

Massachusetts Institute of Technology

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**6.245: MULTIVARIABLE CONTROL SYSTEMS**

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## Problem Set 8 Solutions <sup>1</sup>

The problem set deals with H-Infinity optimization and Hankel Optimal model reduction.

### Problem 8.1T

FOR ALL VALUES OF PARAMETER  $a \in \mathbb{R}$  FOR WHICH THE CT H-INFINITY OPTIMIZATION SETUP

$$\dot{x}(t) = u(t) + w_1(t), \quad e(t) = ax(t) + u(t), \quad y(t) = x(t) + w_2(t)$$

IS WELL-POSED, FIND THE OPTIMAL CONTROLLER AND THE MINIMAL COST.

**Answer:** the setup is not singular if and only if  $a \neq 0$ . The minimal H-Infinity norm  $\gamma_0$  of the closed loop transfer function is  $a$  for  $a > 0$ , and  $\sqrt{r_0}$  for  $a \leq 0$ , where  $r_0$  is the (unique) real root of the polynomial

$$\phi(r) = (r - 1)^2(r - a^2) - 4a^2r$$

satisfying  $r_0 \geq 1$ . The optimal H-Infinity controller has the form  $u(t) = -\gamma_0 y(t)$ , even when  $a = 0$ .

**Reasoning:** for the square  $J$  of the closed loop H-Infinity norm to be smaller than a given number  $r > 0$ , there must exist real  $n$ -by- $n$  matrices  $H = H' > 0$  and  $L$  (where  $n > 1$  is the order of the closed loop system) such that the quadratic form

$$\sigma(x, w, v, z, u, q) = rw^2 + rv^2 - (ax + u)^2 - 2 \begin{bmatrix} x \\ q \end{bmatrix}' H \begin{bmatrix} u + w \\ z \end{bmatrix}$$

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is positive definite on the vector space

$$V_L = \left\{ (x, w, v, z, u, q) \in \mathbb{R} \times \mathbb{R} \times \mathbb{R} \times \mathbb{R}^{n-1} \times \mathbb{R} \times \mathbb{R}^{n-1} : \begin{bmatrix} q \\ u \end{bmatrix} = L \begin{bmatrix} z \\ x + w \end{bmatrix} \right\}.$$

According to the generalized Parrott's lemma, such  $L$  exists for a given  $H = H' > 0$  if and only if

$$\sigma(x, w, -x, 0, 0, 0) > 0 \quad \forall (x, w) \neq 0 \quad (1)$$

(the "zero information" condition), and

$$\sup_{u, q} \sigma(x, w, v, z, u, q) > 0 \quad \forall (x, w, v, z) \neq 0. \quad (2)$$

Let  $P$  and  $Q$  denote the upper left elements of  $H$  and  $H^{-1}$  respectively. Then (1) is equivalent to

$$r(r - a^2) > P^2, \quad (3)$$

and (2) is equivalent to

$$2aQ > \frac{1 - r}{r}. \quad (4)$$

Taking into account that positive numbers  $P, Q$  can be upper left elements of  $H, H^{-1}$  for some  $H = H' > 0$  if and only if  $PQ \geq 1$ , we conclude that  $J$  can be made less than  $r$  if and only if either  $a > 0$  and  $r > a^2$ , or  $a \leq 0$ ,  $r > 1$ , and  $\phi(r) > 0$ . This proves that  $\gamma_0$  is not smaller than the value indicated in the answer.

Given positive numbers  $P, Q$  such that  $PQ > 1$ , one matrix  $H = H' > 0$  such that  $P, Q$  are the upper left elements of  $H, H^{-1}$  is given by

$$H = \begin{bmatrix} P & Q^{-1} - P \\ Q^{-1} - P & P - Q^{-1} \end{bmatrix}.$$

To show that optimality level  $\gamma_0 = a$  can be achieved for  $a > 0$ , consider the matrix  $H$  which corresponds to the "extremal" values of  $r, P, Q^{-1}$ :

$$r = a^2, \quad P = 0, \quad Q = +\infty, \quad Q^{-1} = 0, \quad H = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}.$$

The corresponding

$$\sigma(x, w, v, z, u, q) = a^2 w^2 + a^2 v^2 - (ax + u)^2$$

is made positive semidefinite by

$$u = -ay = -a(x + w),$$

which proves the optimality of the feedback  $u = -ay$  for  $a > 0$ .

