# OPTIMAL CONTROL SYSTEMS MIN-MAX OPTIMAL CONTROL

Harry G. Kwatny

Department of Mechanical Engineering & Mechanics Drexel University



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# OUTLINE

Drexel

Optimal Control

### PROBLEM SETUP

Consider the control system

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

where

- $x \in \mathbb{R}^n$  is the state
- $u \in U \subset R^m$  is the control
- $w \in W \subset R^q$  is an external disturbance

We seek a control u that minimizes the cost J for the worst possible disturbance w and

$$J = \ell(x(T)) + \int_{0}^{T} L(x(t), u(t), w(t)) dt$$



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## HJB EQUATION

We seek both  $u^*(t)$  and  $w^*(t)$  such that

$$\{u^*, w^*\} = \arg\min_{u \in U} \max_{w \in W} J(u, w)$$

Define the cost to go

$$V(x,t) = \min_{u \in U} \max_{w \in W} \left( \ell(x(T)) + \int_{t}^{T} L(x(\tau), u(\tau), w(\tau)) d\tau \right)$$

The principle of optimality leads to

$$\frac{\partial V(x,t)}{\partial t} + H^*\left(x,\frac{\partial V(x,t)}{\partial x}\right) = 0$$

where

$$H^{*}(x,\lambda) = \min_{u \in U} \max_{w \in W} \left\{ L(x,u,w) + \lambda^{T} f(x,u,w) \right\}$$



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Optimal Control

# EXAMPLE

Consider the linear dynamics

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

and cost

$$J(u,w) = \lim_{T \to \infty} \frac{1}{2T} \int_0^T \left[ y^T(t) y(t) + \rho u^T(t) u(t) - \gamma^2 w^T(t) w(t) \right] dt$$

Suppose

Then

$$u(t) = -\frac{1}{\rho} B^T S x(t), \quad w(t) = \frac{1}{\gamma^2} E^T S x(t)$$
$$A^T S + S A - S \left(\frac{1}{\rho} B B^T - \frac{1}{\gamma^2} E E^T\right) S = -C^T C$$



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# EXAMPLE CONT'D: SOME CALCULATIONS

$$H^{*} = \min_{u} \max_{w} \left\{ \frac{1}{2} \left( x^{T} C^{T} C x + \rho u^{T} u - \gamma^{2} w^{T} w \right) + \lambda^{T} \left( A x + B u + E w \right) \right\}$$
$$u^{*} = -\frac{1}{\rho} B^{T} \lambda, \quad w^{*} = \frac{1}{\gamma^{2}} E^{T} \lambda$$
$$H^{*} = \frac{1}{2} x^{T} C^{T} C x + \lambda^{T} A x - \frac{1}{2\rho} \lambda^{T} B B^{T} \lambda + \frac{1}{2\gamma^{2}} \lambda^{T} E E^{T} \lambda$$
$$\dot{x} = \frac{\partial H^{*}}{\partial \lambda} = A x + \frac{1}{\gamma^{2}} E E^{T} \lambda - \frac{1}{\rho} B B^{T} \lambda$$
$$\dot{\lambda} = -\frac{\partial H^{*}}{\partial x} = -C^{T} C x + -A^{T} \lambda$$
$$\left[ \begin{array}{c} \dot{x} \\ \dot{\lambda} \end{array} \right] = \left[ \begin{array}{c} A & \frac{1}{\gamma^{2}} E E^{T} - \frac{1}{\rho} B B^{T} \\ -C^{T} C & -A^{T} \end{array} \right] \left[ \begin{array}{c} x \\ \lambda \end{array} \right]$$

#### EXAMPLE CONT'D

Consider the Hamiltonian matrix

$$H = \begin{bmatrix} A & \frac{1}{\gamma^2} E E^T - \frac{1}{\rho} B B^T \\ -C^T C & -A^T \end{bmatrix}$$

Now, under the above assumptions:

- There exists a γ<sub>min</sub> such that so that there are no eigenvalues of H on the imaginary axis provided γ > γ<sub>min</sub>.
- When  $\gamma \to \infty$  we approach the standard LQR solution.
- For  $\gamma = \gamma_{\min}$ , the controller is the full state feedback  $H_{\infty}$  controller.
- All other values of γ produce valid min-max controllers.
- The condition  $\operatorname{Re}\lambda(H) \neq 0$  is equivalent to stability of the matrix

$$A + \frac{1}{\gamma^2} E E^T - \frac{1}{\rho} B B^T$$

The closed loop system matrix is

$$A - \frac{1}{\rho}BB^{T}$$

The term  $EE^T/\gamma^2$  is destabilizing, so the system is guaranteed some margin of stability.



