

# **Chapter 6**

## **Pontryagin Minimum Principle**

## **6.2 Pontryagin Minimum Principle**

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

$$J = S(\mathbf{x}(t_f), t_f) + \int_{t_0}^{t_f} V(\mathbf{x}(t), \mathbf{u}(t), t) dt$$

$$\|\mathbf{u}(t)\| \leq \mathbf{U}$$

$$\begin{aligned}
\delta J = & \int_{t_0}^{t_f} \left[ \left( \frac{\partial \mathcal{L}}{\partial \mathbf{x}} \right)_* - \frac{d}{dt} \left( \frac{\partial \mathcal{L}}{\partial \dot{\mathbf{x}}} \right)_* \right]' \delta \mathbf{x}(t) dt \\
& + \int_{t_0}^{t_f} \left( \frac{\partial \mathcal{L}}{\partial \mathbf{u}} \right)'_* \delta \mathbf{u}(t) dt \\
& + \mathcal{L}|_{t_f} \delta t_f + \left[ \left( \frac{\partial \mathcal{L}}{\partial \dot{\mathbf{x}}} \right)'_* \delta \mathbf{x}(t) \right] \Big|_{t_f}. \quad (2.7.18)
\end{aligned}$$

$$\begin{aligned}
\mathcal{L}^* = & \mathcal{L}^*(\mathbf{x}^*(t), \dot{\mathbf{x}}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t) \\
= & \mathcal{H}^*(\mathbf{x}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t) \\
& + \left( \frac{\partial S}{\partial \mathbf{x}} \right)'_* \dot{\mathbf{x}}^*(t) + \left( \frac{\partial S}{\partial t} \right)_* - \boldsymbol{\lambda}^{*'}(t) \dot{\mathbf{x}}^*(t). \quad (2.7.28)
\end{aligned}$$

$$\begin{aligned}
\delta J(\mathbf{u}^*(t), \delta \mathbf{u}(t)) = & \int_{t_0}^{t_f} \left\{ \left[ \frac{\partial \mathcal{H}}{\partial \mathbf{x}} + \dot{\boldsymbol{\lambda}}(t) \right]_* \delta \mathbf{x}(t) \right. \\
& + \left[ \frac{\partial \mathcal{H}}{\partial \mathbf{u}} \right]_*' \delta \mathbf{u}(t) + \left[ \frac{\partial \mathcal{H}}{\partial \boldsymbol{\lambda}} - \dot{\mathbf{x}}(t) \right]_*' \delta \boldsymbol{\lambda}(t) \Big\} dt \\
& + \left[ \frac{\partial S}{\partial \mathbf{x}} - \boldsymbol{\lambda}(t) \right]_{*t_f}' \delta \mathbf{x}_f + \left[ \mathcal{H} + \frac{\partial S}{\partial t} \right]_{*t_f} \delta t_f.
\end{aligned} \tag{6.2.6}$$

$$\Rightarrow \delta J(\mathbf{u}^*(t), \delta \mathbf{u}(t)) = \int_{t_0}^{t_f} \left[ \frac{\partial \mathcal{H}}{\partial \mathbf{u}} \right]_*' \delta \mathbf{u}(t) dt. \tag{6.2.7}$$

$$\begin{aligned}
& = \int_{t_0}^{t_f} [\mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t) + \delta \mathbf{u}(t), \boldsymbol{\lambda}^*(t), t) \\
& \quad - \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t)] dt.
\end{aligned}$$

$$\Rightarrow \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t) + \delta \mathbf{u}(t), \boldsymbol{\lambda}^*(t), t) \geq \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t).$$

➡  $\mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t) \leq \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}(t), \boldsymbol{\lambda}^*(t), t)$

➡  $\min_{|\mathbf{u}(t)| \leq \mathbf{U}} \{\mathcal{H}(\mathbf{x}^*(t), \mathbf{u}(t), \boldsymbol{\lambda}^*(t), t)\} = \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t)$

**Table 6.1** Summary of Pontryagin Minimum Principle

A. Statement of the Problem
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Given the plant as

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

the performance index as

$$J = S(\mathbf{x}(t_f), t_f) + \int_{t_0}^{t_f} V(\mathbf{x}(t), \mathbf{u}(t), t) dt,$$

and the boundary conditions as

$\mathbf{x}(t_0) = \mathbf{x}_0$  and  $t_f$  and  $\mathbf{x}(t_f) = \mathbf{x}_f$  are free,  
find the optimal control.

