

Robust and Optimal Control

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Necessary conditions for optimal control

The problem is to find an optimal control (u^*) that causes the system

$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t), \quad (1)$$

to follow a trajectory (x^*) that minimizes the performance measure

$$J(\mathbf{x}(t), t) = h(\mathbf{x}(t_f), t_f) + \int_{t_0}^{t_f} g(\mathbf{x}(\tau), \mathbf{u}(\tau), \tau) d\tau. \quad (2)$$

Please refer to chapter 5, page 184 (Kirk).

Necessary conditions for optimal control

The Hamiltonian is given by

$$\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}(t), t) \triangleq g(\mathbf{x}(t), \mathbf{u}(t), t) + \mathbf{p}^T(t) [\mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t)]. \quad (3)$$

The necessary conditions of optimality are

$$\dot{\mathbf{x}}^*(t) = \frac{\partial \mathcal{H}}{\partial \mathbf{p}}(\mathbf{x}^*(t), \mathbf{u}^*(t), \mathbf{p}^*(t), t), \quad (4)$$

$$\dot{\mathbf{p}}^*(t) = -\frac{\partial \mathcal{H}}{\partial \mathbf{x}}(\mathbf{x}^*(t), \mathbf{u}^*(t), \mathbf{p}^*(t), t), \quad (5)$$

$$0 = \frac{\partial \mathcal{H}}{\partial \mathbf{u}}(\mathbf{x}^*(t), \mathbf{u}^*(t), \mathbf{p}^*(t), t). \quad (6)$$

Example

The system

$$\dot{x}_1(t) = x_2(t) \quad (7)$$

$$\dot{x}_2(t) = -x_2(t) + u(t) \quad (8)$$

is to be controlled so that its control effort is conserved.
Therefore, the performance measure is given by

$$J(u) = \int_{t_0}^{t_f} \frac{1}{2} u^2(t) dt. \quad (9)$$

The Hamiltonian (36) is given by

$$\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}(t), t) = \frac{1}{2} u^2(t) + p_1(t)x_2(t) - p_2(t)x_2(t) + p_2(t)u(t). \quad (10)$$

Example

$$\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}(t), t) = \frac{1}{2}u^2(t) + p_1(t)x_2(t) - p_2(t)x_2(t) + p_2(t)u(t). \quad (11)$$

The necessary conditions for optimality are

$$\dot{p}_1^*(t) = \frac{\partial \mathcal{H}}{\partial x_1} = 0 \quad (12)$$

$$\dot{p}_2^*(t) = -\frac{\partial \mathcal{H}}{\partial x_2} = -p_1^*(t) + p_2^*(t), \quad (13)$$

and

$$0 = \frac{\partial \mathcal{H}}{\partial u} = u^*(t) + p_2^*(t). \quad (14)$$

Linear Regulator Problems

The plant is described by the linear state equations

$$\dot{\mathbf{x}}(t) = \mathbf{A}(t)\mathbf{x}(t) + \mathbf{B}(t)\mathbf{u}(t) \quad (15)$$

which may have time-varying coefficients. The performance measure to be minimized is

$$J = \frac{1}{2}\mathbf{x}^T(t_f)\mathbf{H}\mathbf{x}(t_f) + \int_{t_0}^{t_f} \frac{1}{2} [\mathbf{x}^T\mathbf{Q}\mathbf{x} + \mathbf{u}^T\mathbf{R}\mathbf{u}] dt, \quad (16)$$

where the final time t_f is fixed. Further, \mathbf{H} and \mathbf{Q} are real symmetric positive semi-definite matrices. Finally, \mathbf{R} is a real symmetric positive definite matrix.

The Hamiltonian is

$$\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}(t), t) \triangleq g(\mathbf{x}(t), \mathbf{u}(t), t) + \mathbf{p}^T(t) [\mathbf{f}(\mathbf{x}(t), \mathbf{u}(t), t)]. \quad (17)$$

Linear Regulator Problems

$$\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}(t), t) = \frac{1}{2}\mathbf{x}^T \mathbf{Q} \mathbf{x} + \frac{1}{2}\mathbf{u}^T \mathbf{R} \mathbf{u} + \mathbf{p}^T \mathbf{A}(t)\mathbf{x}(t) + \mathbf{p}^T \mathbf{B}(t)\mathbf{u}(t), \quad (18)$$

and the necessary conditions for optimality are

$$\dot{\mathbf{x}}^*(t) = \mathbf{A}(t)\mathbf{x}^*(t) + \mathbf{B}(t)\mathbf{u}^*(t), \quad (19)$$

$$\dot{\mathbf{p}}^*(t) = -\mathbf{Q}(t)\mathbf{x}^*(t) - \mathbf{A}^T(t)\mathbf{p}^*(t), \quad (20)$$

$$0 = \mathbf{R}(t)\mathbf{u}^*(t) + \mathbf{B}^T(t)\mathbf{p}^*(t). \quad (21)$$

