

# Optimal Control

## Lecture 4

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Reading: Sections 2.1 and 2.2 from Ref.[2]

Optimal control of multi-stage systems over finite horizon

- first-order conditions for free and constrained final state
- Special case of linear discrete-time systems

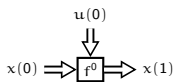
# Optimal control and its connection to constrained optimization

$$\mathbf{u}^* = \underset{\mathbf{u} \in \mathbb{R}^m}{\operatorname{argmin}} F(\mathbf{x}, \mathbf{u}), \quad \text{s.t.},$$
$$f(\mathbf{x}, \mathbf{u}) = 0$$

where  $F: \mathbb{R}^{m+n} \rightarrow \mathbb{R}$  and  $f: \mathbb{R}^{m+n} \rightarrow \mathbb{R}^n$  are differentiable.

## Optimal Control Example

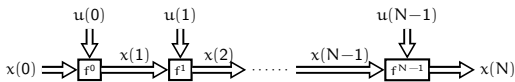
### Single stage system



$$\mathbf{u}^*(0) = \underset{\mathbf{u}(0)}{\operatorname{argmin}} \underbrace{\phi(\mathbf{x}(1)) + L^0(\mathbf{x}(0), \mathbf{u}(0))}_{J(\mathbf{u}(0))}$$

$$\text{s.t. } \mathbf{x}(1) = f^0(\mathbf{x}(0), \mathbf{u}(0)),$$
$$\mathbf{x}(0) = \mathbf{x}_0 \in \mathbb{R}^n.$$

### Multi stage system



$$\mathbf{u}^* = \underset{\mathbf{u}(0), \dots, \mathbf{u}(N-1)}{\operatorname{argmin}} \underbrace{\phi(\mathbf{x}(N)) + \sum_{k=0}^{N-1} L^k(\mathbf{x}(k), \mathbf{u}(k))}_{J(\mathbf{u}(0), \dots, \mathbf{u}(N-1))} \quad \text{s.t.}$$

$$\mathbf{x}(N) = f^{N-1}(\mathbf{x}(N-1), \mathbf{u}(N-1)),$$

$$\vdots$$

$$\mathbf{x}(1) = f^0(\mathbf{x}(0), \mathbf{u}(0)),$$

$$\mathbf{x}(0) = \mathbf{x}_0 \in \mathbb{R}^n.$$

# First order optimality condition for single stage optimal control

$$\begin{aligned} \mathbf{u}(0)^* &= \underset{\mathbf{u}(0) \in \mathbb{R}^m}{\operatorname{argmin}} J(\mathbf{x}(1), \mathbf{u}(0)) = \phi(\mathbf{x}(1)) + L^0(\mathbf{x}(0), \mathbf{u}(0)), \quad \text{s.t.}, \\ \mathbf{x}(1) &= \mathbf{f}^0(\mathbf{x}(0), \mathbf{u}(0)), \quad \mathbf{x}(1) \in \mathbb{R}^n, \quad \mathbf{u}(0) \in \mathbb{R}^m, \\ \mathbf{x}(0) &= \mathbf{x}_0 \in \mathbb{R}^n, \quad (\text{given initial condition}). \end{aligned}$$

- $\bar{J} = J + \lambda(1)^\top (\mathbf{f}^0(\mathbf{x}(0), \mathbf{u}(0)) - \mathbf{x}(1)) = \phi(\mathbf{x}(1)) + L^0(\mathbf{x}(0), \mathbf{u}(0)) + \lambda(1)^\top (\mathbf{f}^0(\mathbf{x}(0), \mathbf{u}(0)) - \mathbf{x}(1))$
- Let  $H^0(\mathbf{x}(0), \mathbf{u}(0), \lambda(1)) = L^0(\mathbf{x}(0), \mathbf{u}(0)) + \lambda(1)^\top (\mathbf{f}^0(\mathbf{x}(0), \mathbf{u}(0)) - \mathbf{x}(1))$ .
- Then, we can rewrite  $\bar{J}$  as  $\bar{J} = (\phi(\mathbf{x}(1)) - \lambda(1)^\top \mathbf{x}(1)) + H^0(\mathbf{x}(0), \mathbf{u}(0), \lambda(1))$ .