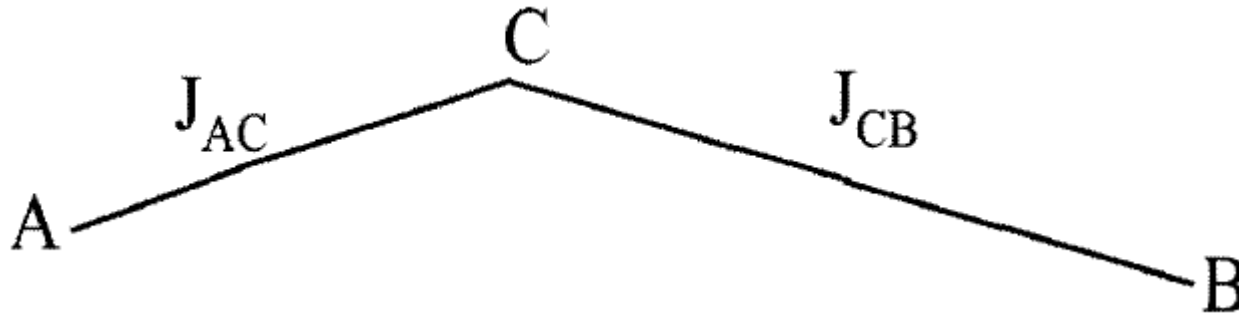
## B. Solution of the Problem

Step 1	Form the Pontryagin $\mathcal{H}$ function $\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \boldsymbol{\lambda}(t), t) = V(\mathbf{x}(t), \mathbf{u}(t), t) + \boldsymbol{\lambda}'(t)\mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t)$
Step 2	Minimize $\mathcal{H}$ w.r.t. <u><math>\mathbf{u}(t) (\leq \mathbf{U})</math></u> <u><math>\mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t), \boldsymbol{\lambda}^*(t), t) \leq \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}(t), \boldsymbol{\lambda}^*(t), t)</math></u>
Step 3	Solve the set of $2n$ state and costate equations $\dot{\mathbf{x}}^*(t) = \left( \frac{\partial \mathcal{H}}{\partial \mathbf{x}} \right)_*$ and $\dot{\boldsymbol{\lambda}}^*(t) = - \left( \frac{\partial \mathcal{H}}{\partial \boldsymbol{\lambda}} \right)_*$ with boundary conditions $\mathbf{x}_0$ and $\left[ \mathcal{H} + \frac{\partial S}{\partial t} \right]_{*t_f} \delta t_f + \left[ \frac{\partial S}{\partial \mathbf{x}} - \boldsymbol{\lambda} \right]'_{*t_f} \delta \mathbf{x}_f = 0.$



## **6.3 Dynamic Programming**

## **6.3.1 Principle of Optimality**



$$J_{AB}^* = J_{AC} + J_{CB}^*$$

An optimal policy has the property that **whatever the previous state and decision** (i.e., control), **the remaining decisions must constitute an optimal policy** with regard to the state resulting from the previous decision.

## **6.3.2 Optimal Control Using Dynamic Programming**

$$\mathbf{x}(k+1) = \mathbf{f}(\mathbf{x}(k), \mathbf{u}(k), k)$$

$$J_i(\mathbf{x}(k_i)) = J = S(\mathbf{x}(k_f), k_f) + \sum_{k=i}^{k_f-1} V(\mathbf{x}(k), \mathbf{u}(k))$$

## Principle of optimality

$$J_k^*(\mathbf{x}(k)) = \min_{\mathbf{u}(k)} [V[\mathbf{x}(k), \mathbf{u}(k)] + J_{k+1}^*(\mathbf{x}^*(k+1))]$$