Solving (54) for $\mathbf{u}^*(t)$ yields

$$\mathbf{u}^*(t) = -\mathbf{R}^{-1}(t)\mathbf{B}^T(t)\mathbf{p}^*(t). \quad (22)$$

Substitution of (55) into (52) yields

$$\dot{\mathbf{x}}^*(t) = \mathbf{A}(t)\mathbf{x}^*(t) - \mathbf{B}(t)\mathbf{R}^{-1}(t)\mathbf{B}^T(t)\mathbf{p}^*(t). \quad (23)$$

Linear Regulator Problems

Putting (53) and (56) into the following matrix format

$$\begin{bmatrix} \dot{\mathbf{x}}^*(t) \\ \vdots \\ \dot{\mathbf{p}}^*(t) \end{bmatrix} = \begin{bmatrix} \mathbf{A}(t) & | & -\mathbf{B}(t)\mathbf{R}^{-1}(t)\mathbf{B}^T(t) \\ \vdots & \vdots & \vdots \\ -\mathbf{Q}(t) & | & \mathbf{p}^*(t) \end{bmatrix} \begin{bmatrix} \mathbf{x}^*(t) \\ \vdots \\ \mathbf{p}^*(t) \end{bmatrix}. \quad (24)$$

The solution of these equations has the following form

$$\begin{bmatrix} \mathbf{x}^*(t_f) \\ \vdots \\ \mathbf{p}^*(t_f) \end{bmatrix} = \varphi(t_f, t) \begin{bmatrix} \mathbf{x}^*(t) \\ \vdots \\ \mathbf{p}^*(t) \end{bmatrix}, \quad (25)$$

where $\varphi(t_f, t)$ is the state-transition matrix of the system (57).

$$\begin{bmatrix} \mathbf{x}^*(t_f) \\ \vdots \\ \mathbf{p}^*(t_f) \end{bmatrix} = \begin{bmatrix} \varphi_{11}(t_f, t) & | & \varphi_{12}(t_f, t) \\ \vdots & \vdots & \vdots \\ \varphi_{21}(t_f, t) & | & \varphi_{22}(t_f, t) \end{bmatrix} \begin{bmatrix} \mathbf{x}^*(t) \\ \vdots \\ \mathbf{p}^*(t) \end{bmatrix}, \quad (26)$$

From the boundary condition, the final co-states are related to the final states using

$$\mathbf{p}^*(t_f) = \mathbf{H}\mathbf{x}^*(t_f). \quad (27)$$

Solving for $\mathbf{p}^*(t_f)$, we obtain

$$\mathbf{p}^*(t) = [\varphi_{22}(t_f, t) - \mathbf{H}\varphi_{12}(t_f, t)]^{-1} [\mathbf{H}\varphi_{11}(t_f, t) - \varphi_{21}(t_f, t)] \mathbf{x}^*(t). \quad (28)$$

$$\mathbf{p}^*(t) = \mathbf{K}(t)\mathbf{x}^*(t). \quad (29)$$

The optimal control law is given by

$$\mathbf{u}^*(t) = -\mathbf{R}^{-1}(t)\mathbf{B}^T(t)\mathbf{K}(t)\mathbf{x}^*(t). \quad (30)$$

Example

It is desired to determine the control law (using the principle of optimality and the Hamilton-Jacobi-Bellman equation) that causes the plant

$$\dot{x}_1 = x_2(t) \quad (31)$$

$$\dot{x}_2 = -x_1(t) - 2x_2(t) + u(t) \quad (32)$$

to minimize the performance measure

$$J = 10x_1^2(T) + \frac{1}{2} \int_0^T [x_1^2(t) + 2x_2^2(t) + u^2(t)] . \quad (33)$$

