

6 Applications of H_∞ Control

6.1 A Simple Frequency Domain Design

Let

$$H(s) = \begin{bmatrix} H_{zw}(s) & H_{zu}(s) \\ H_{yw}(s) & H_{yu}(s) \end{bmatrix} = \left[\begin{array}{c|cc} A & B_w & B_u \\ \hline C_z & D_{zw} & D_{zu} \\ C_y & D_{yw} & 0 \end{array} \right].$$

Assume that the closed loop transfer $\mathcal{F}_l(H(s), C(s))$ is SISO.

Frequency Domain Control Problem: Find a controller $C(s)$ such that $\mathcal{F}_l(H(s), C(s))$ is asymptotically stable and

$$|\mathcal{F}_l(H(j\omega), C(j\omega))| < \phi(\omega), \quad \text{for all } \omega$$

for some given frequency domain specification $\phi(\omega)$.

Let $W(s) \in RH_\infty$ be such that $W^{-1}(s) \in RH_\infty$ and

$$|W^{-1}(j\omega)| = \phi(\omega), \quad \forall \omega$$

Then the above problem is solved if

$$|\mathcal{F}_l(H(j\omega), C(j\omega))| < |W^{-1}(j\omega)| = \phi(\omega), \quad \text{for all } \omega$$

or, equivalently,

$$|W(j\omega)| |\mathcal{F}_l(H(j\omega), C(j\omega))| = |W(j\omega) \mathcal{F}_l(H(j\omega), C(j\omega))| < 1 \quad \text{for all } \omega.$$

The function $W(s)$ is called a *scaling* or *shapping* function. This specification can be translated as the following H_∞ control problem.

H_∞ Control Problem: Given a scaling function $W(s)$ find a controller $C(s)$ such that $\mathcal{F}_l(H(s), C(s))$ is asymptotically stable and

$$\|W(s) \mathcal{F}_l(H(j\omega), C(j\omega))\| < 1.$$

Note that

$$\begin{aligned}
 W(s)\mathcal{F}_l(H(s), C(s)) &= W(s) \left\{ H_{zw}(s) + H_{zu}(s)C(s) [I - H_{yu}(s)C(s)]^{-1} H_{yw}(s) \right\} \\
 &= W(s)H_{zw}(s) + W(s)H_{zu}(s)C(s) [I - H_{yu}(s)C(s)]^{-1} H_{yw}(s) \\
 &= \mathcal{F}_l \left(\begin{bmatrix} W(s) & 0 \\ 0 & I \end{bmatrix} H(s), C(s) \right)
 \end{aligned}$$

so that the scaling is applied in the open loop plant $H(s)$.

6.2 Example #1

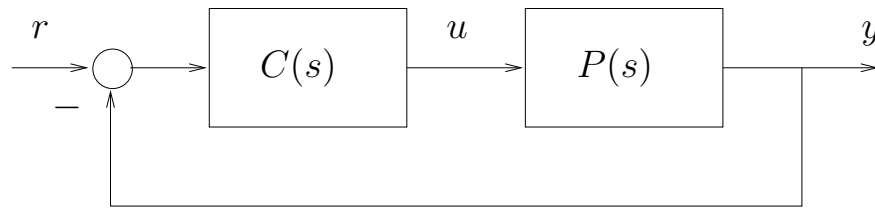


Figure 6: Block diagram

Plant:

$$P(s) = \frac{1}{(s+2)(s-1)}$$

Control objective:

Have a low pass transfer function from r to y . Say, below

$$\phi(\omega) = \frac{1}{\sqrt{(\omega/\omega_b)^2 + 1}}$$

for some $\omega_b > 0$.

Scaling design:

Set

$$W^{-1}(s) = \frac{1}{s/\omega_b + 1}, \quad W(s) = s/\omega_b + 1.$$

Problem is $W(s) \notin RH_\infty$! General solution is to introduce a pole at 'minus infinity', that is

$$W^{-1}(s) = \frac{\epsilon s + 1}{s/\omega_b + 1}, \quad W(s) = \frac{s/\omega_b + 1}{\epsilon s + 1}.$$

Then augment the plant with the scaling.

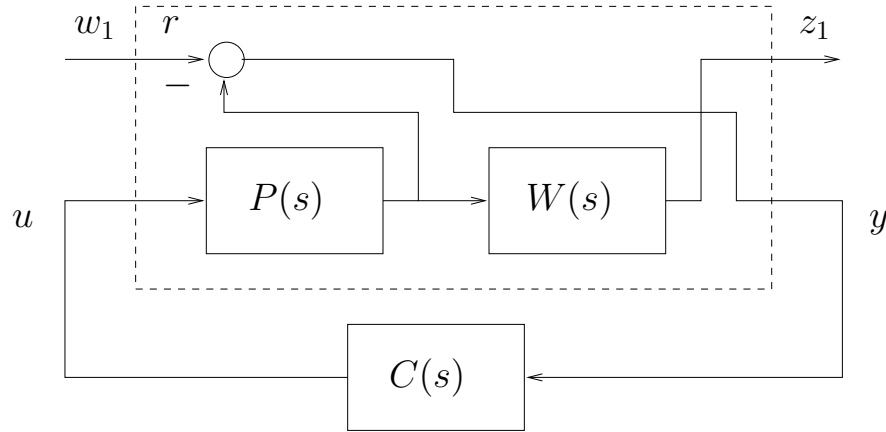


Figure 7: Plant augmented with scaling

The above system does not satisfy the assumptions on the H_∞ control problem for two reasons: $B_w = 0$ ³ and $D_{zu} = 0$ ⁴. This is so because w_1 never gets into the plant before going through the controller and $P(s)$ is strictly proper. In order to overcome that, introduce an extra input w_2 and an extra output $z_2 = \epsilon u$, with $\epsilon > 0$ small, as in the diagram:

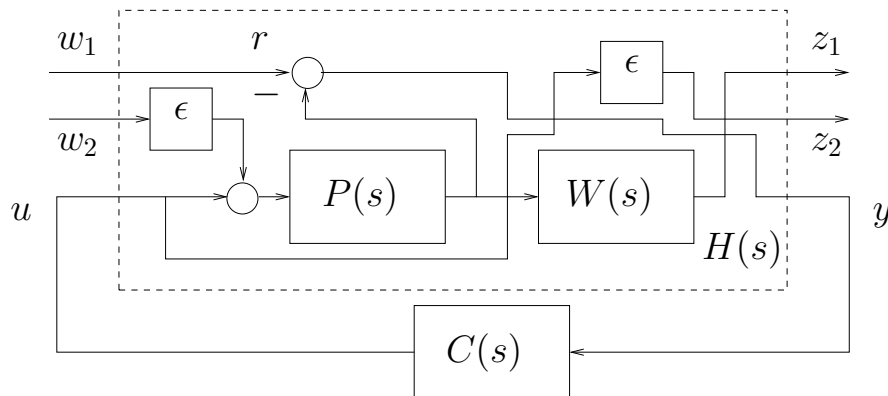


Figure 8: Plant augmented with extra scalings and inputs/outputs

³ (A, B_w) is not controllable.

⁴ $D_{zu}^T D_{zu} \neq 1$. Worse than that, D_{zu} is not full rank.

