

Pr 7.3

$$\dot{x} = Ax + Bu; \quad \min J = \int_0^{t_f} 1 dt$$

$$x_0 \rightarrow [2, 2] = x_f$$

$$H = 1 + \lambda^T (Ax + Bu)$$

→ optimal control:  $u = -\text{sign } B^T \lambda$   
 $\dot{\lambda} = -A^T \lambda$ . (integrators ⇒ max 1 switching)

Optimal policy: find the switching surface. Then

$$u_*(x) = +u_{\max} \quad \text{if } x \text{ below the surface}$$

$$u_*(x) = -u_{\max} \quad \text{if } x \text{ above " " " " .}$$

The switching surface becomes more "vertical" when  $u_{\max}$  increases.

Computational detail: to determine the switching surface solve  $\dot{x} = Ax + Bu$  backwards in time, starting with I.C.  $[2; 2]$ . I.e.,

$$x = \text{lsim}(-A, -B, I, 0, \text{ones}(\text{size}(t)), t, [2; 2])$$

When the control is unconstrained  $\min t_f = 0$  (an infimum, actually!) and the optimal control becomes impulsive.

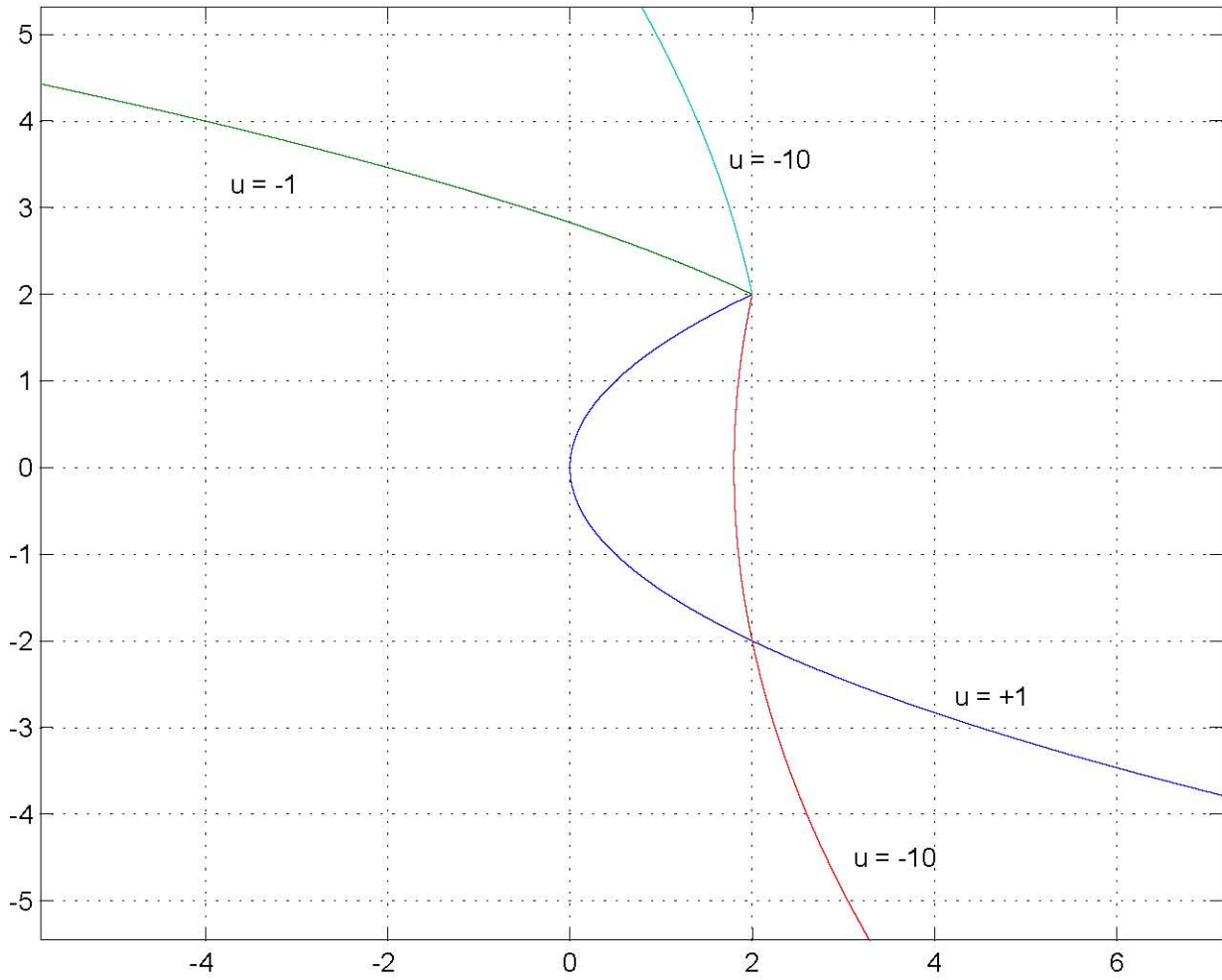
This can also be seen from  $\min J(t_f) = \frac{1}{2} \int_0^{t_f} u^T u$ .

