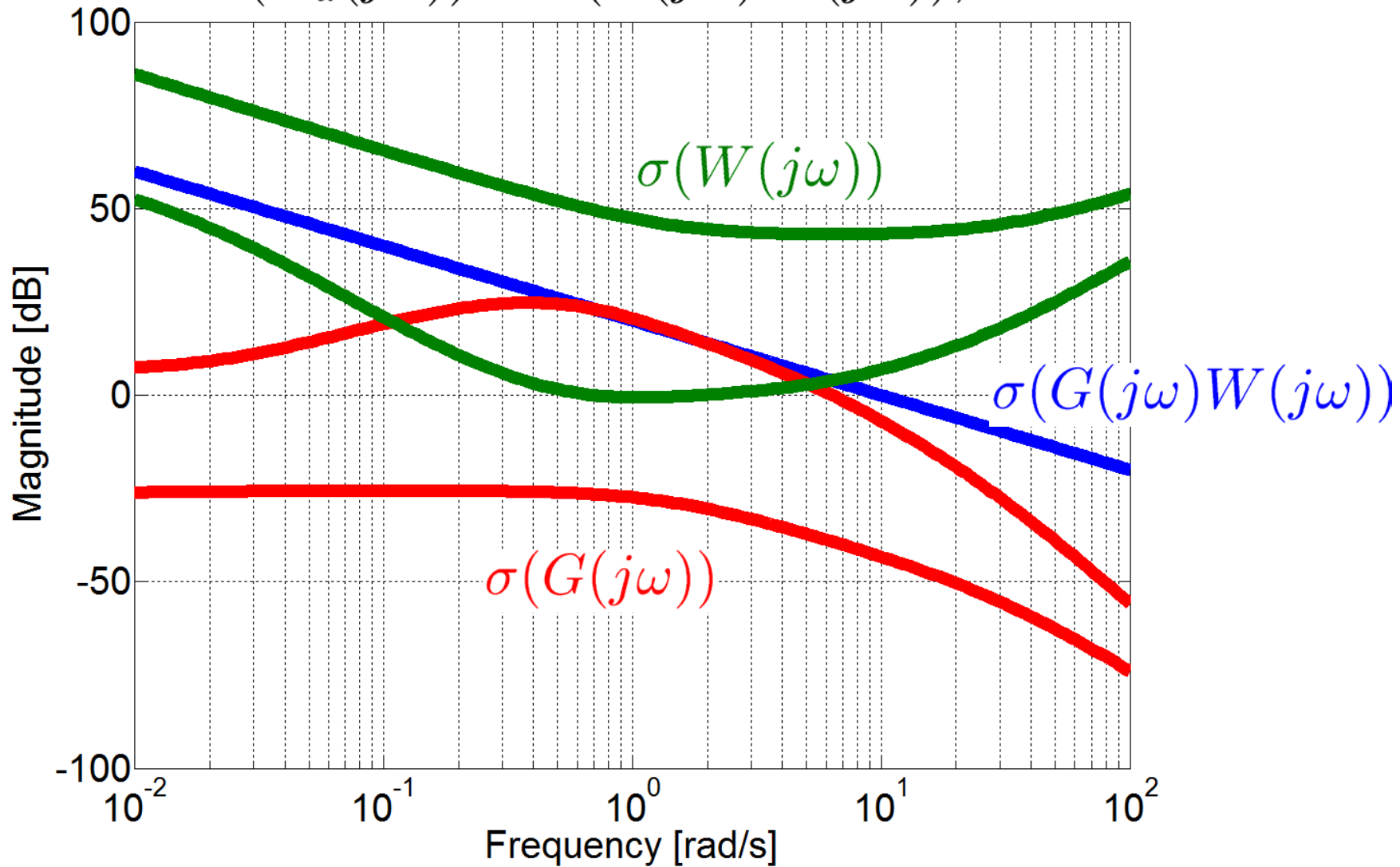


## MATLAB Command

```
T = feedback(G*K,eye(2));  
S = eye(2)-T;  
sigma(inv(S),'m',T,'g',L,'r--',Gd,'b',Gd/GAM,'b:',Gd*GAM,'b:',{.1,100})
```

