Multivariable Robust Control Lecture 1 (Meeting 1)

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MULTIVARIABLE FEEDBACK CONTROL Analysis and design

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Lecture notes are based on material provided by Prof. Skogestad

Content

- Introduction [Chapter 1]
- ② Classical feedback control [Chapter 2]
- Performance limitations SISO [Chapter 5]
- Uncertainty and robustness SISO [Chapter 7]
- Selements of linear systems theory [Chapter 4]
- Introduction to multivariable control [Chapter 3]
- Performance limitations MIMO [Chapter 6]
- Solution Stability and performance MIMO [Chapter 8]
- Ontroller design [Chapter 9]

Another Books

- C. Chen, Linear System Theory and Design, Oxford University Press, 1999.
- G. Dullerud and F. Paganini, A Course in Robust Control Theory, Springer, 2000.
- M. Green and D. Limebeer, Linear Robust Control, Prentice Hall, 1995.
- J. Helton and O. Merino, Classical Control Using H-infinity Methods, SIAM, 1998.
- J. Maciejowski, Multivariable Feedback Design, Addison Wesley, 1989.
- R. Sanchez-Pena and M. Sznaier, Robust Systems Theory and Applications, John Wiley & Sons, 1998.
- K. Zhou, Robust and Optimal Control, Prentice Hall, 1996.
- K. Zhou, Essentials of Robust Control, Prentice Hall, 1998.

https://www.lehigh.edu/~eus204/teaching/ME450_MRC/ME450_MRC.html

- Study the "plant" to be controlled and learn about control objectives
- Ø Model the system and simplify the model (if necessary)
- On Analyze the resulting model; determine its properties
- Oecide which variables are to be controlled (controlled outputs)
- Oecide on measurement (sensors)/manipulated (actuators) variables
- Select the control configuration
- Ø Decide on the type of controller to be used
- Obecide on performance specifications (based on control objectives)
- Oesign a controller
- Analyze resulting controlled system (redesign if necessary)
- Simulate the resulting controlled system
- Repeat from Step 2, if necessary.
- Choose hardware and software and implement the controller
- Test and validate control system and tune on-line if necessary

 \Rightarrow Much of the course will be spent on input-output "controllability analysis" of the plant/process.

Always keep in mind

- Power of control is limited.
- Control quality depends controller AND on plant/process.

$$y = Gu + G_d d \tag{1}$$

- y : output/controlled variable
- u : input/manipulated variable
- d : disturbance
- r : reference/setpoint

 $\begin{array}{rcl} \mbox{Regulator problem} & : & \mbox{counteract } d \\ \mbox{Servo}/\mbox{Tracking problem} & : & \mbox{let } y \mbox{ follow } r \end{array}$

Goal of control: make control error e = y - r "small".

Major difficulties:

Model (G, G_d) inaccurate

 \Rightarrow RealPlant: $G_p = G + E$;

E = "uncertainty" or "perturbation" (unknown but bounded)

- Nominal stability (NS) : system is stable with no model uncertainty.
- Nominal Performance (NP) : system satisfies performance specifications with no model uncertainty.
- Robust stability (RS) : system stable for "all" perturbed plants
- Robust performance (RP) : system satisfies performance specifications for all perturbed plants

The course will be make extensive use of the *transfer function* representation, i.e., we will work on the frequency domain.

Given a state-space representation

$$\dot{x} = Ax + Bu, y = Cx + Du,$$

the transfer-function representation is given by

$$\frac{y(s)}{u(s)} = G(s) = C(sI - A)^{-1}B + D, \quad G(s) = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

$$G(s) = \frac{\beta_{n_z} s^{n_z} + \dots + \beta_1 s + \beta_0}{s^n + a_{n-1} s^{n-1} + \dots + a_1 s + a_0}$$

For multivariable systems, G(s) is a matrix of transfer functions. n = order of denominator (or pole polynomial) or order of the system $n_z = \text{order of numerator (or zero polynomial)}$ $n - n_z = \text{pole excess or relative order.}$ (2)

Definition

- A system G(s) is strictly proper if $G(s) \to 0$ as $s \to \infty$.
- A system G(s) is semi-proper or bi-proper if $G(s) \to D \neq 0$ as $s \to \infty$.
- A system G(s) which is strictly proper or semi-proper is *proper*.
- A system G(s) is improper if $G(s) \to \infty$ as $s \to \infty$.