H_∞ **control problem:** Set $\omega_b = 1$ and find $C(s)$ such that

$$\|\mathcal{F}_l(H(s), C(s))\| < 1$$

State space realization:

```
% MAE 284 - Robust and Multivariable Control
% Mauricio de Oliveira
% April 2006
%
% Example #1
% Frequency domain design
% Low pass cutoff frequency
omega_b = 1;

% epsilon
epsilon = 1e-4;

% Plant:
P = tf(1, conv([1 2],[1 -1]))

Transfer function:
      1
-----
s^2 + s - 2

% Scaling:
S = tf([1/omega_b 1],[epsilon 1])

Transfer function:
      s + 1
-----
0.0001 s + 1

% Convert to state-space
[Ap, Bp, Cp, Dp] = ssdata(P);
[As, Bs, Cs, Ds] = ssdata(S);

% Generalized plant H
A = [As Bs*Cp; zeros(size(Cp'*Bs')) Ap];
Bw1 = zeros(3,1);
Bu = [Bs*Dp; Bp];

Cy = -[zeros(1,1) Cp];
Dyw1 = 1;

Cz1 = [Cs Ds*Cp];
Dzu1 = [Ds*Dp];

% Extra input w2
```

```

Bw2 = epsilon*Bu;
Dyw2 = 0;

% Extra output penalizing control input
Cz2 = zeros(1,3);
Dzu2 = epsilon;

% Augmented plant
Cz = [Cz1; Cz2];
Dzu = [Dzu1; Dzu2];

Bw = [Bw1 epsilon*Bu];
Dyw = [Dyw1 Dyw2];

% Show matrices
A
A =
    1.0e+04 *
    -1.0000         0    0.4096
         0    -0.0001    0.0001
         0    0.0004         0

Bu
Bu =
         0
    0.5000
         0

Bw
Bw =
    1.0e-04 *
         0         0
         0    0.5000
         0         0

Cy
Cy =
         0         0   -0.5000

Dyw
Dyw =
     1         0

Cz
Cz =
    1.0e+04 *
    -1.2206         0    0.5000
         0         0         0

Dzu
Dzu =
    1.0e-04 *
         0
    1.0000

diary off

```

Assumptions:

```
% Checking assumptions
% Stabilizability and detectability
rank(ctrb(A,Bu))
ans =
    3
rank(observ(A,Cy)) % not observable but detectable
ans =
    2

% Normality
Dzu' * Dzu
ans =
    1.0000e-08
Dyw * Dyw'
ans =
    1

% Enforcing Dyw * Dyw' = 1
Bu = Bu / epsilon
Bu =
     0
    5000
     0
Dzu = Dzu / epsilon
Dzu =
     0
     1

Dzu' * Dzu
ans =
    1
Dyw * Dyw'
ans =
    1

% Auxiliary matrices
Ac = A - Bu * Dzu' * Cz;
Cc = (eye(size(Dzu,1)) - Dzu*Dzu') * Cz;

Af = A - Bw * Dyw' * Cy;
Bf = Bw * (eye(size(Dyw,2)) - Dyw'*Dyw);

% Technical observability and controllability
rank(observ(Ac,Cc))
ans =
    3
rank(ctrb(Af,Bf))
ans =
    3

diary off
```

Riccati equations:

```
% Set gamma
gamma = 1;

% Solve control ARE
X = are(Ac, Bu*Bu' - Bw*Bw'/gamma^2, Cc'*Cc);

% Solve filtering ARE
Y = are(Af', Cy'*Cy - Cz'*Cz/gamma^2, Bf*Bf');

% Positivity
mX = min(eig(X))
mX =
    2.5140e-09
mY = min(eig(Y))
mY =
   -3.2815

% Spectral condition
mXY = max(abs(eig(X*Y)))/gamma^2
mXY =
    4.6203e-04

% GAMMA = 1 IS NOT FEASIBLE
diary off
```

Conclusion: $\|\mathcal{F}_l(H(s), C(s))\| \not\leq 1$. Raise γ and try again!

```

% Set gamma
gamma = 2.01;

% Solve control ARE
X = are(Ac, Bu*Bu' - Bw*Bw'/gamma^2, Cc'*Cc);

% Solve filtering ARE
Y = are(Af', Cy'*Cy - Cz'*Cz/gamma^2, Bf*Bf');

% Positivity
mX = min(eig(X))
mX =
    2.5180e-09
mY = min(eig(Y))
mY =
   -5.0831e-14

% Spectral condition
mXY = max(abs(eig(X*Y)))/gamma^2
mXY =
    0.0339

if ( mX > -epsilon^2 && mY > -epsilon^2 && mXY < 1 )

    % Constructing central controller
    Zoo = inv(eye(size(A)) - Y*X/gamma^2);
    Loo = -Y*Cy' - Bw*Dyw';
    Koo = -Bu'*X - Dzu'*Cz;
    Aoo = A + Bw*Bw'*X/gamma^2 + Bu*Koo + Zoo*Loo*(Cy + Dyw*Bw'*X/gamma^2);

    % Build controller (do not forget to scale Cc!)
    C = -ss(Aoo, -Zoo*Loo, Koo/epsilon, zeros(size(Koo*Loo))); zpk(C)

Zero/pole/gain:
    -4090077.8764 (s+2) (s+1e04)
-----
(s+205.5) (s^2 + 1.732e04s + 1e08)

% Build closed loop system
Acl = [Ap, Bp*C.c; C.b*Cp C.a];
Bcl = [zeros(2,1); C.b*1];
Ccl = [Cp, zeros(1,3)];
Dcl = zeros(size(Ccl*Bcl));
CL = ss(Acl,Bcl,Ccl,Dcl);

% Check closed loop stability
max(real(eig(Acl)))
ans =
   -1.0000

diary off

```

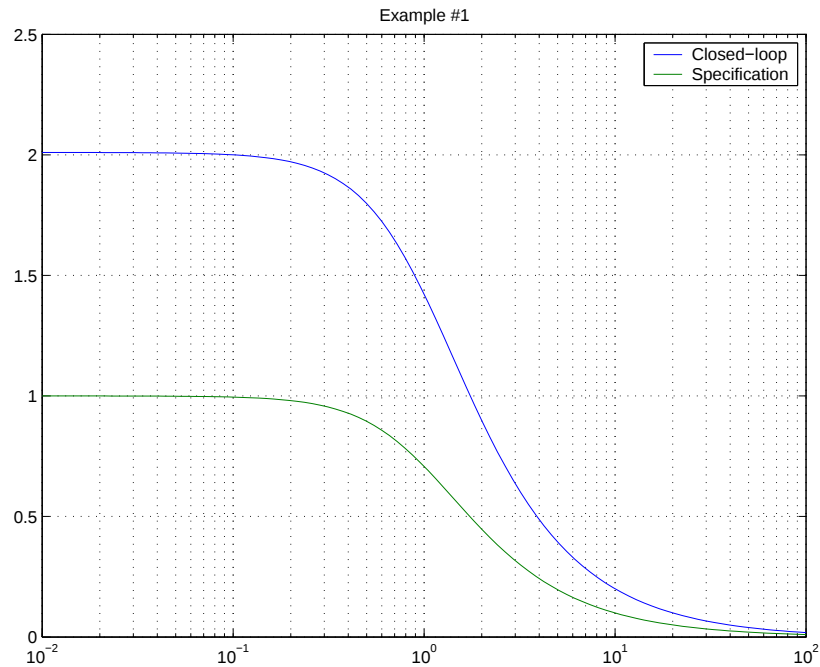


Figure 9: Example #1

6.3 Example #2

Same data as Example #1. Less stringent constraint.