# HIMAT: Designed Weight

$$\sigma(G_d(j\omega)) \approx \sigma(G(j\omega)W(j\omega)), \forall \omega$$

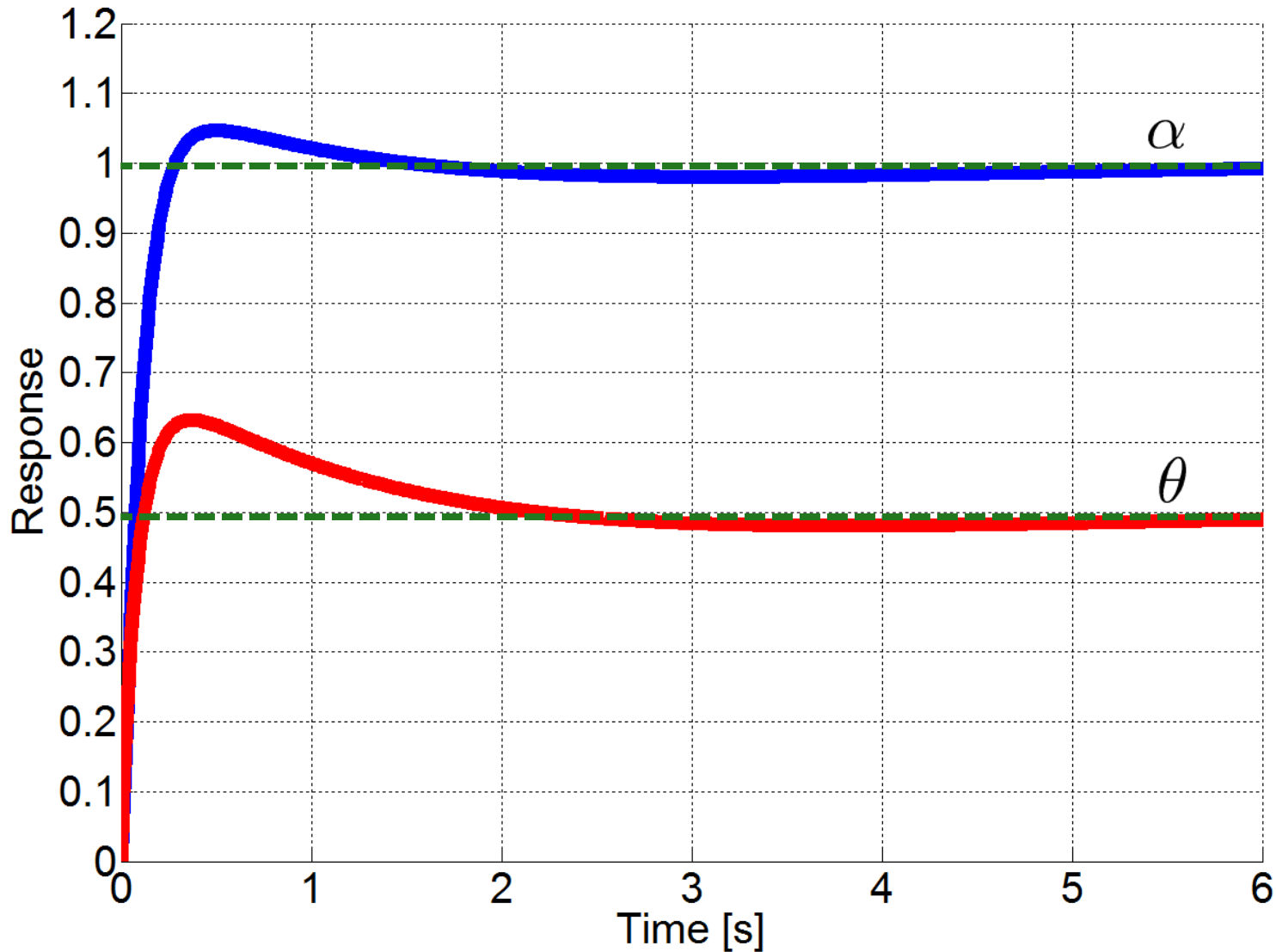
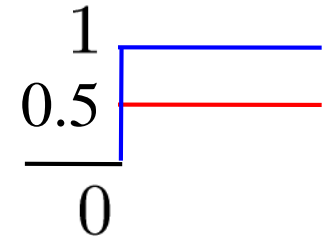


## MATLAB Command

```
sigma(INFO.Gs,'b',G,'r',INFO.W,'g',{.01,100});
```

# HIMAT: Step Response (closed loop)

$$w(t) = \begin{pmatrix} 1 \\ 0.5 \end{pmatrix}$$





# Youla Parameterization

Case 2: **Unstable** Plant  $P(s)$  [SP05, p. 149]

Coprime Factorization [SP05, p. 122]

$P(s) = \frac{N(s)}{M(s)}$  Coprime: No common right-half plane(RHP) zeros  
 $N(s), M(s)$ : **Proper Stable** Transfer Functions

[SP05, Ex. 4.1]  $P(s) = \frac{(s-1)(s+2)}{(s-3)(s+4)}$

➔  $N(s) = \frac{s-1}{s+4}$ ,  $M(s) = \frac{s-3}{s+2}$  (\*)

Bezout Identity  $NX + MY = 1$   $\leftrightarrow$   $M(s), N(s)$  : Coprime

$X(s), Y(s)$  : **Proper Stable** Transfer Functions

[SP05, Ex.]  $M(s), N(s)$  : (\*) ➔  $X(s) = \frac{s+32}{2s+4}$ ,  $Y(s) = \frac{s-16}{2s+8}$

[Ex.]  $5x + 3y = 1$   $x, y$  : Integer ➔  $x = -1 - 3q$ ,  $y = 2 + 5q$

$q$  : Integer



# Youla Parameterization

Case 2: **Unstable** Plants  $P(s)$  [SP05, p. 149]

**A Stabilizing Controller**  $K(s) = \frac{X(s)}{Y(s)} \quad (Q(s) = 0)$

[SP05, Ex.]  $P(s) = \frac{(s-1)(s+2)}{(s-3)(s+4)}, \quad X(s) = \frac{s+32}{2s+4}, \quad Y(s) = \frac{s-16}{2s+8}$

