

Stabilizability of a control system

Note Title

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$$\dot{x}(t) = Ax(t) + Bu(t) \quad A \in \mathbb{C}^{n \times n}, B \in \mathbb{C}^{n \times m}$$
$$x(0) = x_0$$

Def: the system is stabilizable if $\exists u: [0, \infty) \rightarrow \mathbb{C}^m$
s.t. $\lim_{t \rightarrow \infty} x(t) = 0$

Ex:

$$A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix} \quad B = \begin{bmatrix} B_1 \\ 0 \end{bmatrix} \quad \Leftrightarrow \begin{cases} \dot{x}_1 = A_{11}x_1 + A_{12}x_2 + B_1 u \\ \dot{x}_2 = A_{22}x_2 \end{cases}$$

If A_{22} is a stable matrix, i.e. $\Lambda(A_{22}) \subset \text{LHP}$, then
the system is not controllable, but it is stabilizable:

$\lim_{t \rightarrow \infty} x_2(t) = 0$ already, and one can control the first part
(if (A_{11}, B_1) controllable)

Recall: Kalman decomposition: every pair (A, B) can be written
up to a change of basis as $A = \begin{bmatrix} A_{11} & A_{12} \\ 0 & A_{22} \end{bmatrix}$, $B = \begin{bmatrix} B_1 \\ 0 \end{bmatrix}$,
with (A_{11}, B_1) controllable (and $v_2 = 0$ if necessary)

Theorem: (stabilizability conditions)

The following are equivalent:

1. The system $\dot{x} = Ax + Bu$ is stabilizable
2. In the Kalman decomposition, $\Lambda(A_{22}) \subset \text{LHP}$
3. $\text{rk} [A - \lambda I \quad B] = n$ for all $\lambda \in \mathbb{C} \setminus \text{LHP}$

4. There exists $F \in \mathbb{C}^{m \times n}$ s.t. $u(t) = Fx(t)$ is a stabilizing control, i.e., $\lambda(A+BF) \subset \text{LHP}$ and $\lim x(t) = 0$.

Remark: when (A, B) stabilizable, there are an infinite number of stabilizing feedback matrices F , by continuity (if $\lambda(A+BF) \subset \text{LHP}$, the same holds for small perturbations to F)

This means that we can impose more conditions.

Optimal control (linear-quadratic regulator)

Find $u: [0, \infty) \rightarrow \mathbb{C}^m$ that minimizes

$$V(u) = \int_0^{\infty} x(t)^* Q x(t) + u(t)^* R u(t) dt$$

for given $Q \in \mathbb{C}^{n \times n}$, $Q = Q^* \succeq 0$, $R \in \mathbb{C}^{m \times m}$, $R = R^* \succ 0$.

s.t. $\dot{x} = Ax + Bu$

Idea: control reducing the oscillations and the "input" u .

We shall assume $R \succ 0$

Theorem:

Let $Q \succeq 0$, $R \succ 0$, (A, B) stabilizable, (A^*, Q) stabilizable.

Then, there exists $X \in \mathbb{C}^{n \times n}$, $X = X^* \succeq 0$ s.t. $x_0 \in \mathbb{C}^n$ be given

$$1. \quad A^* X + X A + Q - X G X = 0,$$

$$G = B R^{-1} B^* \in \mathbb{C}^{n \times n}$$

$$G = G^* \succeq 0$$

$$2. \quad \lambda(A - GX) \subset \text{LHP}.$$

Then, the optimum of the minimum problem

$$\min \int_0^{\infty} x(t)^* Q x(t) + u(t)^* R u(t) dt$$

$$\text{s.t. } \begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ x(0) = x_0 \in \mathbb{C}^n \end{cases} \quad \lim_{t \rightarrow \infty} x(t) = 0$$

is $x_0^* X x_0$, attained with $u(t) = Fx(t)$
with $F = -R^{-1}B^*X$.

The matrix eqn in 1. is called algebraic Riccati equation
 X that satisfies 2. (in addition to 1.) is called a
stabilizing solution.

$$\text{Note that } A + BF = A + B(-R^{-1}B^*X) = A - \underbrace{BR^{-1}B^*X}_G = A - GX.$$

Proof: We shall prove later the existence of X that
solves 1. + 2., for now let us just give the connection
to the minimum problem.

