Optimal Control Lecture 14

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Note: These slides only cover small part of the lecture. For details and other discussions consult your class notes.
Reading suggestion: Section 5.2 of Ref [1]
For infinite horizon LQR you can study: Section 6.2 of Ref[3] (see the syllabus/class website for the list of the references)".

- Optimal Control
 - Properties of optimal contorl
 - LQR problem for continuous-time systems
 - Dynamic signal tracking in finite-time

Optimal control (Review)

$$\begin{split} u^{\star}(t)\Big|_{t\in[t_0,t_f]} &= \underset{u(t)\in\mathcal{U}}{\operatorname{argmin}}(J=h(x(t_f),t_f)+\int_{t_0}^{t_f}g(x(t),u(t),t))dt, \ s.t.\\ \dot{x}(t) &= a(x(t),u(t),t),\\ x(t_0),\ t_0 \ \text{is given},\\ m(x(t_f),t_f) &= 0 \leftarrow \ \text{when final state is constrained},\\ x(t):\mathbb{R}\to\mathbb{R}^n, \ u(t):\mathbb{R}\to\mathbb{R}^m, \ f:\mathbb{R}^n\times\mathbb{R}^m\times\mathbb{R}\to\mathbb{R}^n. \end{split}$$

• Hamiltonian $H(x, u, p, t) = g(x(t), u(t), t) + p(t)^{\top} a(x(t), u(t), t)$,

first order conditions for extremal solution

$$\begin{split} \dot{p} &= -H_x, & (n \text{ dimensional}) \\ 0 &= H_u, & (m \text{ dimensional}) \\ \dot{x} &= H_p: \quad \dot{x} = a(x, u, t), & (n \text{ dimensional}) \\ \hline x(t_0) &= x_0 & \\ \text{if } t_f \text{ free: } \left. \frac{\partial h}{\partial t} \right|_{t_f} + H(t_f) = 0 & \\ \text{if } x_i(t_f) \text{ is fixed: } x_i(t_f) &= x_{i_f} \\ \text{if } x_i(t_f) \text{ is free: } p_i(t_f) &= \frac{\partial h}{\partial x_i}(t_f) & \\ \text{o } \text{ if } t_f \text{ free: } \left. \frac{\partial w}{\partial t} \right|_{t_f} + H(t_f) = 0 & \\ \text{o } \text{ since } x(t_f) \text{ is not directly given we need} \\ p(t_f) &= \frac{\partial w}{\partial x}(t_f) \\ \text{o } \text{ if } t_f \text{ free: } \left. \frac{\partial w}{\partial t} \right|_{t_f} + H(t_f) = 0 & \\ \end{split}$$

Properties of Hamiltonian

$$\begin{split} u^{\star}(t)\Big|_{t\in[t_0,t_f]} &= \underset{u(t)\in\mathcal{U}}{\operatorname{argmin}}(J = h(x(t_f),t_f) + \int_{t_0}^{t_f} g(x(t),u(t),t))dt, \ s.t.\\ \dot{x}(t) &= a(x(t),u(t),t),\\ x(t_0), \ t_0 \ \text{is given},\\ x(t_f), \ t_f \text{various final state constrained},\\ x(t) &: \mathbb{R} \to \mathbb{R}^n, \quad u(t) : \mathbb{R} \to \mathbb{R}^m, \quad f: \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \to \mathbb{R}^n, \ C: \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R} \to \mathbb{R}^q. \end{split}$$

$$\mathbf{H} = g(\mathbf{x}(t), \mathbf{u}(t), t)) + p(t)^{\top} \mathbf{a}(\mathbf{x}(t), \mathbf{u}(t), t)$$

 If g(x, u) and a(x, u) do not explicitly depend on time t, then the Hamiltonian H is at least piece-wise contract.

$$\frac{dH}{dt} = H_u \dot{u}$$

 \bullet The third necessary condition is $H_{\mathfrak{u}}=0$ so

$$\frac{\mathrm{dH}}{\mathrm{dt}} = (0)\dot{\mathrm{u}} = 0$$

which suggest H is constant

- It might be possible for the value of this constant to change at a discontinuity of u, since then \dot{u} would be infinite, and $0.\infty$ is not defined.
- This H is at least piece-wise constant.
- For free final time problems, the transversality condition gives

$$h_t + H(t_f) = 0$$

- If h is not function of time then $h_t = 0$ and as a result $H(t_f) = 0$.
- \bullet with no jumps in u, H is constant: H(t)=0 for all $t\in [t_0.t_f].$