To show that optimality level  $\gamma_0 = \sqrt{r_0}$  can be achieved for  $a < 0$ , consider the matrix  $H$  which corresponds to the "extremal" values of  $r, P, Q^{-1}$ :

$$r = r_0, \quad P = \sqrt{r_0(r_0 - a^2)} = \frac{-2ar_0}{r_0 - 1}, \quad Q = \frac{r_0 - 1}{-2ar_0}, \quad H = \begin{bmatrix} -2ar_0/(r_0 - 1) & 0 \\ 0 & 0 \end{bmatrix}.$$

The corresponding

$$\sigma(x, w, v, z, u, q) = r_0 w^2 + r_0 v^2 - (ax + u)^2 + \frac{4ar_0(u + w)}{r_0 - 1}$$

is made positive semidefinite by

$$u = -\sqrt{r_0}y = -\sqrt{r_0}(x + w),$$

which proves optimality of the feedback  $u = -\sqrt{r_0}y$  for  $a < 0$ .

The analytical solution can be verified by the following code:

```
function ps81(a)
s=tf('s');
p=ss(0,[1 0 1],[a;1],[0 0 1;0 1 0]);
[K1,~,GAM1]=hinfsyn(p,1,1); % numerical solution
if a>0,
    K=-a;
    GAM=a;
else
    r=roots([1, -2-a^2, 1-2*a^2, -a^2]);
    r=max(r(r==real(r)));
    P=sqrt(r*(r-a^2));
    K=-r/(P+a);
    GAM=sqrt(r);
end
CL=lft(p,K);
GAM2=norm(CL,Inf);
K0=squeeze(freqresp(K1,[0]));
fprintf('-K:      analytical %f, numerical %f\n',-K,-K0)
fprintf('GAMMA: analytical %f, numerical %f, achieved %f\n', ...
    GAM,GAM1,GAM2)
```

### Problem 8.2T

FOR ALL VALUES OF PARAMETER  $a \in \mathbb{R}$  FIND A HANKEL OPTIMAL REDUCED MODEL OF ORDER 1 FOR

$$G(z) = \frac{1 - a^2}{z} + \frac{a}{z^2}.$$

**Answer:** the reduced model  $\hat{G}(z)$  is given by

$$\hat{G}(z) = \begin{cases} \frac{1-a^2}{z-a}, & |a| < 1, \\ \frac{1-a^2}{z+1/a}, & |a| > 1, \\ 0, & |a| = 1. \end{cases}$$

**Reasoning:** in general, in order to find an Hankel norm-optimal reduced model of order  $d$  for a stable controllable and observable state space model

$$x(t+1) = Ax(t) + Bw(t), \quad y(t) = Cx(t) \quad (5)$$

with  $x(t) \in \mathbb{R}^n$ ,  $w(t) \in \mathbb{R}^m$ ,  $y(t) \in \mathbb{R}^k$ , we seek  $(N+k)$ -by- $(N+m)$  real matrix  $L$  and an  $(n+N)$ -by- $(n+N)$  real symmetric matrix  $H = H'$  with not more than  $n+d$  positive eigenvalues, such that the quadratic form

$$\sigma(x, w, v, u, q) = r|w|^2 - |Cx - u|^2 - \begin{bmatrix} Ax + Bw \\ q \end{bmatrix}' H \begin{bmatrix} Ax + Bw \\ q \end{bmatrix} - \begin{bmatrix} x \\ v \end{bmatrix}' H \begin{bmatrix} x \\ v \end{bmatrix}$$

is positive semidefinite on the subspace defined by

$$\begin{bmatrix} q \\ u \end{bmatrix} = L \begin{bmatrix} v \\ w \end{bmatrix}. \quad (6)$$

Once such  $H, L$  are found, with  $r$  being as small as possible, the optimal reduced model is the *stable* part of the system with input  $w$  and output  $u$  defined by the state space equations

$$\begin{bmatrix} v(t+1) \\ u(t) \end{bmatrix} = L \begin{bmatrix} v(t) \\ w(t) \end{bmatrix}.$$