Let us take an arbitrary function u that is a
stabilizing control, i.e., $\lim_{t \rightarrow \infty} x(t) = 0$. We compute

$$\frac{d}{dt} x(t)^* X x(t) = \dot{x}^* X x + x^* X \dot{x} = (Ax + Bu)^* X x + x^* X (Ax + Bu) =$$

$$= x^* \left(\underbrace{A^* X + X A}_{XBR^{-1}B^*X - Q} \right) x + u^* B X x + x^* X B u$$

$$= \underbrace{x^* X B R^{-1} B^* X x}_{\cancel{0}} - x^* Q x + \underbrace{u^* B X x + x^* X B u}_{\cancel{0}} + u^* R u - u^* R u$$

$$= \underbrace{(u + R^{-1}B^*Xx)^* R (u + R^{-1}B^*Xx)}_{\cancel{0}} - \underbrace{x^* Q x + u^* R u}_{\text{positivity under the integral}}$$

Hence,

$$\int_0^{\infty} x^T Q x + u^T R u \, dt = \int_0^{\infty} \underbrace{(u + R^{-1} B^T X x)^T R (u + R^{-1} B^T X x)}_{\substack{\neq \\ 0}} \, dt - \underbrace{x^T X x \Big|_0^{\infty}}_{\substack{\cancel{x(\infty)^T X(\infty) - x_0^T X x_0} \\ 0}}$$

$$= x_0^T X x_0 + (\text{something} \geq 0)$$

So the minimum is $x_0^T X x_0$, attained with $u(t) = \underbrace{-R^{-1} B^T X x(t)}_{\substack{= \\ F}}$ $\forall t$

The most interesting part for us: solving

$$A^T X + X A + Q - X G X = 0, \quad \Lambda(A - G X) \subset \text{LHP}$$

EX: in dimension $n=1$:

$$-g x^2 + (\bar{a} + \alpha) x + q = 0 \quad x_{\pm} = \frac{-(\bar{a} + \alpha) \pm \sqrt{(\bar{a} + \alpha)^2 + 4gq}}{-2g}$$

Since $g, q \geq 0$, the discriminant is real, and this has two real solutions, one positive and one negative, and $\alpha - gx$ is negative in one case and positive in the other. (or LHP/RHP if $\alpha \notin \mathbb{R}$)

Let us assume $\alpha \in \mathbb{R}$, then

$$x_{\pm} = \frac{-\alpha \pm \sqrt{\alpha^2 + gq}}{-g} \begin{cases} \frac{-\alpha + \sqrt{\alpha^2 + gq}}{-g} < 0 & \text{if } g > 0, q > 0 \\ \frac{-\alpha - \sqrt{\alpha^2 + gq}}{-g} > 0 \end{cases}$$

$$\alpha - g x_{\pm} = \pm \sqrt{\alpha^2 - gq}$$

The solution x_{-} has

$$\boxed{\alpha - g x_{-} \in \text{LHP}, x_{-} \rightarrow 0}$$

Linear algebra of the algebraic Riccati equation:

if $A^T X + X A + Q - X G X = 0, \quad \Lambda(A - G X) \subset \text{LHP}$, then

$$\begin{bmatrix} A & -G \\ -Q & -A^T \end{bmatrix} \begin{bmatrix} 1 \\ x \end{bmatrix} = \begin{bmatrix} 1 \\ x \end{bmatrix} (A - G X) \Leftrightarrow \begin{cases} A - G X = A - G X \\ -Q - A^T X = X(A - G X) \end{cases}$$

I.e., $\begin{bmatrix} \bar{I} \\ X \end{bmatrix}$ is an invariant subspace for $\begin{bmatrix} A & -G \\ -Q & -A^* \end{bmatrix} =: \mathcal{H}$

(Hamiltonian matrix), and all the ^{associated} eigenvalues are in the LHP

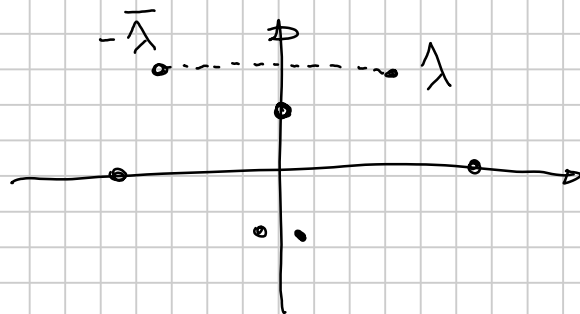
We start by proving some properties of \mathcal{H}

Lemma: let λ be an eigenvalue of \mathcal{H} , then $-\bar{\lambda}$ is also an eigenvalue of \mathcal{H} , with the same alg. multiplicity.

