TSRT09 – Control Theory

Lecture 5: Controller structure and control design

Claudio Altafini

Reglerteknik, ISY, Linköpings Universitet



Summary of lecture 4

Kalman filter

- Optimal observer
- Requires stochastic models of the disturbances
- Requires solving an algebraic Riccati equation (ARE)

$$K = PC^T R_2^{-1}$$
$$AP + PA^T - PC^T R_2^{-1} CP + NR_1 N^T = 0$$



Summary of lecture 4 (cont'd)

Important transfer functions:

- Closed loop:
 - $G_c = (I + GF_y)^{-1}GF_r$
 - $\bullet \quad \text{reference} \to \text{controlled signal}$
 - Sensitivity function: $S = (I + GF_y)^{-1}$
 - $\bullet \quad \text{output disturbance} \to \text{controlled signal}$
 - model error \rightarrow controlled signal $\Delta_z \sim S \Delta_G$
 - Complementary sensitivity function: $T = (I + GF_y)^{-1}GF_y$
 - $\bullet \quad \text{measurement error} \rightarrow \text{controlled signal}$
 - $\bullet \quad {\rm model \ error} \ \to {\rm stability} \quad |T(i\omega)| < \frac{1}{|\Delta_G(i\omega)|} \ \ \forall \ \omega$
 - Input sensitivity function: $S_u = (I + F_y G)^{-1}$
 - $\bullet \quad \text{input disturbance} \to \text{input}$



Specifications for control design

Task: Choose u so that z follows r as close as possible, in spite of the presence of disturbances w, n, and of uncertainty in the system, while at the same time using reasonable values of u

Translate this into transfer functions:

• error e = r - z small

$$e = (I - G_c)r - Sw + Tn$$

• input not too large

$$u = G_{ru}r + G_{wu}(w+n)$$

• effect of uncertainty small

$$\Delta_z \sim S \Delta_G$$

• closed-loop stability with uncertainty

$$|T(i\omega)| < \frac{1}{|\Delta_G(i\omega)|} \quad \forall \; \omega$$



Specifications for control design

Conditions: design F_y , F_r s.t.

- $I G_c$ small \Rightarrow controlled variable z follows reference signal r
- $S \text{ small} \implies$ system disturbance and model errors have small impact on the controlled variable
- $T \text{ small} \implies$ measurement disturbance has small impact on the controlled variable, and model errors do not compromise stability
- G_{ru} and G_{wu} small \implies input u stays moderate

But observe that

$$S + T = I$$

$$G_c = GG_{ru}$$

 \implies conflicts on the specifications!

k



Lecture 5

Controller structure and control design

- MIMO systems: who should control who?
 - RGA (Relative Gain Array)
- From structure to design:
 - Decentralized control
 - Decoupled Control
 - IMC (Internal Model Control)
- Synthesis method 1: Linear Quadratic synthesis

In the book: Ch. 8 and 9



Control design

Most successful controller ever: PID

- Timeline:
 - Boulton and Watt 1788: speed control of steam engines, mechanical implementation
 - Hydraulic and pneumatic implementation: late 1800.
 - Electronic implementation: 1930.
 - Computer implemention: 1950.
 - "PID-on-a-chip": 1990s
- Modeling
 - First systematic approach (poles): Maxwell 1868.
 - Robust shaping of system gain: Aström och Hägglund 2006.



source: Wikipedia – Andy

Dingley

• Applications: all.



PID and beyond

PIDs:

- It pairs an output and an input
- Can be developed with intuition and experimentation. Result are normally sufficient (good in very simple cases)
- Interpretation in a Bode diagram: lead and lag
- Systematic analysis (poles, zeros, *S*, *T*,...) can give control design with high performances
- MIMO systems \implies inputs/outputs must be paired two by two

When is PID not enough:

- most MIMO systems (e.g. when no natural pairing exists)
- in some advanced applications, e.g. when you are given state-based costs to minimize, e.g. linear quadratic design
- nonlinear systems



MIMO systems

Approach: explore structure

- if natural pairings exist ⇒ decentralized control (each loop is controlled independently)
- if natural pairings do not exist:
 - RGA (Relative Gain Array) a way to measure interaction
 - Decoupled control: a way to reduce coupling

More advanced control methods

- IMC (Internal Model Control).
- Minimization of quadratic criterion: LQ, LQG.
- Systematic shaping of transfer functions: \mathcal{H}_2 , \mathcal{H}_∞ .
- Nonlinear methods.



