# AE-777A (Optimal Space Flight Control)

Quiz No. 3 (Solution)

## 1. For the minimization of

$$L(u) = u_1^2 - u_1 u_2 + \frac{1}{3} u_2^3$$

with respect to  $u = (u_1, u_2)^T \in \mathbb{R}^2$ , and subject to

$$u_1 \ge 0$$
;  $u_2 \ge 0$ 

find the minimum points,  $\hat{u}$ , (if any).

#### Ans

Determination of the stationary points:

$$L_u = \frac{\partial L}{\partial u} = (2u_1 - u_2, -u_1 + u_2^2)$$
  
= [0, 0]

which is satisfied for  $u_1 = u_2 = 0$  and  $u_1 = 1/4$ ,  $u_2 = 1/2$ . Hence,  $u^* = (0,0)^T$  and  $u^* = (\frac{1}{4},\frac{1}{2})^T$  are the stationary points of L(u). Both the stationary points lie in the feasible region defined by the inequality constraints.

Evaluation of  $L_{uu}$  at the stationary point,  $u^* = (0,0)^T$ :

$$L_{uu} = \frac{\partial^2 L}{\partial u^2} = \begin{pmatrix} 2 & -1 \\ -1 & 2u_2^* \end{pmatrix}_{u_2^* = 0}$$
$$= \begin{pmatrix} 2 & -1 \\ -1 & 0 \end{pmatrix}$$

The eigenvalues of  $L_{uu}$  at the stationary point,  $u^* = (0,0)^T$ :

$$\det(sI - L_{uu}) = s(s-2) - 1 = s^2 - 2s - 1 = 0$$

or  $s_{1,2} = 1 \pm \sqrt{2}$ . Since one eigenvalue of  $L_{uu}$  is positive, while the other is negative, neither the necessary condition,  $L_{uu} \geq 0$ , nor the sufficient condition of minimization,  $L_{uu} > 0$ , is satisfied at the stationary point,

 $u^* = (0,0)^T$ . It is a saddle point lying on the boundary of the feasible region.

Evaluation of  $L_{uu}$  at the stationary point,  $u^* = (\frac{1}{4}, \frac{1}{2})^T$ :

$$L_{uu} = \frac{\partial^2 L}{\partial u^2} = \begin{pmatrix} 2 & -1 \\ -1 & 2u_2^* \end{pmatrix}_{u_2^* = 1/2}$$
$$= \begin{pmatrix} 2 & -1 \\ -1 & 1 \end{pmatrix}$$

The eigenvalues of  $L_{uu}$  at the stationary point,  $u^* = (\frac{1}{4}, \frac{1}{2})^T$ :

$$\det(sI - L_{uu}) = (s - 2)(s - 1) - 1 = s^2 - 3s + 1 = 0$$

or  $s_{1,2} = (3 \pm \sqrt{5})/2$ . Since both the eigenvalues of  $L_{uu}$  are positive, the sufficient condition of minimization,  $L_{uu} > 0$ , is satisfied at the stationary point,  $u^* = (\frac{1}{4}, \frac{1}{2})^T$ , and it is the only minimum point of L(u) satisfying the inequality constraints.

The inequality constraints can be expressed as follows:

$$f(u) = \begin{pmatrix} -u_1 \\ -u_2 \end{pmatrix} \le \begin{pmatrix} 0 \\ 0 \end{pmatrix} ,$$

with the necessary condition for minimization given by

$$\frac{\partial L}{\partial u} + \lambda^T \frac{\partial f}{\partial u} = 0 \; ,$$

where

$$\lambda \left\{ \begin{array}{ll} = (0,0)^T & \qquad f(u) < (0,0)^T \\ > (0,0)^T, & \qquad f(u) = (0,0)^T \end{array} \right.$$

and

$$\frac{\partial f}{\partial u} = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$$

Since the minimum point,  $\hat{u} = (\frac{1}{4}, \frac{1}{2})^T$ , lies strictly inside the feasible region,  $f(u) < (0,0)^T$ , the necessary condition is satisfied with  $\lambda = (0,0)^T$ .

In contrast, the necessary condition of minimization is *not* satisfied for the stationary point,  $u^* = (0,0)^T$ , for any non-zero value of  $\lambda$ , hence it cannot be a minimum point lying at the boundary of the feasible region with  $f(u) = (0,0)^T$ .