Proof: We shall prove that \mathcal{H} is similar to $-\mathcal{H}^*$

Let $J = \begin{bmatrix} 0 & I \\ -I & 0 \end{bmatrix}$. Then, we can compute

$$J^{-1} \mathcal{H} J = \begin{bmatrix} 0 & -I \\ I & 0 \end{bmatrix} \begin{bmatrix} A & -G \\ -Q & -A^* \end{bmatrix} \begin{bmatrix} 0 & I \\ -I & 0 \end{bmatrix} = \begin{bmatrix} Q & A^* \\ A & -G \end{bmatrix} \begin{bmatrix} 0 & I \\ -I & 0 \end{bmatrix} = \begin{bmatrix} -A^* & Q \\ G & A \end{bmatrix} = -\mathcal{H}^*$$



$\lambda, -\bar{\lambda}$ are symmetric wrt the imaginary axis

The eigenvalues of \mathcal{H} are symmetric wrt the im. axis.

We shall prove that there are no eigenvalues on the im. axis, hence $\mathcal{H} \in \mathbb{C}^{2n \times 2n}$ has n eigenvalues in the LHP and n in the RHP.

Theorem: Assume $Q \succeq 0$, $G = BR^{-1}B^* \succeq 0$,

$(A, B), (A^*, Q)$ stabilizable. \square

\uparrow
 (A, G) stabilizable

$$\begin{aligned} K(A, B) &= \text{lm} [B \ AB \ A^2 B \ \dots] = \\ &= K(A, G) = \text{lm} [G \ AG \ A^2 G \ \dots] \end{aligned}$$

because $\text{Im } B = \text{Im } G$)

Then, H has no purely imaginary eigenvalue.

Proof: assume, by contradiction, that $\lambda \in i\mathbb{R}$
 $v \neq 0 \rightarrow v_1 \neq 0$ or $v_2 \neq 0$

$$0 = (H - \lambda I)v = \begin{bmatrix} A - \lambda I & -G \\ -Q & -(A - \lambda I)^* \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \Leftrightarrow \begin{cases} (A - \lambda I)v_1 - Gv_2 = 0 \\ -Qv_1 - (A - \lambda I)^*v_2 = 0 \end{cases}$$

Multiply the first eq. by v_2^* , and the second transposed by v_1

$$\begin{cases} v_2^* (A - \lambda I)v_1 - v_2^* Gv_2 = 0 \\ -v_1^* Qv_1 - v_2^* (A - \lambda I)v_1 = 0 \end{cases} \rightarrow v_1^* Qv_1 + v_2^* Gv_2 = 0$$

Since $Q \succeq 0$, $G \succeq 0$, it must be $Qv_1 = Gv_2 = 0$

and so $(A - \lambda I)v_1 = (A - \lambda I)^*v_2 = 0$

$$v_2^* [A - \lambda I \quad G] = 0$$

$$v_1^* [A^* + \lambda I \quad Q] = 0$$

Since one among v_1, v_2 must be nonzero,

we contradict stabilizability of (A, G) or (A^*, Q)

Now we can compute a basis for the inv. subspace associated to $\Lambda(H) \cap \text{LHP}$:

$$\begin{bmatrix} A & -G \\ -Q & -A^* \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} s \quad \Lambda(s) \subset \text{LHP}$$

If u_1 is invertible, I can write

$$\begin{bmatrix} A & -G \\ -Q & -A^* \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} u_1^{-1} = \begin{bmatrix} u_1 \\ u_2 \end{bmatrix} u_1^{-1} \underbrace{u_1 s u_1^{-1}}_{\downarrow \Lambda(u_1 s u_1^{-1}) \subset \text{LHP}}$$

$\begin{bmatrix} 1 \\ x \end{bmatrix}$ $\begin{bmatrix} 1 \\ x \end{bmatrix}$

with $x = u_2 u_1^{-1}$

Hence

$$\begin{bmatrix} A & -G \\ -Q & -A^* \end{bmatrix} \begin{bmatrix} 1 \\ X \end{bmatrix} = \begin{bmatrix} 1 \\ X \end{bmatrix} \hat{S} \Leftrightarrow \begin{cases} A - GX = \hat{S} & \text{N}(A-GX) \subset \text{LHP} \\ -Q - A^*X = X\hat{S} = X(A-GX) & \text{ARE} \end{cases}$$

So $X = U_2 U_1^{-1}$ is a stabilizing solution of the ARE.

Theorem: under the same assumptions as the above results
 $Q \succeq 0$, $G \succeq 0$, (A, B) , (A^*, Q) stab.,
 U_1 is invertible

Proof: omitted (see notes if you are curious).

Also, $X = X^* \succeq 0$ (also not proved here)

How to solve AREs?

1. Via Schur form form + reordering + inv. subspace
2. Via sign(H)
3. Newton's method
4. Structured eigenvalue methods