First order analysis:

$$\begin{aligned} \bar{J}(\mathbf{x}(1) + d\mathbf{x}(1), \mathbf{u}(0) + d\mathbf{u}(0)) &= \bar{J}(\mathbf{x}(1), \mathbf{u}(0)) + \\ &\underbrace{\left( \frac{\partial \phi(\mathbf{x}(1))}{\partial \mathbf{x}(1)} - \lambda(1)^\top \right) d\mathbf{x}(1) + \left( \frac{\partial H^0}{\partial \mathbf{x}(0)} \right)^\top d\mathbf{x}(0) + \left( \frac{\partial H^0}{\partial \mathbf{u}(0)} \right)^\top d\mathbf{u}(0)}_{d\bar{J}} \end{aligned}$$

Here  $d\mathbf{x}(0) = 0$  because the initial condition is given (no need for variation). Think of  $d\mathbf{u}(0)$  as free variable and  $d\mathbf{x}(1)$  the dependent variable, which is defined from the constraint equation (constraint equation relates  $d\mathbf{x}(1)$  to  $d\mathbf{u}(0)$ ). Next, pick  $\lambda(1)$  such that

$$\frac{\partial \phi(\mathbf{x}(1))}{\partial \mathbf{x}(1)} - \lambda(1) = 0,$$

which gives us  $\bar{J}(\mathbf{x}(1) + d\mathbf{x}(1), \mathbf{u}(0) + d\mathbf{u}(0)) = \bar{J}(\mathbf{x}(1), \mathbf{u}(0)) + \underbrace{\left( \frac{\partial H^0}{\partial \mathbf{u}(0)} \right)^\top d\mathbf{u}(0)}_{d\bar{J}}$ .

## First order optimality condition for single stage optimal control (cont'd)

- For  $(x(1), u(0))$  to be a minimum point we need  $d\bar{J} = \left(\frac{\partial H^0}{\partial u(0)}\right)^\top du(0) \geq 0$ . Because we are free to vary  $du(0)$  in all directions, then the necessary condition for  $(x(1), u(0))$  to be a minimum point is  $\frac{\partial H^0}{\partial u(0)} = 0$ .

- To summarize:

First order necessary condition for  $(x(1), u(0))$  to be a minimum point:

$$\begin{cases} \lambda(1) = \frac{\partial \phi(x(1))}{\partial x(1)}, & n \text{ eq} \\ \frac{\partial H^0}{\partial u(0)} = 0, & m \text{ eq} \\ x(1) = f^0(x(0), u(0)), & n \text{ eq}^1 \end{cases}$$

Here, we have  $2n + m$  equation for  $2n + m$  unknowns (the unknowns in the set of equations above are  $\lambda(1) \in \mathbb{R}^n$ ,  $x(1) \in \mathbb{R}^n$  and  $u(0) \in \mathbb{R}^m$ ).

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<sup>1</sup>which can also be written as  $x(1) = \frac{\partial H^0}{\partial \lambda(1)}$

# First order optimality condition for multi-stage optimal control

$$\begin{aligned}(\mathbf{u}^*(0), \dots, \mathbf{u}^*(N-1)) &= \underset{(\mathbf{u}(0) \in \mathbb{R}^m, \dots, \mathbf{u}(N-1) \in \mathbb{R}^m)}{\operatorname{argmin}} J = \phi(\mathbf{x}(N)) + \sum_{i=0}^{N-1} L^i(\mathbf{x}(i), \mathbf{u}(i)), \quad \text{s.t.}, \\ \mathbf{x}(1) &= f^0(\mathbf{x}(0), \mathbf{u}(0)), \quad \mathbf{x}(1) \in \mathbb{R}^n, \quad \mathbf{u}(0) \in \mathbb{R}^m, \\ &\vdots \\ \mathbf{x}(N) &= f^{N-1}(\mathbf{x}(N-1), \mathbf{u}(N-1)), \quad \mathbf{x}(N) \in \mathbb{R}^n, \quad \mathbf{u}(N-1) \in \mathbb{R}^m, \\ \mathbf{x}(0) &= \mathbf{x}_0 \in \mathbb{R}^n, \quad (\text{given initial condition}).\end{aligned}$$