H_∞ control problem: Set $\omega_b = 20$ and find $C(s)$ such that

$$\min\{\gamma : \|\mathcal{F}_l(H(s), C(s))\| < \gamma\} = 1.06$$

Solution:

```
% MAE 284 - Robust and Multivariable Control
% Mauricio de Oliveira
% April 2006
%
% Example #2
% Frequency domain design
% Low pass cutoff frequency
omega_b = 20;

% epsilon
epsilon = 1e-4;

% Plant:
P = tf(1, conv([1 2],[1 -1]))
```

```

Transfer function:
      1
-----
s^2 + s - 2

% Scaling:
S = tf([1/omega_b 1],[epsilon 1])

Transfer function:
      0.05 s + 1
-----
      0.0001 s + 1

% Convert to state-space
[Ap, Bp, Cp, Dp] = ssdata(P);
[As, Bs, Cs, Ds] = ssdata(S);

% Generalized plant H
A = [As Bs*Cp; zeros(size(Cp'*Bs')) Ap];
Bw1 = zeros(3,1);
Bu = [Bs*Dp; Bp];

Cy = -[zeros(1,1) Cp];
Dyw1 = 1;

Cz1 = [Cs Ds*Cp];
Dzu1 = [Ds*Dp];

% Extra input w2
Bw2 = epsilon*Bp;
Dyw2 = 0;

% Extra output penalizing control input
Cz2 = zeros(1,3);
Dzu2 = epsilon;

% Augmented plant
Cz = [Cz1; Cz2];
Dzu = [Dzu1; Dzu2];

Bw = [Bw1 epsilon*Bp];
Dyw = [Dyw1 Dyw2];

% Show matrices
A
A =
      1.0e+04 *
      -1.0000           0      0.1024
           0      -0.0001      0.0001
           0      0.0004           0

Bu
Bu =

```

```

    0
    0.5000
    0
Bw
Bw =
    1.0e-04 *
        0      0
        0      0.5000
        0      0
Cy
Cy =
        0      0      -0.5000
Dyw
Dyw =
    1      0
Cz
Cz =
    1.0e+03 *
    -2.4365      0      0.2500
        0      0      0
Dzu
Dzu =
    1.0e-04 *
        0
    1.0000

% Checking assumptions
% Stabilizability and detectability
rank(ctrb(A,Bu))
ans =
    3
rank(observ(A,Cy)) % not observable but detectable
ans =
    2

% Normality
Dzu' * Dzu
ans =
    1.0000e-08
Dyw * Dyw'
ans =
    1

% Enforcing Dyw * Dyw' = 1
Bu = Bu / epsilon
Bu =
        0
    5000
        0
Dzu = Dzu / epsilon
Dzu =
    0
    1

```

```

Dzu' * Dzu
ans =
    1
Dyw * Dyw'
ans =
    1

% Auxiliary matrices
Ac = A - Bu * Dzu' * Cz;
Cc = (eye(size(Dzu,1)) - Dzu*Dzu') * Cz;

Af = A - Bw * Dyw' * Cy;
Bf = Bw * (eye(size(Dyw,2)) - Dyw'*Dyw);

% Technical observability and controllability
rank(observ(Ac,Cc))
ans =
    3
rank(ctrb(Af,Bf))
ans =
    3

% Set gamma
gamma = 1.06;

% Solve control ARE
X = are(Ac, Bu*Bu' - Bw*Bw'/gamma^2, Cc'*Cc);

% Solve filtering ARE
Y = are(Af', Cy'*Cy - Cz'*Cz/gamma^2, Bf*Bf');

% Positivity
mX = min(eig(X))
mX =
    3.3507e-07
mY = min(eig(Y))
mY =
   -1.3188e-15

% Spectral condition
mXY = max(abs(eig(X*Y)))/gamma^2
mXY =
    0.2134

if ( mX > -epsilon^2 && mY > -epsilon^2 && mXY < 1 )

    % Constructing central controller
    Zoo = inv(eye(size(A)) - Y*X/gamma^2);
    Loo = -Y*Cy' - Bw*Dyw';
    Koo = -Bu'*X - Dzu'*Cz;
    Aoo = A + Bw*Bw'*X/gamma^2 + Bu*Koo + Zoo*Loo*(Cy + Dyw*Bw'*X/gamma^2);

    % Build controller (do not forget to scale Cc!)
    C = -ss(Aoo, -Zoo*Loo, Koo/epsilon, zeros(size(Koo*Loo))); zpk(C)

```

```

Zero/pole/gain:
-1409708.2925 (s+2) (s+1e04)
-----
(s+9987) (s+491.9) (s+162.3)

% Build closed loop system
Acl = [Ap, Bp*C.c; C.b*Cp C.a];
Bcl = [zeros(2,1); C.b*1];
Ccl = [Cp, zeros(1,3)];
Dcl = zeros(size(Ccl*Bcl));
CL = ss(Acl,Bcl,Ccl,Dcl);

% Check closed loop stability
max(real(eig(Acl)))
ans =
    -2.0000

diary off

```

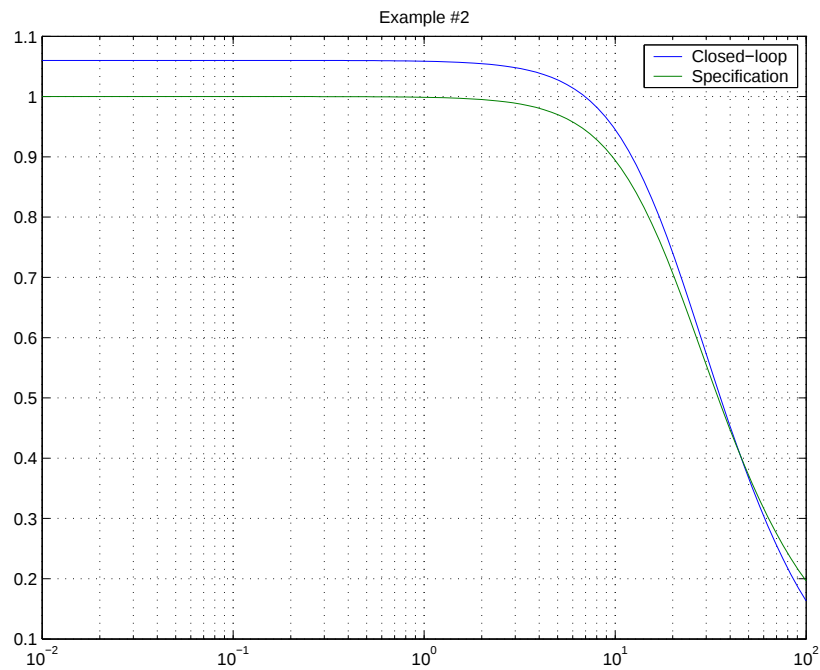


Figure 10: Example #2

6.4 Example #3

Same plant and problem data as Example #1.

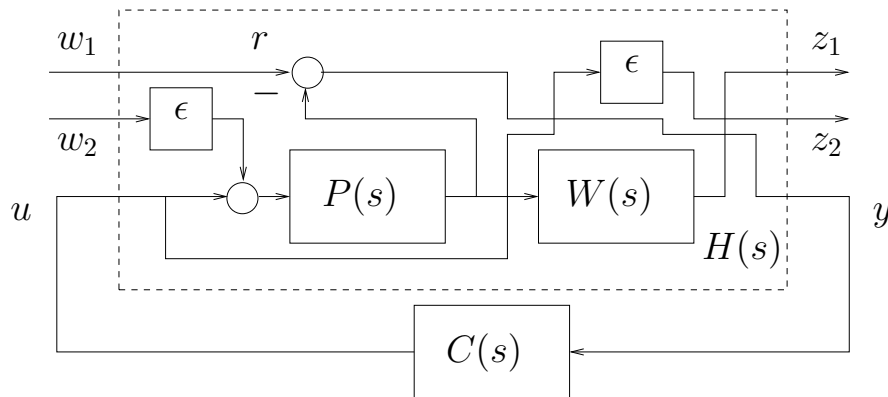


Figure 11: Plant augmented with extra scalings and inputs/outputs

Scaling design:

Set

$$W^{-1}(s) = \frac{1}{s/\omega_b + 1}, \quad W(s) = s/\omega_b + 1.$$

$W(s) \notin RH_\infty$ but $P(s)$ is strictly proper! That means $W(s)P(s)$ is realizable!