The theory suggests using

$$H = \begin{bmatrix} W_o & -\theta \\ -\theta & \theta \end{bmatrix}, \quad \theta = W_o - rW_c^{-1}, \quad (7)$$

where  $W_o, W_c$  are the observability and controllability Gramians of (5), i.e.

$$W_o = A'W_oA + C'C, \quad W_c = AW_cA' + BB',$$

and  $r$  is chosen to control the number of positive eigenvalues of  $H$ , which in this case equals  $n + d$ , where  $d$  is the number of positive eigenvalues of  $\theta$ . For this particular  $H$ , we have

$$\sigma(x, w, v, u, q) = 2x'(A'\theta q + C'u - \theta v - A'W_o Bw) + r|w|^2 - |u|^2 + 2w'B'\theta q - q'\theta q + v'\theta v, \quad (8)$$

which implies that the best pair  $(\tilde{q}, u) = (\theta q, u)$  is uniquely defined as a linear function of  $(\tilde{v}, w) = (\theta v, w)$  by

$$(\tilde{q}, u) = \tilde{L}(\tilde{v}, u) = \arg \min_{\tilde{q}, u: A'\tilde{q} + C'u = \theta v + A'W_o Bw} \{r|w|^2 - |u|^2 + 2w'B'\tilde{q} - \tilde{q}'\theta^+ \tilde{q} + \tilde{v}'\theta^+ \tilde{v}\}, \quad (9)$$

where  $\theta^+$  is a pseudo-inverse of  $\theta$ , i.e. a matrix such that  $\theta = \theta\theta^+\theta$ . This in turn determines the reduced system  $\hat{G}$  as the stable part of the state space model

$$(\tilde{v}(t+1), u(t)) = \tilde{L}(\tilde{v}(t), w(t)),$$

where  $\tilde{v}(t)$  belongs to the range of  $\theta$  for all  $t$ .

In this problem we have  $n = 2$ ,  $m = k = d = 1$ , and a state space model for  $G$  is given by

$$A = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad C = [1 - a^2 \quad a].$$

The corresponding Gramians are

$$W_o = \begin{bmatrix} 1 - a^2 + a^4 & a(1 - a^2) \\ a(1 - a^2) & a^2 \end{bmatrix}, \quad W_c = I, \quad A'W_o B = \begin{bmatrix} a(1 - a^2) \\ 0 \end{bmatrix}.$$

The Hankel singular numbers of system  $G$  are square roots of the eigenvalues of  $W_o = W_o W_c$ , i.e. 1 and  $a^2$ . When  $|a| > 1$ , the largest possible value of  $r$  for which matrix  $H$  in (7) has a single positive eigenvalue is  $r = 1$ , which corresponds to

$$\theta = (a^2 - 1) \begin{bmatrix} -a \\ 1 \end{bmatrix} \begin{bmatrix} -a \\ 1 \end{bmatrix}', \quad A'\theta = (a^2 - 1) \begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} -a \\ 1 \end{bmatrix}'.$$

Representing the elements  $\tilde{v}$  in the range of  $\theta$  by

$$\tilde{v} = \frac{1}{1 + a^2} \begin{bmatrix} -a \\ 1 \end{bmatrix} \tilde{x}, \quad \tilde{x} \in \mathbb{R}$$

in

$$A'\tilde{v}(t+1) + C'u(t) = \theta\tilde{v}(t+1) + A'W_o Bw(t) \quad (10)$$

yields

$$\tilde{x}(t+1) = -\frac{1}{a}\tilde{x}(t) - aw(t), \quad u(t) = \frac{a^2 - 1}{a}\tilde{x}(t),$$

which means that the optimal reduced model has transfer function

$$\hat{G}(z) = \frac{1 - a^2}{z + 1/a}.$$

When  $|a| < 1$ , the largest possible value of  $r$  for which matrix  $H$  in (7) has a single positive eigenvalue is  $r = a^2$ , which corresponds to

$$\theta = (1 - a^2) \begin{bmatrix} 1 \\ a \end{bmatrix} \begin{bmatrix} 1 \\ a \end{bmatrix}', \quad A'\theta = (1 - a^2) \begin{bmatrix} a \\ 0 \end{bmatrix} \begin{bmatrix} 1 \\ a \end{bmatrix}'.$$

