

# Optimal Control

Focus so far: stabilization, controllability, etc.

Optimality: accomplishing these tasks in "best possible way" (according to some criterion)

## General setting:

System:  $\dot{x} = f(t, x, u)$ ,  $x(t_0) = x$   $x(t) \in \mathbb{R}^n$   
 $u(t) \in \mathbb{R}^m$   
 $t_0 \leq t \leq t_1$

Control law (input)  $u: [t_0, t_1] \rightarrow \mathbb{R}^m$

Cost:  $J(t_1, t_0, x, u(\cdot)) \leftarrow \min$

- instantaneous cost:  $q(t, x, u)$   $t_0 \leq t < t_1$   
 $x \in \mathbb{R}^n, u \in \mathbb{R}^m$
- terminal cost:  $\phi(x)$ ,  $x \in \mathbb{R}^n$

$$J(t_1, t_0, x, u(\cdot)) := \int_{t_0}^{t_1} q(t, x(t), u(t)) dt + \phi(x(t_1))$$

subject to  $\dot{x}(t) = f(t, x(t), u(t))$   
 $x(t_0) = x$ ,  $t \in [t_0, t_1]$

## Optimal Control Problem:

$$\min J(t_1, t_0, x, u(\cdot)) \text{ over } u(\cdot) \in \mathcal{U},$$

where  $\mathcal{U}$  is a given class of admissible controls.

Example (to keep in mind): Linear Quadratic Regulator (LQR)

$$\dot{x}(t) = A(t)x(t) + B(t)u(t) \quad (\text{LTV})$$

$$q(t, x, u) = u^T R(t)u + x^T Q(t)x \quad R(t) = R(t)^T > 0$$

$$p(x) = x^T S x \quad S = S^T \geq 0, \text{ constant}$$

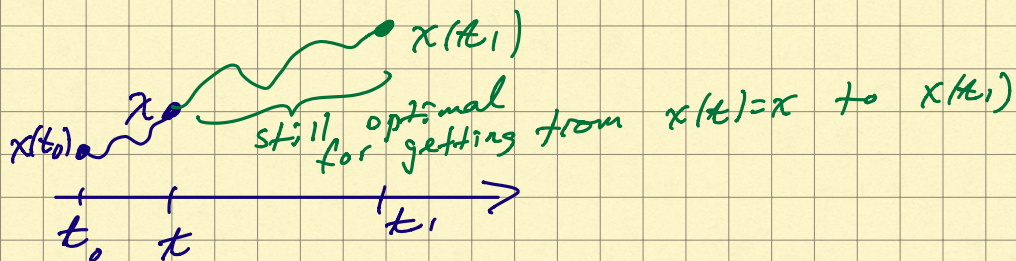
$$Q(t) = Q(t)^T \geq 0$$

Cost to be minimized:

$$J(t_1, t_0, x, u(\cdot)) = \int_{t_0}^{t_1} \{ u(t)^T R(t) u(t) + \lambda(t)^T Q(t) x(t) \} dt + \lambda(t_1)^T S x(t_1)$$

s.t.  $\dot{x}(t) = A(t)x(t) + B(t)u(t)$   
 $x(t_0) = x, \quad t_0 \leq t \leq t_1$

## Dynamic Programming framework (R. Bellman)



Cost-to-go:  $t_0 \leq t \leq t_1, x, u(\cdot)$ :

$$J(t_1, t, x, u(\cdot)) := \int_t^{t_1} q(s, x(s), u(s)) ds + p(x(t_1))$$

s.t.  $\dot{x}(s) = f(s, x(s), u(s)), \quad t \leq s \leq t_1$   
 $x(t) = x$

Bellman fcn (value fcn, optimal cost-to-go):

$$V(t, x), \quad t \in [t_0, t_1], \quad x \in \mathbb{R}^n$$

$$V(t, x) := \inf_{u(\cdot)} J(t_1, t, x, u(\cdot))$$

Assumption: the Bellman fcn is achieved, i.e.

$$\forall t \in [t_0, t_1], \forall x \in \mathbb{R}^n: \exists u^*(\cdot) \text{ s.t. } V(t, x) = J(t_1, t, x, u^*(\cdot))$$

# Dynamic Programming Lemma

Notation: •  $\varphi(s, t, x, u(\cdot))$  for  $t_0 \leq t \leq s \leq t_1$ ,  
 $x \in \mathbb{R}^n$   
 $u(\cdot)$  - adm. control  
 $:= \pi(s)$

$$\text{s.t. } \begin{cases} \dot{x}(r) = f(r, x(r), u(r)) & s \geq r \geq t \\ x(t) = x & \end{cases}$$

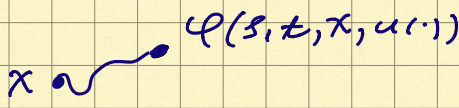
$$\bullet J(s, t, x, u(\cdot)) := \int_t^s g(r, x(r), u(r)) dr$$

starting at  $x(t) = x$

Two key properties of  $V$ :

1) for all  $t_0 \leq t \leq s \leq t_1$ , all  $x \in \mathbb{R}^n$ , all  $u(\cdot)$ ,

$$V(t, x) \leq J(s, t, x, u(\cdot)) + V(s, \varphi(s, t, x, u(\cdot)))$$

$x$  

$t_0 \quad t \quad s \quad t_1$

2) If the Bellman fcn is achieved, then

$$V(t, x) = J(s, t, x, u_{[t, s]}) + V(s, \varphi(s, t, x, u(\cdot))),$$

where  $u(\cdot)$  is an optimal strategy that achieves  $V$ .

