



COLLEGE OF ENGINEERING  
SCHOOL OF AERONAUTICAL AND ASTRONAUTICAL ENGINEERING

AAE 666: NONLINEAR DYNAMICS, SYSTEMS, AND CONTROL

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## HW7

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## Exercise 1

Prove the following result: Suppose there exists a positive-definite symmetric matrix  $P$  and a positive scalar  $\alpha$  which satisfy

$$\begin{bmatrix} PA + A^T P + C^T C + 2\alpha P & PB \\ B^T P & -\gamma^{-2} I \end{bmatrix} \leq 0.$$

Then the system (11.18)-(11.19) is globally asymptotically stable about the origin with rate of convergence  $\alpha$ .

Where (11.18)

$$\dot{x} = Ax + B\phi(Cx)$$

and (11.19)

$$\|\phi(z)\| \leq \gamma \|z\|.$$

### Solution:

From the Schur complement result we can rewrite the given matrix as

$$\begin{aligned} -\gamma^{-2} I &< 0 \\ PA + A^T P + C^T C + 2\alpha P - PB(-\gamma^{-2} I)^{-1} B^T P &< 0 \end{aligned}$$

and the second equation can be organized to be

$$\begin{aligned} PA + A^T P + C^T C + 2\alpha P + \gamma^2 P B B^T P &< 0 \\ PA + A^T P + C^T C + \gamma^2 P B B^T P &< -2\alpha P < 0 \end{aligned}$$

and from **Theorem 18** we can say that if

$$PA + A^T P + C^T C + \gamma^2 P B B^T P < 0$$

is satisfied the system (11.18)-(11.19) is globally asymptotically stable about the origin with Lyapunov matrix  $P$ . And now if we let  $Q = \alpha P > 0$ , we can see that

$$\begin{aligned} \lambda_{\min}(P^{-1}Q) &= \lambda_{\min}(P^{-1}\alpha P) \\ &= \alpha. \end{aligned}$$

Hence, the rate of convergence is  $\alpha$ .

q.e.d

## Exercise 2

Recall the double inverted pendulum of Exercise 34. Using the results of this section, obtain a value of the spring constant  $k$  which guarantees that this system is globally exponentially stable about the zero solution.

The double inverted pendulum is described as

$$\begin{aligned}\ddot{\theta}_1 + 2\dot{\theta}_1 - \dot{\theta}_2 + 2k\theta_1 - k\theta_2 - \sin \theta_1 &= 0 \\ \ddot{\theta}_2 - \dot{\theta}_1 + \dot{\theta}_2 - k\theta_1 + k\theta_2 - \sin \theta_2 &= 0\end{aligned}$$

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### Solution:

The given system equations can be modified as

$$\begin{aligned}\ddot{\theta}_1 &= -2\dot{\theta}_1 + \dot{\theta}_2 - 2k\theta_1 + k\theta_2 + \sin \theta_1 \\ \ddot{\theta}_2 &= \dot{\theta}_1 - \dot{\theta}_2 + k\theta_1 - k\theta_2 + \sin \theta_2\end{aligned}$$

In space-state representation it becomes

$$\begin{bmatrix} \dot{\theta}_1 \\ \dot{\theta}_2 \\ \ddot{\theta}_1 \\ \ddot{\theta}_2 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -2k & k & -2 & 1 \\ k & -k & 1 & -1 \end{bmatrix} \begin{bmatrix} \theta_1 \\ \theta_2 \\ \dot{\theta}_1 \\ \dot{\theta}_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \sin \theta_1 \\ \sin \theta_2 \end{bmatrix}$$

Now if we define  $x_1 := \theta_1$ ,  $x_2 := \theta_2$ ,  $x_3 := \dot{\theta}_1$ , and  $x_4 := \dot{\theta}_2$ , we can rewrite this as

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \\ \ddot{x}_3 \\ \ddot{x}_4 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -2k & k & -2 & 1 \\ k & -k & 1 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \\ \sin x_1 \\ \sin x_2 \end{bmatrix}.$$