- $\bar{J} = J + \lambda_1^\top (f^0(\mathbf{x}_0, \mathbf{u}_0) - \mathbf{x}_1) + \dots + \lambda_N^\top (f^{N-1}(\mathbf{x}_{N-1}, \mathbf{u}_{N-1}) - \mathbf{x}_N) = \phi(\mathbf{x}_N) + \sum_{i=0}^{N-1} (L^i(\mathbf{x}_i, \mathbf{u}_i) + \lambda_{i+1}^\top (f^i(\mathbf{x}_i, \mathbf{u}_i) - \mathbf{x}_{i+1}))$
- Let  $H^i(\mathbf{x}_i, \mathbf{u}_i, \lambda_{i+1}) = L^i(\mathbf{x}_i, \mathbf{u}_i) + \lambda_{i+1}^\top (f^i(\mathbf{x}_i, \mathbf{u}_i) - \mathbf{x}_{i+1})$ .
- Then, we can rewrite  $\bar{J}$  as  $\bar{J} = (\phi(\mathbf{x}_N) - \lambda_N^\top \mathbf{x}_N) + \sum_{i=1}^{N-1} (H^i(\mathbf{x}_i, \mathbf{u}_i) - \lambda_i^\top \mathbf{x}_i) + H^0$ .

First order analysis: Here  $d\mathbf{x}(0) = 0$  because the initial condition is given (no need for variation).

$$\begin{aligned}d\bar{J} &= \left( \frac{\partial \phi(\mathbf{x}_N)}{\partial \mathbf{x}_N} - \lambda_N \right)^\top d\mathbf{x}_N + \sum_{i=1}^{N-1} \left( \left( \frac{\partial H^i}{\partial \mathbf{x}_i} \right)^\top d\mathbf{x}_i + \left( \frac{\partial H^i}{\partial \mathbf{u}_i} \right)^\top d\mathbf{u}_i - \lambda_i^\top d\mathbf{x}_i \right) + \left( \frac{\partial H^0}{\partial \mathbf{x}_0} \right)^\top d\mathbf{x}_0 + \left( \frac{\partial H^0}{\partial \mathbf{u}_0} \right)^\top d\mathbf{u}_0 \\ &= \left( \frac{\partial \phi(\mathbf{x}_N)}{\partial \mathbf{x}_N} - \lambda_N \right)^\top d\mathbf{x}_N + \sum_{i=1}^{N-1} \left( \left( \frac{\partial H^i}{\partial \mathbf{x}_i} - \lambda_i \right)^\top d\mathbf{x}_i + \left( \frac{\partial H^i}{\partial \mathbf{u}_i} \right)^\top d\mathbf{u}_i \right) + \left( \frac{\partial H^0}{\partial \mathbf{u}_0} \right)^\top d\mathbf{u}_0\end{aligned}$$

# First order optimality condition for multi-stage optimal control (cont'd)

- Think of  $du(i)$ ,  $i = 0, \dots, N-1$  as free variable and  $x(i+1)$  the dependent variable, which is defined from the constraint equation (constraint equation relates  $x(i+1)$  to  $du(i)$ ).

- Next, pick  $\lambda_N$  such that 
$$\frac{\partial \phi(x_N)}{\partial x_N} - \lambda_N = 0,$$

- also pick  $\lambda_i$ , such that 
$$\frac{\partial H^i}{\partial x_i} - \lambda_i = 0, \quad i = 1, \dots, N-1.$$

- Then, we have

$$d\bar{J} = \sum_{i=1}^{N-1} \left( \left( \frac{\partial H^i}{\partial u_i} \right)^T du_i \right) + \left( \frac{\partial H^0}{\partial u_0} \right)^T du_0$$

For  $(u(0), \dots, u(N-1), x(1), \dots, x(N))$  to be a minimum point we need

$$d\bar{J} = \sum_{i=1}^{N-1} \left( \left( \frac{\partial H^i}{\partial u_i} \right)^T du_i \right) + \left( \frac{\partial H^0}{\partial u_0} \right)^T du_0 \geq 0. \text{ Because we are free to vary } du_i,$$

$i = 0, \dots, N-1$  in all directions, then the necessary condition for

$(u(0), \dots, u(N-1), x(1), \dots, x(N))$  to be a minimum point is  $\frac{\partial H^i}{\partial u_i} = 0$ ,  $i = 0, \dots, N-1$ .