State space realization:

Let

$$P(s) = \left[\begin{array}{c|c} A_p & B_p \\ \hline C_p & 0 \end{array} \right].$$

Then

$$W(s)P(s) = \left[\begin{array}{c|c} A_p & B_p \\ \hline C_p(A_p/\omega_b + I) & C_p B_p/\omega_b \end{array} \right] = \left[\begin{array}{c|c} A_p & B_p \\ \hline C_p(A_p/\omega_b + I) & 0 \end{array} \right]$$

Note that $C_p B_p = 0$ because the original system has relative degree 2!

ADVANTAGE: smaller plant, smaller controller!

```

% MAE 284 - Robust and Multivariable Control
% Mauricio de Oliveira
% April 2006
%
% Example #3
% Frequency domain design
% Low pass cutoff frequency
omega_b = 1;

% epsilon
epsilon = 1e-4;

% Plant:
P = tf(1, conv([1 2],[1 -1]))

Transfer function:
      1
-----
s^2 + s - 2

% Scaling:
S = tf([1/omega_b 1],1)

Transfer function:
s + 1

% Convert to state-space
[Ap, Bp, Cp, Dp] = ssdata(P);

% Generalized plant H
A = Ap;
Bw1 = zeros(2,1);
Bu = Bp;

Cy = -Cp;
Dyw1 = 1;

Cz1 = Cp*(Ap/omega_b + eye(2));
Dzu1 = Cp*Bp;
Dzu1 =
    0

% Extra input w2
Bw2 = epsilon*Bp;
Dyw2 = 0;

% Extra output penalizing control input
Cz2 = zeros(1,2);
Dzu2 = epsilon;

% Augmented plant

```

```

Cz = [Cz1; Cz2];
Dzu = [Dzu1; Dzu2];

Bw = [Bw1 epsilon*Bu];
Dyw = [Dyw1 Dyw2];

% Show matrices
A
A =
    -1.0000    0.5000
     4.0000         0

Bu
Bu =
    0.5000
         0

Bw
Bw =
    1.0e-04 *
         0    0.5000
         0         0

Cy
Cy =
         0   -0.5000

Dyw
Dyw =
     1         0

Cz
Cz =
     2.0000    0.5000
         0         0

Dzu
Dzu =
    1.0e-04 *
         0
     1.0000

diary off

```

H_∞ **control problem:** Set $\omega_b = 1$ and find $C(s)$ such that

$$\min\{\gamma : \quad \|\mathcal{F}_l(H(s), C(s))\| < \gamma\} = 2.01$$

Assumptions:

```
% Checking assumptions
% Stabilizability and detectability
rank(ctrb(A,Bu))
ans =
     2
rank(observ(A,Cy)) % This time it is observable!
ans =
     2

% Normality
Dzu' * Dzu
ans =
 1.0000e-08
Dyw * Dyw'
ans =
     1

% Enforcing Dyw * Dyw' = 1
Bu = Bu / epsilon;
Dzu = Dzu / epsilon;

Dzu' * Dzu
ans =
     1
Dyw * Dyw'
ans =
     1

% Auxiliary matrices
Ac = A - Bu * Dzu' * Cz;
Cc = (eye(size(Dzu,1)) - Dzu*Dzu') * Cz;

Af = A - Bw * Dyw' * Cy;
Bf = Bw * (eye(size(Dyw,2)) - Dyw'*Dyw);

% Technical observability and controllability
rank(observ(Ac,Cc))
ans =
     2
rank(ctrb(Af,Bf))
ans =
     2

diary off
```

Solution:

```
% Set gamma
gamma = 2.01;

% Solve control ARE
X = are(Ac, Bu*Bu' - Bw*Bw'/gamma^2, Cc'*Cc);

% Solve filtering ARE
Y = are(Af', Cy'*Cy - Cz'*Cz/gamma^2, Bf'*Bf');

% Positivity
mX = min(eig(X))
mX =
    4.7048e-09
mY = min(eig(Y))
mY =
    5.8822e-10

% Spectral condition
mXY = max(abs(eig(X*Y)))/gamma^2
mXY =
    0.0200

if ( mX > -epsilon^2 && mY > -epsilon^2 && mXY < 1 )

    % Constructing central controller
    Zoo = inv(eye(size(A)) - Y*X/gamma^2);
    Loo = -Y*Cy' - Bw*Dyw';
    Koo = -Bu'*X - Dzu'*Cz;
    Aoo = A + Bw*Bw'*X/gamma^2 + Bu*Koo + Zoo*Loo*(Cy + Dyw*Bw'*X/gamma^2);

    % Build controller (do not forget to scale Cc!)
    C = -ss(Aoo, -Zoo*Loo, Koo/epsilon, zeros(size(Koo*Loo))); zpk(C)

Zero/pole/gain:
-4112493.4645 (s+2)
-----
(s+1e04) (s+206.6)

% Build closed loop system
Acl = [Ap, Bp*C.c; C.b*Cp C.a];
Bcl = [zeros(2,1); C.b*1];
Ccl = [Cp, zeros(1,2)];
Dcl = zeros(size(Ccl*Bcl));
CL = ss(Acl,Bcl,Ccl,Dcl);

% Check closed loop stability
max(real(eig(Acl)))
ans =
    -1.0000

diary off
```