We structure the nonlinearity to be

$$\begin{aligned}\psi_1(x) &= \sin x_1 \\ \psi_2(x) &= \sin x_2\end{aligned}$$

and since

$$\begin{aligned}-1 &\leq \sin x_1 \leq 1 \\ -1 &\leq \sin x_2 \leq 1\end{aligned}$$

The system matrices become

$$A = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -2k & k & -2 & 1 \\ k & -k & 1 & -1 \end{bmatrix}, \quad B_1 = \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \quad B_2 = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$C_1 = [1 \ 0 \ 0 \ 0], \quad C_2 = [0 \ 1 \ 0 \ 0]$$

Now if  $z_1 := x_1$ ,  $z_2 := x_2$ ,  $\lambda_1 = 1$ , and  $\lambda_2 = 1$ , we can say that

$$\tilde{\phi}(z) = \begin{bmatrix} \lambda_1 \phi_1(\lambda_1^{-1} z_1) \\ \lambda_2 \phi_2(\lambda_2^{-1} z_2) \end{bmatrix} \quad \text{where} \quad z = \begin{bmatrix} z_1 \\ z_2 \end{bmatrix} \quad \text{with} \quad z_i \in \mathbb{R}^{p_i}.$$

Then the system can also be expressed as

$$\dot{x} = Ax + \tilde{B}\tilde{\phi}(\tilde{C}x)$$

with

$$\tilde{B} := [\lambda_1^{-1} B_1 \quad \lambda_2^{-1} B_2], \quad \tilde{C} := \begin{bmatrix} \lambda_1 C_1 \\ \lambda_2 C_2 \end{bmatrix}.$$

Provided what we have so far we can setup the LMI to be

$$\begin{bmatrix} PA + A^T P + \lambda_1^2 C_1^T C_1 + \lambda_2^2 C_2^T C_2 & \gamma P B_1 & \gamma P B_2 \\ \gamma B_1^T P & -\lambda_1^2 I & 0 \\ \gamma B_2^T P & 0 & -\lambda_2^2 I \end{bmatrix} < 0.$$

Since  $\gamma = 1$ ,  $\lambda_1 = 1$ , and  $\lambda_2 = 1$

$$\begin{bmatrix} PA + A^T P + C_1^T C_1 + C_2^T C_2 & P B_1 & P B_2 \\ B_1^T P & -I & 0 \\ B_2^T P & 0 & -I \end{bmatrix} < 0$$

$$0 < P$$

Now we solve this using MATLAB's LMI Toolbox, and the code is as follows.

```

1 % AAE 666 HW7 Exercise 2
2 % Tomoki Koike
3 close all; clear all; clc;
4 %%
5 echo off;
6 %k = 1
7 k = 17.9;
8 while true

```

```

9      % Quadratic stability LMI of the problem
10     A = [ 0 0 1 0;
11           0 0 0 1;
12           -2*k k -2 1;
13           k -k 1 -1];
14
15     B1 = [0; 0; 1; 0];
16     B2 = [0; 0; 0; 1];
17     C1 = [1 0 0 0];
18     C2 = [0 1 0 0];
19
20     % Setup LMI
21     setlmis([]);
22     % P matrix
23     p=lmivar(1, [4,1]);
24     % Equation 1
25     lmi1=newlmi;
26     lmiterm([lmi1,1,1,p],1,A,'s'); % PA + A'P
27     lmiterm([lmi1,1,1,0],C1'*C1); % C1'C1
28     lmiterm([lmi1,1,1,0],C2'*C2); % C2'C2
29     lmiterm([lmi1,1,2,p],1,B1); % PB1
30     lmiterm([lmi1,1,3,p],1,B2); % PB1
31     lmiterm([lmi1,2,2,0],-1); % -I
32     lmiterm([lmi1,3,3,0],-1); % -I
33     % Equation 2
34     lmi2=newlmi;
35     lmiterm([-lmi2,1,1,p],1,1); % 0 < P
36     % Configure for solver
37     lmis = getlmis;
38     % Results
39     [tfeas, xfeas] = feasp(lmis);
40     P = dec2mat(lmis,xfeas,p);
41
42     if tfeas < 0
43         break;
44     end
45     % Increment gamma value
46     %k = k + 0.1;
47     k = k + 0.0001;
48 end
49
50 % Save file as .m
51 matlab.internal.liveeditor.openAndConvert('aee666_hw7_ex1.mlx', ...
52     convertStringsToChars(fullfile(pwd, 'aee666_hw7_ex1.m')));