→  $K(s) = \frac{X(s)}{Y(s)} = \frac{s^2 + 36s + 128}{s^2 - 14s - 32}$

**All Stabilizing Controllers**  $K(s) = \frac{X(s) + M(s)Q(s)}{Y(s) - N(s)Q(s)}$

$\left[ N = P, M = 1, X = 0, Y = 1 \rightarrow K = \frac{Q}{1 - PQ} \right]$

**Gang of Four**

$$\begin{aligned} S &= \underline{M(Y - NQ)} & T &= \underline{N(X + MQ)} \\ PS &= \underline{N(Y - NQ)} & KS &= \underline{M(X + MQ)} \end{aligned}$$

**Affine Functions of  $Q$**



# Coprime Factorization: State-space Procedure

## Observer design

$$\begin{cases} \dot{x} = Ax + Bu - L(y - Cx) \\ y = Cx \end{cases}$$

$$y = C(sI - A - LC)^{-1} \begin{bmatrix} B & -L \end{bmatrix} \begin{bmatrix} u \\ y \end{bmatrix}$$

$$= \begin{bmatrix} N_l & I - M_l \end{bmatrix} \begin{bmatrix} u \\ y \end{bmatrix} \quad \longrightarrow \quad y = M_l(s)^{-1} N_l(s) u$$

## State feedback pole assignment

$$\begin{cases} \dot{x} = (A + BF)x + Be \\ e = u + z, \quad z = -Fx \\ y = Cx \end{cases}$$

$$\begin{bmatrix} y \\ z \end{bmatrix} = \begin{bmatrix} C \\ -F \end{bmatrix} (sI - A - BF)^{-1} Be = \begin{bmatrix} N_r \\ I - M_r \end{bmatrix} e$$

$$\longrightarrow e = u + (I - M_r(s))e = M_r(s)^{-1} u$$

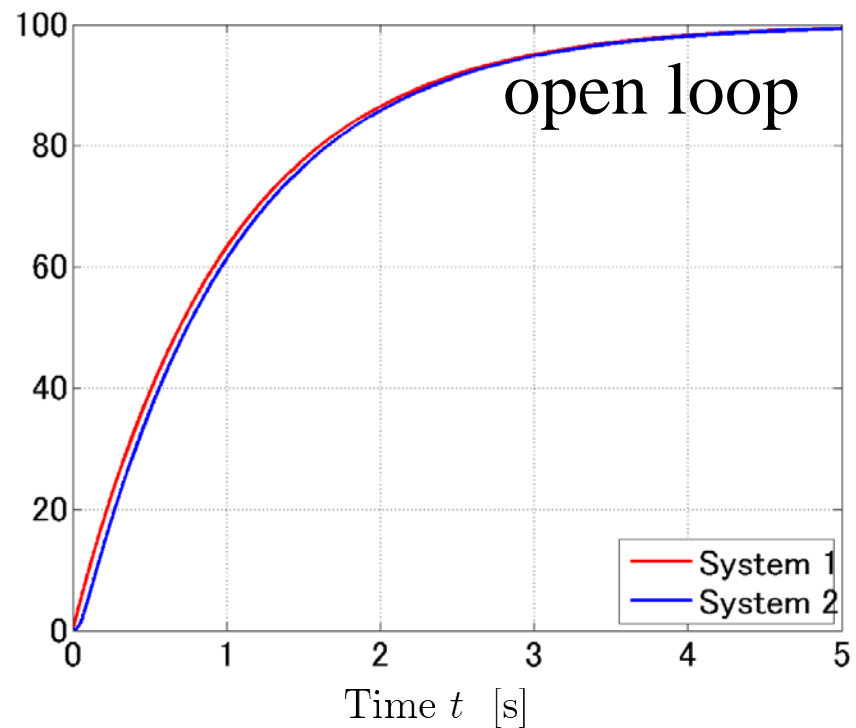
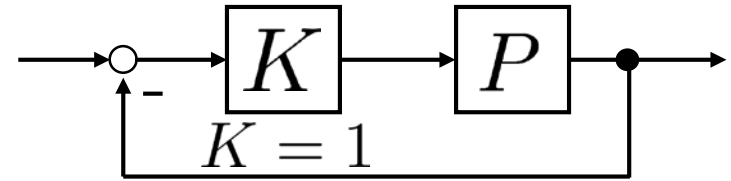
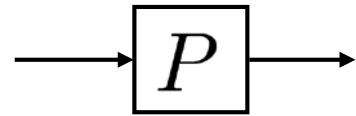
$$\longrightarrow y = N_r(s)e = N_r(s)M_r(s)^{-1} u$$