Representing the elements  $\tilde{v}$  in the range of  $\theta$  by

$$\tilde{v} = \frac{1}{1 + a^2} \begin{bmatrix} 1 \\ a \end{bmatrix} \tilde{x}, \quad \tilde{x} \in \mathbb{R}$$

in (10) yields

$$\tilde{x}(t+1) = a\tilde{x}(t) + w(t), \quad u(t) = (1 - a^2)\tilde{x}(t),$$

which means that the optimal reduced model has transfer function

$$\hat{G}(z) = \frac{1 - a^2}{z - a}.$$

The analytical solution can be checked by MATLAB code

```
function ps82(a)
z=tf('z');
G=(1-a^2+a/z)/z;
sig=hsvd(G);
if a>1,
    Gr=(1-a^2)/(z+1/a);
    GAM=1;
elseif a<1,
    Gr=(1-a^2)/(z-a);
    GAM=a^2;
else
    Gr=0;
```

```

    GAM=1;
end
Gr1=tf(hankelmr(G,1))
er=max(hsvd(Gr-Gr1));
fprintf('GAMMA:    analytical %f,    numerical %f\n',GAM,sig(2))
fprintf('MISMATCH: %e\n',er)
fprintf('HinfErr:  analytical %f,    numerical %f\n',norm(G-Gr,Inf),norm(G-Gr1,Inf))

```

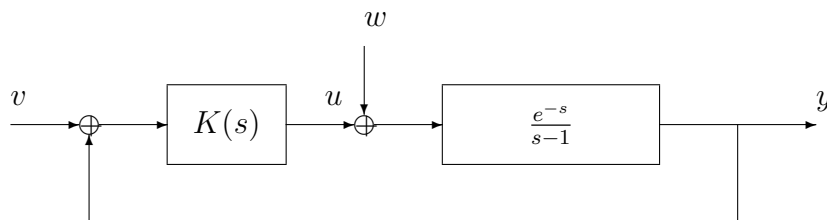


Figure 1: Problem 8.3 setup

### Problem 8.3P

CONSIDER THE TASK OF FINDING THE MINIMAL CONTROL EFFORT REQUIRED TO STABILIZE THE CT SYSTEM WITH TRANSFER FUNCTION

$$P(s) = \frac{e^{-s}}{s-1}$$

USING A FIXED ORDER CONTROLLER  $K = K(s)$ , AS SHOWN ON FIGURE 1. MORE SPECIFICALLY, YOU ARE ASKED TO USE HANKEL OPTIMAL MODEL REDUCTION AND H-INFINITY OPTIMIZATION TO DEVELOP AN ALGORITHM FOR FINDING CONTROLLERS  $K = K(s)$  OF A GIVEN ORDER WHICH STABILIZE THE SYSTEM FROM FIGURE 1 (SO THAT THE CLOSED LOOP L2 GAIN FROM  $[v; w]$  TO  $[u, y]$  FINITE), AND MAKE THE CLOSED LOOP L2 GAIN FROM  $w$  TO  $u$  AS SMALL AS POSSIBLE.

YOU ARE NOT EXPECTED TO LOOK FOR THE TRUE OPTIMAL SOLUTION (THERE ARE NO EFFICIENT METHODS FOR DOING THIS), BUT RATHER TO EXPLORE WAYS IN WHICH A GOOD LOW ORDER APPROXIMATION OF AN  $H_\infty$  CLASS TRANSFER FUNCTION CONTAINING  $e^{-s}$ , SUCH AS

$$G(s) = \frac{e^{-s}}{s+1}, \quad \text{OR} \quad G(s) = \frac{e^{-s} - e^{-1}}{s-1},$$

CAN BE USED TO SET UP SOME H-INFINITY OPTIMIZATION TO AID IN DESIGNING A GOOD CONTROLLER.

PRODUCE A TABLE OF VALUES OF BEST  $w \rightarrow u$  L2 GAINS ACHIEVED BY YOUR ALGORITHM WITH CONTROLLERS OF ORDER  $m$  FOR  $m \in \{1, 2, 4, 8, 16\}$  (IT'S OK IF SOME OF THE NUMBERS IN THE TABLE EQUAL  $\infty$ ).