```

As a result, we obtain the minimal spring constant  $k$  that guarantees that this system is GES about the zero solution to be

$$k = 18.0398$$

with a corresponding  $P$  matrix of

$$P = \begin{bmatrix} 119.9267 & -72.4921 & 1.1087 & -0.6401 \\ -72.4921 & 47.4346 & -0.6401 & 0.4686 \\ 1.1087 & -0.6401 & 2.6295 & -1.3891 \\ -0.6401 & 0.4686 & -1.3891 & 1.2405 \end{bmatrix}.$$

### Exercise 3

Prove the following result: Suppose there exists a positive-definite symmetric matrix  $P$  and a positive scalar  $\alpha$  which satisfy

$$\begin{aligned} PA + A'P + 2\alpha P &\leq 0 \\ B'P &= C \end{aligned}$$

Then the system (11.37)-(11.38) is globally exponentially stable about the origin with rate  $\alpha$  and with Lyapunov matrix  $P$ .

Where (11.37) is

$$\dot{x} = Ax - B\phi(Cx)$$

and (11.38) is

$$z'\phi(z) \leq 0$$

for all  $z$ .

---

#### Solution:

If  $V = x'Px$

$$\begin{aligned} \dot{V} &= \dot{x}'Px + x'P\dot{x} \\ &= 2x'P\dot{x} \\ &= 2x'P(Ax - B\phi(Cx)) \\ &= 2x'PAx - 2x'PB\phi(Cx) \\ &= x'(PA + A'P)x - 2x'C'\phi(Cx) \\ &= x'(PA + A'P)x - 2(Cx)'\phi(Cx) \\ &\leq x'(PA + A'P)x. \end{aligned}$$

Since from the given conditions we know that

$$PA + A'P < -2\alpha P$$

we can posit that

$$\dot{V} < -2\alpha V.$$

Hence, the system (11.37)-(11.38) is globally exponentially stable about the origin with rate  $\alpha$  and with a Lyapunov matrix  $P$ .

q.e.d



## Exercise 4

Consider the transfer function

$$\hat{g}(s) = \frac{\beta s + 1}{s^2 + s + 2}$$

Using Lemma 12, determine the range of  $\beta$  for which this transfer function is SPR. Verify your results with the KYSPR lemma.

---

### Solution:

This transfer function can be expressed as the following state space model

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

From Lemma 12 we first check the stability of this transfer function, so

$$\det(\lambda I - A) = \lambda^2 + \lambda + 2 = 0$$

which gives us eigenvalues of

$$\text{eig}(A) = \frac{-1 \pm \sqrt{7}i}{2}.$$

Since the eigenvalues have a negative real part **this system is stable**.

Next, we check the dissipativity of the transfer function.

$$\begin{aligned} \hat{g}(j\omega) &= \frac{\beta\omega j + 1}{-\omega^2 + j\omega + 2} \\ &= \frac{1 + \beta\omega j}{(2 - \omega^2) + j\omega} \\ &= \frac{(1 + \beta\omega j)((2 - \omega^2) - j\omega)}{((2 - \omega^2) + j\omega)((2 - \omega^2) - j\omega)} \\ &= \frac{2 + (\beta - 1)\omega^2 - ((2 - \omega^2)\beta\omega - \omega)j}{(2 - \omega^2)^2 + \omega^2} \end{aligned}$$

and therefore,

$$\hat{g}(j\omega) + \hat{g}(j\omega)' = \frac{2 + (\beta - 1)\omega^2}{(2 - \omega^2)^2 + \omega^2}$$

which is greater than 0 when  $\beta \geq 1$ , thus

$$\hat{g}(j\omega) + \hat{g}(j\omega)' > 0 \quad \text{if } \beta \geq 1$$

Finally, we check the asymptotic side condition

$$\begin{aligned}\lim_{\omega \rightarrow \infty} \omega^2 \frac{2 + (\beta - 1)\omega^2}{(2 - \omega^2)^2 + \omega^2} &= \lim_{\omega \rightarrow \infty} \frac{\frac{2}{\omega^2} + (\beta - 1)}{(\frac{2}{\omega^2} - 1)^2 + \frac{1}{\omega^2}} \\ &= \beta - 1.\end{aligned}$$