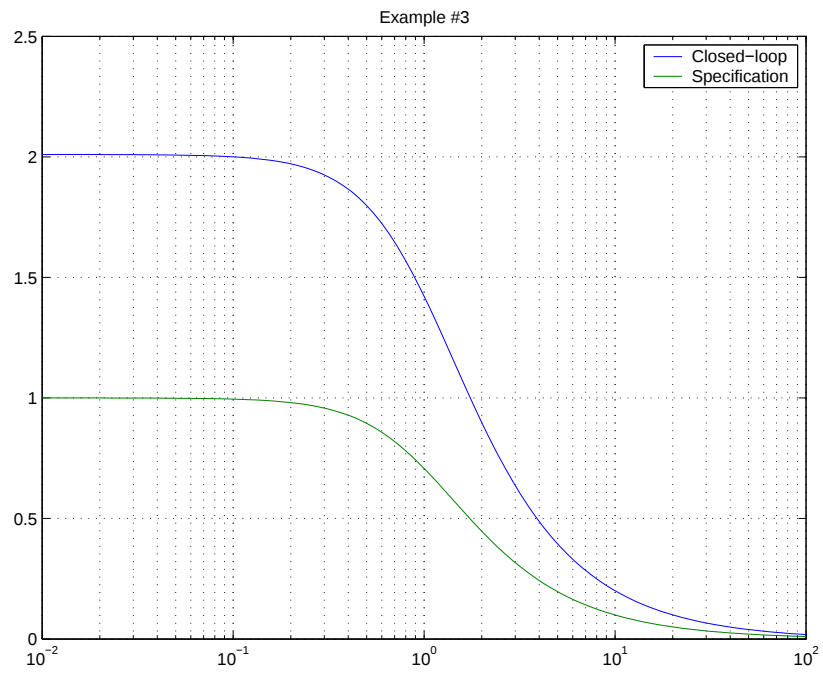


Figure 12: Example #3

6.5 Robust Analysis

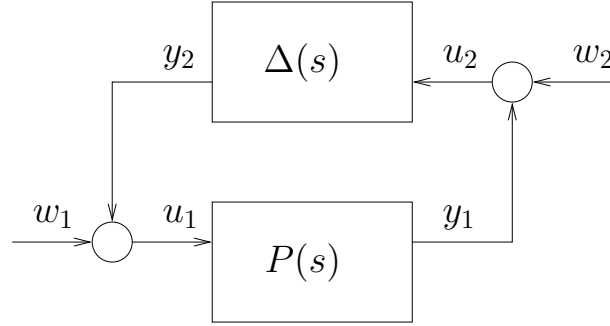


Figure 13: Feedback connection

Theorem [Small Gain]: Let $P(s) \in RH_\infty$. Then the feedback connection in Figure 13 is well posed and internally stable for all $\Delta(s) \in RH_\infty$ such that $\|\Delta(s)\| < 1$ if and only if $\|P(s)\| \leq 1$.

In the proof we will make use of the following technical lemma.

Lemma: $\underline{\sigma}(A + B) \geq \underline{\sigma}(A) - \bar{\sigma}(B)$.

Proof: First note that

$$\frac{\|Ax\| - \|Bx\|}{\|x\|} \geq \frac{\|Ax\|}{\|x\|} - \max_{\|x\| \neq 0} \frac{\|Bx\|}{\|x\|} = \frac{\|Ax\|}{\|x\|} - \bar{\sigma}(B).$$

Consequently

$$\min_{\|x\| \neq 0} \frac{\|Ax\| - \|Bx\|}{\|x\|} \geq \min_{\|x\| \neq 0} \frac{\|Ax\|}{\|x\|} - \bar{\sigma}(B) = \underline{\sigma}(A) - \bar{\sigma}(B)$$

and

$$\begin{aligned} \underline{\sigma}(A + B) &= \min_{\|x\| \neq 0} \frac{\|Ax + Bx\|}{\|x\|} \\ &\geq \min_{\|x\| \neq 0} \frac{\|Ax\| - \|Bx\|}{\|x\|} \\ &\geq \underline{\sigma}(A) - \bar{\sigma}(B) \end{aligned}$$

Proof of Small Gain Theorem: We prove only sufficiency. Since $P(s) \in RH_\infty$, the feedback connection is asymptotically stable if and only if

$$\Delta(s)[I - P(s)\Delta(s)]^{-1} \in RH_\infty.$$

But as $\Delta(s) \in RH_\infty$ the above condition is equivalent to

$$[I - P(s)\Delta(s)]^{-1} \in RH_\infty,$$

or in other words, $\det[I - P(s)\Delta(s)]$ has no poles in the closed right-half plane. That is

$$\inf_{\operatorname{Re}(s) \geq 0} \underline{\sigma}[I - P(s)\Delta(s)] > 0$$

Using the previous lemma

$$\begin{aligned} \inf_{\operatorname{Re}(s) \geq 0} \underline{\sigma}[I - P(s)\Delta(s)] &\geq \inf_{\operatorname{Re}(s) \geq 0} \{\underline{\sigma}(I) - \bar{\sigma}[P(s)\Delta(s)]\} \\ &= 1 + \inf_{\operatorname{Re}(s) \geq 0} \{-\bar{\sigma}[P(s)\Delta(s)]\} \\ &= 1 - \sup_{\operatorname{Re}(s) \geq 0} \bar{\sigma}[P(s)\Delta(s)] \\ &= 1 - \|P(s)\Delta(s)\| \end{aligned}$$

Now if $\|\Delta(s)\| < 1$

$$\begin{aligned} \inf_{\operatorname{Re}(s) \geq 0} \underline{\sigma}[I - P(s)\Delta(s)] &\geq 1 - \|P(s)\Delta(s)\| \\ &\geq 1 - \|P(s)\| \|\Delta(s)\| \\ &> 1 - \|P(s)\| \geq 0 \end{aligned}$$

A similar argument can show that the connection is well-posed. For a proof of necessity see Zhou p. 138.

6.5.1 Unstructured Additive Uncertainty

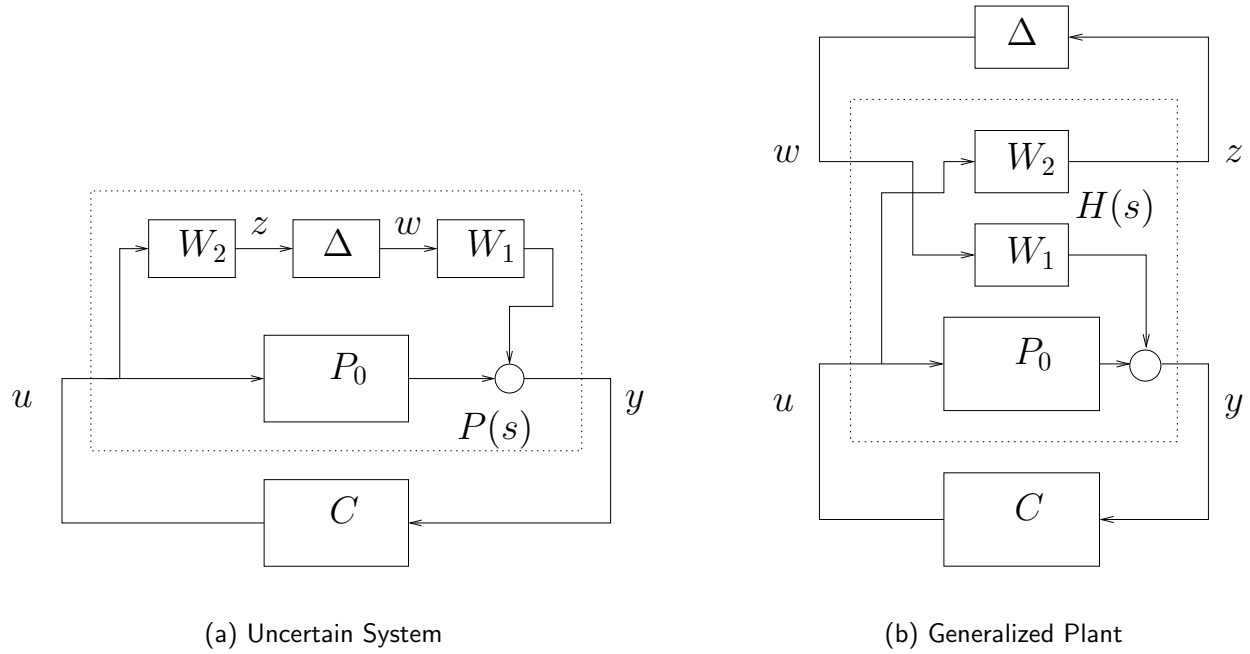


Figure 14: Additive Uncertainty

Theorem: The feedback connection in Figure 14(a) where

$$P(s) = P_0(s) + W_1(s)\Delta(s)W_2(s)$$

and the uncertainty $\Delta(s) \in RH_\infty$ is such that $\|\Delta(s)\| < 1$ is robustly stable if and only if

$$\|W_2(s)C(s)[I - P_0(s)C(s)]^{-1}W_1(s)\| \leq 1.$$

Proof: Define the input w and output z as in Fig.14(b), where the generalized plant $H(s)$ is given by

$$H(s) = \begin{bmatrix} H_{zw}(s) & H_{zu}(s) \\ H_{yw}(s) & H_{yu}(s) \end{bmatrix} = \begin{bmatrix} 0 & W_2(s) \\ W_1(s) & P_0(s) \end{bmatrix}$$