The optimal control exists for arbitrarily small  $t_f$ , and can be computed from

$$\begin{pmatrix} \dot{x} \\ \dot{\lambda} \end{pmatrix} = \begin{pmatrix} A & -BB^T \\ 0 & -A^T \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix}$$

with B.C.  $x_0, x_f$  (⇒  $\lambda_0 = \Phi_{12}^{-1}(x_f - \Phi_{11} x_0) \dots$ )

Pr. 7.3, Switching surface for different input amplitudes

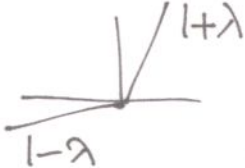


Pr. 7.6  $\dot{x} = u, \quad |u| \leq 1 \quad J = \int_0^{t_f} |u|, \quad t_f \text{ free.}$

$\min J, \quad x_0 \rightarrow 0.$

$H = |u| + \lambda u \quad \Rightarrow \quad u_{\text{opt}} = \begin{cases} -1 & \text{if } \lambda > 1 \\ 0 & \text{if } \lambda \in (-1, 1) \\ 1 & \text{if } \lambda < -1 \\ ? & \text{if } \lambda = \pm 1 \end{cases}$

$-\dot{\lambda} = \frac{\partial H}{\partial x} = 0$



$t_f$  is free, system and objective are autonomous  
 $\Rightarrow H_{\text{opt}} = 0 \Rightarrow \lambda = \pm 1$  (singular interval)

Notice:  $A=0 \Rightarrow$  singular intervals may exist.

Indeed, the solution is not unique. For any  $t_f$ ,

$u = -\frac{x_0}{t_f}$  satisfies the B.C. and  $J = \int_0^{t_f} \frac{|x_0|}{t_f} dt = |x_0|$

(all have the same, optimal cost).

Pr. 7.8

$$\dot{x} = \begin{pmatrix} 0 & 1 \\ 0 & -a \end{pmatrix} x + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u \quad ; \quad a > 0 \\ |u| \leq 1$$

$$J = \int_0^{t_f} \beta + |u| \quad ; \quad t_f \text{ free, } \beta > 0 \\ x_0 \rightarrow 0$$

(weighted min-time - min fuel problem)

$$H = \beta + |u| + \lambda^T (Ax + Bu)$$

$$u_{\text{opt}} = \begin{cases} -1 & \text{if } \lambda_2 > 1 \\ 0 & \text{if } \lambda_2 \in [-1, 1] \\ 1 & \text{if } \lambda_2 < -1 \\ ? & \text{if } \lambda_2 = \pm 1 \end{cases} \quad \left\| \quad -\dot{\lambda} = \frac{\partial H}{\partial x} = A^T \lambda \right.$$

$$\Rightarrow \dot{\lambda}_1 = 0 \quad \dot{\lambda}_2 = -\lambda_1 + a\lambda_2 \Rightarrow \lambda(t) = \begin{bmatrix} \lambda_{10} \\ e^{at}\lambda_{20} + \frac{\lambda_{10}}{a}[1 - e^{at}] \end{bmatrix}$$

Depending on the sign and magnitude of the initial costate, we may have up to two switchings ( $-1 \rightarrow 0 \rightarrow 1$  or vice-versa). There can be no singular intervals.