**Conclusion:** for  $m \in \{1, 2, 4, 8, 16\}$ , robust stabilization is possible for  $m \in \{4, 8, 16\}$ , with L2 gain bounds of (respectively) 11.7, 5.0, 4.2. This does not mean that better L2 gains are not achievable.

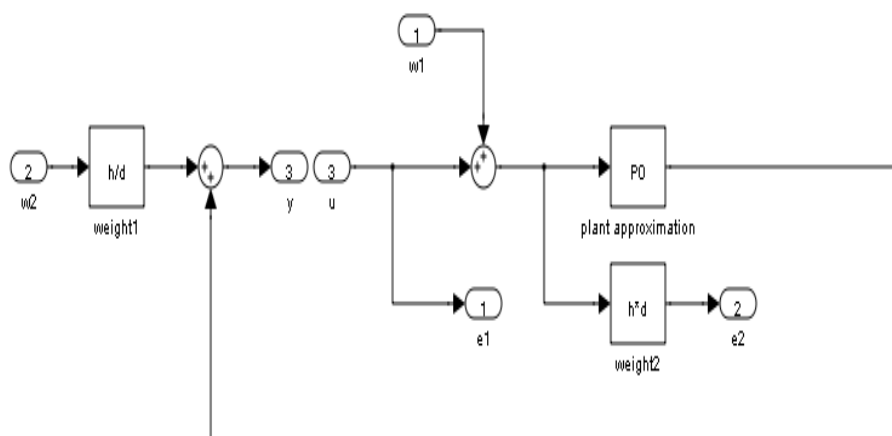


Figure 2: Open loop model for Problem 8.3

**Code:** uses design block diagram ps83des.mdl shown on Figure 2. The essential part of ps83.m is shown below:

```
function ps83(g,m,n,dd)
% function ps83(g,m,n,dd)
%
% g - desired closed loop L2 gain bound
% m - desired controller order
% n - order of the initial approximation for (exp(-s)-exp(-1))/(s-1)
% dd - vector of scaling factors to use
if nargin<1, g=100; end
if nargin<2, m=2; end
if nargin<3, n=100; end
if nargin<4, dd=logspace(-1,1,10)'; end
```



```

dd=dd(:);
nd=length(dd);
s=tf('s');
[A,B,C,D]=ssdata(ss((1-s*(0.5/n))/(1+s*(0.5/n)))^n);
sys=pck(A,B,C/(A-eye(n)),0);
[sys,sig]=sysbal(sys);
[Ar,Br,Cr,Dr]=unpck(hankmr(sys,sig,m-1,'d'));
L0=ss(Ar,Br,Cr,Dr); % approximation for (exp(-s)-exp(-1))/(s-1)
%L0=hankelmr(ss(A,B,C/(A-eye(n)),0),m);
P0=exp(-1)/(s-1)+L0;
ww=logspace(-2,2,10000);
sss=1i*ww(:);
mm=abs(squeeze(freqresp(L0,sss))-(exp(-sss)-exp(-1))./(sss-1));
r=max(mm);
%close(gcf);semilogx(ww,mm,[ww(1) ww(end)],[r r]);grid;pause
fprintf('approximation error: %f\n',r)

h=sqrt(g*r);
assignin('base','P0',P0)
assignin('base','h',h)
gg=zeros(nd,1);
for i=1:length(dd),
    d=dd(i);
    assignin('base','d',d) % export variables
    p=linmod('ps83des'); % extract open loop
    p=ss(p.a,p.b,p.c,p.d);
    [K,~,GAM]=hinfsyn(p,1,1); % optimize controller
    gg(i)=GAM;
    fprintf('.')
end
Af=ssdata(K);
fprintf('order: %d\n',size(Af,1))
close(gcf); plot(dd,gg/g); grid
return

assignin('base','K',K) % export controller
p=linmod('ps54test'); % extract closed loop
S=ss(p.a,p.b,p.c,p.d);

```

```
fprintf('T=%f]  stability: %f<0,  small gain: %f<%f\n', ...  
        T,max(real(eig(p.a))),norm(S,Inf),1/r)  
close(gcf);plot(ww,mm,[ww(1),ww(end)],[r r]);grid
```