2. Consider a system with the following state equations  $[(x_1, x_2)^T \in \mathbb{R}^2, u(t) \in \mathbb{R}]$ :

$$\begin{aligned}
\dot{x}_1 &= x_2 \\
\dot{x}_2 &= -x_1 + u
\end{aligned}$$

Given the initial state vector to be  $x(0) = (0,0)^T$ , determine the extremal control history,  $u^*(t)$ , such that the following objective function is minimized w.r.t. u(t) in the control interval  $0 \le t \le 1$ :

$$J = \frac{1}{2}[x_1(1) - 1]^2 + \frac{1}{2} \int_0^1 u^2(t) dt$$

### Ans.

The state equations are expressed in the following form:

$$\dot{x} = Ax + Bu$$

where the coefficient matrices are the following:

$$A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}; \qquad B = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

The Hamiltonian of the optimization problem is given by

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

The co-state equations of the system are next derived as follows:

$$\dot{\lambda}^* = \left\{ \begin{array}{c} \dot{\lambda}_1^* \\ \dot{\lambda}_2^* \end{array} \right\} = -\left(\frac{\partial H}{\partial x}\right)^{*T} = -A^T \lambda^*$$

where

$$-A^T = \left( \begin{array}{ccc} 0 & & 1 \\ & & \\ -1 & & 0 \end{array} \right)$$

The solution to the co-state equations subject to the initial condition,  $\lambda^*(0) = (c_1, c_2)^T$ , is expressed by

$$\lambda^{*}(t) = e^{-A^{T}t}\lambda^{*}(0) = \mathcal{L}^{-1}[sI - (-A^{T})]^{-1}\lambda^{*}(0)$$

$$= \mathcal{L}^{-1}\frac{1}{s^{2}+1}\begin{pmatrix} s & 1 \\ -1 & s \end{pmatrix}\lambda^{*}(0)$$

$$= \begin{pmatrix} \cos t & \sin t \\ -\sin t & \cos t \end{pmatrix} \begin{cases} c_{1} \\ c_{2} \end{cases} \quad (t \ge 0)$$

$$= \begin{cases} c_{1}\cos t + c_{2}\sin t \\ -c_{1}\sin t + c_{2}\cos t \end{cases} \quad (t \ge 0)$$

The necessary conditions for optimality include the following stationary condition of the Hamiltonian with respect to the control history, u(t):

$$H_u = \left(\frac{\partial H}{\partial u}\right)^* = u^*(t) + \lambda_2^*(t) = 0$$

which results in the following extremal control history

$$u^*(t) = -\lambda_2^*(t) = c_1 \sin t - c_2 \cos t \quad (0 \le t \le 1)$$

In order to determine the constants  $c_1, c_2$ , the state equations must be solved for  $x(t) = (x_1(t), x_2(t))^T$ , and the following boundary condition applied on the co-state,  $\lambda(t) = (\lambda_1(t), \lambda_2(t))^T$ :

$$\lambda^*(1) = \left\{ \begin{array}{c} c_1 \cos 1 + c_2 \sin 1 \\ -c_1 \sin 1 + c_2 \cos 1 \end{array} \right\} = \left( \frac{\partial \phi}{\partial x} \right)_{t=1}^T = \left( \begin{array}{c} x_1^*(1) - 1 \\ 0 \end{array} \right)$$

To apply the terminal boundary condition, the extremal trajectory must be solved from the state equations, given  $x(0) = (0,0)^T$ , and it happens that  $A = -A^T$  for this system:

$$x^{*}(t) = e^{At}x(0) + \int_{0}^{t} e^{A(t-\tau)}Bu^{*}(\tau)d\tau$$

$$= -\int_{0}^{t} e^{A(t-\tau)}B\lambda_{2}^{*}(\tau)d\tau$$

$$= \int_{0}^{t} \begin{pmatrix} \cos(t-\tau) & \sin(t-\tau) \\ -\sin(t-\tau) & \cos(t-\tau) \end{pmatrix} \begin{cases} 0 \\ 1 \end{cases} (c_{1}\sin\tau - c_{2}\cos\tau)d\tau$$

$$= \int_{0}^{t} \begin{cases} \sin(t-\tau) \\ \cos(t-\tau) \end{cases} (c_{1}\sin\tau - c_{2}\cos\tau)d\tau$$

To apply the terminal boundary condition, only the first state variable needs to be solved for:

$$x_1^*(t) = \int_0^t [c_1 \sin(t - \tau) \sin \tau - c_2 \sin(t - \tau) \cos \tau] d\tau$$

