

LQ with Frequency Shaped Cost Function (FSLQ)

Background

Parseval's Theorem

Frequency-shaped LQ cost function

Transformation to a standard LQ

Big picture

why are we learning this:

- ▶ in standard LQ, Q and R are constant matrices in the cost function

$$J = \int_0^{\infty} \left(x^T(t) Q x(t) + \rho u^T(t) R u(t) \right) dt \quad (1)$$

- ▶ how can we introduce more design freedom for Q and R ?

Connection between time and frequency domains

Theorem (Parseval's Theorem)

For a square integrable signal $f(t)$ defined on $[0, \infty)$

$$\int_0^{\infty} f^T(t) f(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} F^T(-j\omega) F(j\omega) d\omega$$

1D case:
$$\int_0^{\infty} |f(t)|^2 dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} |F(j\omega)|^2 d\omega$$

Intuition: energy in time-domain equals energy in frequency domain

For the general case, $f(t)$ can be acausal. We have

$$\int_{-\infty}^{\infty} f^T(t) f(t) dt = \frac{1}{2\pi} \int_{-\infty}^{\infty} F^T(-j\omega) F(j\omega) d\omega$$

Discrete-time version:

$$\sum_{k=-\infty}^{\infty} f^T(k) f(k) = \frac{1}{2\pi} \int_{-\infty}^{\infty} F^T(e^{-j\omega}) F(e^{j\omega}) d\omega$$

History

Marc-Antoine Parseval (1755-1836):

- ▶ French mathematician
- ▶ published just five (but important) mathematical publications in total (source: Wikipedia.org)

Frequency-domain LQ cost function

From Parseval's Theorem, the LQ cost in frequency domain is

$$J = \int_0^{\infty} \left(x^T(t) Q x(t) + \rho u^T(t) R u(t) \right) dt \quad (2)$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \left(X^T(-j\omega) Q X(j\omega) + \rho U^T(-j\omega) R U(j\omega) \right) d\omega \quad (3)$$

Frequency-shaped LQ expands Q and R to frequency-dependent functions:

$$J = \frac{1}{2\pi} \int_{-\infty}^{\infty} \left(X^T(-j\omega) Q(j\omega) X(j\omega) + \rho U^T(-j\omega) R(j\omega) U(j\omega) \right) d\omega \quad (4)$$

Frequency-domain LQ cost function

Let

$$Q(j\omega) = Q_f^T(-j\omega)Q_f(j\omega) \succeq 0, \quad X_f(j\omega) = Q_f(j\omega)X(j\omega)$$

$$R(j\omega) = R_f^T(-j\omega)R_f(j\omega) \succ 0, \quad U_f(j\omega) = R_f(j\omega)U(j\omega)$$

(4) becomes

$$J = \frac{1}{2\pi} \int_{-\infty}^{\infty} \left(X_f^T(-j\omega)X_f(j\omega) + \rho U_f^T(-j\omega)U_f(j\omega) \right) d\omega$$

which is equivalent to (using Parseval's Theorem again)

$$\boxed{J = \int_0^{\infty} \left(x_f^T(t)x_f(t) + \rho u_f^T(t)u_f(t) \right) dt} \quad (5)$$

Frequency-domain LQ cost function

Summarizing, we have:

► plant:

$$\begin{cases} \dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) \end{cases} \quad (6)$$

► new cost:

$$J = \int_0^\infty \left(x_f^T(t) x_f(t) + \rho u_f^T(t) u_f(t) \right) dt \quad (7)$$

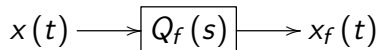
► filtered states and inputs:

$$x(t) \longrightarrow \boxed{Q_f(s)} \longrightarrow x_f(t), \quad u(t) \longrightarrow \boxed{R_f(s)} \longrightarrow u_f(t)$$

We just need to translate the problem to a standard one [which we know (very well) how to solve]

Frequency-domain weighting filters

state filtering

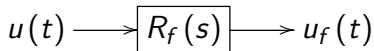


- ▶ a MIMO process in general: if $x(t) \in \mathbb{R}^n$ and $x_f(t) \in \mathbb{R}^q$, then $Q_f(s)$ is a $q \times n$ transfer function matrix
- ▶ $Q_f(s)$: state filter; designer's choice; can be selected to meet the desired control action and the performance requirements
- ▶ write $Q_f(s) = C_1(sI - A_1)^{-1}B_1 + D_1$ in the general state-space realization:

$$\begin{cases} \dot{z}_1(t) = A_1 z_1(t) + B_1 x(t) \\ x_f(t) = C_1 z_1(t) + D_1 x(t) \end{cases} \quad (8)$$

Frequency-domain weighting filters

input filtering



- ▶ $R_f(s)$: input filter; designer's choice; can be selected to meet the robustness requirements
- ▶ write $R_f(s) = C_2(sI - A_2)^{-1}B_2 + D_2$ in the general state-space realization:

$$\begin{cases} \dot{z}_2(t) &= A_2 z_2(t) + B_2 u(t) \\ u_f(t) &= C_2 z_2(t) + D_2 u(t) \end{cases} \quad (9)$$

Back to time-domain design

Combining (6), (8) and (9) gives the enlarged system

$$\frac{d}{dt} \underbrace{\begin{bmatrix} x(t) \\ z_1(t) \\ z_2(t) \end{bmatrix}}_{x_e(t)} = \underbrace{\begin{bmatrix} A & 0 & 0 \\ B_1 & A_1 & 0 \\ 0 & 0 & A_2 \end{bmatrix}}_{A_e} \begin{bmatrix} x(t) \\ z_1(t) \\ z_2(t) \end{bmatrix} + \underbrace{\begin{bmatrix} B \\ 0 \\ B_2 \end{bmatrix}}_{B_e} u(t)$$

and

$$x_f(t) = \underbrace{[D_1 \ C_1 \ 0]}_{C_e} \begin{bmatrix} x(t) \\ z_1(t) \\ z_2(t) \end{bmatrix}$$
$$u_f(t) = [0 \ 0 \ C_2]x_e(t) + D_2 u(t)$$

