

Lecture 7: LQR from E-L eqns

Note Title

11-06-2008

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

$$x(t_0) \longrightarrow x(t_f) \equiv 0$$

t_0 & t_f fixed.

using acceptable levels of control
+ not exceeding acceptable levels
of state on the way

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

$$S_f, Q, R > 0$$

E-L Necessary Conditions:

$$\dot{x} = Ax + Bu$$
$$\dot{\lambda}^T = -\frac{\partial H}{\partial x}$$

$$\lambda(t_f) = S_f x(t_f)$$

$$0 = \frac{\partial H}{\partial u}$$

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

$$\dot{\lambda} = -Qx - A^T \lambda$$

$$0 = Ru + B^T \lambda \Rightarrow u = -R^{-1} B^T \lambda$$

using ①:

$$\begin{bmatrix} \dot{x} \\ \dot{\lambda} \end{bmatrix} = \begin{bmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix} \Bigg| \begin{array}{l} x(t_0) \text{ given} \\ \lambda(t_f) = S_f x(t_f) \end{array}$$

if $x(t_f)$ is known,
 Now, the solutions to these eqns
 can be written as:

$$\begin{aligned} x(t) &= X(t, t_f) x(t_f) \\ \lambda(t) &= \Lambda(t, t_f) x(t_f) \end{aligned}$$

where $X(t_f, t_f) = I$
 & $\Lambda(t_f, t_f) = S_f$

At $t = t_0$: $x(t_0) = X(t_0, t_f) x(t_f)$
 or $x(t_f) = [X(t_0, t_f)]^{-1} x(t_0)$

$$x(t) = X(t, t_f) [X(t_0, t_f)]^{-1} x(t_0)$$

$$\lambda(t) = \Lambda(t, t_f) [X(t_0, t_f)]^{-1} x(t_0)$$

Substituting into $u(t) = -R^{-1} B^T \lambda$

$$u(t) = -R^{-1} B^T \Lambda(t, t_f) [X(t_0, t_f)]^{-1} x(t_0)$$

Clearly, we can replace t_0 by any
 t , hence

$$u(t) = -R^{-1} B^T \Lambda(t, t_f) [X(t, t_f)]^{-1} x(t)$$

Similarly $\lambda(t) = \Lambda(t, t_f) [X(t, t_f)]^{-1} x(t)$

Differential Riccati Eqn $S(t)$

Use $\lambda(t) = S(t) x(t)$

in $\dot{\lambda} = -Qx - A^T \lambda$ to get

$$\dot{x}x + S\dot{x} = -Qx - A^T Sx$$

$$\left[\text{Now substituting } \dot{x} = Ax - BR^{-1}B^T \lambda \right. \\ \left. = Ax - BR^{-1}B^T Sx \right]$$

$$S\dot{x} + S[Ax - BR^{-1}B^T Sx] = -Qx - A^T Sx$$

$$\left[S + SA + A^T S - SBR^{-1}B^T S + Q \right] x = 0$$

Since $x(t) \neq 0$:

$$S + SA + A^T S - SBR^{-1}B^T S + Q = 0$$

$$S(t_f) = S_f \quad \hookrightarrow \text{DRE}$$

* must be solved backwards in time.

to find $S(t_0)$

$$* \text{ then } \lambda(t_0) = S(t_0)x(t_0)$$

$$* \text{ then } \begin{bmatrix} \dot{x} \\ \dot{\lambda} \end{bmatrix} = \begin{bmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{bmatrix} \begin{bmatrix} x \\ \lambda \end{bmatrix}$$

can be solved to compute optimal trajectories

* Otherwise $S(t)$ can be used to generate the feedback law

Example: $\dot{x} = u$ $x(t_0), t_0, t_f$ sp.

$$\min J = \frac{1}{2} c[x(t_f)]^2 + \frac{1}{2} \int_{t_0}^{t_f} u^2 dt$$

x, u - values

$$H = \frac{1}{2} u^2 + \lambda u \quad (\lambda - \text{scalar})$$

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

$$\frac{\partial H}{\partial u} = 0 \Rightarrow u + \lambda = 0 \Rightarrow u = -\lambda$$

$$\lambda(t_f) = c \lambda(t_0) \quad (\text{odd condition})$$

$$x(t) = -[c \lambda(t_f)](t - t_0) + x(t_0)$$

$$\Rightarrow x(t_f) = \frac{x(t_0)}{1 + c(t_f - t_0)}$$

$$u(t, t_f) = -\frac{1}{\frac{1}{c} + (t_f - t_0)} x(t_0)$$

$$u(t) = -\frac{1}{\frac{1}{c} + t_f - t} x(t)$$

Note: $x(t_f) \rightarrow 0$ as $c \rightarrow \infty$.