

LECTURE 14, 4. 20. 23

LINEAR-QUADRATIC CONTROL :

$$(S) \quad \begin{cases} \dot{y} = Ay + Bu \\ y(t) = x \end{cases} \quad A \in \mathbb{M}_{n \times n}, \quad B \in \mathbb{B}_{n \times m} \\ u(t) \in L^1_{\text{loc}}([0, T], \mathbb{R}^m) \\ R > 0$$

$$(JQ) \quad J(x, t, u(\cdot)) := \int_t^T [y(s)^T M y(s) + u(s)^T R u(s)] ds + y(T)^T Q y(T).$$

$$S = B R^{-1} B^T \quad \boxed{w(x(t)) = x^T K(t) x} \quad K(\cdot) \in S(n)$$

$$(RT) \quad \begin{cases} \dot{K} = KSK - A^T K + K^T A - M \\ K(T) = Q \end{cases} \quad \text{RICCATI MATRIX} \times \\ \text{ODE.}$$

Theorem. If $K \in C^1((t_0, T), \text{Sym}(n))$ cont. at $t = T$ sol. of (RT),

then $w(t, x) = x^T K(t) x$ is a C^1 sol. of (CT), the

feedback $\boxed{\Phi(y, s) = -R^{-1} B^T K(s) y}$ is ADMISSIBLE \Leftarrow

OPTIMAL $\Leftarrow (x, t)$ &

$$w(x, t) = \text{Value function} := \inf_{u \in L^1_{\text{loc}}([0, T], \mathbb{R}^m)} J(x, t, u(\cdot))$$

Proof. $\arg \min_u H(p, x, u) = \arg \min_u \tilde{H}(p, u) = \left\{ -\frac{R^{-1} B^T p}{2} \right\}$

$$p = D_x w = 2Kx$$

$$\Phi(y, s) = -\frac{R^{-1} B^T}{2} 2Ky = -R^{-1} B^T K(s) y$$

Use verification thm.: if Φ is admissible then it is

$$\text{OPTIMAL : } \left. \begin{array}{l} \dot{y} = Ay - \underbrace{BR^{-1}B^T K(s)y}_S \\ y(t) = x \end{array} \right\} t \leq T$$

this is a LINEAR HOMOGENEOUS SYSTEM of ODE

$\Rightarrow \forall x, t \in [t_0, T] \exists \text{ UNIQUE sol. } y \in C^1([t_0, T], \mathbb{R}^n)$

$\Rightarrow \Phi$ ADMISS. $\Rightarrow \Phi$ optimal. \square

Rank. Can be proved $\Phi = -R^{-1}B^T Kg$ is the UNIQUE OPT. FEEDBACK [Engv. p. 178].

Q: $\exists?$ sol. of (RT)?

Prop. (local \exists). In our ass. ($R > 0$) $\exists t_0 < T$: (RT) has a unique sol. $K \in C^1([t_0, T], S(n))$.

Pf.. R.H.S. of ODE is C^1 , so local \exists thm. $\Rightarrow \exists$ sol. $K \in C^1([t_0, T], M_{n \times n})$. Remains to prove $K(t) \in \text{Sym}(n) \forall t$.

Take the transpose of RICCATI E.Q.: $\dot{K} = KSK - A^T K - KA - M$

$$\left. \begin{array}{l} \dot{K}^T = K^T SK^T - K^T A - A^T K - M \\ K^T(T) = Q \end{array} \right\}$$

$\Rightarrow K^T$ solves the same ODE and can. calc. as K !

\Rightarrow by uniq. of ODE we get $K = K^T \forall t \Rightarrow K \in \text{Sym}(n)$ \square .

Recall. $\ddot{y} = y^2 \quad y(0) > 0 \quad$ blows up. in finite time!

Thm: (global \exists) Supp. $R > 0$, $Q, \alpha \geq 0$ (pos. semidef.)

\Rightarrow (RT) has a sol $K \in C^1((-\infty, T], \text{Sym}(n))$ cont. up to $t=T$.

Pf [Fl.-Rish.; my notes] Know that $K:]t_0, T] \rightarrow \text{Sym}(n)$

& if $t_0 > -\infty$ $\lim_{t \rightarrow t_0^+} \|K(t)\| = +\infty$ by fl properties

of the maximal interval of \exists of sol. of ODE.

Use "trace norm" of $\tilde{X} \in \text{Sym}(n)$:

$$\|\tilde{X}\|_* := \sum_{i=1}^n |\lambda_i|, \lambda_i \text{ eigenvalues of } \tilde{X}.$$

Ex HW!: $\|\cdot\|$ is a norm on $\text{Sym}(n)$.

Notation $\|\tilde{X}\|_2 = \text{euclidean norm}$.

Goal: $\forall \tau < T \quad \exists \tilde{\tau}_\tau : \|K(t)\| \leq \tilde{c}_{\tilde{\tau}} \quad \forall \tau < t \leq T$.

