

GIAN course: Singular optimal control

Slide collection 8

The Stability radius and an application of KYP

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The stability radius

Linear system: $\dot{x} = Ax + Bu, \quad y = Cx$

Output feedback: $u = \Delta y \quad (\Delta = \text{feedback matrix})$

Closed loop: $\dot{x} = \underbrace{(A + B\Delta C)}_{=: A_\Delta} x$

Problem (distance to instability): Suppose A is stable (all eigenvalues in left half plane).

What is the smallest Δ , such that A_Δ is unstable? Small means $\|\Delta\|_2$ small.

Stability radius:

$$\begin{aligned} r_{\mathbb{C}}(A, B, C) &:= \inf\{\|\Delta\|_2 \mid \Delta \in \mathbb{C}^{p \times q}, A_\Delta \text{ unstable}\} \\ &= \sup\{\rho > 0 \mid A_\Delta \text{ stable for all } \Delta \in \mathbb{C}^{p \times q} \text{ with } \|\Delta\|_2 < \rho\}. \end{aligned}$$

Since eigenvalues of A_Δ depend continuously on Δ :

$$r_{\mathbb{C}}(A, B, C) = \inf\{\rho > 0 \mid \exists \Delta : (\|\Delta\|_2 = \rho \text{ and } A_\Delta \text{ has eigenvalue on imaginary axis})\}$$

Moving eigenvalues by minimal perturbations

Lemma. 1 Suppose $\lambda \in \mathbb{C}$ is not an eigenvalue of A . Then

$$\lambda \text{ is eigenvalue of } A_\Delta = A + B\Delta C \quad \Leftrightarrow \quad \exists v \neq 0 : \Delta G(\lambda) v = v,$$

where

$$G(\lambda) = C(\lambda I - A)^{-1}B \quad (\text{transfer function})$$

Proof. If λ is eigenvalue then there is $w \neq 0$ such that

$$\begin{aligned} (A + B\Delta C)w = \lambda w &\Rightarrow B\Delta Cw = (\lambda I - A)w \\ &\Rightarrow (\lambda I - A)^{-1}B\Delta Cw = w \quad (w \neq 0 \text{ implies } \Delta Cw \neq 0) \\ &\Rightarrow \underbrace{\Delta C(\lambda I - A)^{-1}B}_{G(\lambda)} \underbrace{\Delta Cw}_v = \underbrace{\Delta Cw}_v. \end{aligned}$$

Proof of other direction analogous: if $\Delta G(\lambda) v = v$ then $w = (\lambda I - A)^{-1}Bv$ is eigenvector.

Lemma 2. If λ is eigenvalue of A_Δ then since $\Delta G(\lambda)v = v$ we have $\|\Delta\|_2 \|G(\lambda)\|_2 \geq 1$.

Lemma 3. A smallest Δ such that A_Δ has eigenvalue λ is given by

$$\Delta := \frac{v G(\lambda)^*}{\|G(\lambda)\|_2^2},$$

where v is a normalized singular vector to the largest singular value of $G(\lambda)$:

$$\|v\|_2 = 1, \quad \|G(\lambda)v\|_2 = \|G(\lambda)\|_2, \quad \|\Delta\|_2 = \frac{1}{\|G(\lambda)\|_2}.$$

Formula for the complex stability radius $r_{\mathbb{C}}(A, B, C)$

From the lemmas on the slide before we have:

If $\lambda \in \mathbb{C}$ is not an eigenvalue of A then there exists a matrix Δ with

$$\|\Delta\|_2 = \frac{1}{\|G(\lambda)\|_2} = \frac{1}{\|C(\lambda I - A)^{-1}B\|_2}$$

such that λ is an eigenvalue of $A_{\Delta} = A + B\Delta C$.

There is no matrix of smaller norm with the latter property.

Consequence for stability radius of stable A :

$$\begin{aligned} r_{\mathbb{C}}(A, B, C) &= \inf\{\rho > 0 \mid \exists \Delta : (\|\Delta\|_2 = \rho \text{ and } A_{\Delta} \text{ has eigenvalue on imaginary axis})\} \\ &= \inf\{\|G(i\omega)\|^{-1} \mid \omega \in \mathbb{R}\} \\ &= \frac{1}{\sup_{\omega \in \mathbb{R}} \|G(i\omega)\|_2} \end{aligned}$$

Stability of nonlinear feedback (here the KYP-Lemma comes into play)

We considered stability for linear systems of the form

$$\dot{x}(t) = (A + B\Delta C) x(t).$$

Replacing here the linear map $y \mapsto \Delta y$ by an arbitrary function f we obtain the ODE

$$\dot{x}(t) = Ax(t) + Bf(t, Cx(t)).$$

Question: under which condition is

$$\lim_{t \rightarrow \infty} x(t) = 0 \quad ?$$

Conjecture: sufficient condition is

$$\|f(t, y)\|_2 \leq r \|y\|_2, \quad \text{for some } r < r_{\mathbb{C}}(A, B, C).$$

Conjecture is true. The proof on following pages via KYP-Lemma.