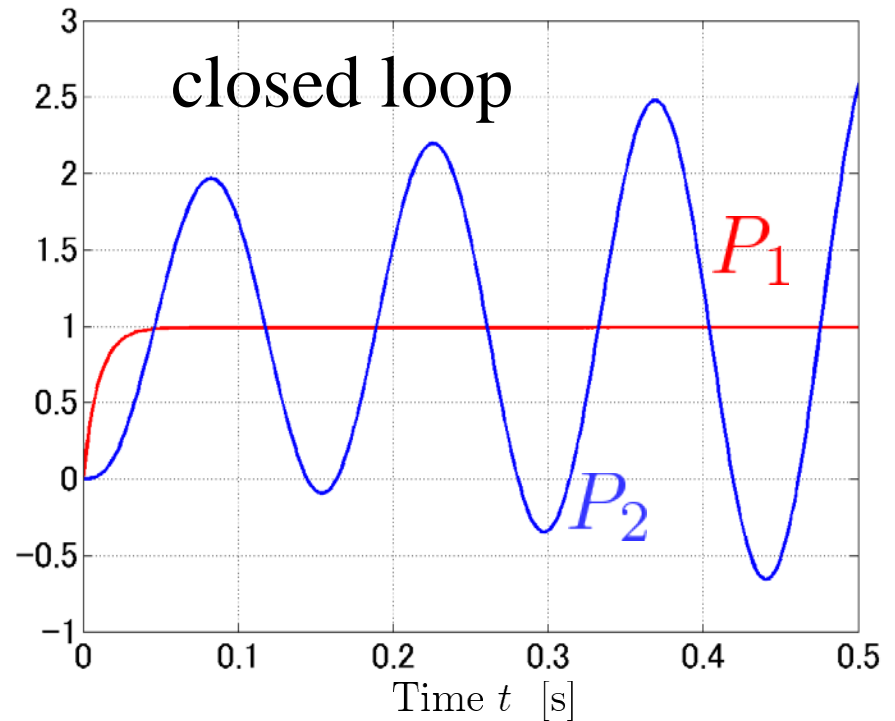


# When Are Two Systems Similar ? [AM09, pp. 349-352]

[AM09, Ex 12.2]  $P_1(s) = \frac{100}{s + 1}$  ,  $P_2(s) = \frac{100}{(s + 1)(0.025s + 1)^2}$



(a) Step response (open loop)



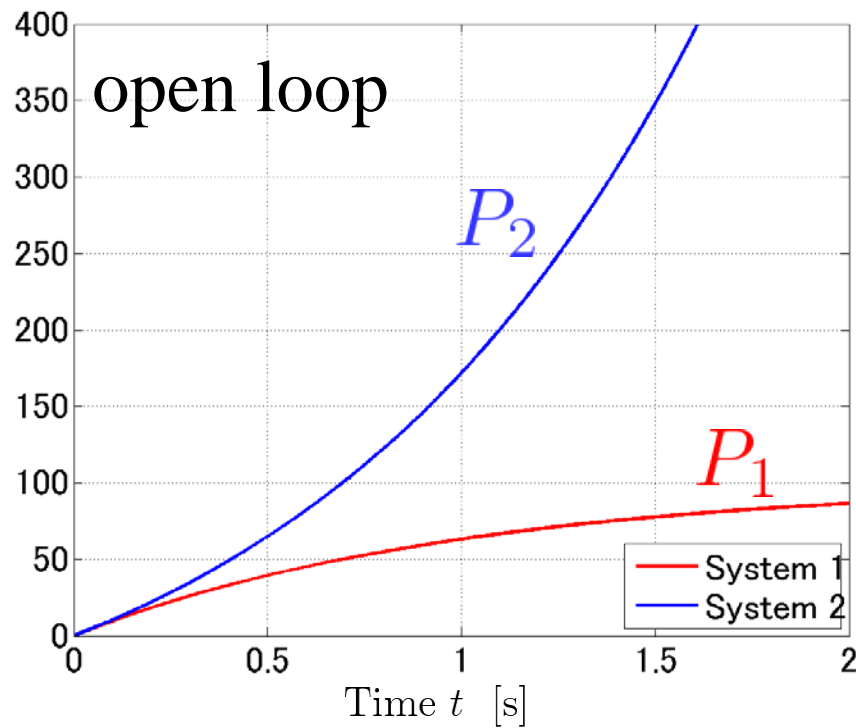
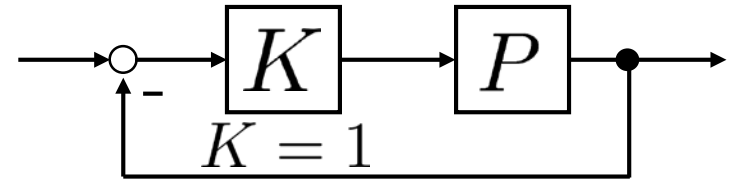
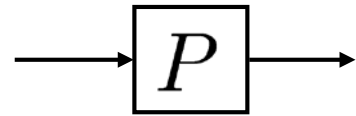
(b) Step response (closed loop)