Proper scaling simplifies controller design and performance analysis. $\ensuremath{\textbf{SISO}}$:

unscaled:

$$\hat{y} = \hat{G}\hat{u} + \hat{G}_d\hat{d}; \quad \hat{e} = \hat{y} - \hat{r}$$
(3)

scaled:

$$d = \hat{d}/\hat{d}_{\max}, \quad u = \hat{u}/\hat{u}_{\max} \tag{4}$$

where:

- \hat{d}_{\max} largest expected change in disturbance
- \hat{u}_{\max} largest allowed input change

Scale \hat{y} , \hat{e} and \hat{r} by:

- \hat{e}_{\max} largest allowed control error, or
- $\hat{r}_{\rm max}$ largest expected change in reference value

Usually:

$$y = \hat{y}/\hat{e}_{\max}, \quad r = \hat{r}/\hat{e}_{\max}, \quad e = \hat{e}/\hat{e}_{\max}$$
 (5)

MIMO :

$$d = D_d^{-1}\hat{d}, \quad u = D_u^{-1}\hat{u}, \quad y = D_e^{-1}\hat{y}$$
(6)

$$e = D_e^{-1}\hat{e}, \quad r = D_e^{-1}\hat{r}$$
 (7)

where $D_e = \hat{e}_{max}$, $D_u = \hat{u}_{max}$, $D_d = \hat{d}_{max}$ and $D_r = \hat{r}_{max}$ are diagonal scaling matrices

Substituting (6) and (7) into (3):

$$D_e y = \hat{G} D_u u + \hat{G}_d D_d d; \quad D_e e = D_e y - D_e r$$

and introducing the scaled transfer functions

$$G = D_e^{-1} \hat{G} D_u, \quad G_d = D_e^{-1} \hat{G}_d D_d$$
 (8)

Model in terms of scaled variables:

$$y = Gu + G_d d; \quad e = y - r \tag{9}$$

Often also:

$$\tilde{r} = \hat{r}/\hat{r}_{\max} = D_r^{-1}\hat{r} \tag{10}$$

so that:

$$r = R\tilde{r}$$
 where $R \stackrel{\Delta}{=} D_e^{-1} D_r = \hat{r}_{\max} / \hat{e}_{\max}$ (11)



Objective:

 $\begin{array}{l} \text{for } |d(t)| \leq 1 \text{ and } |\tilde{r}(t)| \leq 1, \\ \text{manipulate } u \text{ with } |u(t)| \leq 1 \\ \text{such that } |e(t)| = |y(t) - r(t)| \leq 1. \end{array}$

Given the general nonlinear model

$$\dot{x} = f(x, u) \tag{12}$$

where (x^*, u^*) is an equilibrium, i.e., $f(x^*, u^*) = 0$. The linearization around this equilibrium is given by

$$\dot{\tilde{x}} \approx \left(\frac{\partial f}{\partial x}\right)^* \tilde{x} + \left(\frac{\partial f}{\partial u}\right)^* \tilde{u} \triangleq A\tilde{x} + B\tilde{u}$$
(13)

where $\tilde{x} \triangleq x - x^*$ and $\tilde{u} \triangleq u - u^*$. Therefore,

$$\tilde{x}(s) = (sI - A)^{-1} B\tilde{u}$$
(14)







(c) General control configuration

Figure 2: Control configurations

Table 1: Nomenclature

K controller, in whatever configuration. Sometimes broken down into parts. For example, in Figure 2(b), $K = \begin{bmatrix} K_r & K_y \end{bmatrix}$ where K_r is a prefilter and K_y is the feedback controller.

Conventional configurations (Fig 2(a), 2(b)):

- G plant model
- G_d disturbance model
- *r* reference inputs (commands, setpoints)
- d disturbances (process noise)
- n measurement noise
- y plant outputs. (include the variables to be controlled ("primary" outputs with reference values r) and possibly additional "secondary" measurements to improve control)
- y_m measured y
- *u* control signals (manipulated plant inputs)

General configuration (Fig 2(c)):

- P generalized plant model. Includes G and G_d and the interconnection structure between the plant and the controller. May also include weighting functions.
- w $\,$ exogenous inputs: commands, disturbances and noise $\,$
- z exogenous outputs; "error" signals to be minimized, e.g. y-r
- v controller inputs for the general configuration, e.g. commands, measured plant outputs, measured disturbances, etc. For the special case of a one degree-of-freedom controller with perfect measurements we have v = r y.
- u control signals