Use the small gain theorem to conclude that the system is robustly stable if and only if

$$\|\mathcal{F}_l(H(s), C(s))\| \leq 1$$

where

$$\mathcal{F}_l(H(s), C(s)) = W_2(s)C(s)[I - P_0(s)C(s)]^{-1}W_1(s).$$

6.5.2 Unstructured Multiplicative Uncertainty

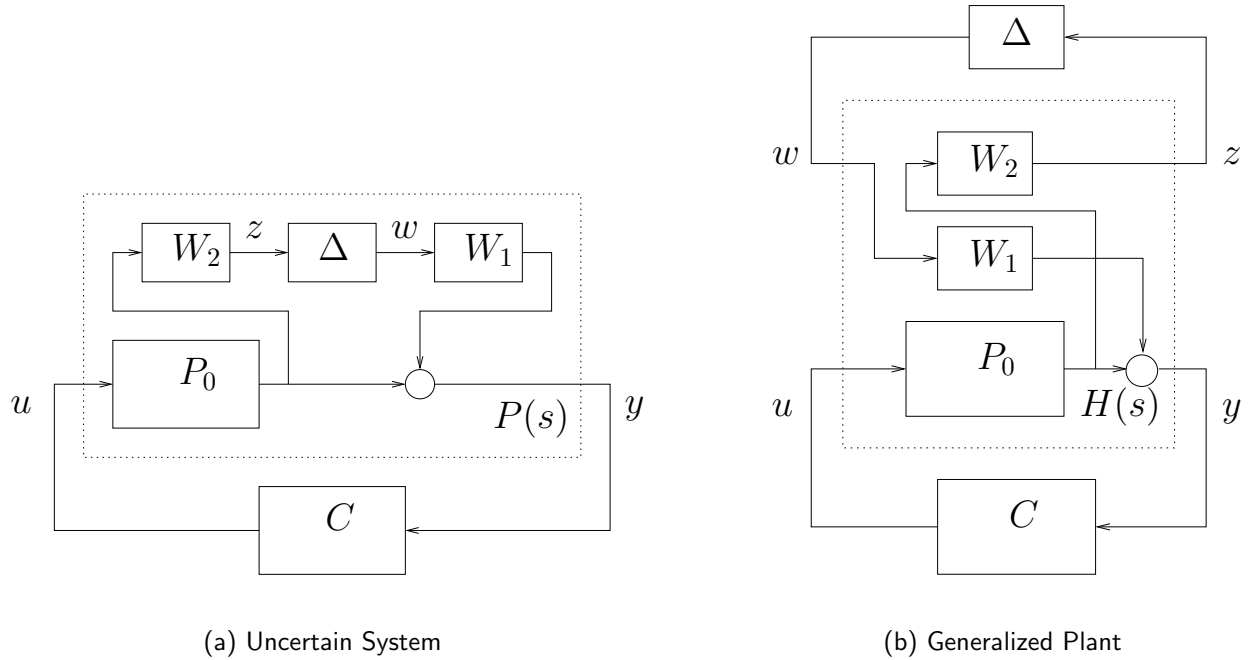


Figure 15: Multiplicative Uncertainty

Theorem: The feedback connection in Figure 15(a) where

$$P(s) = [I + W_1(s)\Delta(s)W_2(s)]P_0(s)$$

and the uncertainty $\Delta(s) \in RH_\infty$ is such that $\|\Delta(s)\| < 1$ is robustly stable if and only if

$$\|W_2(s)P_0(s)C(s)[I - P_0(s)C(s)]^{-1}W_1(s)\| \leq 1.$$

Proof: Define the input w and output z as in Fig.15(b), where the generalized plant $H(s)$ is given by

$$H(s) = \begin{bmatrix} H_{zw}(s) & H_{zu}(s) \\ H_{yw}(s) & H_{yu}(s) \end{bmatrix} = \begin{bmatrix} 0 & W_2(s)P_0(s) \\ W_1(s) & P_0(s) \end{bmatrix}$$

Use the small gain theorem to conclude that the system is robustly stable if and only if

$$\|\mathcal{F}_l(H(s), C(s))\| \leq 1$$

where

$$\mathcal{F}_l(H(s), C(s)) = W_2(s)P_0(s)C(s)[I - P_0(s)C(s)]^{-1}W_1(s).$$

6.5.3 Unstructured Fractional Uncertainty

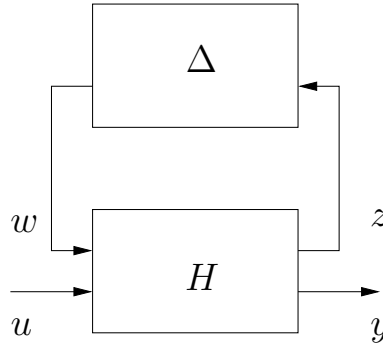


Figure 16: Uncertainty System

Lemma: The uncertain system

$$P(s) = [I + W_1(s)\Delta(s)W_2(s)]^{-1}P_0(s)$$

can be represented as the feedback connection of the plant

$$H(s) = \begin{bmatrix} H_{zw}(s) & H_{zu}(s) \\ H_{yw}(s) & H_{yu}(s) \end{bmatrix} = \begin{bmatrix} -W_2(s)W_1(s) & W_2(s)P_0(s) \\ -W_1(s) & P_0(s) \end{bmatrix}$$

as in Figure 16.

Proof: Use the identity

$$(A + BCD)^{-1} = A^{-1} - A^{-1}B(C^{-1} + DA^{-1}B)^{-1}DA^{-1}$$

to rewrite

$$\begin{aligned} [I + W_1(s)\Delta(s)W_2(s)]^{-1}P_0(s) \\ = P_0(s) - W_1(s)\Delta(s)[I + W_2(s)W_1(s)\Delta(s)]^{-1}W_2(s)P_0(s). \end{aligned}$$

Equivalently

$$\begin{aligned} \mathcal{F}_l \left(\begin{bmatrix} H_{yu}(s) & H_{yw}(s) \\ H_{zu}(s) & H_{zw}(s) \end{bmatrix}, \Delta(s) \right) \\ = \underbrace{P_0(s)}_{H_{yu}(s)} + \underbrace{[-W_1(s)]}_{H_{yw}(s)} \Delta(s) \{ I - \underbrace{[-W_2(s)W_1(s)]}_{H_{zw}(s)} \Delta(s) \}^{-1} \underbrace{W_2(s)P_0(s)}_{H_{zu}(s)} \end{aligned}$$

from which the blocks of $H(s)$ are defined.