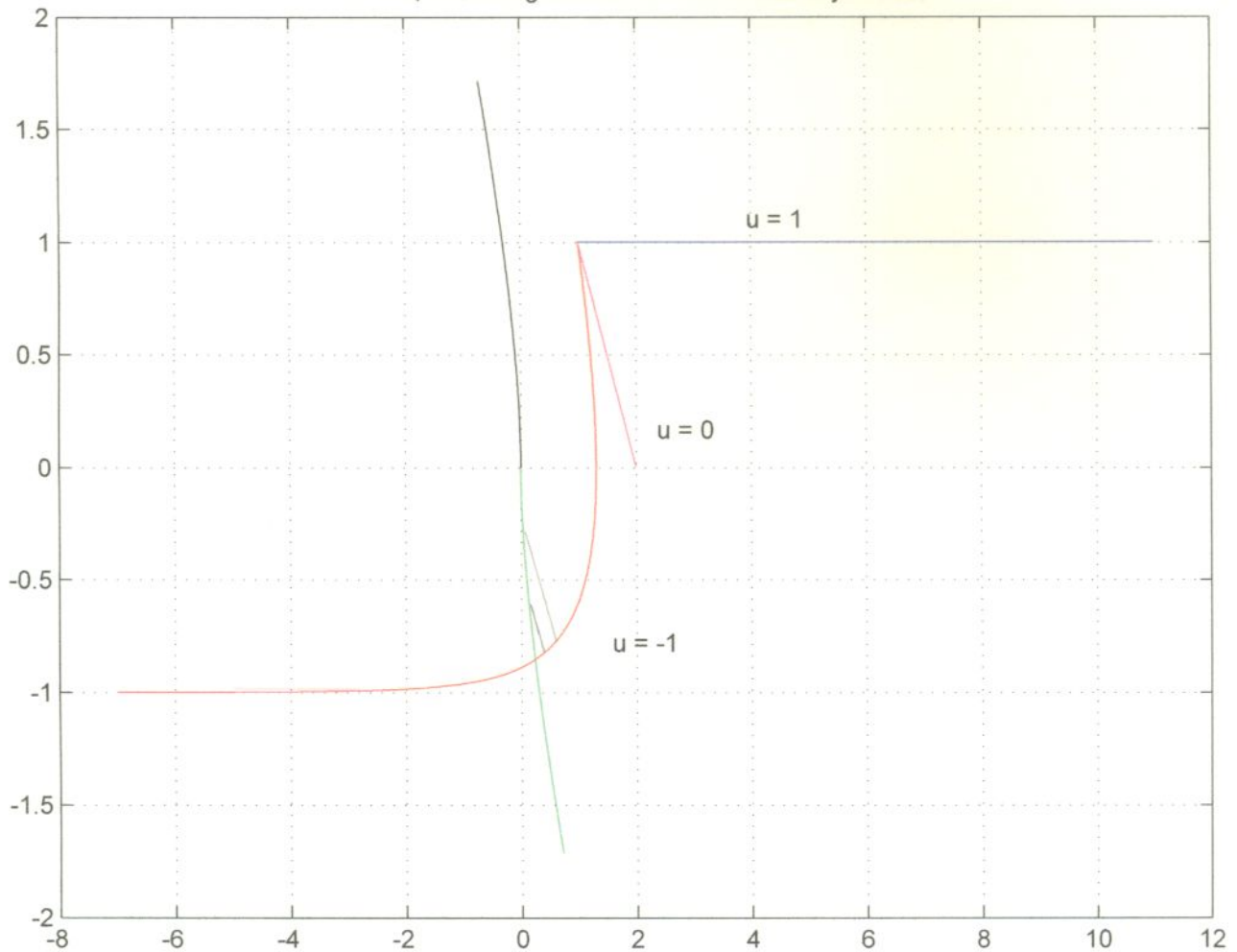
Critical values for  $\lambda, x$  at the switching times can be obtained from  $H=0$  e.g. switching

$$\begin{aligned} -1 \rightarrow 0 : \quad \lambda_2 = 1 \quad @ \quad t^- \quad u = -1 &\Rightarrow \beta + 1 + (\lambda_{10}, 1) \begin{pmatrix} Ax + \begin{pmatrix} 0 \\ -1 \end{pmatrix} \end{pmatrix} \\ &= 0 \\ @ \quad t^+ \quad u = 0 &\Rightarrow \beta + (\lambda_{10}, 1) Ax = 0 \end{aligned}$$

etc

Alternatively, we can sketch the switching surface to 0 with inputs  $\pm 1$ , and the 3 possible trajectories from  $x_0$  (ie.  $u = +1, -1, 0$ ). Note: for  $u=0$ ,  $x_1(t) = c - \frac{x_2(t)}{a}$ , a straight line.

