

Automatic Control LTH, 2016

Course Outline

L1-L5 Specifications, models and loop-shaping by hand

L6-L8 Limitations on achievable performance

- 6. Controllability/observability, multivariable poles/zeros, realizations
- 7. Fundamental limitations
- 8. Multivariable and decentralized control

L9-L11 Controller optimization: Analytic approach

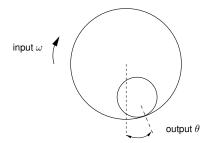
L12-L14 Controller optimization: Numerical approach

Lecture 6 - Outline

- 1. Controllability and observability, gramians
- 2. Multivariable poles and zeros
- 3. Minimal state-space realizations

[Glad & Ljung] Ch. 3.2-3.3, beg. of 3.5; Lecture notes on course web page

Example: Ball in the Hoop

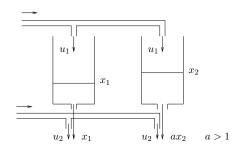


 $\ddot{\theta} + c\dot{\theta} + k\theta = \dot{\omega}$

Can you reach $\theta=\pi/4, \dot{\theta}=0$?

Can you stay there?

Example: Two water tanks



$$\begin{aligned} \dot{x}_1 &= -x_1 + u_1 \\ \dot{x}_2 &= -ax_2 + u_1 \end{aligned}$$

$$y_1 = x_1 + u_2$$
$$y_2 = ax_2 + u_2$$

Can you reach $y_1 = 1, y_2 = 2$?

Can you stay there?

Remember: Controllability and state feedback

Process

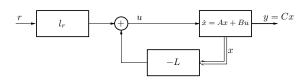
$$\begin{cases} \dot{x} = Ax + Bu \\ u = Cx \end{cases}$$

$$\begin{cases} \dot{x} = (A - BL)x + Bl_r r \\ y = Cx \end{cases}$$

Closed-loop system

State-feedback control

$$u = -Lx + l_r r = -[l_1 \ l_2 \dots l_n]x + l_r r$$



If the system (A,B) is $\mbox{\it controllable}$ we can find a state feedback gain vector \boldsymbol{L} to place the poles of the closed-loop system where we want

Remember: Observability and observers

Process

$$\begin{cases} \frac{dx}{dt} = Ax + Bu\\ y = Cx \end{cases}$$

$$\begin{cases} \frac{dx}{dt} = Ax + Bu \\ y = Cx \end{cases} \begin{cases} \frac{d\hat{x}}{dt} = \underbrace{A\hat{x} + Bu}_{\text{prediction}} + \underbrace{K(y - \hat{y})}_{\text{correction}} \end{cases}$$

Estimation/observer error $\tilde{x} = x - \hat{x}$:

$$\begin{split} \dot{\hat{x}} &= \dot{x} - \dot{\hat{x}} \\ &= Ax + Bu - A\hat{x} - Bu - K(Cx - C\hat{x}) \\ &= (A - KC)\tilde{x} \end{split}$$

If the system (A,C) is *observable* we can find an observer gain vector

$$K = \begin{bmatrix} k_1 \\ \cdot \\ k_n \end{bmatrix}$$
 which assigns desired eigenvalues for $(A - KC)$.

Controllability

The system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

is **controllable**, if for every $x_1 \in \mathbf{R}^n$ there exists $u(t), t \in [0, t_1]$, such that $x(t_1) = x_1$ is reached from x(0) = 0.

The collection of vectors x_1 that can be reached in this way is called the controllable subspace.

(Matlab: orth(ctrb(A,B)))

Controllability criteria

The following statements regarding a system $\dot{x}(t)=Ax(t)+Bu(t)$ of order n are equivalent:

(i) The system is controllable

(ii) rank
$$[A - \lambda I \ B] = n$$
 for all $\lambda \in \mathbf{C}$

(iii) rank
$$[B \ AB \dots A^{n-1}B] = n$$

If A is exponentially stable, define the ${f controllability}$ ${f Gramian}$

$$S = \int_0^\infty e^{At} B B^T e^{A^T t} dt$$

For such systems there is a fourth equivalent statement:

(iv) The controllability Gramian is non-singular

Interpretation of the controllability Gramian

The controllability Gramian measures how difficult it is to reach a certain state.