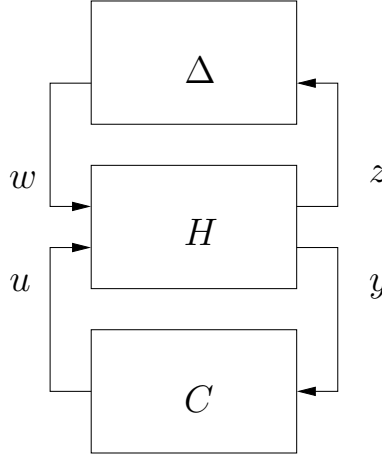


Figure 17: Uncertainty System

Theorem: The feedback connection in Figure 17 where

$$P(s) = [I + W_1(s)\Delta(s)W_2(s)]^{-1}P_0(s)$$

and the uncertainty $\Delta(s) \in RH_\infty$ is such that $\|\Delta(s)\| < 1$ is robustly stable if and only if

$$\|W_2(s)[I - P_0(s)C(s)]^{-1}W_1(s)\| \leq 1.$$

Proof: From the previous lemma and the small gain theorem the system is robustly stable if and only if

$$\|\mathcal{F}_l(H(s), C(s))\| \leq 1$$

where

$$\begin{aligned} \mathcal{F}_l(H(s), C(s)) &= -W_2(s)W_1(s) - W_2(s)P_0(s)C(s)[I - P_0(s)C(s)]^{-1}W_1(s) \\ &= -W_2(s) \{I + P_0(s)C(s)[I - P_0(s)C(s)]^{-1}\} W_1(s) \\ &= -W_2(s)[I - P_0(s)C(s)]^{-1}W_1(s), \end{aligned}$$

after using the same identity used in the proof of the previous lemma.

6.5.4 Structured Uncertainty

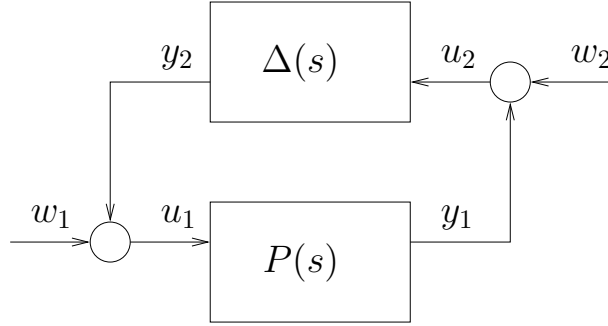


Figure 18: Uncertain System

Structured Uncertainty Robust Analysis: Is the feedback system in Figure 18 robustly stable for all $\Delta(s) \in RH_\infty$ such that $\|\Delta(s)\| < 1$ in the presence of some structural constraint $\Delta(s) \in \Delta_S(s)$?

Without loss of generality

$$\Delta_S(s) := \{\Delta(s) : \Delta(s) = \text{diag}[\delta_1(s)I, \dots, \delta_s(s)I, \Delta_1(s), \dots, \Delta_F(s)] \in RH_\infty\}$$

or

$$\Delta_S := \{\Delta : \Delta = \text{diag}[\delta_1 I, \dots, \delta_s I, \Delta_1, \dots, \Delta_F], \quad \delta_i \in \mathbb{C}, \quad \Delta_j \in \mathbb{C}^{m_j \times m_j}\}.$$

6.5.5 μ Analysis

In this section we focus on uncertainty in Δ_S . Results are presented without proof. See Zhou, Chapter 10.

Definition: For $M \in \mathbb{C}^{n \times n}$, $\Delta \in \Delta_S$, the structured singular value of M is the scalar

$$\mu_{\Delta_S}(M) := \frac{1}{\min\{\|\Delta\| : \Delta \in \Delta_S, \det(I - M\Delta) = 0\}}$$

If $\det(I - M\Delta) \neq 0$ for all $\Delta \in \Delta_S$ then $\mu_{\Delta_S}(M) := 0$.

Lemma: $\max_i |\lambda_i(M)| \leq \mu_{\Delta_S}(M) \leq \bar{\sigma}(M)$.

Scalings:

$$\mathbf{D}_S := \{\Delta : \Delta = \text{diag}[D_1, \dots, D_s, d_1 I, \dots, d_F I], \\ D_i \in \mathbb{C}^{r_i \times r_i}, D_i = D_i^* > 0, \quad d_j \in \mathbb{R}, d_j > 0\}.$$

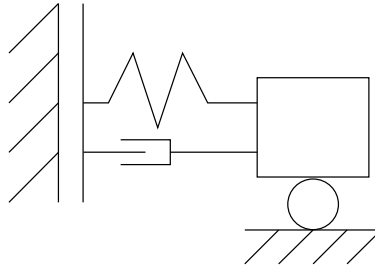
Lemma: $\max_{i, UU^*=I} |\lambda_i(UM)| = \mu_{\Delta_S}(M) \leq \inf_{D \in \mathbf{D}_S} \bar{\sigma}(DM D^{-1}).$

Structured Small Gain Theorem: Let $P(s) \in RH_\infty$. Then the feedback connection in Figure 18 is well posed and internally stable for all $\Delta \in \Delta_S$ such that $\|\Delta\| < 1$ if and only if

$$\sup_{\omega \in \mathbb{R}} \mu_{\Delta_S}[P(j\omega)] \leq 1.$$

6.6 Example #4 [Parametric Uncertainty]

Mass/Spring/Damper system from Zhou, p. 171.



Uncertain System:

$$m\ddot{x} + c\dot{x} + kx = f$$

where $m > 0$, $c > 0$, $k > 0$.

Uncertainty:

$$c = \bar{c}(1 + \delta_c)$$

$$k = \bar{k}(1 + \delta_k)$$

where

$$|c - \bar{c}| < \delta_c \bar{c}$$

$$|k - \bar{k}| < \delta_k \bar{k}$$

and $|\delta_c| < \bar{\delta}_c \leq 1$, $|\delta_k| < \bar{\delta}_k \leq 1$.

Pulling out uncertainty:

Write system equation

$$\begin{aligned} \ddot{x} &= -m^{-1}c\dot{x} - m^{-1}kx + m^{-1}f \\ &= -m^{-1}\bar{c}(1 + \delta_c)\dot{x} - m^{-1}\bar{k}(1 + \delta_k)x + m^{-1}f \\ &= -m^{-1}\bar{c}\dot{x} - m^{-1}\bar{k}x - m^{-1}\delta_c\bar{c}\dot{x} - m^{-1}\delta_k\bar{k}x + m^{-1}f \end{aligned}$$

Define auxiliary outputs

$$z_c = \bar{c}\dot{x},$$

$$z_k = \bar{k}x$$

System with auxiliary outputs

$$\begin{aligned}\ddot{x} &= -m^{-1}\bar{c}\dot{x} - m^{-1}\bar{k}x - m^{-1}\delta_c z_c - m^{-1}\delta_k z_k + m^{-1}f \\ z_c &= \bar{c}\dot{x} \\ z_k &= \bar{k}x\end{aligned}$$

Define auxiliary inputs in feedback

$$\begin{aligned}w_c &= \delta_c z_c \\ w_k &= \delta_k z_k\end{aligned}$$

System with auxiliary inputs and outputs

$$\begin{aligned}\ddot{x} &= -m^{-1}\bar{c}\dot{x} - m^{-1}\bar{k}x - m^{-1}w_c - m^{-1}w_k + m^{-1}f \\ z_c &= \bar{c}\dot{x} \\ z_k &= \bar{k}x\end{aligned}$$

Uncertain System in State Space: ($x_2 = \dot{x}$, $x_1 = x$)

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ z_c \\ z_k \end{pmatrix} = \left[\begin{array}{cc|ccc} 0 & 1 & 0 & 0 & 0 \\ -m^{-1}\bar{k} & -m^{-1}\bar{c} & -m^{-1} & -m^{-1} & m^{-1} \\ \hline 0 & \bar{c} & 0 & 0 & 0 \\ \bar{k} & 0 & 0 & 0 & 0 \end{array} \right] \begin{pmatrix} x_1 \\ x_2 \\ w_c \\ w_k \\ f \end{pmatrix}$$
$$\begin{pmatrix} w_c \\ w_k \end{pmatrix} = \begin{bmatrix} \delta_c & 0 \\ 0 & \delta_k \end{bmatrix} \begin{pmatrix} z_c \\ z_k \end{pmatrix}$$

6.7 Example #5 [Parametric Uncertainty]

Mass/Spring/Damper system from Zhou, p. 171.