State Transition Matrix

Consider the scalar case

$$\dot{x}(t) = ax(t). \quad (34)$$

Taking the Laplace transform of (34), we obtain

$$sX(s) - x(0) = aX(s), \quad (35)$$

$$X(s) = \frac{x(0)}{s - a} = (s - a)^{-1}x(0). \quad (36)$$

Finally, inverse Laplace transform of (36) yields

$$x(t) = e^{at}x(0). \quad (37)$$

State Transition Matrix

Now consider the following homogenous state equation

$$\dot{\mathbf{x}}(t) = \mathbf{A}\mathbf{x}(t). \quad (38)$$

$$s\mathbf{X}(s) - \mathbf{x}(0) = \mathbf{A}\mathbf{X}(s), \quad (39)$$

$$\mathbf{X}(s) = (s\mathbf{I} - \mathbf{A})^{-1}\mathbf{x}(0). \quad (40)$$

The inverse Laplace transform yields

$$\mathbf{x}(t) = \mathcal{L}^{-1} [(s\mathbf{I} - \mathbf{A})^{-1}] \mathbf{x}(0) = e^{\mathbf{A}t}\mathbf{x}(0). \quad (41)$$

Therefore, the state transition matrix ($e^{\mathbf{A}t}$) is given by

$$e^{\mathbf{A}t} = \mathcal{L}^{-1} [(s\mathbf{I} - \mathbf{A})^{-1}]. \quad (42)$$

State Transition Matrix

- Calculate the state transition matrix of the following system

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ 2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad (43)$$

$$[s\mathbf{I} - \mathbf{A}] = \begin{bmatrix} (s+1) & 0 \\ -2 & (s+3) \end{bmatrix} \quad (44)$$

$$[s\mathbf{I} - \mathbf{A}]^{-1} = \begin{bmatrix} \frac{(s+3)}{(s+1)(s+3)} & 0 \\ \frac{2}{(s+1)(s+3)} & \frac{(s+1)}{(s+1)(s+3)} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{(s+1)} & 0 \\ \left(\frac{1}{(s+1)} - \frac{1}{(s+1)} \right) & \frac{1}{(s+3)} \end{bmatrix}$$

$$e^{\mathbf{A}t} = \mathcal{L}^{-1} [(s\mathbf{I} - \mathbf{A})^{-1}], \quad (45)$$

$$e^{\mathbf{A}t} = \begin{bmatrix} e^{-t} & 0 \\ (e^{-t} - e^{-3t}) & e^{-3t} \end{bmatrix}.$$

State Transition Matrix

- Calculate the state transition matrix of the following system

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad (46)$$

$$[s\mathbf{I} - \mathbf{A}] = \begin{bmatrix} s & -1 \\ 2 & (s+3) \end{bmatrix} \quad (47) \quad e^{\mathbf{A}t} = \mathcal{L}^{-1} [(s\mathbf{I} - \mathbf{A})^{-1}], \quad (48)$$
$$= \begin{bmatrix} 2e^t - e^{-2t} & e^{-t} - e^{-2t} \\ -2e^{-t} + 2e^{-2t} & -e^{-t} + 2e^{-2t} \end{bmatrix}.$$

$$[s\mathbf{I} - \mathbf{A}]^{-1} = \begin{bmatrix} \frac{(s+3)}{(s+1)(s+2)} & \frac{1}{(s+1)(s+2)} \\ \frac{-2}{(s+1)(s+2)} & \frac{s}{(s+1)(s+2)} \end{bmatrix}$$

State Transition Matrix

If the matrix \mathbf{A} can be transformed into a diagonal form, then the state transition matrix $e^{\mathbf{A}t}$ is given by

$$e^{\mathbf{A}t} = \mathbf{P}e^{\mathbf{D}t}\mathbf{P}^{-1} = \mathbf{P} \begin{bmatrix} e^{\lambda_1 t} & 0 & \dots & 0 \\ 0 & e^{\lambda_2 t} & \dots & 0 \\ \vdots & \dots & \ddots & 0 \\ 0 & \dots & 0 & e^{\lambda_n t} \end{bmatrix} \mathbf{P}^{-1}, \quad (49)$$

where \mathbf{P} is a diagonalizing matrix for \mathbf{A} . Further, λ_i is the i th eigenvalue of the matrix \mathbf{A} , for $i = 1, \dots, n$.