- Putting all the conditions we stated and derived, we obtain:

First order necessary condition for  $(x(1), u(0))$  to be a minimum point:

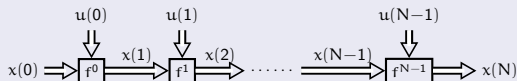
$$\begin{cases} \lambda_N = \frac{\partial \phi(x_N)}{\partial x_N}, & n \text{ eq} \\ \lambda_i = \frac{\partial H^i}{\partial x_i} = 0, \quad i = 1, \dots, N-1, & (N-1)n \text{ eq} \\ \frac{\partial H^i}{\partial u_i} = 0, \quad i = 0, \dots, N-1 & Nm \text{ eq} \\ x_{i+1} = f^i(x_i, u_i), \quad i = 0, \dots, N-1 & Nn \text{ eq}^2 \end{cases}$$

Here, we have  $2Nn + Nm$  equation for  $2Nn + m$  unknowns (the unknowns in the set of equations above are  $\lambda_i \in \mathbb{R}^n$ ,  $x_i \in \mathbb{R}^n$  and  $u_{i-1} \in \mathbb{R}^m$ ,  $i = 1, \dots, N$ ).

<sup>2</sup>which can also be written as  $x_{i+1} = \frac{\partial H^i}{\partial \lambda_{i+1}}$

# Optimal control of multi-stage systems over finite horizon

$$u^* = \operatorname{argmin} \underbrace{\phi(x(N)) + \sum_{k=0}^{N-1} L^k(x(k), u(k))}_{J(u(0), \dots, u(N-1))} \quad \text{s.t.}$$



$$H^k = L^k(x(k), u(k)) + \lambda(k+1)^T f^k(x(k), u(k)), \quad k = 0, 1, \dots, N-1$$

Free final state

Constrained final state, i.e.,

$$\psi(x(N)) = 0, \quad \psi: \mathbb{R}^n \rightarrow \mathbb{R}^p, \quad p \leq n$$

$$\lambda(N) = \frac{\partial \phi(x(N))}{\partial x(N)},$$

$$\psi(x(N)) = 0,$$

$$\lambda(k) = \frac{\partial H^k}{\partial x(k)}, \quad k = 1, \dots, N-1,$$

$$\lambda(N) = \frac{\partial (\phi(x(N)) + \sum_{i=1}^p v_i \psi_i(x(N)))}{\partial x(N)},$$

$$0 = \frac{\partial H^k}{\partial u(k)}, \quad k = 0, \dots, N-1,$$

$$\lambda(k) = \frac{\partial H^k}{\partial x(k)}, \quad k = 1, \dots, N-1,$$

$$x(k+1) = \frac{\partial H^k}{\partial \lambda(k+1)}, \quad k = 1, \dots, N-1,$$

$$0 = \frac{\partial H^k}{\partial u(k)}, \quad k = 0, \dots, N-1,$$

$$\begin{cases} x(0) = x_0, & \text{given initial condition,} \\ 0 = \frac{\partial H^0}{\partial x(0)}, & \text{free initial condition.} \end{cases}$$

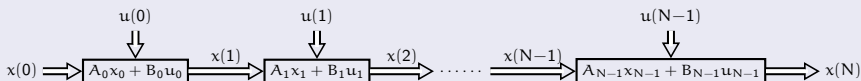
$$x(k+1) = \frac{\partial H^k}{\partial \lambda(k+1)}, \quad k = 1, \dots, N-1,$$

$$\begin{cases} x(0) = x_0, & \text{given initial condition,} \\ 0 = \frac{\partial H^0}{\partial x(0)}, & \text{free initial condition.} \end{cases}$$