This becomes positive when only  $\beta > 1$ . Hence,

$$\lim_{|\omega| \rightarrow \infty} \omega^2 \hat{g}(j\omega) + \hat{g}(j\omega)' \neq 0.$$

Thus, from Lemma 12 we have proven this transfer function to be **strictly positive real (SPR)**.

Let us verify this using the KYSPR lemma. First we check the observability and controllability of the system when  $\beta = 1.2$ .

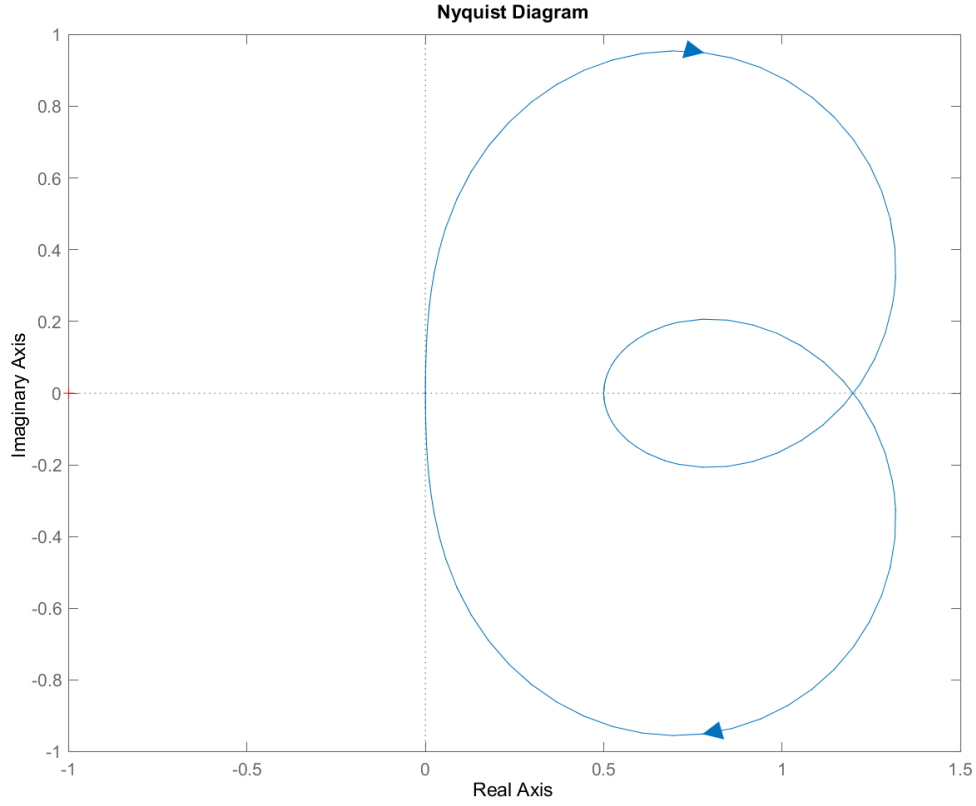
$$\begin{aligned}Q_c &= [B \quad AB] = \begin{bmatrix} 0 & 1 \\ 1 & -1 \end{bmatrix} \\ \text{rank}(Q_c) &= 2.\end{aligned}$$

Hence the system is controllable.

$$\begin{aligned}Q_o &= \begin{bmatrix} C \\ CA \end{bmatrix} = \begin{bmatrix} 1 & 1.2 \\ -2.4 & -0.2 \end{bmatrix} \\ \text{rank}(Q_o) &= 2.\end{aligned}$$

Hence the system is observable.