## Example 6.2

$$x(k+1) = x(k) + u(k)$$

$$J = \frac{1}{2}x^2(k_f) + \frac{1}{2} \sum_{k=k_0}^{k_f-1} [x^2(k) + u^2(k)] \quad k_f = 2$$

$$\begin{aligned} -1.0 \leq u(k) \leq +1.0, \quad k = 0, 1, 2 \quad \text{or} \\ u(k) = -1.0, \quad -0.5, \quad 0, \quad +0.5, \quad +1.0 \end{aligned}$$

$$\begin{aligned} 0 \leq x(k) \leq +2.0, \quad k = 0, 1 \quad \text{or} \\ x(k) = 0, \quad 0.5, \quad 1.0 \quad 1.5 \quad 2.0. \end{aligned}$$

## Solution:

$$J_k(x(k)) = \min_{u(k)} \left[ \frac{1}{2}u^2(k) + \frac{1}{2}x^2(k) + J_{k+1}^* \right]$$

## Principle of optimality

$$J^*(x(0)) = \min_{u(0)} [V(x(0), u(0)) + J^*(x(1))]$$

$J_{02}$

$J_{12}$

**Table 6.2** Computation of Cost during the Last Stage  $k = 2$

Current State $x(1)$	Current Control $u(1)$	Next State $x(2)$	Cost $J_{12}$	Optimal Cost $J_{12}^*(x(1))$	Optimal Control $u^*(x(1), 1)$
2.0	-1.0	1.0	3.0	$J_{12}^*(2.0) = 2.25$	$u^*(1.5, 1) = -0.5$
	-0.5	1.5	2.25		
	0	2.0	4.0		
	0.5	<del>2.5</del>			
	1.0	<del>3.0</del>			
1.5	-1.0	0.5	1.75	$J_{12}^*(1.5) = 1.75$	$u^*(1.5, 1) = -1.0$
	-0.5	1.0	1.75	$J_{12}^*(1.5) = 1.75$	$u^*(1.5, 1) = -0.5$
	0	1.5	2.25		
	0.5	2.0	3.25		
	1.0	<del>2.5</del>			
1.0	-1.0	0	1.0	$J_{12}^*(1.0) = 0.75$	$u^*(1, 1) = -0.5$
	-0.5	0.5	0.75		
	0	1.0	1.0		
	0.5	1.5	1.75		
	1.0	2.0	3.0		

Use these to calculate the above:  $x(2) = x(1) + u(1)$ ;

$$J_{12} = 0.5x^2(2) + 0.5u^2(1) + 0.5x^2(1)$$

A strikeout ( $\longrightarrow$ ) indicates the value is not admissible.



Current State $x(1)$	Current Control $u(1)$	Next State $x(2)$	Cost $J_{12}$	Optimal Cost $J_{12}^*(x(1))$	Optimal Control $u^*(x(1), 1)$
0.5	-1.0	<del>-0.5</del>			
	-0.5	0	0.25	$J_{12}^*(0.5)=0.25$	$u^*(0.5,1)=-0.5$
	0	0.5	0.25	$J_{12}^*(0.5)=0.25$	$u^*(0.5,1)=0$
	0.5	1.0	0.75		
	1.0	1.5	1.75		
0	-1.0	<del>-1.0</del>			
	-0.5	<del>-0.5</del>			
	0	0	0	$J_{12}^*(0)=0$	$u^*(0,1)=0$
	0.5	0.5	0.25		
	1.0	1.0	1.0		
Use these to calculate the above: $x(2) = x(1) + u(1)$ ; $J_{12} = 0.5x^2(2) + 0.5u^2(1) + 0.5x^2(1)$ A strikeout ( $\longrightarrow$ ) indicates the value is not admissible.					

**Table 6.3** Computation of Cost during the Stage  $k = 1, 0$ 

Current State $x(0)$	Current Control $u(0)$	Next State $x(1)$	Cost $J_{02}$	Optimal Cost $J_{02}^*(x(0))$	Optimal Control $u^*(x(0), 0)$
2.0	-1.0	1.0	3.25	$J_{02}^*(2.0) = 3.25$	$u^*(2.0, 0) = -1.0$
	-0.5	1.5	3.875		
	0	2.0	4.25		
	0.5	<del>2.5</del>			
	1.0	<del>3.0</del>			
1.5	-1.0	0.5	1.875	$J_{02}^*(1.5) = 1.875$	$u^*(1.5, 0) = -1.0$
	-0.5	1.0	2.0		
	0	1.5	2.875		
	0.5	2.0	3.25		
	1.0	<del>2.5</del>			
1.0	-1.0	0	1.0	$J_{02}^*(1) = 1$	$u^*(1, 0) = -1.0$
	-0.5	0.5	0.875		
	0	1.0	1.25		
	0.5	1.5	2.375		
	1.0	2.0	3.0		