Similar in Open Loop but Large Differences in Closed Loop

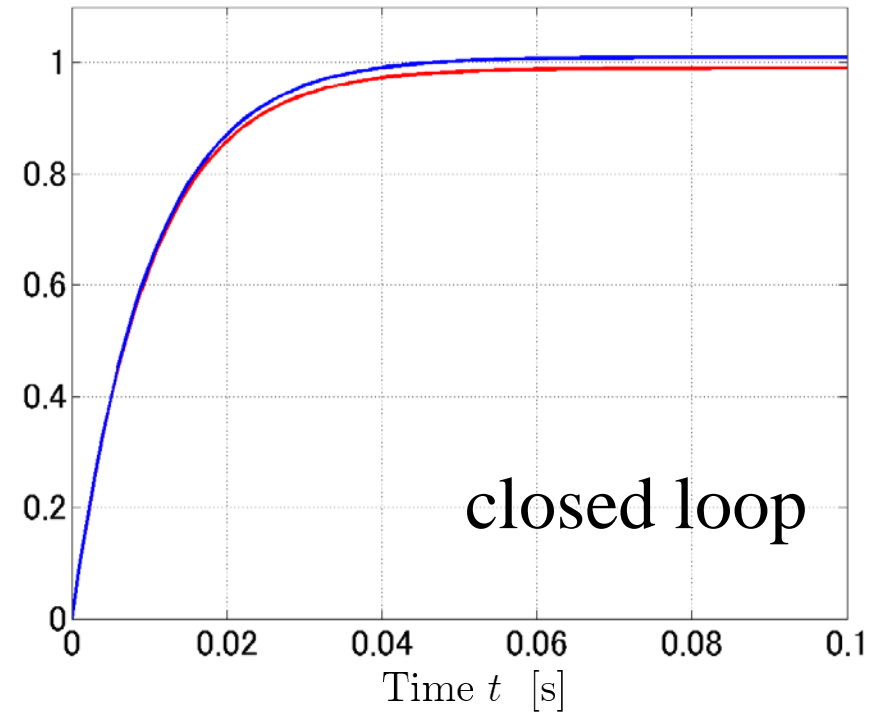


# When Are Two Systems Similar ? [AM09, pp. 349-352]

[AM09, Ex 12.3]  $P_1(s) = \frac{k}{s+1}$  ,  $P_2(s) = \frac{k}{(s-1)}$   $k = 100$



(a) Step response (open loop)



(b) Step response (closed loop)

Different in Open Loop but Similar in Closed Loop

# Vinnicombe Metric ( $\nu$ -gap Metric) [Zhou98, Chap.17]



$\delta_g(G_o, G_1)$  : the smallest value of  $\| [\Delta_N(j\omega) \quad \Delta_M(j\omega)] \|_\infty$   
that perturbs  $G_o$  into  $G_1$  is called the *gap* between  $G_o$  and  $G_1$

If  $\delta_g(G_o, G_1) < b(G_o, K)$  , then the closed loop system  
with  $G_1$  and  $K$  will also be stable

$\delta_\nu(G_o, G_1)$  : the  $\nu$ -gap between  $G_o$  and  $G_1$

If  $\delta_\nu(G_o, G_1) < b(G_o, K)$  ,

then we have closed loop stability of  $G_1$  and  $K$

$b(G_o, K)$  gives the radius (in terms of the distance in the  $\nu$ -gap metric)  
of the largest “ball” of plants stabilized by  $K$

- Note:** Both  $\delta_g$  and  $\delta_\nu$  are metrics (i.e. distance measures)
- (1)  $0 \leq \delta_\nu(G_o, G_1) \leq 1$
  - (2)  $\delta_\nu(G_o, G_1) = 0 \Rightarrow G_o = G_1$
  - (3)  $\delta_\nu(G_o, G_1) = \delta_\nu(G_1, G_o)$
  - (4)  $\delta_\nu(G_o, G_2) \leq \delta_\nu(G_o, G_1) + \delta_\nu(G_1, G_2)$  (Triangle inequality)





# NLCF Robust Control Problem

Nominal Plant Model  $G = \left[ \begin{array}{c|c} A & B \\ \hline C & D \end{array} \right]$

Sub-optimal Solution ( $H_\infty$  controller)

$X > 0$  ,  $Z > 0$  satisfying that

Riccati equalities

$$X \tilde{A}_X + \tilde{A}_X^T X - X B S^{-1} B^T X + C^T R^{-1} C = 0$$

$$Z \tilde{A}_Z^T + \tilde{A}_Z Z - Z C^T R^{-1} C Z + B S^{-1} B^T = 0$$

where  $\tilde{A}_X = A - B S^{-1} D^T C$  ,  $\tilde{A}_Z = A - B D^T R^{-1} C$

$$R = D D^T , S = I + D^T D$$

$$\begin{bmatrix} K_{11} & K_{12} \\ K_{21} & K_{22} \end{bmatrix} = \left[ \begin{array}{c|cc} A + B F & -\gamma^2 W_1^{-T} B S^{-1/2} & \gamma^2 \zeta^{-1} W_1^{-T} Z C^T R^{-1/2} \\ \hline F & S^{-1/2} & \zeta^{-1} D^T R^{-1/2} \\ C + D F & D S^{-1/2} & -\zeta^{-1} R^{-1/2} \end{array} \right]$$

where  $\zeta = \sqrt{\gamma^2 - 1}$  ,  $W_1 = (1 - \gamma^2) I + X Z$  ,

$$F = -S^{-1} (D^T C + B^T X) , S = I + D^T D$$



# Observer-based Structure

Central controller ( $\Phi = 0$ )

$$K = \left[ \begin{array}{c|c} \frac{A + BF + \gamma^2 W_1^{-T} Z C^T (C + DF)}{B^T X} & \frac{\gamma^2 W_1^{-T} X C^T}{-D^T} \end{array} \right]$$