MIMO systems: structure-based control design

• G(s) is a $p \times m$ matrix:

$$G(s) = \left[G_{ij}(s)\right]_{\substack{i = 1, \dots, p \\ j = 1, \dots, m}}$$

• Control design

$$u = F_r(s)r - F_y(s)y$$

where $F_r(s)$ and $F_y(s)$ are $m \times p$ matrices

$$F_y(s) = \left[F_{y,ij}(s)\right]_{\substack{i=1,\dots,m\\j=1,\dots,p}}$$

- Q: who controls who?
- Q: Which y_k use to design u_j ?
- Q: How to design F_r and F_y ?

MIMO systems: problems

- 1. Who should control who?
 - G square: p = m
 - 'Tall' system: p > m

$$G = \begin{bmatrix} \cdots \\ \cdots \\ \cdots \\ \cdots \end{bmatrix}$$

Less inputs than outputs: Not all outputs can be controlled perfectly $-\mbox{ must prioritize}.$

• "Fat" system:
$$p < m$$

$$G = \begin{bmatrix} \cdots & \cdots & \cdots \\ \cdots & \cdots & \cdots \end{bmatrix}$$

More inputs than outputs. How should the control effort be distributed among the inputs?

2. Interactions / cross couplings



Interaction / Cross coupling

• Two-handle water mixer, a system with "hard" crosscoupling.

- Multiple inputs influence (heavily) each output.
- Multiple outputs are influenced (heavily) by any input.





Interaction / Cross coupling

- Two-handle water mixer, a system with "hard" crosscoupling.
 - Multiple inputs influence (heavily) each output.
 - Multiple outputs are influenced (heavily) by any input.





Interaction / Cross coupling (cont'd)

• Single-handle water mixer, a system with "gentle" crosscoupling.

- Every input influences (almost) just one output.
- Every output is influenced (almost) by only one input.





Interaction / Cross coupling (cont'd)

- Single-handle water mixer, a system with "gentle" crosscoupling.
 - Every input influences (almost) just one output.
 - Every output is influenced (almost) by only one input.





Decentralised regulation

- Build a regulator for a MIMO system by letting *one* output be used by *one* input.
- Result is a number of scalar loops

$$u_j = F_{r,jk} r_k - F_{y,jk} y_k$$

where each regulator is not "aware" of the presence of the others.

- If G square transfer function matrix ⇒ can be done If G rectangular: disregard some of the signals (?)
- Works when there are natural pairings and cross-couplings are small.
- "Pairing problem": one would like to pair the input and outputs with strongest coupling.
- How do you determine which coupling exist between inputs and outputs?



Example: Temperature regulator



- Two rooms with a wall separating them.
- Temperatures T_1 and T_2 are states and measured variables.
- Each room can warm or cool via U_1 and U_2 .



Example: Temperature regulator





Example: Temperature regulator



Which sensor should be used by which heat/cold source?



Example: Temperature regulator (cont'd)

Regulator. First pairing: emperatur [grad. C] 6 15 25 16 16 25 т2 • T_1 is used by U_1 . • T_2 is used by U_2 . 10 Tid [h] 15 20 Decentralized PI-regulator x 10⁴ Effekt [W] $F_y(s) = \begin{bmatrix} 1000 + \frac{500}{s} & 0\\ 0 & 1000 + \frac{500}{s} \end{bmatrix}$ 10 15 5 20 Tid [h] Effekt störning [W] 5 After 10 hours, 10 people enter into room 1. 0 10 15 5 20 Stable Tid [h]



Example: Temperature regulator (cont'd) Regulator. Second pairing:



problem "analytically"?

RGA (Relative Gain Array)

Consider the following extreme cases

1. output k is controlled from input j and no other input is active

2. output k is controlled from input j and all other outputs are perfectly regulated (e.g. forced to 0).