## Optimal control of multi-stage systems over finite horizon: regulator problem

$$\mathbf{u}^* = \operatorname{argmin} \frac{1}{2} \mathbf{x}_N^\top \mathbf{S}_N \mathbf{x}_N + \frac{1}{2} \sum_{k=0}^{N-1} \mathbf{x}_k^\top \mathbf{Q}_k \mathbf{x}_k + \mathbf{u}_k^\top \mathbf{R}_k \mathbf{u}_k \quad \text{s.t.}$$



$$\mathbf{H}^k = \frac{1}{2} \mathbf{x}_k^\top \mathbf{Q}_k \mathbf{x}_k + \frac{1}{2} \mathbf{u}_k^\top \mathbf{R}_k \mathbf{u}_k + \lambda_{k+1}^\top (\mathbf{A}_k \mathbf{x}_k + \mathbf{B}_k \mathbf{u}_k), \quad k = 0, 1, \dots, N-1$$

Free final state: Linear systems with given initial condition

$$\lambda(N) = \frac{\partial \phi(\mathbf{x}(N))}{\partial \mathbf{x}(N)} \quad \Rightarrow \quad \lambda_N = \mathbf{S}_N \mathbf{x}_N,$$

$$\lambda(k) = \frac{\partial \mathbf{H}^k}{\partial \mathbf{x}(k)}, \quad k = 1, \dots, N-1 \quad \Rightarrow \quad \lambda_k = \mathbf{Q}_k \mathbf{x}_k + \mathbf{A}_k^\top \lambda_{k+1}, \quad k = 1, \dots, N,$$

$$0 = \frac{\partial \mathbf{H}^k}{\partial \mathbf{u}(k)}, \quad k = 0, \dots, N-1 \quad \Rightarrow \quad 0 = \mathbf{R}_k \mathbf{u}_k + \mathbf{B}_k^\top \lambda_{k+1}, \quad k = 0, \dots, N-1,$$

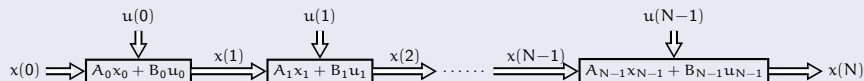
$$\mathbf{x}(k+1) = \frac{\partial \mathbf{H}^k}{\partial \lambda(k+1)}$$

$$= \mathbf{f}^k(\mathbf{x}(k), \mathbf{u}(k)), \quad k = 0, \dots, N-1 \quad \Rightarrow \quad \mathbf{x}_{k+1} = \mathbf{A}_k \mathbf{x}_k + \mathbf{B}_k \mathbf{u}_k, \quad k = 1, \dots, N-1,$$

$$\mathbf{x}(0) = \mathbf{x}_0 \quad \Rightarrow \quad \mathbf{x}(0) = \mathbf{x}_0.$$

## Optimal control of multi-stage systems over finite horizon: regulator problem

$$\mathbf{u}^* = \operatorname{argmin} \frac{1}{2} \mathbf{x}_N^\top \mathbf{S}_N \mathbf{x}_N + \frac{1}{2} \sum_{k=0}^{N-1} \mathbf{x}_k^\top \mathbf{Q}_k \mathbf{x}_k + \mathbf{u}_k^\top \mathbf{R}_k \mathbf{u}_k \quad \text{s.t.}$$



$$\lambda_N = \mathbf{S}_N \mathbf{x}_N,$$

$$\lambda_k = \mathbf{Q}_k \mathbf{x}_k + \mathbf{A}_k^\top \lambda_{k+1}, \quad k = 1, \dots, N-1,$$

$$\mathbf{0} = \mathbf{R}_k \mathbf{u}_k + \mathbf{B}_k^\top \lambda_{k+1}, \Rightarrow \mathbf{u}^* = -\mathbf{R}_k^{-1} \mathbf{B}_k^\top \lambda_{k+1}, \quad k = 0, \dots, N-1$$

$$\mathbf{x}_{k+1} = \mathbf{A}_k \mathbf{x}_k + \mathbf{B}_k \mathbf{u}_k, \Rightarrow \mathbf{x}_{k+1} = \mathbf{A}_k \mathbf{x}_k - \mathbf{B}_k \mathbf{R}_k^{-1} \mathbf{B}_k^\top \lambda_{k+1} \quad k = 1, \dots, N-1,$$

$$\mathbf{x}(0) = \mathbf{x}_0.$$

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$$\begin{bmatrix} \mathbf{x}(k+1) \\ \lambda(k) \end{bmatrix} = \begin{bmatrix} \mathbf{A}_k & -\mathbf{B}_k \mathbf{R}_k^{-1} \mathbf{B}_k^\top \\ \mathbf{Q}_k & \mathbf{A}_k^\top \end{bmatrix} \begin{bmatrix} \mathbf{x}(k) \\ \lambda(k+1) \end{bmatrix}, \quad \mathbf{x}(0) = \mathbf{x}_0, \lambda_N = \mathbf{S}_N \mathbf{x}_N.$$