➔ 
$$\begin{cases} \dot{x}_k = Ax_k + Bu + \hat{H}(y - Cx_k - Du) \\ u = -B^T X x_k - D^T y \end{cases}$$

where 
$$\hat{H} = (BD^T + V^{-1}C^T)R^{-1},$$
$$V = Z^{-1} - \gamma^{-2}(X + Z^{-1})$$

## Features

Observer gain  $\hat{H}$  is **automatically** designed

Observer gain  $\hat{H}$  is related to both  $Z$  and  $X$  (Riccati Solutions)

$\gamma \rightarrow \infty$  ➔ 
$$\hat{H}_\infty = (BD^T + ZC^T)R^{-1}$$

Cost Function 
$$J = \int_0^\infty (y^T \hat{Q}y + u^T \hat{R}u) dt$$

➔ 
$$W_1(s) = \hat{R}^{-1/2}(s), \quad W_2(s) = \hat{Q}^{1/2}(s)$$



# Computation

$$G = \left[ \begin{array}{c|c} A & B \\ \hline C & D \end{array} \right] \quad G \text{ is stable} \Leftrightarrow \operatorname{Re}[\lambda(A)] < 0$$

$\mathcal{H}_2$  -norm      Controllability/Observability

$\mathcal{H}_\infty$  -norm      Hamiltonian Matrix

$$H = \left[ \begin{array}{c|c} A + BS^{-1}D^T C & BS^{-1}B^T \\ \hline -C^T(I + DS^{-1}D^T)C & -(A + BS^{-1}D^{-1}C)^T \end{array} \right]$$

$$s = \gamma^2 I - D^T D$$

## Theorem

$$\|G(s)\|_\infty < \gamma \Leftrightarrow \begin{cases} \bar{\sigma}(D) < \gamma \\ H \text{ has no eigenvalues on imaginary axis} \end{cases}$$

## Graphical Test

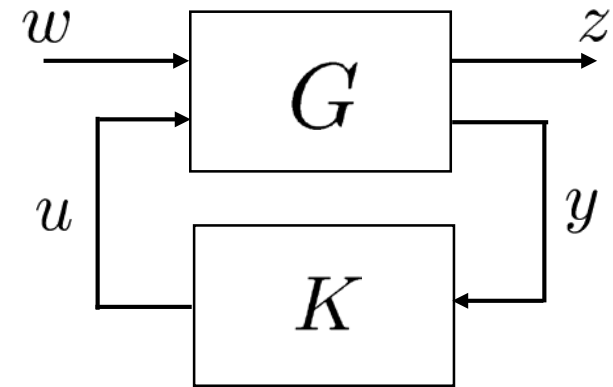
$$\max_{\omega} \bar{\sigma}[G(j\omega)] < \gamma$$



# $H_\infty$ Frobenius synthesis with Hadamard weight

Given  $\gamma > \gamma_{min}$ , find all stabilizing controllers  $K$  such that

$$\|W \circ F_l(G, K)\|_{\infty F} \leq \gamma$$







# Stability Margin

$$b(G, K) := \left\| \begin{bmatrix} K \\ I \end{bmatrix} (I - GK)^{-1} \begin{bmatrix} I & G \end{bmatrix} \right\|_{\infty}^{-1}$$

The closed loop will be stable for all  $\| \begin{bmatrix} \Delta_N & \Delta_M \end{bmatrix} \|_{\infty} < \epsilon \Leftrightarrow b(G, K) \geq \epsilon$

$b(G, K) > 0.2 \sim 0.3$  (for good robustness)

## In SISO Systems

$$GM \geq \frac{1 + b(G, K)}{1 - b(G, K)}, \quad PM \geq 2 \arcsin(b(G, K))$$

$$\left[ \begin{array}{l} \text{Proof: } \bar{\sigma}^2 \left\{ \begin{bmatrix} K \\ 1 \end{bmatrix} (1 - GK)^{-1} \begin{bmatrix} 1 & G \end{bmatrix} \right\} \\ = (1 + |K|^2) |1 - GK|^{-2} (1 + |G|^2) \leq \frac{1}{b^2(G, K)}, \quad \forall \omega \\ \therefore b^2(1 + |K|^2) \left(1 + \frac{GM^2}{|K|^2}\right) \leq |1 - GM|^2 \\ \Rightarrow b^2(1 + GM)^2 \leq (1 - GM)^2 \quad \text{for } 0 \leq GM \leq 1 \end{array} \right]$$