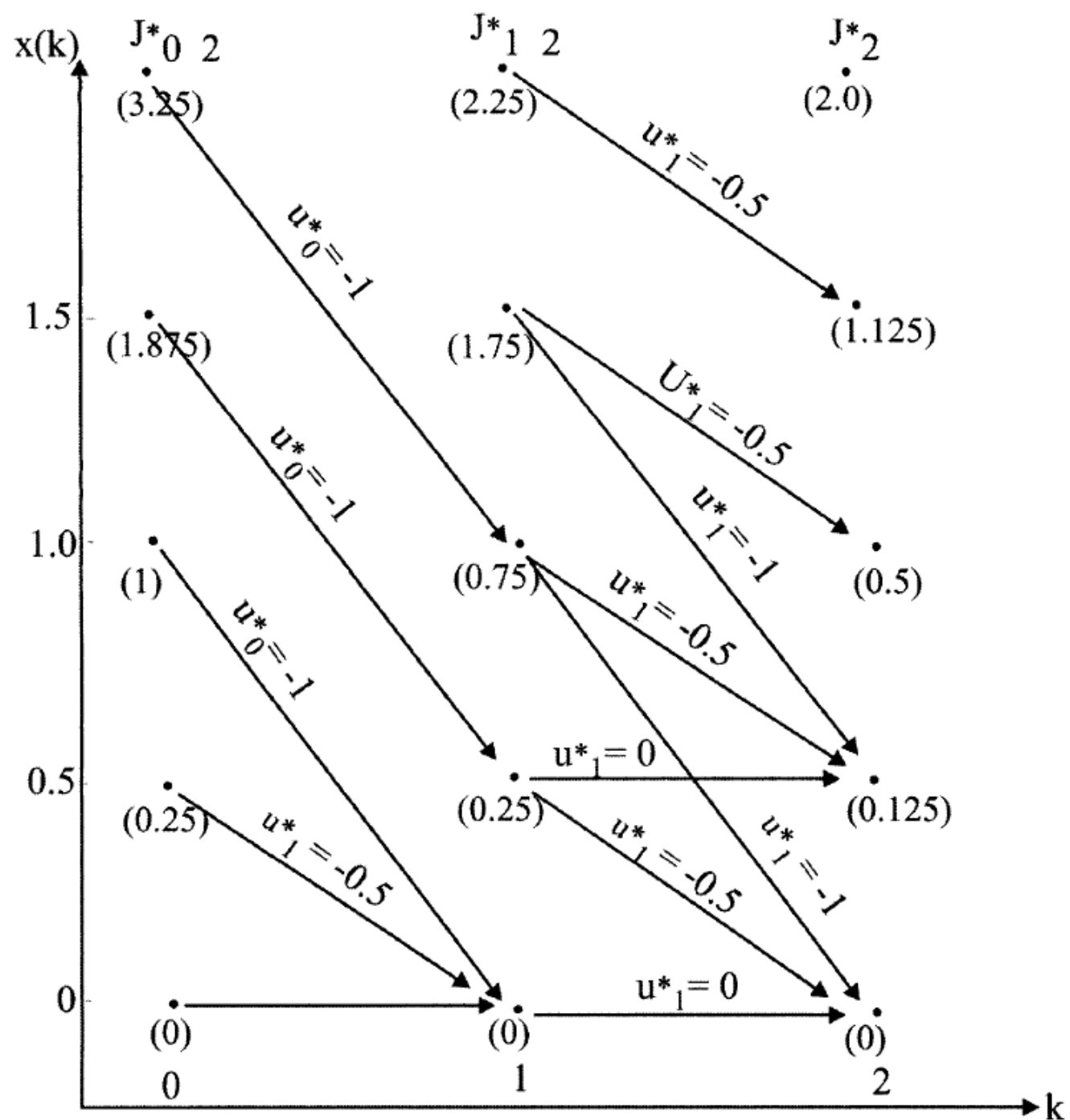
Use these to calculate the above:  $x(1) = x(0) + u(0)$ ;

$$J_{02} = 0.5u^2(0) + 0.5x^2(0) + J_{12}^*(x(1))$$

A strikeout ( $\longrightarrow$ ) indicates the value is not admissible.