The Nyquist plot for this is



Now we solve the following LMI to prove KYSPR,

$$\begin{aligned} 0 &\leq \begin{bmatrix} \alpha I & B'P - C \\ PB - C' & \alpha I \end{bmatrix} \\ PA + A'P &< 0 \\ 0 &< P \end{aligned}$$

We minimize this LMI optimization problem for the parameter  $\alpha$  and we obtain,

$$\alpha = 0.$$

This shows that the following relation is satisfied

$$\begin{aligned} PA + A'P &< 0 \\ B'P &= C \end{aligned}$$

with

$$\begin{aligned} P &= \begin{bmatrix} 3.3731 & 0.5415 \\ 0.5415 & 1.5927 \end{bmatrix} > 0 \\ eig(P) &= \begin{bmatrix} 1.4409 \\ 3.5249 \end{bmatrix} \end{aligned}$$

and therefore, **this system is SPR from the KYSPR lemma.**

MATLAB Code:

```
1 % AAE 666 HW7 Exercise 4
2 % Tomoki Koike
3 close all; clear all; clc;
4 %%
5 % Control matrices
6 beta = 1.2;
7 A = [0 1; -2 -1];
8 B = [0; 1];
9 C = [1, beta];
10 D = 0;
11
12 % Nyquist Plot
13 sys = ss(A,B,C,D);
14 fig = figure("Renderer","painters","Position",[60 60 900 700]);
15 nyquist(sys);
16 saveas(fig, "ex4_nyquist.png")
17
18 % Observability and Controllability
19 Qc = ctrb(A,B);
20 rankQc = rref(Qc);
21 Qo = obsv(A,C);
22 rankQo = rref(Qo);
23 %%
24 echo off;
25 % Quadratic stability LMI of the problem
26
27 % Setup LMI
28 setlmis([]);
29 % P matrix
30 p = lmivar(1, [2, 1]); % P
31 a = lmivar(1, [1, 1]); % alpha
32
33 if D == 0
34     % Equation 1
35     lmi1 = newlmi;
36     lmiterm([-lmi1, 1, 1, 0], a); % aI
37     lmiterm([-lmi1, 1, 2, 1], B', p); % B'P
38     lmiterm([-lmi1, 1, 2, 0], -C); % -C
39     lmiterm([-lmi1, 2, 2, 0], a); % aI
40
41     % Equation 2
42     lmi2 = newlmi;
43     lmiterm([lmi2, 1, 1, p], 1, A, 's'); % PA + A'P
```

```

44
45 % Equation 3
46 lmi3 = newlmi;
47 lmiterm([-lmi3, 1, 1, p], 1, 1); %  $0 < P$ 
48
49 lmi4 = newlmi;
50 lmiterm([lmi4, 1, 1, a], 1, 1);
51 else
52 % Equation 1
53 lmi1 = newlmi;
54 lmiterm([lmi1, 1, 1, p], 1, A, 's'); %  $PA + A'P$ 
55 lmiterm([lmi1, 1, 1, a], 2, p); %  $2aP$ 
56 lmiterm([lmi1, 1, 2, p], 1, B); %  $PB$ 
57 lmiterm([lmi1, 1, 2, 0], -C'); %  $-C'$ 
58
59 lmiterm([lmi1, 2, 2, 0], -D); %  $-D$ 
60 lmiterm([lmi1, 2, 2, 0], -D'); %  $-D'$ 
61
62 % Equation 2
63 lmi2 = newlmi;
64 lmiterm([-lmi2, 1, 1, p], 1, 1); %  $0 < P$ 
65
66 % Equation 3
67 lmi3 = newlmi;
68 lmiterm([-lmi3, 1, 1, a], 1, 1); %  $0 < a$ 
69 end
70
71 % Configure for solver
72 lmis = getlmis;
73 %%
74
75 % Results
76 [tfeas, xfeas] = feasp(lmis);
77 P = dec2mat(lmis, xfeas, p);
78 v1 = defcx(lmis, 2, a);
79 c = mat2dec(lmis, v1);
80 options = [1e-5 0 0 0 0];
81 [alpha, xopt] = mincx(lmis, c, options);
82
83 assert(tfeas < 0, "This results is infeasible.");
84 %%
85 % Save file as .m
86 matlab.internal.liveeditor.openAndConvert('aee666_hw7_ex4.mlx', ...
87     convertStringsToChars('aee666_hw7_ex4.m'));

```