Summary of solution

With the enlarged system, the cost

$$J = \int_0^\infty \left(x_f^T(t) x_f(t) + \rho u_f^T(t) u_f(t) \right) dt \quad (10)$$

translates to

$$J = \int_0^\infty \left(x_e^T(t) Q_e x_e(t) + 2u^T(t) \underbrace{\begin{bmatrix} 0 & 0 & \rho D_2^T C_2 \end{bmatrix}}_{N_e} x_e(t) + u^T(t) \underbrace{\rho D_2^T D_2}_{R_e} u(t) \right) dt$$

$$Q_e = \begin{bmatrix} D_1^T D_1 & D_1^T C_1 & 0 \\ C_1^T D_1 & C_1^T C_1 & 0 \\ 0 & 0 & \rho C_2^T C_2 \end{bmatrix}$$

- ▶ solution (see appendix for more details):

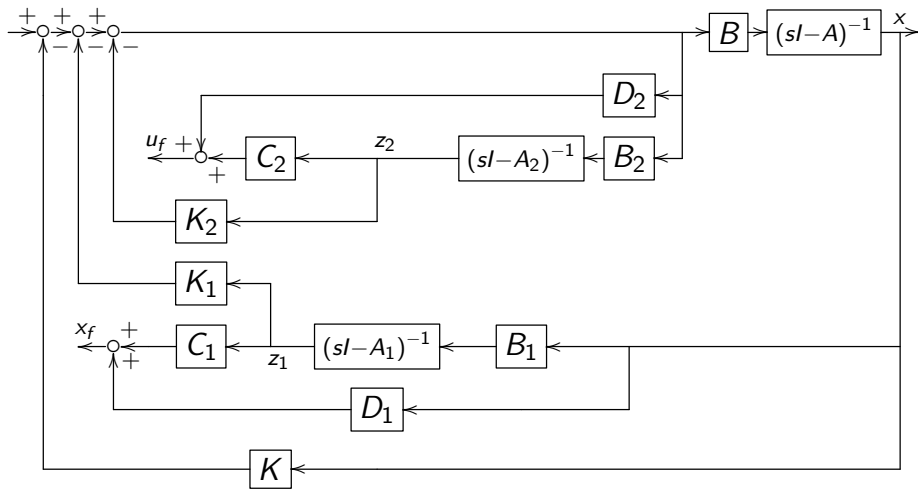
$$u(t) = -R_e^{-1} (B_e^T P_e + N_e) x_e(t) = -Kx(t) - K_1 z_1(t) - K_2 z_2(t)$$

- ▶ algebraic Riccati equation:

$$A_e^T P_e + P_e A_e - (B_e^T P_e + N_e)^T R_e^{-1} (B_e^T P_e + N_e) + Q_e = 0$$

Implementation

structure of the FSLQ system:



Appendix: general LQ solution

Consider LQ problems with cost

$$J = \int_0^\infty \left(x^T(t) \underbrace{C^T C}_Q x(t) + 2u^T(t) N x(t) + u^T(t) R u(t) \right) dt \quad (11)$$

and system dynamics

$$\dot{x}(t) = Ax(t) + Bu(t)$$

- ▶ assume (A, B) is controllable/stabilizable and (A, C) is observable/detectable
 - ▶ the solution of the problem is

$$u(t) = -R^{-1}(B^T P + N)x(t)$$

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

Appendix: general LQ solution

Intuition: under the assumptions, we know we can stabilize the system and drive $x(t)$ to zero. Consider Lyapunov function

$$V(t) = x^T(t) P x(t), \quad P = P^T \succ 0$$

$$\begin{aligned} \cancel{V(\infty)} - V(0) &= \int_0^\infty \dot{V}(t) dt \\ &= \int_0^\infty \left(x^T(t) (PA + A^T P) x(t) + 2x^T(t) P B u(t) \right) dt \end{aligned}$$

Adding (11) on both sides yields

$$\begin{aligned} V(\infty) - V(0) + J &= \\ \int_0^\infty \left(x^T(t) (Q + PA + A^T P) x(t) + 2x^T(t) (PB + N^T) u(t) + u^T(t) R u(t) \right) dt \end{aligned} \quad (12)$$

- to minimize the cost, we are going to re-organize the terms in (12) into some “squared” terms

Appendix: general LQ solution

“completing the squares”:

$$2x^T(t) \left(PB + N^T \right) u(t) + u^T(t) R u(t) = \left\| R^{1/2} u(t) + R^{-1/2} \left(B^T P + N \right) x(t) \right\|_2^2 \\ - x^T(t) \left(PB + N^T \right) R^{-1} \left(B^T P + N \right) x(t)$$

hence (12) is actually

$$\cancel{V(\infty)} \overset{0}{-} V(0) + J \\ = \int_0^\infty \left[x^T(t) \left(Q + PA + A^T P - \left(PB + N^T \right) R^{-1} \left(B^T P + N \right) \right) x(t) \right. \\ \left. + \left\| R^{1/2} u(t) + R^{-1/2} \left(B^T P + N \right) x(t) \right\|_2^2 \right] dt$$

hence $J_{\min} = V(0) = x^T(0) P x(0)$ is achieved when

$$Q + PA + A^T P - \left(PB + N^T \right) R^{-1} \left(B^T P + N \right) = 0$$

$$\text{and } u(t) = -R^{-1} \left(B^T P + N \right) x(t)$$