In fact, let $S_1=\int_0^{t_1}e^{At}BB^Te^{A^Tt}dt$. Then, for the system $\dot{x}(t)=Ax(t)+Bu(t)$ to reach $x(t_1)=x_1$ from x(0)=0 it is necessary that

$$\int_0^{t_1} |u(t)|^2 dt \ge x_1^T S_1^{-1} x_1$$

Equality is attained with

$$u(t) = B^T e^{A^T (t_1 - t)} S_1^{-1} x_1$$

(For proof, see the lecture notes.)

Computing the controllability Gramian

The controllability Gramian $S=\int_0^\infty e^{At}BB^Te^{A^Tt}dt$ can be computed by solving the Lyapunov equation

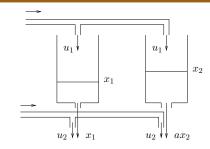
$$AS + SA^T + BB^T = 0$$

(For proof, see the lecture notes.)

Matlab: S = lyap(A, B*B')

Q: Where have we seen this equation before?

Example: Two water tanks



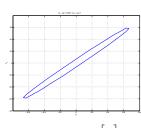
$$\dot{x}_1 = -x_1 + u_1 \qquad \qquad \dot{x}_2 = -ax_2 + u_1$$

$$\text{Controllability Gramian: } S = \int_0^\infty \begin{bmatrix} e^{-t} \\ e^{-at} \end{bmatrix} \begin{bmatrix} e^{-t} \\ e^{-at} \end{bmatrix}^T dt = \begin{bmatrix} \frac{1}{2} & \frac{1}{a+1} \\ \frac{1}{a+1} & \frac{1}{2a} \end{bmatrix}$$

S close to singular when $a \approx 1$. Interpretation?

Example cont'd

Matlab:



Plot of
$$\begin{bmatrix} x_1 & x_2 \end{bmatrix} \cdot S^{-1} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = 1$$
 corresponds to the states we can reach by
$$\int_0^\infty |u(t)|^2 dt = 1.$$

Observability

The system

$$\dot{x}(t) = Ax(t)$$
$$y(t) = Cx(t)$$

is **observable**, if the initial state $x(0)=x_0\in\mathbf{R}^n$ can be uniquely determined by the output $y(t),t\in[0,t_1].$

The collection of vectors x_0 that cannot be distinguished from x=0 is called the ${\bf unobservable}$ subspace.

(Matlab: null(obsv(A,C)))

Observability criteria

The following statements regarding a system $\dot{x}(t)=Ax(t),$ y(t)=Cx(t) of order n are equivalent:

(i) The system is observable

$$\begin{array}{l} \text{(ii)} \ \ \text{rank} \begin{bmatrix} A-\lambda I \\ C \end{bmatrix} = n \ \text{for all} \ \lambda \in \mathbf{C} \\ \\ \text{(iii)} \ \ \text{rank} \begin{bmatrix} C \\ CA \\ \vdots \end{bmatrix} = n \\ \end{array}$$

If A is exponentially stable, define the **observability Gramian**

$$O = \int_0^\infty e^{A^T t} C^T C e^{At} dt$$

For such systems there is a fourth equivalent statement:

(iv) The observability Gramian is non-singular

Interpretation of the observability Gramian

The observability Gramian measures how difficult it is to distinguish two initial states from each other by observing the output.

In fact, let $O_1=\int_0^{t_1}e^{A^Tt}C^TCe^{At}dt$. Then, for $\dot{x}(t)=Ax(t)$, the influence from the initial state $x(0)=x_0$ on the output y(t)=Cx(t) satisfies

$$\int_0^{t_1} |y(t)|^2 dt = x_0^T O_1 x_0$$

Computing the observability Gramian

The observability Gramian $O=\int_0^\infty e^{A^Tt}C^TCe^{At}dt$ can be computed by solving the Lyapunov equation

$$A^TO + OA + C^TC = 0$$

Matlab: 0 = lyap(A',C'*C)

Mini-problem

Is the water tank system with a=1 observable?