Current State $x(0)$	Current Control $u(0)$	Next State $x(1)$	Cost $J_{02}$	Optimal Cost $J_{02}^*(x(0))$	Optimal Control $u^*(x(0), 0)$
0.5	-1.0	<del>-0.5</del>		$J_{02}^*(0.5)=0.25$	$u^*(1,0)=-0.5$
	-0.5	0	0.25		
	0	0.5	0.375		
	0.5	1.0	1.0		
	1.0	1.5	2.375		
0	-1.0	<del>-1.0</del>		$J_{02}^*(0)=0$	$u^*(0,0) = 0$
	-0.5	<del>-0.5</del>			
	0	0	0		
	0.5	0.5	0.375		
	1.0	1.0	1.25		

Use these to calculate the above:  $x(1) = x(0) + u(0)$ ;  
 $J_{02} = 0.5u^2(0) + 0.5x^2(0) + J_{12}^*(x(1))$   
A strikeout ( $\longrightarrow$ ) indicates the value is not admissible.




## **6.4 The Hamilton-Jacobi-Bellman Equation**

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


$$J(\mathbf{x}(t_0), t_0) = \int_{t_0}^{t_f} V(\mathbf{x}(t), \mathbf{u}(t), t) dt$$

Define

$$J^*(\mathbf{x}^*(t), t) = \int_t^{t_f} V(\mathbf{x}^*(\tau), \mathbf{u}^*(\tau), \tau) d\tau$$


$$\frac{dJ^*(\mathbf{x}^*(t), t)}{dt} = -V(\mathbf{x}^*(t), \mathbf{u}^*(t), t)$$

$$\begin{aligned}
\underline{\frac{dJ^*(\mathbf{x}^*(t), t)}{dt}} &= \left( \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial \mathbf{x}^*} \right)' \dot{\mathbf{x}}^*(t) + \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial t}, \\
&= \left( \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial \mathbf{x}^*} \right)' \mathbf{f}(\mathbf{x}^*(t), \mathbf{u}^*(t), t) + \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial t}.
\end{aligned}$$



$$\begin{aligned}
\frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial t} + \underline{V(\mathbf{x}^*(t), \mathbf{u}^*(t), t)} \\
+ \left( \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial \mathbf{x}^*} \right)' \mathbf{f}(\mathbf{x}^*(t), \mathbf{u}^*(t), t) = 0.
\end{aligned}$$

Define

$$\mathcal{H} = V(\mathbf{x}(t), \mathbf{u}(t), t) + \left( \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial \mathbf{x}^*} \right)' \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t)$$



$$\frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial t} + \mathcal{H} \left( \mathbf{x}^*(t), \frac{\partial J^*(\mathbf{x}^*(t), t)}{\partial \mathbf{x}^*}, \mathbf{u}^*(t), t \right) = 0; \forall t \in [t_0, t_f]$$

with boundary condition

$$J^*(\mathbf{x}^*(t_f), t_f) = 0$$

$$\text{or } J^*(\mathbf{x}^*(t_f), t_f) = S(\mathbf{x}^*(t_f), t_f)$$



**Table 6.4** Procedure Summary of Hamilton-Jacobi-Bellman (HJB) Approach

A. Statement of the Problem
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Given the plant as
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$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t),$
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the performance index as
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$J = S(\mathbf{x}(t_f), t_f) + \int_{t_0}^{t_f} V(\mathbf{x}(t), \mathbf{u}(t), t)dt,$
--

and the boundary conditions as
--------------------------------

$\mathbf{x}(t_0) = \mathbf{x}_0; \quad \mathbf{x}(t_f) \text{ is free}$
---

find the optimal control.
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## B. Solution of the Problem