RGA (Relative Gain Array)

• Construct the ratios between the two gains in the two cases (for all input/output pairs)

• Matematically: elementwise multiplication of G and G^{-T} (or $(G^{\dagger})^{T}$ if G rectangular)

 $\mathsf{RGA}(G) = G.*(G^\dagger)^T$



Decentralized control with RGA

- Properties of RGA
 - function of ω
 - row sum = 1
 - column sum =1

- Estimating coupling
 - Ideal case (after permutation of inputs/outputs):

 $\mathsf{RGA}(G(i\omega)) = I \;\; \forall \, \omega$

• Deviation of $RGA(G(i\omega))$ from $I \implies$ coupling



Decentralized control with RGA

- Practical rules: pair y_k and u_j so that
 - diagonal elements in $RGA(G(i\omega_c))$ are near 1 (ω_c = closed-loop cross-over freq.)

• diagonal elements in RGA(G(0)) do not become negative (can give instability).

- Decentralized control:
 - treat each loop independently

$$u_j = F_{r,jk} r_k - F_{y,jk} y_k$$



Example: Temperature regulator with RGA

Example: room temperature regulation

First pairing

- T_1 is used by U_1
- T_2 is used by U_2

$$\mathsf{RGA}(G(i5)) \approx \begin{bmatrix} 1 & 0\\ 0 & 1 \end{bmatrix}$$
$$\mathsf{RGA}(G(0)) = \begin{bmatrix} 1.17 & -0.17\\ -0.17 & 1.17 \end{bmatrix}$$

OK with both rules

Second pairing

- T_2 is used by U_1
- T_1 is used by U_2

$$\mathsf{RGA}(G(i5)) \approx \begin{bmatrix} 1 & 0\\ 0 & 1 \end{bmatrix}$$
$$\mathsf{RGA}(G(0)) = \begin{bmatrix} 1.17 & -0.17\\ -0.17 & 1.17 \end{bmatrix}$$

Violates both rules



Decoupled regulation

- Decentralized control: need "natural pairs" of inputs and outputs
- What if this is not the case? Create them!
- Change of variables for inputs and outputs:

$$\tilde{u} = W_1^{-1}u \qquad \qquad \tilde{y} = W_2 y$$

• Task: design W_1 and W_2 such that the "virtual system":

$$\tilde{G}(s) = W_2(s)G(s)W_1(s)$$

is as decoupled (i.e., diagonal) as possible

- Then design diagonal regulator $\tilde{F}_y(s)$
- Resulting regulator: $F_y(s) = W_1(s)\tilde{F}_y(s)W_2(s)$



Decoupled regulation (cont'd)

How to choose W_1 and W_2 ?

- 1. "Dynamical" decoupling
 - completely decoupled "virtual system", over all ω
 - s-dependent matrices $W_1(s)$ and $W_2(s)$ are needed
 - often not possible (complicated or non-linear regulator)
- 2. "Static" decoupling
 - choose one frequency at which the system becomes decoupled
 - 1. steady state $\omega = 0$
 - 2. cross-over frequency $\omega = \omega_c$
 - Example of steady state decoupling

$$W_1 = G^{-1}(0)$$
 $W_2 = I$



Example: two handle water mixer



With the right choice of W_1 and W_2 you can make a two-handle water mixer behave as a single-handle mixer. More easily controlled!

$$W_1 = G^{-1}(0)$$
 $W_2 = I$



Internal Model Control (IMC)

- true system: G_0
- model (assumed stable): G
- "new information": y Gu
- Idea: feed back the "new information" y Gu
- block diagram:





25 / 37

Internal Model Control (IMC)

• IMC: use new information y - Gu as feedback, with a T.F. Q

$$u = -Q \underbrace{(y - Gu)}_{\text{new information}} + Q\tilde{F}_r r$$

new information

or

$$u = -\underbrace{(I - QG)^{-1}Q}_{F_y} y + \underbrace{(I - QG)^{-1}Q\tilde{F}_r}_{F_r} r$$

• If G stable: all stabilizing controllers $u = -F_y y$ are of the form

$$F_y = (I-QG)^{-1}Q \qquad orall \; Q \;\;$$
 stable



IMC design rules: basic idea

• Ideally: choose
$$Q = G^{-1} \implies G_c = I$$

• Complication: unfeasible $F_y \equiv \infty$

• Solution: approximate the inverse appropriately, e.g.

$$Q(s) = \frac{1}{(\lambda s + 1)^n} G^{-1}(s)$$

• Different system properties lead to different suitable approximations of the inverse.