What if only y_1 is available?

Poles and zeros

$Y(s) = \underbrace{[C(sI - A)^{-1}B + D]}_{G(s)}U(s)$

For scalar systems, the points $p\in\mathbf{C}$ where $G(s)=\infty$ are called **poles** G. They are eigenvalues of A and determine stability.

The poles of $G(s)^{-1}$ are called **zeros** of G.

This definition can be used also for square MIMO systems, but we will next give a more general definition, involving also multiplicity.

Pole and zero polynomials

- ightharpoonup The **pole polynomial** is the least common denominator of all minors (sub-determinants) to G(s).
- $\,\blacktriangleright\,$ The zero polynomial is the greatest common divisor of the maximal minors of G(s).

The **poles** of ${\cal G}$ are the roots of the pole polynomial.

The (transmission) zeros of G are the roots of the zero polynomial.

Poles and zeros - example

Calculate the poles and zeros of

$$G(s) = \begin{pmatrix} \frac{2}{s+1} & \frac{3}{s+2} \\ \frac{1}{s+1} & \frac{1}{s+1} \end{pmatrix}$$

 $\textbf{Poles: Minors: } \tfrac{2}{s+1}, \tfrac{3}{s+2}, \tfrac{1}{s+1}, \tfrac{1}{s+1}, \tfrac{2}{(s+1)^2} - \tfrac{3}{(s+1)(s+2)} = \tfrac{-(s-1)}{(s+1)^2(s+2)}$

The least common denominator is $(s+1)^2(s+2)$, giving the poles -2 and -1 (double)

Zeros: Maximal minor: $\frac{-(s-1)}{(s+1)^2(s+2)}$

The greatest common divisor is s-1, giving the (transmission) zero 1.

Zeros of square systems

When G(s) is square, the only maximal minor is $\det G(s)$, so the zeros are determined from the equation

$$\det G(s) = 0$$

For a square system with minimal state-space realization, the zeros are the solutions to

$$\det \begin{bmatrix} sI - A & B \\ -C & D \end{bmatrix} = 0$$

Interpretation of poles and zeros

Poles:

- ▶ A pole s = a is associated with a time function $x(t) = x_0 e^{at}$
- $\qquad \qquad \mathbf{A} \text{ pole } s = a \text{ is an eigenvalue of } A$

Zeros:

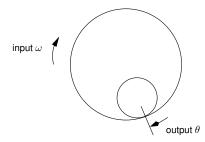
- $\,\blacktriangleright\,$ A zero s=a means that an input $u(t)=u_0e^{at}$ is blocked
- A zero describes how inputs and outputs couple to states







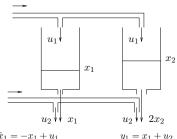
Example: Ball in the Hoop



$$\ddot{\theta}+c\dot{\theta}+k\theta=\dot{\omega}$$

The transfer function from ω to θ is $\frac{s}{s^2+cs+k}.$ The zero in s=0 makes it impossible to control the stationary position of the ball.

Example: Two water tanks



$$\dot{x}_1 = -x_1 + u_1$$
 $y_1 = x_1 + u_2$
 $\dot{x}_2 = -2x_2 + u_1$ $y_2 = 2x_2 + u_2$

$$G(s) = \begin{bmatrix} \frac{1}{s+1} & 1\\ \frac{2}{s+2} & 1 \end{bmatrix} \qquad \det G(s) = \frac{-s}{(s+1)(s+2)}$$

The system has a zero in the origin! At stationarity $y_1 = y_2$.

» s=tf('s') » sigma(G) ; plot singular values

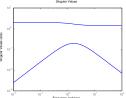
Plot singular values of $G(i\omega)$ vs frequency

» G=[1/(s+1) 1; 2/(s+2) 1]

% Alt. for a certain frequency:

» A = freqresp(G,i*w);

"[U,S,V] = svd(A)



The largest singular value of $G(i\omega)=\begin{bmatrix} \frac{1}{i\frac{\omega+1}{\omega+1}} & 1\\ \frac{2}{i\omega+2} & 1 \end{bmatrix}$ is fairly constant.

