AMS 231: Nonlinear Control Theory: Winter 2018 Homework #4

Name:

Due: February 22, 2018

NOTE: Please show all the steps in your solution. Turn in a hard copy of your HW stapled with this as cover sheet with your name written in the above field. Please submit your HW in class on the due date.

Problem 1

State Space Computation of \mathcal{H}_{∞} Norm

$$(25+20+15+20=80 \text{ points})$$

In class (Lecture 12), we derived that the worst-case \mathcal{L}_2 gain of a stable LTI system is

$$\gamma_{\text{LTI}} = \| G(j\omega) \|_{\infty} := \sup_{\omega \in \mathbb{R}} \sigma_{\text{max}} (G(j\omega)), \qquad j := \sqrt{-1}, \qquad G(s) = C(sI - A)^{-1}B + D,$$

where $G(j\omega)$ is the associated transfer matrix. However, this frequency domain formula is inconvenient for computing γ_{LTI} , since it requires solving a nonlinear optimization problem in ω . The purpose of this exercise is to demonstrate an alternate method for computing γ_{LTI} using state space formulation, for the case D=0 (no direct feedthrough).

(a) By specializing the \mathcal{L}_2 gain theorem for nonlinear systems (Lecture 12 notes, page 8 and 9) for $f(\underline{x}) = A\underline{x}$, $g(\underline{x}) = B$, $h(\underline{x}) = C\underline{x}$, and $V(\underline{x}) = \underline{x}^{\top}P\underline{x}$ where $P \succ 0$, prove that **if** the following optimization problem:

minimize
$$\gamma$$

subject to $\gamma > 0$, $P \succ 0$, $PA + A^{\top}P + \frac{1}{\gamma^2}PBB^{\top}P + C^{\top}C \preceq 0$,

has unique solution, **then** the answer of this optimization problem gives the tightest upper bound of \mathcal{L}_2 gain γ_{LTI} . (In fact, when the triple (A, B, C) is minimal, meaning both controllable and observable, then the answer of this optimization problem equals γ_{LTI} , and hence equals $\|G(j\omega)\|_{\infty}$. But you can ignore this detail).

(b) At first glance, it may seem that the optimization problem in part (a) is nonlinear in both variables: scalar γ and matrix P, due to the last inequality constraint. However, this difficulty can be overcome via the following lemma.

Lemma: Consider real square matrices Q, R, S with Q and R symmetric. The linear matrix inequality $\begin{bmatrix} Q & S \\ S^\top & R \end{bmatrix} \succeq 0$ is equivalent to (if and only if) $R \succ 0$ and $Q - SR^{-1}S^\top \succeq 0$.

Prove this lemma.

(c) Using the lemma in part (b), and introducing $\sigma := \gamma^2$, show that the optimization problem derived in part (a) is equivalent to the following optimization problem:

subject to
$$\sigma > 0$$
, $P \succ 0$,
$$\begin{bmatrix} A^{\top}P + PA + C^{\top}C & PB \\ B^{\top}P & -\sigma I \end{bmatrix} \preceq 0,$$

which is linear in both variables σ and P. Here I denotes the identity matrix of appropriate dimension.

(d) The type of optimization problem derived in part (c) is called semi-definite programming (SDP) problem that minimizes linear objective subject to linear matrix inequalities. SDPs are convex optimization problems, and can be solved efficiently via software like cvx in MATLAB.

Download cvx from http://cvxr.com/cvx/download/ and follow installation instructions in http://cvxr.com/cvx/doc/install.html. To understand how to specify an optimization problem in cvx, you may want to take a look at: http://cvxr.com/cvx/examples/

Then write a MATLAB code to compute the \mathcal{H}_{∞} norm of the following stable, controllable and observable linear system (see partial code) in two ways: by using cvx to solve the optimization in part (c), and by using MATLAB command norm(sys,inf) to solve the frequency domain optimization problem. Report the \mathcal{H}_{∞} norms computed from the two methods, and submit your code.

Partial MATLAB Code

```
1 clear; clc;

2
3 A = \begin{bmatrix} -2 & 1 & 0 & 0; \\ -6 & -6 & 0 & 3; \\ 0 & 0 & -1 & 1; \\ 6 & 0 & 0 & -2]; \\ 7 B = \begin{bmatrix} 0; 1; 1; 2]; \\ 8 C = \begin{bmatrix} 0 & 6 & 2 & -8; & 2 & -3 & 4 & 5 \end{bmatrix}; \end{cases}
```

```
9 D = 0;
10
11    sys = ss(A,B,C,D);
12
13    eig(A) % is stable?
14    rank(ctrb(sys))==length(A) % is controllable?
15    rank(obsv(sys))==length(A) % is observable?
16    dim=size(B);    n_x = dim(1);    n_u = dim(2);
17    dim=size(B);    n_x = dim(1);    n_u = dim(2);
18    dim=size(B);    n_x = dim(1);    n_u = dim(2);
19    dim=size(B);    n_x = dim(1);    n_u = dim(2);
10    dim=size(B);    n_x = dim(1);    n_u = dim(2);
10    dim=size(B);    n_x = dim(1);    n_u = dim(2);
11    dim=size(B);    n_x = dim(1);    n_u = dim(2);
12    dim=size(B);    n_x = dim(1);    n_u = dim(2);
13    dim=size(B);    n_x = dim(1);    n_u = dim(2);
13    dim=size(B);    n_x = dim(1);    n_u = dim(2);
14    dim=size(B);    n_x = dim(1);    n_u = dim(2);
15    dim=size(B);    n_x = dim(1);    n_u = dim(2);
16    dim=size(B);    n_x = dim(1);    n_u = dim(2);
17    dim=size(B);    n_x = dim(1);    n_u = dim(2);
18    dim=size(B);    n_x = dim(1);    n_u = dim(2);
18    dim=size(B);    n_x = dim(1);    n
```

Problem 2

Input-to-State Stability (ISS)

 $(4 \times 5 = 20 \text{ points})$

Consider the scalar nonlinear systems

(a)
$$\dot{x} = -(1+u)x^3$$
, (b) $\dot{x} = -(1+u)x^3 - x^5$, (c) $\dot{x} = -x + x^2u$, (d) $\dot{x} = x - x^3 + u$.

Which systems are input-to-state stable (ISS) and which are not? Give reasons.