Step 1	Form the Pontryagin $\mathcal{H}$ function $\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), J_{\mathbf{x}}^*, t) = V(\mathbf{x}(t), \mathbf{u}(t), t) + J_{\mathbf{x}}^{*'} \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t).$
Step 2	Minimize $\mathcal{H}$ w.r.t. $\mathbf{u}(t)$ as $\left( \frac{\partial \mathcal{H}}{\partial \mathbf{u}} \right)_* = 0 \text{ and obtain } \mathbf{u}^*(t) = \mathbf{h}(\mathbf{x}^*(t), J_{\mathbf{x}}^*, t).$
Step 3	Using the result of Step 2, find the optimal $\mathcal{H}^*$ function $\mathcal{H}^*(\mathbf{x}^*(t), \mathbf{h}(\mathbf{x}^*(t), J_{\mathbf{x}}^*, t), J_{\mathbf{x}}^*, t) = \mathcal{H}^*(\mathbf{x}^*(t), J_{\mathbf{x}}^*, t)$ and obtain the HJB equation.
Step 4	Solve the HJB equation $J_t^* + \mathcal{H}(\mathbf{x}^*(t), J_{\mathbf{x}}^*, t) = 0.$ with boundary condition $J^*(\mathbf{x}^*(t_f), t_f) = S(\mathbf{x}(t_f), t_f).$
Step 5	Use the solution $J^*$ , from Step 4 to evaluate $J_{\mathbf{x}}^*$ and substitute into the expression for $\mathbf{u}^*(t)$ of Step 2, to obtain the optimal control.

### Example 6.3

$$\dot{x}(t) = -2x(t) + u(t)$$

$$J = \frac{1}{2}x^2(t_f) + \frac{1}{2} \int_0^{t_f} [x^2(t) + u^2(t)] dt$$

**Solution:**

$$\begin{aligned} V(\mathbf{x}(t), \mathbf{u}(t), t) &= \frac{1}{2}u^2(t) + \frac{1}{2}x^2(t); \quad S(\mathbf{x}(t_f), t_f) = \frac{1}{2}x^2(t_f) \\ f(\mathbf{x}(t), \mathbf{u}(t), t) &= -2x(t) + u(t). \end{aligned} \tag{6.4.19}$$

$$\begin{aligned}\mathcal{H}[\mathbf{x}^*(t), J_{\mathbf{x}}, \mathbf{u}^*(t), t] &= V(\mathbf{x}(t), \mathbf{u}(t), t) + J_{\mathbf{x}}\mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t) \\ &= \frac{1}{2}u^2(t) + \frac{1}{2}x^2(t) + J_{\mathbf{x}}(-2x(t) + u(t)).\end{aligned}$$

$$\frac{\partial \mathcal{H}}{\partial u} = 0 \longrightarrow u(t) + J_x = 0 \quad \Rightarrow \quad u^*(t) = -J_x$$

$$\begin{aligned}\Rightarrow \quad \mathcal{H} &= \frac{1}{2}(-J_x)^2 + \frac{1}{2}x^2(t) + J_x(-2x(t) - J_x) \\ &= -\frac{1}{2}J_x^2 + \frac{1}{2}x^2(t) - 2x(t)J_x.\end{aligned}$$

HJB Eq.

$$\Rightarrow \quad J_t - \frac{1}{2}J_x^2 + \frac{1}{2}x^2(t) - 2x(t)J_x = 0$$

$$\text{BC: } J(x(t_f), t_f) = S(x(t_f), t_f) = \frac{1}{2}x^2(t_f)$$

Assume  $J(x(t)) = \frac{1}{2}p(t)x^2(t)$

$$J(x(t_f)) = \frac{1}{2}x^2(t_f) = \frac{1}{2}p(t_f)x^2(t_f) \quad \Rightarrow \quad p(t_f) = 1$$

$$\Rightarrow \quad J_x = p(t)x(t); \quad J_t = \frac{1}{2}\dot{p}(t)x^2(t)$$

$$\Rightarrow \quad u^*(t) = -p(t)x^*(t)$$

HJB Eq.