Uncertain System:

$$m\ddot{x} + c\dot{x} + kx = f$$

where $m > 0$, $c > 0$, $k > 0$.

Uncertainty:

$$m = \bar{m}(1 + \delta_m) \quad c = \bar{c}(1 + \delta_c) \quad k = \bar{k}(1 + \delta_k)$$

where

$$|m - \bar{m}| < \delta_m \bar{m} \quad |c - \bar{c}| < \delta_c \bar{c} \quad |k - \bar{k}| < \delta_k \bar{k}$$

and $|\delta_m| < \bar{\delta}_m \leq 1$, $|\delta_c| < \bar{\delta}_c \leq 1$, $|\delta_k| < \bar{\delta}_k \leq 1$.

Auxiliary inputs/outputs:

Define auxiliary outputs

$$z_c = \bar{c}\dot{x},$$
$$z_k = \bar{k}x$$

Define auxiliary inputs in feedback

$$w_c = \delta_c z_c$$
$$w_k = \delta_k z_k$$

System with auxiliary inputs and outputs

$$\ddot{x} = m^{-1} (-\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f)$$
$$z_c = \bar{c}\dot{x}$$
$$z_k = \bar{k}x$$

The fractional uncertainty:

Use the lemma of the inverse

$$\begin{aligned}
 m^{-1} &= \bar{m}^{-1}(1 + \delta_m)^{-1} \\
 &= \bar{m}^{-1}[1 - \delta_m(1 + \delta_m)^{-1}] \\
 &= \bar{m}^{-1} \mathcal{F}_l \left(\begin{bmatrix} 1 & -1 \\ 1 & -1 \end{bmatrix}, \delta_m \right)
 \end{aligned}$$

Since for any u

$$y = \mathcal{F}_l(M, \delta) u$$

is the feedback connection of

$$\begin{pmatrix} y \\ z \end{pmatrix} = M \begin{pmatrix} u \\ w \end{pmatrix}$$

with

$$w = \delta z$$

we have

$$\begin{aligned}
 \begin{pmatrix} m^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f) \\ z_m \end{pmatrix} &= \begin{bmatrix} \bar{m}^{-1} & -\bar{m}^{-1} \\ 1 & -1 \end{bmatrix} \begin{pmatrix} -\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f \\ w_m \end{pmatrix} \\
 &= \begin{pmatrix} \bar{m}^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k - w_m + f) \\ -\bar{c}\dot{x} - \bar{k}x - w_c - w_k - w_m + f \end{pmatrix}
 \end{aligned}$$

System with auxiliary inputs and outputs

$$\begin{aligned}
 \ddot{x} &= \bar{m}^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k - w_m + f) \\
 z_c &= \bar{c}\dot{x} \\
 z_k &= \bar{k}x \\
 z_m &= -\bar{c}\dot{x} - \bar{k}x - w_c - w_k - w_m + f
 \end{aligned}$$

Uncertain System in State Space: ($x_2 = \dot{x}$, $x_1 = x$)

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ z_c \\ z_k \\ z_m \end{pmatrix} = \left[\begin{array}{cc|cccc} 0 & 1 & 0 & 0 & 0 & 0 \\ -\bar{m}^{-1}\bar{k} & -\bar{m}^{-1}\bar{c} & -\bar{m}^{-1} & -\bar{m}^{-1} & -\bar{m}^{-1} & \bar{m}^{-1} \\ \hline 0 & \bar{c} & 0 & 0 & 0 & 0 \\ \bar{k} & 0 & 0 & 0 & 0 & 0 \\ -\bar{k} & -\bar{c} & -1 & -1 & -1 & 1 \end{array} \right] \begin{pmatrix} x_1 \\ x_2 \\ w_c \\ w_k \\ w_m \\ f \end{pmatrix}$$

$$\begin{pmatrix} w_c \\ w_k \\ w_m \end{pmatrix} = \begin{bmatrix} \delta_c & 0 & 0 \\ 0 & \delta_k & 0 \\ 0 & 0 & \delta_m \end{bmatrix} \begin{pmatrix} z_c \\ z_k \\ z_m \end{pmatrix}$$

6.8 Example #6 [Parametric Uncertainty]

Mass/Spring/Damper system from Zhou, p. 171.

Uncertain System:

$$\ddot{x} + m^{-1}c\dot{x} + m^{-1}kx = m^{-1}f$$

where $m > 0$, $c > 0$, $k > 0$.

Uncertainty:

$$m^{-1} = \bar{m}^{-1}(1 + \delta_m) \quad c = \bar{c}(1 + \delta_c) \quad k = \bar{k}(1 + \delta_k)$$

where

$$|m^{-1} - \bar{m}^{-1}| < \delta_m \bar{m}^{-1} \quad |c - \bar{c}| < \delta_c \bar{c} \quad |k - \bar{k}| < \delta_k \bar{k}$$

and $|\delta_m| < \bar{\delta}_m \leq 1$, $|\delta_c| < \bar{\delta}_c \leq 1$, $|\delta_k| < \bar{\delta}_k \leq 1$.

Auxiliary inputs/outputs:

Define auxiliary outputs

$$z_c = \bar{c}\dot{x},$$
$$z_k = \bar{k}x$$

Define auxiliary inputs in feedback

$$w_c = \delta_c z_c$$
$$w_k = \delta_k z_k$$

System with auxiliary inputs and outputs

$$\ddot{x} = m^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f)$$
$$z_c = \bar{c}\dot{x}$$
$$z_k = \bar{k}x$$

The multiplicative uncertainty:

Since

$$m^{-1} = \bar{m}^{-1}(1 + \delta_m)$$

we have

$$m^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f) = \bar{m}^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k + w_m + f)$$

where

$$\begin{aligned} w_m &= \delta_m z_m \\ z_m &= -\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f \end{aligned}$$

System with auxiliary inputs and outputs

$$\begin{aligned} \ddot{x} &= \bar{m}^{-1}(-\bar{c}\dot{x} - \bar{k}x - w_c - w_k + w_m + f) \\ z_c &= \bar{c}\dot{x} \\ z_k &= \bar{k}x \\ z_m &= -\bar{c}\dot{x} - \bar{k}x - w_c - w_k + f \end{aligned}$$

Uncertain System in State Space: ($x_2 = \dot{x}$, $x_1 = x$)

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \\ z_c \\ z_k \\ z_m \end{pmatrix} = \left[\begin{array}{cc|cccc} 0 & 1 & 0 & 0 & 0 & 0 \\ -\bar{m}^{-1}\bar{k} & -\bar{m}^{-1}\bar{c} & -\bar{m}^{-1} & -\bar{m}^{-1} & \bar{m}^{-1} & \bar{m}^{-1} \\ \hline 0 & \bar{c} & 0 & 0 & 0 & 0 \\ \bar{k} & 0 & 0 & 0 & 0 & 0 \\ -\bar{k} & -\bar{c} & -1 & -1 & 0 & 1 \end{array} \right] \begin{pmatrix} x_1 \\ x_2 \\ w_c \\ w_k \\ w_m \\ f \end{pmatrix}$$
$$\begin{pmatrix} w_c \\ w_k \\ w_m \end{pmatrix} = \begin{bmatrix} \delta_c & 0 & 0 \\ 0 & \delta_k & 0 \\ 0 & 0 & \delta_m \end{bmatrix} \begin{pmatrix} z_c \\ z_k \\ z_m \end{pmatrix}$$