State Transition Matrix

Derivation: Consider the following homogenous state equation

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x}, \quad (50)$$

and the following similarity transformation:

$$\mathbf{x} = \mathbf{P}\boldsymbol{\xi}, \quad \dot{\mathbf{x}} = \mathbf{P}\dot{\boldsymbol{\xi}}. \quad (51)$$

Substituting (51) in (50) yields

$$\dot{\boldsymbol{\xi}} = \mathbf{P}^{-1}\mathbf{A}\mathbf{P}\boldsymbol{\xi} = \mathbf{D}\boldsymbol{\xi}. \quad (52)$$

Solution of (52) is

$$\boldsymbol{\xi}(t) = e^{\mathbf{D}t}\boldsymbol{\xi}(0), \quad (53)$$

using (51)

$$\mathbf{x}(t) = \mathbf{P}\boldsymbol{\xi}(t) = \mathbf{P}e^{\mathbf{D}t}\boldsymbol{\xi}(0), \quad \mathbf{x}(0) = \mathbf{P}\boldsymbol{\xi}(0). \quad (54)$$

Therefore

$$\mathbf{x}(t) = \mathbf{P}e^{\mathbf{D}t}\mathbf{P}^{-1}\mathbf{x}(0) = e^{\mathbf{A}t}\mathbf{x}(0). \quad (55)$$

State Transition Matrix

- Calculate the state transition matrix of the following system

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad (56)$$

The eigenvalues of \mathbf{A} are $\lambda_1 = 0$ and $\lambda_2 = -2$. A similarity transformation matrix \mathbf{P} is

$$\mathbf{P} = \begin{bmatrix} 1 & 1 \\ 0 & -2 \end{bmatrix}. \quad (57)$$

Using (49) to calculate the state transition matrix

$$e^{\mathbf{A}t} = \mathbf{P}e^{\mathbf{D}t}\mathbf{P}^{-1} \quad (58)$$

$$= \begin{bmatrix} 1 & 1 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} e^0 & 0 \\ 0 & e^{-2t} \end{bmatrix} \begin{bmatrix} 1 & \frac{1}{2} \\ 0 & -\frac{1}{2} \end{bmatrix}$$
$$e^{\mathbf{A}t} = \begin{bmatrix} 1 & \frac{1}{2}(1 - e^{-2t}) \\ 0 & e^{-2t} \end{bmatrix}. \quad (59)$$

Linear Tracking Problem

Consider the state equations

$$\dot{\mathbf{x}}(t) = \mathbf{A}(t)\mathbf{x}(t) + \mathbf{B}(t)\mathbf{u}(t), \quad (60)$$

and the performance measure to be minimized is

$$J = \frac{1}{2} [\mathbf{x}(t_f) - \mathbf{r}(t_f)]^T \mathbf{H} [\mathbf{x}(t_f) - \mathbf{r}(t_f)] + \int_{t_0}^{t_f} \frac{1}{2} [(\mathbf{x} - \mathbf{r})^T \mathbf{Q}(\mathbf{x} - \mathbf{r}) + \mathbf{u}^T \mathbf{R} \mathbf{u}] dt, \quad (61)$$

where $r(t)$ is the desired reference value and the final time t_f is fixed. Further, \mathbf{H} and \mathbf{Q} are real symmetric positive semi-definite matrices. Finally, \mathbf{R} is a real symmetric positive definite matrix. The Hamiltonian is given by

$$\mathcal{H}(\mathbf{x}(t), \mathbf{u}(t), \mathbf{p}(t), t) = \frac{1}{2} [(\mathbf{x} - \mathbf{r})^T \mathbf{Q}(\mathbf{x} - \mathbf{r}) + \mathbf{u}^T \mathbf{R} \mathbf{u}] + \mathbf{p}^T(t) \mathbf{A}(t) \mathbf{x}(t) + \mathbf{p}^T(t) \mathbf{B}(t) \mathbf{u}(t). \quad (62)$$

Linear Tracking Problem

The necessary conditions of optimality are

$$\begin{aligned}\dot{\mathbf{x}}^*(t) &= \frac{\partial \mathcal{H}}{\partial \mathbf{p}}(\mathbf{x}^*(t), \mathbf{u}^*(t), \mathbf{p}^*(t), t), \\ &= \mathbf{A}(t)\mathbf{x}^*(t) + \mathbf{B}(t)\mathbf{u}^*(t).\end{aligned}\quad (63)$$

$$\begin{aligned}\dot{\mathbf{p}}^*(t) &= -\frac{\partial \mathcal{H}}{\partial \mathbf{x}}(\mathbf{x}^*(t), \mathbf{u}^*(t), \mathbf{p}^*(t), t), \\ &= -\mathbf{Q}(t)\mathbf{x}^*(t) - \mathbf{A}^T(t)\mathbf{p}^*(t) + \mathbf{Q}(t)\mathbf{r}(t).\end{aligned}\quad (64)$$

$$\begin{aligned}0 &= \frac{\partial \mathcal{H}}{\partial \mathbf{u}}(\mathbf{x}^*(t), \mathbf{u}^*(t), \mathbf{p}^*(t), t), \\ &= \mathbf{R}(t)\mathbf{u}^*(t) + \mathbf{B}^T(t)\mathbf{p}^*(t).\end{aligned}\quad (65)$$