(Verif. hint.)

Know that $w(x, t) = x^T K(t) x = \inf_{\alpha(\cdot)} J(x, t, \alpha(\cdot))$

$$J = \int_t^T (y^T H y + \alpha^T R \alpha) dt + y(T)^T Q y(T) \geq 0$$

$\bigvee \quad \bigvee \quad \bigvee$
 $0 \quad 0 \quad 0$

$\Rightarrow w(x, t) \geq 0 \quad \forall (x, t) \Rightarrow K(t) \geq 0 \text{ semidef. } \forall t$

$\Rightarrow \lambda_{\min}(K(t)) \geq 0 \quad \forall t$

Estimate from above of $\|K(t)\|$. Use $\alpha \equiv 0$

$$\begin{cases} \dot{y} = Ay \\ y(t) = x \end{cases} \quad \text{the sol } y_x^\circ(s) = y_x(s; t, 0) \text{ s.t. } y_x^\circ(0) = 0 \quad L = \|A\|_2$$

$$|y_x^0(s) - 0| \leq e^{L(s-t)} |x - 0| \leq e^{L(T-\varepsilon)} |x| \quad \forall \tau \leq t \leq s \leq T.$$

$$\Rightarrow \boxed{|y_x^0(s)| \leq C_\varepsilon |x|} \quad \text{for } \varepsilon \in [\varepsilon, T]$$

$$\begin{aligned} \boxed{x^T K(t) x \leq J(x, t, 0) = \int_t^T y_x^0(\tau)^T M y_x^0(\tau) d\tau + y_x^0(T)^T Q y_x^0(T)} \\ &\leq (T-t) \|M\|_2 \sup_{t \leq \tau \leq T} |y_x^0(\tau)|^2 + \|Q\|_2 |y_x^0(T)|^2 \\ &\leq C_\varepsilon^2 |x|^2 ((T-\varepsilon) \|M\|_2 + \|Q\|_2) = \boxed{\tilde{C}_\varepsilon |x|^2} \end{aligned}$$

$$\Rightarrow K(t) - \tilde{C}_\varepsilon I_n \leq 0 \Rightarrow \max_k K(t) \leq \tilde{C}_\varepsilon$$

$$\begin{aligned} \Rightarrow \|K(t)\| &\leq n \max_i |\lambda_i| = n \max_i \lambda_i = \\ &= n \max_k (K(k)) \leq n \tilde{C}_\varepsilon := \bar{C}_\varepsilon. \quad \square \end{aligned}$$

Rank: cond. of this are not necessary: see HW
at the end of the lecture.: example where the sol. of (RT)

$\exists i \in [0, T]$ even if $\lambda_i \neq 0$.

_____ \Rightarrow _____

How to solve explicitly systems of Riccati ODEs.

1. By reduction to LINEAR ODES.

Thm.: If $V, V \in C^1([t_0, T], \mathbb{R}_{n \times n})$ sol. of

$$\left\{ \begin{array}{l} \dot{V} = AV - S^T V \\ \dot{V} = -HV - A^T V \\ V(T) = I, \quad V(T) = Q \end{array} \right. \quad \text{linear homog sys.}$$

where $A, M, Q, S = B R^{-1} B^T$ are known. Then t :

$\det V(t) \neq 0$, $K(t) = V(t)V^{-1}(t)$ is a sol. of (RT)
(Riccati terminal value pt.).

Pf. w/o: Lw at least for $n=1$. □

2. Other method for $n=1$ all data are scalar.

$$\begin{cases} \dot{K} = SK^2 - 2AK - M \\ K(T) = Q \end{cases}$$

$$\int \frac{dK}{SK^2 - 2AK - M} = \int 1 dt = t + \text{const.}$$

can be solved explicitly! □.

Next time: results & comments on the Ex. L. 2 on
H-L formula:

$$u_t + \frac{(u_x)^2}{2} = 0, \quad (\Theta) \quad g(x) = -x^-$$

(b) $g(x) = x^+$ draw graph of sol. u .

$$(C) \quad v = u_x \quad v_t + \left(\frac{v^2}{2}\right)_x = 0 \quad \text{draw graph of } v.$$

Try to do it yourself!

INTRODUCTION TO GAME THEORY.

Refs.: A. Bressan on Michel J. Math.

E.N. Barron, Game theory,

L.C. Evans & P. Souganidis: Introduction U. Math. J.

• [BcD] chp. VII.

PLAN: ➤ 2-person 0-sum } "STATIC" or
 • " Non 0 sum, } ONE SHOT
 ($N \geq 2$ players is similar). GAMES.

➤ DIFFERENTIAL games. $\rightarrow \neq$ sum.
 \rightarrow 0-sum.
 $\rightarrow "N \rightarrow \infty"$ MEAN-FIELD GAMES

— o —

ZERO-SUM GAMES

DATA one: two sets A & B

A = decisions or strategies of pl. 1

B = $a \in \subset^2$.