This is due to the second input. The first input makes it possible to control the difference between the two tanks, but mainly near $\omega=1\,$ where the dynamics make a difference.

Realization in diagonal form

Consider a transfer matrix with partial fraction expansion

$$G(s) = \sum_{i=1}^{n} \frac{C_i B_i}{s - p_i} + D$$

This has the realization

$$\dot{x}(t) = \begin{bmatrix} p_1 I & & 0 \\ & \ddots & \\ 0 & & p_n I \end{bmatrix} x(t) + \begin{bmatrix} B_1 \\ \vdots \\ B_n \end{bmatrix} u(t)$$

$$y(t) = \begin{bmatrix} C_1 & \dots & C_n \end{bmatrix} x(t) + Du(t)$$

The rank of the matrix $C_i B_i$ determines the necessary number of columns in B_i and the multiplicity of the pole p_i .

(Warning: Matlab has no good command for doing this)

Realization of multivariable system - example 1

To find a minimal state-space realization for the system

$$G(s) = \begin{pmatrix} \frac{2}{s+1} & \frac{3}{s+2} \\ \frac{1}{s+1} & \frac{1}{s+1} \end{pmatrix}$$

with poles in -2 and -1 (double), write the transfer matrix as (e.g.) $\,$

$$G(s) = \frac{\begin{bmatrix} 2\\1 \end{bmatrix} \begin{bmatrix} 1 & 0 \end{bmatrix}}{s+1} + \frac{\begin{bmatrix} 0\\1 \end{bmatrix} \begin{bmatrix} 0 & 1 \end{bmatrix}}{s+1} + \frac{\begin{bmatrix} 3\\0 \end{bmatrix} \begin{bmatrix} 0 & 1 \end{bmatrix}}{s+2}$$

giving the realization

$$\dot{x} = \begin{pmatrix} -1 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -2 \end{pmatrix} x + \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 1 \end{pmatrix} u$$
$$y = \begin{pmatrix} 2 & 0 & 3 \\ 1 & 1 & 0 \end{pmatrix} x$$

Realization of multivariable system - example 2

To find state space-realization for the system

$$G(s) = \begin{bmatrix} \frac{1}{s+1} & \frac{2}{(s+1)(s+3)} \\ \frac{6}{(s+2)(s+4)} & \frac{1}{s+2} \end{bmatrix}$$

write the transfer matrix as

$$\begin{bmatrix} \frac{1}{s+1} & \frac{1}{s+1} - \frac{1}{s+3} \\ \frac{3}{s+2} - \frac{3}{s+4} & \frac{1}{s+2} \end{bmatrix} = \frac{\begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix}}{s+1} + \frac{\begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} 3 & 1 \end{bmatrix}}{s+2} + \frac{\begin{bmatrix} 1 \\ 0 \end{bmatrix} \begin{bmatrix} 0 & -1 \end{bmatrix}}{s+3} + \frac{\begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} -3 & 0 \end{bmatrix}}{s+4}$$

$$\begin{bmatrix} \dot{x}_1(t) \\ \dot{x}_2(t) \\ \dot{x}_3(t) \\ \dot{x}_4(t) \end{bmatrix} = \begin{bmatrix} -1 & 0 & 0 & 0 \\ 0 & -2 & 0 & 0 \\ 0 & 0 & -3 & 0 \\ 0 & 0 & 0 & -4 \end{bmatrix} \begin{bmatrix} x_1(t) \\ x_2(t) \\ x_3(t) \\ x_4(t) \end{bmatrix} + \begin{bmatrix} 1 & 1 \\ 3 & 1 \\ 0 & -1 \\ -3 & 0 \end{bmatrix} \begin{bmatrix} u_1(t) \\ u_2(t) \end{bmatrix} \\ \begin{bmatrix} y_1(t) \\ y_2(t) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix} x(t)$$