# LSDP in SISO Systems

For Low frequencies

$$\underline{\sigma}(WG) \gg 1$$

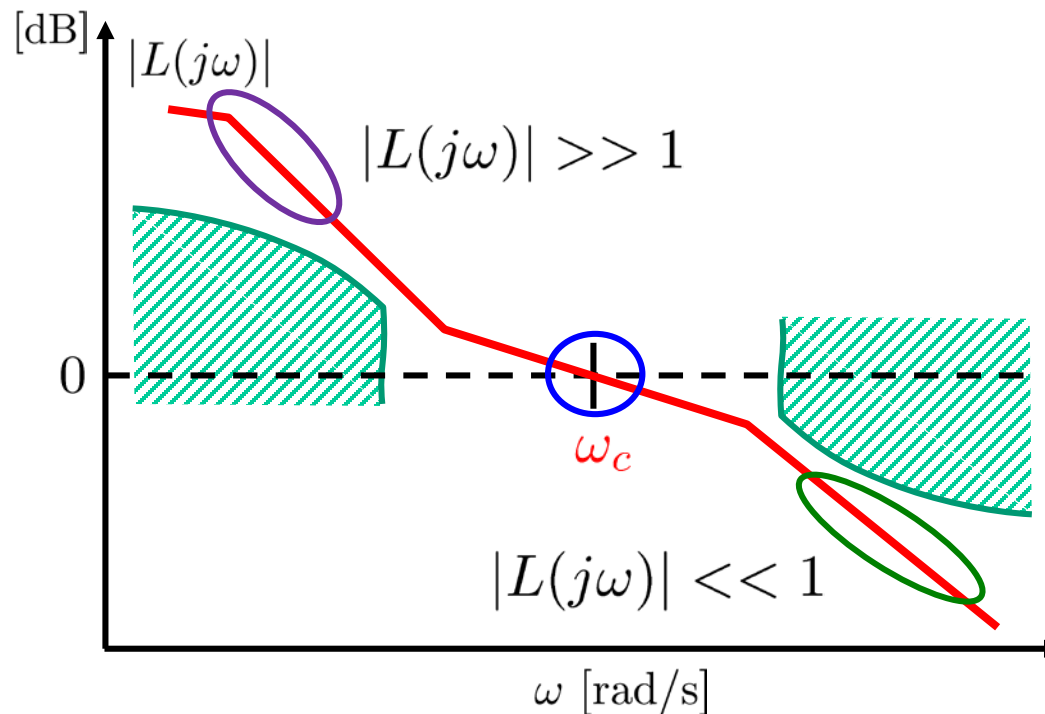
$$\rightarrow \begin{cases} \left| \frac{1}{1 - GK} \right| \leq \frac{\gamma}{|G||W|}, \\ \left| \frac{G}{1 - GK} \right| \leq \frac{\gamma}{|W|} \end{cases}$$

For High frequencies

$$\bar{\sigma}(WG) \ll 1$$

$$\rightarrow \begin{cases} \left| \frac{K}{1 - GK} \right| \leq \gamma|W|, \\ \left| \frac{GK}{1 - GK} \right| \leq \gamma|G||W| \end{cases}$$

$$W := W_2 W_1$$





# Robust Performance in the $\nu$ -Gap Metric

$$\left. \begin{aligned} G_0 &= M^{-1}N \\ G_1 &= (M + \Delta_M)^{-1}(N + \Delta_N) \\ &\| [\Delta_N \quad \Delta_M] \|_\infty < \beta \end{aligned} \right\} \delta_\nu(G_0, G_1) < \beta$$

If  $K$  stabilizes  $G_0$  with  $b(G_0, K) \geq \beta$  then  $K$  will also stabilize  $G_1$

A bound on the robust performance

$$\arcsin(b(G_1, K_1)) \geq \arcsin(b(G_0, K_0)) - \arcsin(\delta_\nu(G_1, G_0)) - \arcsin(\delta_\nu(K_1, K_0))$$

(The derivation of this is due to Vinnicombe and is non-trivial)

This inequality is a slightly stronger inequality than

$$b(G_1, K_1) \geq b(G_0, K_0) - \delta_\nu(G_1, G_0) - \delta_\nu(K_1, K_0)$$

Perturbed performance	Nominal performance	Plant perturbation	Controller perturbation
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which is also true and shows clearly how the performance can be degraded by perturbations to the plant and controller



# MATLAB command: “magshape”

(graphical user interface)