Therefore

$$\mathbf{u}^*(t) = -\mathbf{R}^{-1}(t)\mathbf{B}^T(t)\mathbf{p}^*(t).\quad (66)$$

Linear Tracking Problem

Substituting (66) in (63) yields

$$\begin{bmatrix} \dot{\mathbf{x}}^*(t) \\ \dot{\mathbf{p}}^*(t) \end{bmatrix} = \begin{bmatrix} \mathbf{A}(t) & | & -\mathbf{B}(t)\mathbf{R}^{-1}(t)\mathbf{B}^T(t) \\ \hline -\mathbf{Q}(t) & | & \mathbf{p}^*(t) \end{bmatrix} \begin{bmatrix} \mathbf{x}^*(t) \\ \mathbf{p}^*(t) \end{bmatrix} + \begin{bmatrix} 0 \\ \mathbf{Q}(t)\mathbf{r}(t) \end{bmatrix}. \quad (67)$$

Solution of (67) is

$$\begin{bmatrix} \mathbf{x}^*(t_f) \\ \mathbf{p}^*(t_f) \end{bmatrix} = \varphi(t_f, t) \begin{bmatrix} \mathbf{x}^*(t) \\ \mathbf{p}^*(t) \end{bmatrix} + \int_t^{t_f} \varphi(t_f, \tau) \begin{bmatrix} 0 \\ \mathbf{Q}(\tau)\mathbf{r}(\tau) \end{bmatrix} d\tau, \quad (68)$$

where φ is the state transition matrix. If φ is partitioned, we can replace the integral with a $2n$ vector

$$\begin{bmatrix} \mathbf{x}^*(t_f) \\ \mathbf{p}^*(t_f) \end{bmatrix} = \varphi(t_f, t) \begin{bmatrix} \mathbf{x}^*(t) \\ \mathbf{p}^*(t) \end{bmatrix} + \begin{bmatrix} \mathbf{f}_1(t) \\ \mathbf{f}_2(t) \end{bmatrix}. \quad (69)$$

Linear Tracking Problem

$$\begin{bmatrix} \mathbf{x}^*(t_f) \\ \mathbf{p}^*(t_f) \end{bmatrix} = \begin{bmatrix} \varphi_{11}(t_f, t) & | & \varphi_{12}(t_f, t) \\ \hline \varphi_{21}(t_f, t) & | & \varphi_{22}(t_f, t) \end{bmatrix} \begin{bmatrix} \mathbf{x}^*(t) \\ \mathbf{p}^*(t) \end{bmatrix} + \begin{bmatrix} \mathbf{f}_1(t) \\ \mathbf{f}_2(t) \end{bmatrix}. \quad (70)$$

The boundary conditions are

$$\mathbf{p}^*(t_f) = \mathbf{H}\mathbf{x}^*(t_f) - \mathbf{H}\mathbf{r}(t_f). \quad (71)$$

Solving for $\mathbf{p}^*(t_f)$, we obtain

$$\mathbf{p}^*(t) = [\varphi_{22}(t_f, t) - \mathbf{H}\varphi_{12}(t_f, t)]^{-1} [\mathbf{H}\varphi_{11}(t_f, t) - \varphi_{21}(t_f, t)] \mathbf{x}^*(t) + [\varphi_{22}(t_f, t) - \mathbf{H}\varphi_{12}(t_f, t)]^{-1} [\mathbf{H}\mathbf{f}_1(t) - \mathbf{H}\mathbf{r}(t_f) - \mathbf{f}_2(t)] \quad (72)$$

$$\mathbf{p}^*(t) = \mathbf{K}(t)\mathbf{x}^*(t) + \mathbf{s}(t). \quad (73)$$

The optimal control law is given by

$$\mathbf{u}^*(t) = -\mathbf{R}^{-1}(t)\mathbf{B}^T(t)\mathbf{K}(t)\mathbf{x}(t) - \mathbf{R}^{-1}(t)\mathbf{B}^{-1}(t)\mathbf{s}(t). \quad (74)$$

Questions please