Theorem. Suppose the function $(t, y) \mapsto f(t, y)$ is locally Lipschitz with respect to y (sufficient cond. for local existence of solutions of ODE) and satisfies

$$\|f(t, y)\|_2 \leq r \|y\|_2 \quad \text{for some } r < r_{\mathbb{C}}(A, B, C).$$

Suppose further, that A is (Hurwitz) stable. Then

$$\dot{x}(t) = Ax(t) + Bf(t, Cx(t)) \quad \Rightarrow \quad \lim_{t \rightarrow \infty} x(t).$$

Proof.

$$r < r_{\mathbb{C}}(A, B, C) = \frac{1}{\sup_{\omega \in \mathbb{R}} \|G(i\omega)\|_2} \quad \Rightarrow \quad r^{-2} > \sup_{\omega \in \mathbb{R}} \|G(i\omega)\|_2^2$$

Thus, for all $\omega \in \mathbb{R}$,

$$0 < r^{-2} I - G(i\omega)^* G(i\omega) = \begin{bmatrix} (i\omega I - A)B \\ I \end{bmatrix}^* \begin{bmatrix} -C^*C & 0 \\ 0 & r^{-2} \end{bmatrix} \begin{bmatrix} (i\omega I - A)B \\ I \end{bmatrix} \quad (\text{Popov func.})$$

By KYP there exists $P = P^*$ such that

$$0 < \begin{bmatrix} A^*P + PA - C^*C & PB \\ B^*P & -r^{-2}I \end{bmatrix}$$

Taking the Schur complement we conclude

$$(A^*P + PA - C^*C) - r^2 PBB^*P > 0 \quad (\text{Riccati inequality})$$

This implies $A^*P + PA > 0$, and hence since A is stable: $P < 0$.

Proof continued: The condition $\|f(y, t)\|_2 \leq r \|y\|_2$ implies

$$\|f(t, Cx(t))\|_2^2 \leq r^2 \|Cx\|_2^2 = r^2 x^* C^* C x.$$

In the following computation $f(t, Cx(t))$ is abbreviated by f . With this notation we have

$$r^{-2} \|f\|^2 \leq x^* C^* C x \quad (*)$$

From $\dot{x} = Ax + Bf$ it follows that

$$\begin{aligned} \frac{d}{dt} x^* P x &= x^* P \dot{x} + \dot{x}^* P x \\ &= x^* P (Ax + Bf) + (Ax + Bf)^* P x \\ &= f^* (B^* P x) + ((B^* P x)^* f + x^* (A^* P + PA) x) \\ &= \|r B^* P x + r^{-1} f\|_2^2 - r^2 x^* P B B^* P x - \underbrace{r^{-2} \|f\|^2}_{\geq -x^* C^* C x \text{ by } (*)} + x^* (A^* P + PA) x \\ &\geq \|r B^* P x + r^{-1} f\|_2^2 \underbrace{- r^2 x^* P B B^* P x - x^* C^* C x + x^* (A^* P + PA) x}_{>0 \text{ by the first part of the proof}} \end{aligned}$$

Summary: We have along the solutions of the ODE,

$$x^* P x < 0, \quad \frac{d}{dt} x^* P x > 0, \quad \text{if } x \neq 0.$$

This implies

$$\lim_{t \rightarrow \infty} x(t) = 0.$$

Formula for the real stability radius $r_{\mathbb{R}}(A, B, C)$

In the theorem below $\sigma_2(M)$ denotes the second largest singular value of M .

Theorem. If $\lambda \in \mathbb{C}$ is not an eigenvalue of A then there exists a **real** matrix Δ with

$$\|\Delta\|_2 = \left(\inf_{\gamma \in (0,1]} \sigma_2 \left(\begin{bmatrix} \Re(G(\lambda)) & -\gamma^{-1} \Im(G(\lambda)) \\ \gamma \Im(G(\lambda)) & \Re(G(\lambda)) \end{bmatrix} \right) \right)^{-1}$$

such that λ is an eigenvalue of $A_{\Delta} = A + B\Delta C$.

There is no matrix of smaller norm with the latter property.

Consequence for the **real** stability radius of stable A :

$$\begin{aligned} r_{\mathbb{R}}(A, B, C) &= \inf\{\rho > 0 \mid \exists \text{real } \Delta : (\|\Delta\|_2 = \rho \text{ and } A_{\Delta} \text{ has eigenvalue on imaginary axis})\} \\ &= \frac{1}{\sup_{\omega \in \mathbb{R}} \inf_{\gamma \in (0,1]} \sigma_2 \left(\begin{bmatrix} \Re(G(\lambda)) & -\gamma^{-1} \Im(G(\lambda)) \\ \gamma \Im(G(\lambda)) & \Re(G(\lambda)) \end{bmatrix} \right)}. \end{aligned}$$

Unfortunately:

For nonlinear real feedback the same statement holds as for nonlinear complex feedback. I.e. the complex stability radius and **not** the real applies for real nonlinear feedback.

Literature

- (1) Book: Hinrichsen, Pritchard. Mathematical Systems Theory 1.
- (2) Qiu, Bernhardson, Rantzer, Davison, Young, Doyle.
A formula for computation of the real stability radius.
Automatica 31,6 (1995)
- (3) Bernhardson, Rantzer, Qiu.
Real perturbation values and real quadratic forms
in a complex vector space.
Linear Alg. Appl. 270 (1998)
- (4) Genin, Stefan, van Dooren.
Real and complex stability radii of polynomial matrices.
Linear Alg. Appl. 351-352 (2002)

Perturbing Matrix polynomials

Perturbed companion form of monic matrix polynomial:

$$\begin{bmatrix} 0 & I & & \\ & & \ddots & \\ & & & I \\ -P_0 - \Delta_0 & \dots & \dots & -P_{n-1} - \Delta_{n-1} \end{bmatrix} = \begin{bmatrix} 0 & I & & \\ & & \ddots & \\ & & & I \\ -P_0 & \dots & \dots & -P_{n-1} \end{bmatrix} - \underbrace{\begin{bmatrix} 0 \\ \vdots \\ 0 \\ I \end{bmatrix}}_C [\Delta_0 \quad \dots \quad \Delta_{n-1}]$$