To carry out the integration, the following trigonometric identities are employed:

$$\sin(t-\tau)\sin\tau = \frac{1}{2}[\cos(t-2\tau) - \cos t]$$
$$\sin(t-\tau)\cos\tau = \frac{1}{2}[\sin(t-2\tau) + \sin t]$$

which yields

$$x_1^*(t) = \frac{c_1}{2} \int_0^t \cos(t - 2\tau) d\tau - \frac{c_1}{2} t \cos t - \frac{c_2}{2} t \sin t - \frac{c_2}{2} \int_0^t \sin(t - 2\tau) d\tau$$

$$= \frac{c_1}{4} \int_{-t}^t \cos y dy - \frac{c_1}{2} t \cos t - \frac{c_2}{2} t \sin t - \frac{c_2}{4} \int_{-t}^t \sin y dy$$

$$= \frac{c_1}{2} (\sin t - t \cos t) - \frac{c_2}{2} t \sin t$$

Finally, the boundary condition is applied to produce the following linear equations:

$$\begin{pmatrix} x_1^*(1) - 1 \\ 0 \end{pmatrix} = \begin{cases} c_1 \cos 1 + c_2 \sin 1 \\ -c_1 \sin 1 + c_2 \cos 1 \end{cases}$$

or,

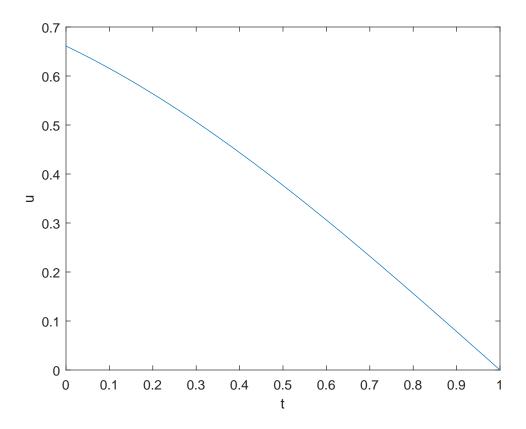
$$c_2 = c_1 \tan 1$$

$$c_1 \left(\frac{3}{2} \cos 1 - \frac{1}{2} \sin 1\right) + \frac{3}{2} c_2 \sin 1 = -1$$

which are solved to give  $c_1 = -0.42454$  and  $c_2 = -0.6611826$ . Therefore, the extremal control is the following:

$$u^*(t) = -0.42454 \sin t + 0.6611826 \cos t \quad (0 \le t \le 1)$$

which is plotted in the figure on the next page.



3. For the system of Problem 2, find a state-feedback regulator (if any) such that the following objective function is minimized w.r.t. u(t):

$$J = \frac{1}{2} \int_0^\infty \left\{ x_1^2(t) + u^2(t) \right\} dt$$

## Ans.

The system in Problem 2 has the following state-space coefficients:

$$A = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix}; \qquad B = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

For minimizing the following quadratic objective function w.r.t. the control input, u(t),

$$J = \frac{1}{2} \int_0^\infty \left\{ x_1^2(t) + u^2(t) \right\} dt \tag{1}$$

the following cost-coefficient matrices are selected (see Lecture 11 for details):

$$Q = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} , \quad R = 1 , \quad S = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$
 (2)

The optimal LQR feedback law is sought

$$u(t) = -R^{-1}[B^T P + S^T]x(t) = -(p_{12}, p_{22})x(t)$$
(3)

where the constant, real matrix

$$P = \begin{pmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{pmatrix}, \tag{4}$$

is the solution to the algebraic Riccati equation:

$$0 = Q + A^{T}P + PA - PBR^{-1}B^{T}P$$
 (5)

or

$$\begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$= \begin{pmatrix} 1 - p_{12}^2 - 2p_{12} & p_{11} - p_{22} - p_{12}p_{22} \\ p_{11} - p_{22} - p_{12}p_{22} & 2p_{12} - p_{22}^2 \end{pmatrix}$$

$$= \frac{1}{p_{11}} \left( \frac{1}{p_{12}} - \frac{1}{p_{12}} + \frac{1}{p_{12}}$$

We look for a positive-definite matrix, P, which satisfies the algebraic Riccati equation. For this purpose, we first solve the quadratic equation for  $p_{12}$  to obtain

$$p_{12} = -1 \pm \sqrt{2} \tag{7}$$

Then we express  $p_{22}$  in terms of  $p_{12}$  as follows:

$$p_{22} = \sqrt{2p_{12}} \tag{8}$$

Since  $p_{22}$  must be real, we require  $p_{12} \ge 0$ , thus the negative sign in Eq.(7) must be rejected. Therefore, we have

$$p_{12} = -1 + \sqrt{2} \tag{9}$$

and

$$p_{22} = \sqrt{2\sqrt{2} - 2} \tag{10}$$

Solving for  $p_{11}$  we get

$$p_{11} = \sqrt{2p_{12}} + \sqrt{2}p_{12}^{3/2} \tag{11}$$

where  $p_{12}$  is given by Eq.(9). Hence, P is unique and positive definite. Therefore, the required LQR feedback law is the following:

$$u(t) = -\left(-1 + \sqrt{2}, \sqrt{2\sqrt{2} - 2}\right)x(t)$$
 (12)