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### TWO PERSON DIFFERENTIAL GAME

Once again consider the system

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

However, now consider both u and w to be controls associated, respectively, with two players A and B. Player A wants to minimize the cost

$$J = \ell(x(T)) + \int_{0}^{T} L(x(t), u(t), w(t)) dt$$

and player *B* wants to maximize it. This is a two person, zero-sum, differential game.



#### DEFINITIONS

Instead of the 'cost to go' define two value functions:

DEFINITION (VALUE FUNCTIONS)

The lower value function is

$$V_{A}(x,t) = \min_{u \in U} \max_{w \in W} \left( \ell(x(T)) + \int_{t}^{T} L(x(\tau), u(\tau), w(\tau)) d\tau \right)$$

and the upper value function is

$$V_{B}(x,t) = \max_{w \in W} \min_{u \in U} \left( \ell(x(T)) + \int_{t}^{T} L(x(\tau), u(\tau), w(\tau)) d\tau \right)$$

- ► if V<sub>A</sub> = V<sub>B</sub>, a unique solution exists. This solution is called a pure strategy.
- ▶ If  $V_A \neq V_B$ , a unique strategy does not exist. In general, the player who 'plays second' has an advantage.



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#### **ISAACS' EQUATIONS**

Define

$$H_{A}(x,\lambda) = \min_{u \in U} \max_{w \in W} \left\{ L(x,u,w) + \lambda^{T} f(x,u,w) \right\}$$
$$H_{B}(x,\lambda) = \min_{u \in U} \max_{w \in W} \left\{ L(x,u,w) + \lambda^{T} f(x,u,w) \right\}$$

Then we get Isaacs' equations

$$\frac{\partial V_A(x,t)}{\partial t} + H_A\left(x, \frac{\partial V_A(x,t)}{\partial x}\right) = 0$$
$$\frac{\partial V_B(x,t)}{\partial t} + H_B\left(x, \frac{\partial V_B(x,t)}{\partial x}\right) = 0$$
$$H_A\left(x, \frac{\partial V_A(x,t)}{\partial x}\right) \equiv H_B\left(x, \frac{\partial V_B(x,t)}{\partial x}\right) \Rightarrow V_A(x,t) \equiv V_B(x,t)$$

and the game is said to satisfy the Isaacs' condition. Notice that if  $u^*, w^*$  are optimal, the the pair  $(u^*, w^*)$  is a saddle point in the sense that

$$J(u^*, w) \le J(u^*, w^*) \le J(u, w^*)$$



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#### Well-stirred Heating Tank Startup

$$\dot{x} = -x + w + u, \quad x(0) = x_0, \quad 0 \le u \le 2, \quad |w| \le 1$$

- x, dimensionless tank temperature
- *u*, dimensionless heat input
- w, dimensionless input water flow

$$J = \int_0^T (x - \bar{x})^2 dt, \quad T \text{ fixed}$$

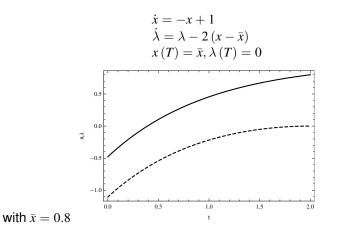
$$H = (x - \bar{x})^2 + \lambda (-x + w + u)$$

$$\{u^*, w^*\} = \arg\min_u \max_w H (x, u, w, \lambda)$$

$$\Rightarrow u^* = (1 - \operatorname{sgn}(\lambda)), w^* = \operatorname{sgn}(\lambda)$$

$$\dot{\lambda} = -\frac{\partial H}{\partial x} \Rightarrow \dot{\lambda} = \lambda - 2 (x - \bar{x}), \quad \lambda (T) = 0$$

## EXAMPLE, CONT'D





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Image: A matrix