Implications: •  $V(t_1, x) = p(x) \quad \forall x \in \mathbb{R}^n$

⋮

$$V(t_0, x) = \inf_{u(\cdot)} J(t_1, t_0, x, u(\cdot))$$

• Verification lemma: any  $\tilde{V}(t, x)$  s.t.

0)  $\tilde{V}(t_1, x) = p(x)$

1)  $\tilde{V}(t, x) \leq J(s, t, x, u(\cdot)) + \tilde{V}(s, \varphi(s, t, x, u(\cdot)))$   
for all  $t_0 \leq t \leq s \leq t_1$ ,  $x, u(\cdot)$

2)  $\exists$  a control  $\bar{u}(\cdot)$  s.t.

$$\tilde{V}(t, x) = Q(s, t, x, \bar{u}(\cdot)) + \tilde{V}(s, \varphi(s, t, x, \bar{u}(\cdot)))$$

for all  $t_0 \leq t \leq s \leq t_1$ ,  $x \in \mathbb{R}^n$ ,

then  $\tilde{V} = V$ , and  $\bar{u}(\cdot)$  is an optimal control.

## Hamilton-Jacobi-Bellman (HJB) equation

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

$$J(t_0, t_1, x, u(\cdot)) = \int_t^{t_1} q(s, x(s), u(s)) ds + p(x(t_1))$$

Let  $u(\cdot)$  be an optimal control; then

$$V(t, x(t)) = \int_t^s q(r, x(r), u(r)) dr + V(s, x(s))$$

$$\frac{d}{ds} V(s, x(s)) = -q(s, x(s), u(s)) \quad \text{[along the optimal trajectory]}$$

$$\frac{d}{ds} V(s, x(s)) = \frac{\partial V}{\partial s}(s, x(s)) + \frac{\partial V}{\partial x}(s, x(s)) \dot{x}(s)$$

$$-q(s, x(s), u(s)) = \frac{\partial V}{\partial s}(s, x(s)) + \frac{\partial V}{\partial x}(s, x(s)) f(s, x(s), u(s))$$

Let  $s \downarrow t$ ,  $x(s) \downarrow x$  (given  $x(t_1) = x$ )

$$\frac{\partial V}{\partial t}(t, x) = -q(t, x, u(t)) - \frac{\partial V}{\partial x}(t, x) f(t, x, u(t))$$

$$t_0 \leq t \leq t_1,$$

$$V(t_1, x) = p(x)$$

Let  $v(\cdot)$  be another admissible control; then

$$V(t, x) \leq \int_t^s q(r, x(r), v(r)) dr + V(s, x(s))$$

$$\text{where } \begin{cases} \dot{x}(r) = f(r, x(r), v(r)) \\ x(t) = x, \quad t \leq r \leq s \end{cases}$$

$$\Rightarrow \frac{d}{dt} V(t, x) \geq -q(t, x, v(t))$$

$$\text{or } \frac{\partial}{\partial t} V(t, x) \geq - \frac{\partial}{\partial x} V(t, x) f(t, x, v(t)) - q(t, x, v(t))$$

for all other  $v(\cdot) \neq u(\cdot)$

Thus, we "obtain" the HJB equation

$$\frac{\partial}{\partial t} V(t, x) = - \min_{u \in \mathbb{R}^m} \left\{ q(t, x, u) + \frac{\partial}{\partial x} V(t, x) f(t, x, u) \right\}$$

for all  $x \in \mathbb{R}^n$   
 $t \in (t_0, t_1]$

s.t.  $V(t_1, x) = p(x)$

Thm If there exists a  $C^1$  fn  $V(t, x)$  that solves the HJB equation, then:

$$V(t_0, x) = \min_{u(\cdot)} J(t_0, t_0, x, u(\cdot))$$

for all  $x \in \mathbb{R}^n$  and the optimal control is given by  $u(t) = k(t, x)$ , where

$$k(t, x) = \operatorname{argmin}_{u \in \mathbb{R}^m} \left\{ q(t, x, u) + \frac{\partial}{\partial x} V(t, x) f(t, x, u) \right\}.$$

Note: implicitly assuming uniqueness:

$$q(t, x, u) + \frac{\partial}{\partial x} V(t, x) f(t, x, u) > q(t, x, k(t, x)) + \frac{\partial}{\partial x} V(t, x) f(t, x, k(t, x))$$

for all  $u \neq k(t, x)$ .

Back to LQR:  $\dot{x}(t) = A(t)x(t) + B(t)u(t)$

$$J(t_1, t_0, x, u(\cdot)) = \int_{t_0}^{t_1} \{u(t)^T R(t)u(t) + x(t)^T Q(t)x(t)\} dt + x(t_1)^T S x(t_1)$$

$$V(t, x) := \min_{u(\cdot)} J(t_1, t, x, u(\cdot))$$

Thm (LQR optimality). Fix  $t_0 < t_1$ . Then (under mild regularity conditions)  $\exists P(t) = P(t)^T$  in  $\mathbb{R}^{n \times n}$  that solves the Riccati Diff. Eq. (RDE)

$$\dot{P} = PBR^{-1}B^T P - PA - A^T P - Q \quad t_0 \leq t \leq t_1, \\ P(t_1) = S$$

Let  $F(t) := -R(t)^{-1}B(t)^T P(t)$ . Then the feedback law  $k(t, x) = F(t)x$  is optimal, and

$$V(t, x) = x^T P(t)x.$$

In particular,  $V(t_0, x) = x^T P(t_0)x$  is the opt. value of the LQR problem.

Proof sketch: next lecture.