Finite time LQR problem

$$\begin{split} u^{\star}(t)\Big|_{t\in[t_0,t_f]} &= \underset{u(t)\in\mathcal{U}}{\text{argmin}}(J = \frac{1}{2}x(t_f)^{\top}\mathsf{H}^f x(t_f) + \frac{1}{2}\int_0^{t_f}(x^{\top}Q(t)x + u^{\top}\mathsf{R}(t)u)dt, \ \text{ s.t.} \\ \dot{x}(t) &= A(t)x + B(t)u, \\ x(t_0), \ t_0 \ \text{is given}, \\ x(t_f) \ \text{is free} \\ x(t) : \mathbb{R} \to \mathbb{R}^n, \quad u(t) : \mathbb{R} \to \mathbb{R}^m, \quad Q(t) \ge 0, \ \mathsf{H}^f \ge 0, \ \mathsf{R}(t) > 0. \end{split}$$

Hamiltonian

$H(x, u, p, t) = \frac{1}{2} \Big(x(t)^{\top} Q(t) x(t) + u^{\top}(t) R(t) u(t) \Big) + p(t)^{\top} (A(t) x(t) + B(t) u(t)),$ FONC:

•
$$\dot{x} = H_p \Rightarrow \dot{x} = A(t)x + B(t)u$$

•
$$\dot{p} = -H_x \Rightarrow \dot{p} = -Q(t)x - A(t)^\top p$$
,

- $0 = H_u \Rightarrow R(t)u(t) + B(t)^\top p(t) = 0 \Rightarrow u^{\star}(t) = -R(t)^{-1}B(t)^\top p(t)^{\star}$
- Boundary condition

$$p(t_f) = \frac{\partial \frac{1}{2} x^\top H^f x}{\partial x}|_{t_f} = H^f x(t_f)$$

with BC

$$\begin{bmatrix} \dot{x} \\ \dot{p} \end{bmatrix} = \begin{bmatrix} A(t) & -B(t)R(t)^{-1}B(t)^{\top} \\ -Q(t) & -A(t)^{\top} \end{bmatrix} \begin{bmatrix} x \\ p \end{bmatrix}$$

$$\begin{cases} x(0) = x_0, \\ p(t_f) = H^f x(t_f) \\ \\ \hline p(t_f) \end{bmatrix} = \underbrace{\begin{bmatrix} \varphi_{11}(t_f, t) & \varphi_{12}(t_f, t) \\ \varphi_{21}(t_f, t) & \varphi_{22}(t_f, t) \end{bmatrix}}_{\varphi(t_f, t)} \begin{bmatrix} x(t) \\ p(t) \end{bmatrix}$$

$$\begin{split} p^{\star}(t) = \underbrace{\left(\varphi_{22}(t_{f},t) - H^{f} \varphi_{12}(t_{f},t) \right)^{-1} \left(H^{f} \varphi_{11}(t_{f},t) - \varphi_{21}(t_{f},t) \right)}_{K(t)} x^{\star}(t) \\ p^{\star}(t) = K(t) x^{\star}(t) \end{split}$$

$$p^{\star}(t) = K(t)x^{\star}(t)$$
$$u^{\star}(t) = -R(t)^{-1}B(t)^{\top}p^{\star}(t)$$
$$u^{\star} = -R(t)^{-1}B(t)^{\top}K(t)x^{\star}(t)$$

$$K(t) = \left(\varphi_{22}(t_{f},t) - H^{f}\varphi_{12}(t_{f},t)\right)^{-1} \left(H^{f}\varphi_{11}(t_{f},t) - \varphi_{21}(t_{f},t)\right)$$

or

Riccati type differential equation: $\dot{p}(t)=\dot{K}(t)x(t)+K(t)\dot{x}(t)\Rightarrow$

$$\begin{aligned} & \left(\dot{K}(t) = -K(t)A(t) - A^{\top}(t)K(t) - Q(t) + K(t)B(t)R^{-1}(t)B^{\top}(t)K(t), \\ & \left(K(t_f) = H^f. \right. \end{aligned}$$

- Optimal tracking control is a state feedback control
- The state feedback gain is time-varying
- Optimal cost to go $J_{i \to N}(x(t_i)) = \frac{1}{2}x(t_i)^\top K(t_i)x(t_i).$

For LTI systems, Kalman has shown that the Riccati equation has a steady state solution if (see Kirk)

- A, B, R, Q are constant
- H^f = 0,
- (A, B) is controllable,

SS Riccati equation is $A^{\top}K + KA + Q - KBR^{-1}B^{\top}K = 0$

- You can find the SS feedback using F = lqr(A, B, Q, R) in Matlab. For further information see https://www.mathworks.com/help/control/ref/lqr.html