$$\Rightarrow \quad \left( \frac{1}{2}\dot{p}(t) - \frac{1}{2}p^2(t) - 2p(t) + \frac{1}{2} \right) x^{*2}(t) = 0$$

$$\Rightarrow \quad \frac{1}{2}\dot{p}(t) - \frac{1}{2}p^2(t) - 2p(t) + \frac{1}{2} = 0$$

$$\Rightarrow \quad p(t) = \frac{(\sqrt{5} - 2) + (\sqrt{5} + 2) \left[ \frac{3 - \sqrt{5}}{3 + \sqrt{5}} \right] e^{2\sqrt{5}(t - t_f)}}{1 - \left[ \frac{3 - \sqrt{5}}{3 + \sqrt{5}} \right] e^{2\sqrt{5}(t - t_f)}}$$

*Note:* Let us note that as  $t_f \rightarrow \infty$ ,  $p(t)$  in (6.4.33) becomes  $p(\infty) = \bar{p} = \sqrt{5} - 2$ , and the optimal control (6.4.30) is

$$\underline{u(t) = -(\sqrt{5} - 2)x(t).} \quad (6.4.34)$$

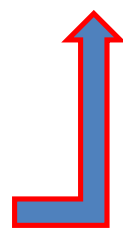
## **6.5 LQR System Using H-J-B Equation**

$$\dot{\mathbf{x}}(t) = \mathbf{A}(t)\mathbf{x}(t) + \mathbf{B}(t)\mathbf{u}(t)$$

$$J = \frac{1}{2}\mathbf{x}'(t_f)\mathbf{F}\mathbf{x}(t_f) + \frac{1}{2} \int_{t_0}^{t_f} [\mathbf{x}'(t)\mathbf{Q}(t)\mathbf{x}(t) + \mathbf{u}'(t)\mathbf{R}(t)\mathbf{u}(t)] dt$$

$$\begin{aligned} \mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), J_{\mathbf{x}}^*, t) &= \frac{1}{2}\mathbf{x}'(t)\mathbf{Q}(t)\mathbf{x}(t) + \frac{1}{2}\mathbf{u}'(t)\mathbf{R}(t)\mathbf{u}(t) \\ &\quad + J_{\mathbf{x}}^{*'}(\mathbf{x}(t), t)[\mathbf{A}(t)\mathbf{x}(t) + \mathbf{B}(t)\mathbf{u}(t)] \end{aligned}$$

$$\begin{aligned} \frac{\partial \mathcal{H}}{\partial \mathbf{u}} = 0 &\longrightarrow \mathbf{R}(t)\mathbf{u}(t) + \mathbf{B}'(t)J_{\mathbf{x}}^{*'}(\mathbf{x}(t), t) = 0 \\ &\longrightarrow \mathbf{u}^*(t) = -\mathbf{R}^{-1}(t)\mathbf{B}'(t)J_{\mathbf{x}}^*(\mathbf{x}(t), t) \end{aligned}$$




note  $\frac{\partial^2 \mathcal{H}}{\partial \mathbf{u}^2} = \mathbf{R}(t) \Rightarrow$  *minimum control.*



$$\begin{aligned}
\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), J_{\mathbf{x}}^*, t) &= \frac{1}{2} \mathbf{x}'(t) \mathbf{Q}(t) \mathbf{x}(t) + \frac{1}{2} J_{\mathbf{x}}^{*'} \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}'(t) J_{\mathbf{x}}^* \\
&\quad + J_{\mathbf{x}}^{*'} \mathbf{A}(t) \mathbf{x}(t) - J_{\mathbf{x}}^{*'} \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}'(t) J_{\mathbf{x}}^* \\
&= \frac{1}{2} \mathbf{x}'(t) \mathbf{Q}(t) \mathbf{x}(t) - \frac{1}{2} J_{\mathbf{x}}^{*'} \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}'(t) J_{\mathbf{x}}^* \\
&\quad + J_{\mathbf{x}}^{*'} \mathbf{A}(t) \mathbf{x}(t). \tag{6.5.7}
\end{aligned}$$



HJB eq:  $J_t^* + \mathcal{H}(\mathbf{x}^*(t), \mathbf{u}^*(t), J_{\mathbf{x}}^*, t) = 0$