Two main approaches:

1. Quadratic weights on the variables + optimization. \implies "Linear quadratic synthesis": "LQ", "LQG"

2. Direct construction of S, T, in the frequency domain: $\implies "`\mathcal{H}_2"$, $'\mathcal{H}_{\infty}"$



A warm-up problem

- State space model
- fully deterministic problem

$$\dot{x} = Ax + Bu,$$
 $x(0)$ given
 $z = Mx$
 $y = x$



Minimization of a quadratic criterion

Problem: Full state LQ (Linear Quadratic) optimal control Find a u = -Lx that solves

$$\min J = \int_0^\infty (z^T Q_1 z + u^T Q_2 u) dt$$

s. t. $\dot{x} = Ax + Bu$, $x(0)$ given
 $z = Mx$
and $A - BL$ is stable

Solution: u = -Lx, $L = Q_2^{-1}B^TS$, where S solves the ARE $A^TS + SA + M^TQ_1M - SBQ_2^{-1}B^TS = 0$









Solving the algebraic Riccati equation

• Is it always possible to solve

$$A^{T}S + SA + \bar{Q}_{1} - SBQ_{2}^{-1}B^{T}S = 0, \qquad \bar{Q}_{1} = M^{T}Q_{1}M$$

so that the state matrix of the regulated system

$$A - BQ_2^{-1}B^T S$$

has its eigenvalues in the left half of the complex plane?



Solving the algebraic Riccati equation (cont'd)

• Example:

$$A = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \bar{Q}_1 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \quad Q_2 = 1$$



Solvability of the algebraic Riccati equation If

1.
$$Q_1 \ge 0$$
, $Q_2 > 0$

2. (A, B) is stabilizable (unstable states are controllable)

3. (A, \bar{Q}_1) is detectable (unstable states can be "seen" in the cost) then there exists a unique solution $S = S^T \ge 0$ of

$$A^TS + SA + \bar{Q}_1 - SBQ_2^{-1}B^TS = 0$$

with $L = Q_2^{-1} B^T S$, that minimizes the criterion

$$J = \int_0^\infty (x^T \bar{Q}_1 x + u^T Q_2 u) dt$$

and

$$A - BQ_2^{-1}B^T S$$

has all eigenvalues strictly in the left half of the complex plane.



Solving the algebraic Riccati equation (cont'd)

• Example:

$$A = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \quad \bar{Q}_1 = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \quad Q_2 = 1$$

• Example:

$$A = \begin{bmatrix} -1 & 0\\ 1 & 1 \end{bmatrix} \qquad B = \begin{bmatrix} 1\\ 0 \end{bmatrix}$$



Similarities between Kalman filter and LQ Kalman filter:

$$AP + PA^{T} + NR_{1}N^{T} - PC^{T}R_{2}^{-1}CP = 0$$

$$K = PC^{T}R_{2}^{-1} \quad (\Leftrightarrow K^{T} = R_{2}^{-1}CP)$$

LQ-regulator:

$$A^{T}S + SA + M^{T}Q_{1}M - SBQ_{2}^{-1}B^{T}S = 0 \qquad R_{1}$$
$$L = Q_{2}^{-1}B^{T}S \qquad R_{2}$$

$$C^T \longleftrightarrow B$$

$$N^T \longleftrightarrow M$$

$$R_1 \longleftrightarrow Q_1$$

$$R_2 \longleftrightarrow Q_2$$

$$K^T \longleftrightarrow L$$

 $\begin{array}{c} P \longleftrightarrow S \\ A^T \longleftrightarrow A \end{array}$

Both problems require solving an ARE



TSRT09 Control Theory 2022, Lecture 5 www.liu.se