Finite time linear tracking problem

Hamiltonian

$$H(x, u, p, t) = \frac{1}{2} \Big((x(t) - r(t))^{\top} Q(t)(x(t) - r(t)) + u^{\top}(t) R(t) u(t) \Big) + p(t)^{\top} (A(t)x(t) + B(t)u(t)),$$

FONC:

•
$$\dot{\mathbf{x}} = \mathbf{H}_{\mathbf{p}} \Rightarrow \dot{\mathbf{x}} = \mathbf{A}(\mathbf{t})\mathbf{x} + \mathbf{B}(\mathbf{t})\mathbf{u}$$

•
$$\dot{\mathbf{p}} = -\mathbf{H}_{\mathbf{x}} \Rightarrow \dot{\mathbf{p}} = -\mathbf{Q}(\mathbf{t})\mathbf{x} - \mathbf{A}(\mathbf{t})^{\top}\mathbf{p}$$
,

• $0 = H_u \Rightarrow R(t)u(t) + B(t)^\top p(t) = 0 \Rightarrow u^{\star}(t) = -R(t)^{-1}B(t)^\top p(t)^{\star}$

• Boundary condition: $p(t_f) = \frac{\partial \frac{1}{2} (x - r(t_f))^\top H^f(x - r(t_f))}{\partial x}|_{t_f} = H^f(x(t_f) - r(t_f))$

$$\begin{bmatrix} \dot{x} \\ \dot{p} \end{bmatrix} = \begin{bmatrix} A(t) & -B(t)R(t)^{-1}B(t)^{\top} \\ -Q(t) & -A(t)^{\top} \end{bmatrix} \begin{bmatrix} x \\ p \end{bmatrix} + \begin{bmatrix} 0 \\ Q(t)r(t) \end{bmatrix}$$
with BC
$$\begin{cases} x(0) = x_0, \\ p(t_f) = H^f x(t_f) - H^f r(t_f) \\ \hline p(t_f) \end{bmatrix} = \underbrace{\begin{bmatrix} \varphi_{11}(t_f, t) & \varphi_{12}(t_f, t) \\ \varphi_{21}(t_f, t) & \varphi_{22}(t_f, t) \end{bmatrix} \\ \downarrow \phi(t_f, t) \end{bmatrix} \begin{bmatrix} x(t) \\ p(t) \end{bmatrix} + \underbrace{\int_{t_0}^{t_f} \varphi(t_f, \tau) \begin{bmatrix} 0 \\ Q(\tau)r(\tau) \end{bmatrix} d\tau} \\ \begin{bmatrix} f_1(t) \\ f_2(t) \end{bmatrix}$$

$$+ p(t_f) = H^f x(t_f) - H^f r(t_f)$$

$$\downarrow$$

$$p^*(t) = K(t)x^*(t) + s(t)$$

$$p^{\star}(t) = K(t)x^{\star}(t) + s(t)$$

$$\mathbf{u}^{\star}(t) = -\mathbf{R}(t)^{-1}\mathbf{B}(t)^{\top}\mathbf{p}(t)^{\star}$$

 $\mathbf{u}^{\star} = -\mathbf{R}(t)^{-1}\mathbf{B}(t)^{\top}\mathbf{K}(t) \, \mathbf{x}^{\star}(t) - \mathbf{R}(t)^{-1}\mathbf{B}(t)^{\top}\mathbf{s}(t) = \mathbf{F}(t)\mathbf{x}^{\star}(t) + \mathbf{v}(t)$

where K(t) and s(t) are computed from (use $\dot{p}(t)=\dot{K}(t)x(t)+K(t)\dot{x}(t)+\dot{s}(t))$

$$\begin{cases} \dot{\mathsf{K}} = -\mathsf{K}(t)\mathsf{A}(t) - \mathsf{A}^{\top}(t)\mathsf{K}(t) - Q(t) + \mathsf{K}(t)\mathsf{B}(t)\mathsf{R}^{-1}(t)\mathsf{B}^{\top}(t)\mathsf{K}(t), & \mathsf{K}(t_{\mathsf{f}}) = \mathsf{H}^{\mathsf{f}}, \\ \dot{\mathsf{s}} = -(\mathsf{A}(t)^{\top} - \mathsf{K}(t)\mathsf{B}(t)\mathsf{R}(t)^{-1}\mathsf{B}(t)^{\top})\mathsf{s}(t) + Q(t)\mathsf{r}(t), & \mathsf{s}(t_{\mathsf{f}}) = -\mathsf{H}^{\mathsf{f}}\mathsf{r}(t_{\mathsf{f}}) \end{cases}$$