Pr. 7.8, Switching surface to 0 and initial trajectories



The <sup>time-</sup>optimal trajectory ( $\beta \rightarrow \infty$ ) switches  $\pm 1 \rightarrow \mp 1$  (red-green in our plot). The weighted time-fuel optimal connects the two segments with a straight line of slope  $-a$ . The length of the segment depends on  $\beta$  (but not straightforward to compute)



Pr 7.9

$$\dot{x} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} x + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u \quad |u| \leq 1$$

$$x_0 = (1; 1) \rightarrow x_f = (0; 0) \quad t_f = 4$$

$$J = \frac{1}{2} \int_0^4 u^2$$

$$H = \frac{1}{2} u^2 + \lambda^T (Ax + Bu)$$

$$u_{\text{opt}} : \frac{\partial H}{\partial u} = 0 \Rightarrow u = -\frac{\partial}{\partial u} B^T \lambda \quad (\text{unconstr.})$$

$$\text{when } |u| \leq 1 : u_{\text{opt}} = \text{sat} \left[ -\frac{\partial}{\partial u} B^T \lambda \right] \Rightarrow \text{SAT}$$

$$\dot{\lambda} = -\frac{\partial H}{\partial x} = -A^T \lambda \Rightarrow \begin{aligned} \lambda_1 &= c_1 \\ \lambda_2 &= c_2 - \lambda_2 t \end{aligned} \Rightarrow \lambda = \begin{pmatrix} 1 & 0 \\ -t & 1 \end{pmatrix} \lambda_0$$

The unconstrained solution is given by

$$u_{\text{opt}} = -\lambda_2(t) \quad \text{and}$$

$$\begin{pmatrix} x \\ \lambda \end{pmatrix} = \begin{pmatrix} A & -BB^T \\ 0 & -A^T \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix} \Rightarrow \begin{pmatrix} x(t) \\ \lambda(t) \end{pmatrix} = e^{\hat{H}t} \begin{pmatrix} x_0 \\ \lambda_0 \end{pmatrix}$$

$$\leftarrow \hat{H} \rightarrow \quad \text{with } x_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, x_f = 0$$

$$\Rightarrow \lambda_0 = -\phi_{12}^{-1}(4) \phi_{11} x_0 = \begin{pmatrix} 0.5625 \\ 1.3750 \end{pmatrix}$$

$$\lambda_2(t) = (t, -1) \lambda_0 \rightarrow \text{violates the bound}$$

$\Rightarrow$  the soln. is not unconstrained  $\Rightarrow$  it changes entirely. Without solving the TPBVP, we can observe that the soln will be equal to the unconstrained ( $u_{\text{opt}} = -\lambda_2$ ) after some time  $t_1$  (if the problem is feasible).

Then, in  $(0, t_1)$   $\begin{pmatrix} \dot{x} \\ \dot{\lambda} \end{pmatrix} = \begin{pmatrix} A & \\ & -A^T \end{pmatrix} \begin{pmatrix} x \\ \lambda \end{pmatrix} + \begin{pmatrix} B \\ 0 \end{pmatrix} (-1)$

and in  $(t_1, 4)$   $\begin{pmatrix} \dot{x} \\ \dot{\lambda} \end{pmatrix} = \hat{H} \begin{pmatrix} x \\ \lambda \end{pmatrix}$

with continuity at  $t_1$  being the boundary condition.

In the interval  $(0, t_1)$ ,  $x(t) = \begin{pmatrix} x_{10} - \frac{1}{2}t^2 + x_{20}t \\ x_{20} - t \end{pmatrix}$

where  $x_{10} = x_{20} = 1$ .

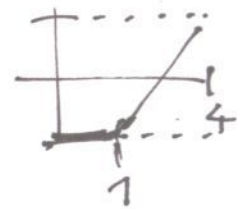
In the interval  $(t_1, 4)$ ,  $\begin{pmatrix} x \\ \lambda \end{pmatrix}(t) = e^{H(t-t_1)} \begin{pmatrix} x \\ \lambda \end{pmatrix}(t_1)$

$$\Rightarrow \lambda(t_1) = -\Phi_{12}^{-1}(4-t_1) \Phi_{11}(4-t_1) x(t_1)$$

$$\Rightarrow u(t) = -\lambda_2(t) = [(t-t_1), -1] \Phi_{12}^{-1}(4-t_1) \Phi_{11}(4-t_1) \cdot \begin{bmatrix} 1 - \frac{t_1^2}{2} + t_1 \\ 1 - t_1 \end{bmatrix}$$

It is now a 1-parameter search to find  $t_1$  s.t.  $|u(t_1)| \leq 1$ . The first (and only) time is at  $t_1 = 1$ . At that time  $x(t_1) = \begin{pmatrix} 3/2 \\ 0 \end{pmatrix}$

We compute & sketch the rest of the solution to check if  $|u| \leq 1 \forall t$ .