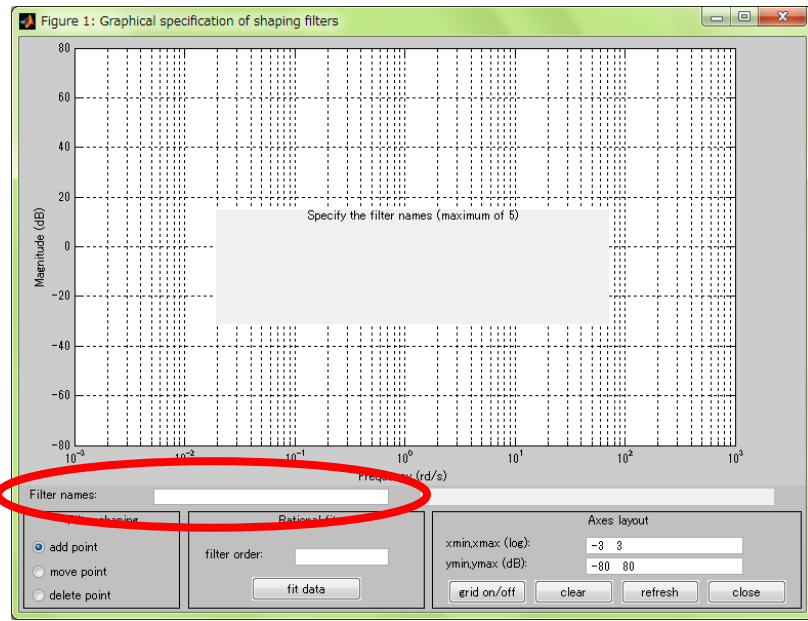


## Design Method

1. Express the design specifications in terms of loop shapes and shaping filters.
2. Specify the shaping filters by their magnitude profile. This is done interactively with the graphical user interface `magshape`.
3. Specify the control loop structure with the functions `sconnect` and `smult`, or alternatively with Simulink.
4. Solve the resulting  $H_\infty$  problem with one of the  $H_\infty$  synthesis functions.

To impose a given roll-off rate in the open-loop response, it is often desirable to use nonproper shaping filters. Meanwhile, “magshape” approximates them by high-pass filters. A drawback of this approximation is the introduction of fast parasitic modes in the filter and augmented plant realizations, which in turn may cause numerical difficulties.

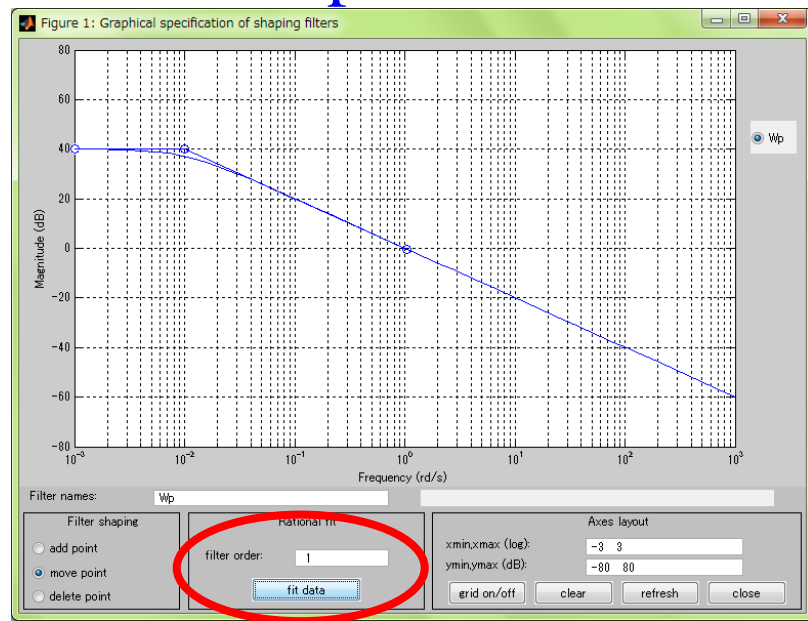
# STEP 1 Input filter name



# STEP 2 Put desired points



# STEP 3 shape a Filter



ワークスペース

名前	値	最小値	最大値
Active...	1	1	1
Filt_who	'Wp'		
HDL_filt	29.0029	29.00...	29.0029
HDL_fix	<12x1 double>	0	24.0022
Order_...	1	1	1
Wp	[-0.0100,1,1;1.0122,...	-Inf	1.0122
ans	[0,-1.4211e-14,105...	-1.42...	105.3584

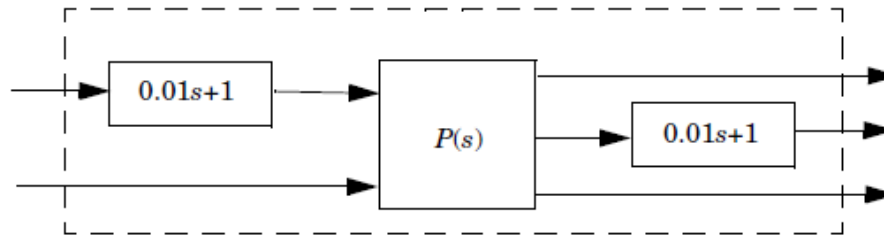
Output the filter automatically  
(SYSTEM matrix form)

# MATLAB command: “sderiv”

Alternatively, you can use “sderiv” to include nonproper shaping filters in the loop-shaping criterion. This function appends a SISO PD component to selected Input/Output of given LTI system.

**Design Method** To specify more complex nonproper filters,

1. Specify the proper “low-frequency” part of the filters with magshape.
2. Augment the plant with these low-pass filters.
3. Add the derivative action of the nonproper filters by applying sderiv to the augmented plant.



is specified by

```
Pd = sderiv(P,[ 1 2],[0.01 1])
```

In the calling list, [ 1 2] lists the input and output channels to be filtered by  $ns + d$  (here 1 for “first input” and 2 for “second output”) and the vector [0.01 1] lists the values of  $n$  and  $d$ . An error is generated if the resulting system Pd is not proper.