$\Phi: A \times B \rightarrow \mathbb{R}$ $\Phi(a, b) =$ payoff of pl. 1
 = cost of pl. 2.

Goals: 1st player wants to MAXIMISE Φ
 2nd a $\in \subset$ - MINIMISE Φ .

Example. Most simple: $A = \{1, \dots, n\}$

$B = \{1, \dots, n\}$ MATRIX GAMES. $\Phi(i, j) = \phi_{ij}$

$M = \begin{pmatrix} \phi_{11} & \dots & \phi_{1n} \\ \vdots & & \vdots \\ \phi_{m1} & \dots & \phi_{mn} \end{pmatrix}$ A chooses row
 B in column.

Ex. 1 "Head or tail". Each player chooses H or T
 if choices are = A wins, else B wins.

	T	H
T	1	-1
H	-1	1

Ex. 2. "splitting a cake" 2 children must split cake.

A cuts, B chooses

		B
		Small big.
A		$\frac{1}{2} + \varepsilon$ $\frac{1}{2} - \varepsilon$
$\frac{1}{2}$ piece		$\frac{3}{4}\varepsilon$ $\frac{1}{4}\varepsilon$
big piece		Small piece

Hypothesis I : $\begin{cases} A, B \text{ compact metric spaces} \\ \Phi \in C(A \times B) \end{cases}$

Def. MARGINAL FUNCTIONS.

$$\Phi^{\max}(b) = \max_a \Phi(a, b), \quad \Phi^{\min}(a) = \min_b \Phi(a, b)$$

BEST RESPONSE MAPS :

$$R^A(b) := \arg\max_a \Phi(a, b) = \left\{ \bar{a} \in A : \Phi(\bar{a}, b) = \max_a \Phi(a, b) \right\}$$

$$R^B(a) := \arg\min_b \Phi(a, b).$$

Lemma $\Phi^{\max} \in C(B)$, $\Phi^{\min} \in C(A)$.

Pf. $b, \bar{b} \in B$ choose $\bar{a} : \Phi^{\max}(\bar{b}) = \Phi(\bar{a}, \bar{b})$

$$\Phi^{\max}(\bar{b}) - \Phi^{\max}(b) \leq \Phi(\bar{a}, \bar{b}) - \Phi(\bar{a}, b) \leq \omega_{\Phi}(\text{dist}_B(b, \bar{b}))$$

$$\rightarrow 0 \text{ as } b \rightarrow \bar{b}$$

Let's invert the roles of b & \bar{b} &

$$\Phi^{\max}(b) - \underline{\Phi}^{\max}(\bar{b}) \rightarrow 0 \text{ as } b \rightarrow \bar{b} . \quad \square$$

Def : Upper value of game = $v^+ := \min_{b \in B} \underline{\Phi}^{\max}(b) =$
 $= \min_b \max_a \underline{\Phi}(a, b).$

Lower value = $v^- := \max_{a \in A} \underline{\Phi}^{a, L}(a) = \max_a \min_b \underline{\Phi}(a, b)$

Prop. $v^- \leq v^+$

Pf $\forall b' \in B \quad \underline{\Phi}(a, b') \geq \min_b \underline{\Phi}(a, b) \quad \forall a$

$\Rightarrow \max_a \underline{\Phi}(a, b') \geq \max_a \min_b \underline{\Phi}(a, b) = v^- \quad \forall b'$

$\Rightarrow \min_{b'} \max_a \underline{\Phi}(a, b') \geq v^-$
 $\leq v^+ \quad \square ,$

Def : If $v^+ = v^-$ then $v = v^+ = v^-$ is THE VALUE
of the GAME.

$$\overbrace{\hspace{10em}}^0 \overbrace{\hspace{10em}}$$

HW. ADDED AFTER THE LECTURE .

Example of LP control . $n=1, m=1$

$y(t)$ = invested capital of a firm

$x(t)$ = investment at time t

Dynamics: $\dot{y}(t) = -\alpha y(t) + \alpha(t)$, $\alpha > 0$ given.

$0 < p$ = price of the product/unit of capital

Problem: $\max_{\alpha(\cdot)} \int_0^T (py^2(s) - r\alpha^2(s)) ds$, $r > 0$ given.

i.e. $\min_{\alpha(\cdot)} J(x, t, \alpha) = \int_t^T (r\alpha^2(s) - py^2(s)) ds$.

N.B.: In the previous notations $H = -p < 0$, so it is NOT known if J sol. of (RT) is $[0, T]$, and not just locally in $(t_0, T]$.

Q: Find values of the parameters st. the sol $k(t)$ of (RT) is defined in $[0, T]$.

(see [Engwerda, p. 184-5]).