If  $\mathbf{A}_k$  is invertible:  $\mathbf{x}_k = \mathbf{A}_k^{-1} \mathbf{x}_{k+1} + \mathbf{A}_k^{-1} \mathbf{B}_k \mathbf{R}_k^{-1} \mathbf{B}_k^\top \lambda_{k+1}$ . Then, we can write

$$\begin{bmatrix} \mathbf{x}(k) \\ \lambda(k) \end{bmatrix} = \begin{bmatrix} \mathbf{A}_k^{-1} & \mathbf{A}_k^{-1} \mathbf{B}_k \mathbf{R}_k^{-1} \mathbf{B}_k^\top \\ \mathbf{Q}_k \mathbf{A}_k^{-1} & \mathbf{A}_k^\top + \mathbf{Q}_k \mathbf{A}_k^{-1} \mathbf{B}_k \mathbf{R}_k^{-1} \mathbf{B}_k^\top \end{bmatrix} \begin{bmatrix} \mathbf{x}(k+1) \\ \lambda(k+1) \end{bmatrix}, \quad \mathbf{x}(0) = \mathbf{x}_0, \lambda_N = \mathbf{S}_N \mathbf{x}_N.$$

If we had  $\mathbf{x}_N$  and  $\lambda_N$ , we could solve the equation above backward in time, but unfortunately we have  $\mathbf{x}_0$  and  $\lambda_N$ .

# Optimal control of multi-stage systems over finite horizon

## Minimum energy control for linear LTI systems with fix final state

$$\mathbf{u}^* = \operatorname{argmin} \frac{1}{2} \sum_{k=0}^{N-1} \mathbf{u}_k^T \mathbf{R}_k \mathbf{u}_k, \text{ s.t.}$$

$$\mathbf{x}_{k+1} = \mathbf{A} \mathbf{x}_k + \mathbf{B} \mathbf{u}_k,$$

$$\mathbf{x}_0 = \mathbf{x}_0, \quad \mathbf{x}_N = \mathbf{r}_N.$$

$$\mathbf{u}_k^* = \mathbf{R}^{-1} \mathbf{B}^T (\mathbf{A}^T)^{N-k-1} \mathbf{G}_{0,N}^{-1} (\mathbf{r}_N - \mathbf{A}^N \mathbf{x}_0).$$

where

$$\begin{aligned} \mathbf{G}_{0,N} &= \sum_{i=0}^{N-1} \mathbf{A}^{N-i-1} \mathbf{B} \mathbf{R}^{-1} \mathbf{B}^T (\mathbf{A}^T)^{N-i-1} \\ &= \underbrace{[\mathbf{B} \quad \mathbf{A}\mathbf{B} \quad \dots \quad \mathbf{A}^{N-1}\mathbf{B}]}_{\mathbf{U}_N} \begin{bmatrix} \mathbf{R}^{-1} & & 0 \\ & \ddots & \\ 0 & & \mathbf{R}^{-1} \end{bmatrix} [\mathbf{B} \quad \mathbf{A}\mathbf{B} \quad \dots \quad \mathbf{A}^{N-1}\mathbf{B}]^T \end{aligned}$$

- if  $|\mathbf{R}| \neq 0$ , solution exists ( $\mathbf{G}_{0,N}$  is invertible) if system is reachable ( $\mathbf{U}_N$  is full rank)
  - $N \geq n$  (recall Cayley-Hamilton theorem)
- This is an **open-loop** control (depends only on  $\mathbf{r}_N$  and  $\mathbf{x}_0$ )
  - If system deviates, there is no way to notice the deviation and respond to it. This is not a robust controller.