 $J_t^* + \frac{1}{2} \mathbf{x}^{*'}(t) \mathbf{Q}(t) \mathbf{x}^*(t) - \frac{1}{2} J_{\mathbf{x}}^{*'} \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}'(t) J_{\mathbf{x}}^* \\
+ J_{\mathbf{x}}^{*'} \mathbf{A}(t) \mathbf{x}^*(t) = 0$

with BC  $J^*(\mathbf{x}^*(t_f), t_f) = \frac{1}{2} \mathbf{x}^{*'}(t_f) \mathbf{F}(t_f) \mathbf{x}^*(t_f)$


assume  $J^*(\mathbf{x}(t), t) = \frac{1}{2} \mathbf{x}'(t) \mathbf{P}(t) \mathbf{x}(t)$

where,  $\mathbf{P}(t)$  is a real, symmetric, positive-definite matrix

$$\begin{cases} \frac{\partial J^*}{\partial t} = J_t = \frac{1}{2} \mathbf{x}(t) \dot{\mathbf{P}}(t) \mathbf{x}(t) \\ \frac{\partial J^*}{\partial \mathbf{x}} = J_{\mathbf{x}} = \mathbf{P}(t) \mathbf{x}(t) \end{cases}$$

HJB eq



$$\begin{aligned} & \frac{1}{2} \mathbf{x}'(t) \dot{\mathbf{P}}(t) \mathbf{x}(t) + \frac{1}{2} \mathbf{x}(t) \mathbf{Q}(t) \mathbf{x}(t) \\ & - \frac{1}{2} \mathbf{x}'(t) \mathbf{P}(t) \mathbf{B}(t) \mathbf{R}^{-1}(t) \mathbf{B}'(t) \mathbf{P}(t) \mathbf{x}(t) \\ & + \mathbf{x}'(t) \mathbf{P}(t) \mathbf{A}(t) \mathbf{x}(t) = 0. \end{aligned} \quad (6.5.13)$$



$$\dot{\mathbf{P}}(t) = -\mathbf{P}(t)\mathbf{A}(t) - \mathbf{A}'(t)\mathbf{P}(t) + \mathbf{P}(t)\mathbf{B}(t)\mathbf{R}^{-1}(t)\mathbf{B}'(t)\mathbf{P}(t) - \mathbf{Q}(t).$$

with BC  $\mathbf{P}(t_f) = \mathbf{F}(t_f).$


Also,  $\mathbf{u}^*(t) = -\mathbf{R}^{-1}(t)\mathbf{B}'(t)\mathbf{P}(t)\mathbf{x}^*(t)$


## Example 6.4

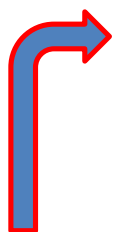
$$\dot{x}(t) = -2x(t) + u(t)$$

$$J = \int_0^\infty [x^2(t) + u^2(t)] dt$$


**Solution:**



$$\begin{aligned} V(x(t), u(t)) &= x^2(t) + u^2(t), \\ f(x(t), u(t)) &= -2x(t) + u(t). \end{aligned}$$


$$\begin{aligned} \mathcal{H}(x(t), u(t), J_x^*) &= V(x(t), u(t)) + J_x^* f(x(t), u(t)) \\ &= x^2(t) + u^2(t) + 2fx(t)[-2x(t) + u(t)] \\ &= x^2(t) + u^2(t) - 4fx^2(t) + 2fx(t)u(t) \end{aligned}$$



Assume  $J^* = fx^2(t)$

 
$$\frac{\partial \mathcal{H}}{\partial u} = 2u^*(t) + 2fx^*(t) = 0 \longrightarrow u^*(t) = -fx^*(t).$$


 
$$\mathcal{H}^*(x^*(t), J_x^*, t) = x^{*2}(t) - 4fx^{*2}(t) - f^2x^{*2}(t)$$

HJB eq  $\mathcal{H}^*(x^*(t), J_x^*) + J_t^* = 0$

$$\longrightarrow x^{*2}(t) - 4fx^{*2}(t) - f^2x^{*2}(t) = 0$$

$$\longrightarrow f^2 + 4f - 1 = 0$$

$$\longrightarrow f = -2 \pm \sqrt{5}$$

 
$$u^*(t) = -fx^*(t) = -(\sqrt{5} - 2)x^*(t)$$