The optimal cost is  $J_* = 1$

(analytically or numerically)

Pr 7.11

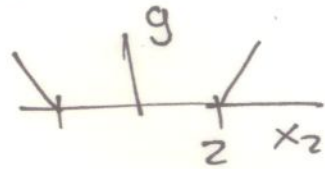
$$\dot{x} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} x + \begin{pmatrix} 0 \\ 1 \end{pmatrix} u$$

$\leftarrow A \rightarrow \quad \leftarrow B \rightarrow$

$$J = \int 1$$

$$|u| \leq 1, \quad |x_2| \leq 2 \quad (\text{Velocity constraint})$$

Let  $g(x_2-2)$  be such that



Define the new state  $\dot{z} = g^2$  and the isoperimetric constraint  $z(0) = 0, z(t_f) = 0$ .

Then, the Hamiltonian is

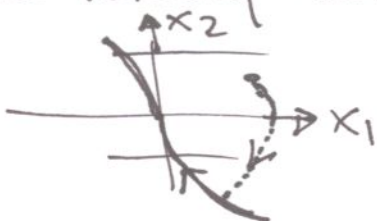
$$H = 1 + \lambda^T (Ax + Bu) + \mu g^2 \quad \lambda, \mu: \text{costates}$$

Optimality:  $u^* = -\text{sign}(B^T \lambda)$

$$\dot{\lambda} = -\frac{\partial H}{\partial x} = -A^T \lambda + 2\mu g \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$\dot{\mu} = 0$$

When  $g \equiv 0$ , so that the solution satisfies the constraints,  $\dot{\lambda} = -A^T \lambda \Rightarrow u^*$  is the true optimal input. This soln is Bang-Bang and may violate the velocity constraint. Intuitively, the optimal



soln. will use the  $\pm 1$  controls when  $|x_2| < 2$  and 0 control

(constant velocity) when  $x_2 = 2$



The way this solution arises from our equations is by considering some time where  $x_2 > 0$ , but small. Then for large  $\mu$ ,  $x_2$  will change sign very quickly. Thus, the solution will chatter around the constraint, which is expected since the extremals still have  $u = \pm 1$ . In an average sense, the trajectory moves in the correct direction (intuitively the same as  $u = 0$ ), see Fillipou for theory. Notice that any  $x_2 > 0$  will make  $z_2(t_f) > 0$ . But for  $\mu \rightarrow \infty$ ,  $z_2(t_f) \downarrow 0$ .

PROBLEM 6.3

Build the table with entries:

STAGE, STATE & OPTIMAL COST-TO-GO ( $J_*(x)$ ),  
STATE & OPTIMAL CONTROL ( $u_*(x)$ ).

Starting from the last stage we get:

STAGE	STATE	$J_*(x)$	$u_*$	NOTES
5	N	0	N-N	
4	H	3	H-N	
	I	4	I-N	
3	E	10	E-H	
	F	6	$\begin{cases} F-H \\ F-I \end{cases}$	$J_*(F) = \min \begin{cases} J_*(H)+3 \\ J_*(I)+2 \end{cases} = 6$ Multiple minima.
	G	9	G-I	
2	C	12	C-F	$J_*(C) = \min \begin{cases} J_*(E)+4 \\ J_*(F)+6 \end{cases} = \min(14, 12)$
	D	13	D-F	$J_*(D) = \min \begin{cases} J_*(G)+8 \\ J_*(F)+7 \end{cases}$
1	L	15	$\begin{cases} L-C \\ L-D \end{cases}$	$J_*(L) = \min \begin{cases} J_*(C)+3 \\ J_*(D)+2 \end{cases}$

Minimum Cost L-N = 15

Optimal paths:  $\left. \begin{array}{l} L-C-F-H-N \\ L-C-F-I-N \\ L-D-F-H-N \\ L-D-F-I-N \end{array} \right\} \text{ (Just go through F)}$