Mass Spring Model with LQR Control

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Model and Cost Rate 1

Model:

$$X_{n+1} = AX_n + BU_n + Wn$$

Cost rate:

$$J_n(G) = \mathbb{E}[|X_{n+1}|^2 + r|U_n|^2]$$

We are looking for the control in the form of $U_n = GX_n$, where $G = \begin{pmatrix} G_1 & G_2 \end{pmatrix}$, . Therefore, our aim is to find ${\cal G}$ such that the cost rate is minimized in the steady state.

Covariance matrix S of X

• Let S_n be the covariance matrix of X_n , R be the covariance matrix of noise W_n ,

$$S_{n+1} = (A + BG)S_n(A + BG)^T + R$$
 (*)

• In the steady state, let M = A + BG, we will have

$$S = (A + BG)S(A + BG)^T + R = MSM^T + R$$

ullet As S is a symmetric matrix, we can rewrite the above equation as

$$\begin{pmatrix} S_{11} \\ S_{12} \\ S_{22} \end{pmatrix} = \begin{pmatrix} M_{11}^2 & 2M_{11}M_{12} & M_{12}^2 \\ M_{11}M_{21} & M_{11}M_{22} + M_{12}M_{21} & M_{22}M_{12} \\ M_{21}^2 & 2M_{21}M_{22} & M_{22}^2 \end{pmatrix} \begin{pmatrix} S_{11} \\ S_{12} \\ S_{22} \end{pmatrix} + \begin{pmatrix} R_{11} \\ R_{12} \\ R_{22} \end{pmatrix}$$

• Let
$$\xi_S = \begin{pmatrix} S_{11} \\ S_{12} \\ S_{22} \end{pmatrix}$$
, $D = \begin{pmatrix} M_{11}^2 & 2M_{11}M_{12} & M_{12}^2 \\ M_{11}M_{21} & M_{11}M_{22} + M_{12}M_{21} & M_{22}M_{12} \\ M_{21}^2 & 2M_{21}M_{22} & M_{22}^2 \end{pmatrix}$, $\xi_R = \begin{pmatrix} R_{11} \\ R_{12} \\ R_{22} \end{pmatrix}$, then
$$\xi_S = D\xi_S + \xi_{P_S} (I - D)\xi_S = \xi_P$$

$$\xi_S = D\xi_S + \xi_R, (I - D)\xi_S = \xi_R$$

2.1 Differentiate S w.r.t G

ullet If differentiating S with respect to parameter heta in the steady state, according to chain rule we will get

$$\dot{S} = M\dot{S}M^T + \dot{M}SM^T + MS\dot{M}^T$$

• Let $u = \dot{M}SM^T + MS\dot{M}^T$, then by the similar method above, we will have

$$(I-D)\xi_{\dot{S}} = \xi_u$$

• Another way to get $\xi_{\dot{S}}$ is to differentiate $(I-D)\xi_S$:

$$0 = -\dot{D}\xi_S + (I - D)\xi_{\dot{S}}$$

$$\xi_{\dot{S}} = (I - D)^{-1}\dot{D}\xi_S$$

$$= (I - D)^{-1}\dot{D}(I - D)^{-1}\xi_R$$

• Let's denote $E = (I - D)^{-1}$, then

$$\xi_{\dot{S}} = E\dot{D}E\xi_R$$

2.2 Use S to represent the cost rate

- $\mathbb{E}[X^T X] = \operatorname{tr}(S)$
- $\mathbb{E}[U^T r U] = \operatorname{tr}(X^T G^T r G X) = \operatorname{tr}(G^T r G X X^T) = \operatorname{tr}(G^T r G S) = \operatorname{tr}(r G S G^T)$ (cyclic propery of trace)
- Then the cost rate can be represented as

$$J_n(G) = \operatorname{tr}(S) + \operatorname{tr}(rGSG^T) \tag{**}$$

• Differentiate J_n w.r.t θ gives

$$\begin{split} 0 &= \frac{\partial J}{\partial \theta} \\ &= \operatorname{tr}(\dot{S}) + \operatorname{tr}(r(\dot{G}SG^T + G\dot{S}G^T + GS\dot{G}^T)) \\ &= \operatorname{tr}(\dot{S}) + \operatorname{tr}(r(2GS\dot{G}^T + G\dot{S}G^T)) \end{split}$$

where θ is G_1 or G_2

Then, we want to use G to represent S and \dot{S} in the above equation:

- $\xi_S = E\xi_R$
- $\xi_{\dot{S}} = E\dot{D}E\xi_R$
- $\operatorname{tr}(\dot{S}) = \begin{pmatrix} 1 & 0 & 1 \end{pmatrix} \xi_{\dot{S}} := q \xi_{\dot{S}}$

- $\operatorname{tr}(r(2GS\dot{G}^T + G\dot{S}G^T)) = r(2GS\dot{G}^T + G\dot{S}G^T)$
- $G\dot{S}G^T = \begin{pmatrix} G_1^2 & 2G_1G_2 & G_2^2 \end{pmatrix} \xi_{\dot{S}} := H\xi_{\dot{S}}$
- $\bullet \ \ {\rm Let} \ \xi_G = \begin{pmatrix} G_1 & G_2 & 0 \\ 0 & G_1 & G_2 \end{pmatrix}, \ {\rm then} \ S\dot{G}^T = \xi_{\dot{G}}\xi_S$
- Plugging them in the derivative of cost rate function:

$$0 = q\xi_{\dot{S}} + r(H\xi_{\dot{S}} + 2G\xi_{\dot{G}}\xi_S)$$

• Plugging in the formulas for $\xi_{\dot{S}}$ and ξ_{S} :

$$0 = [qE\dot{D} + r(HE\dot{D} + 2G\xi_{\dot{G}})]E\xi_R$$

The equation is only about G_1 and G_2 , in the form of $f(G_1, G_2) = 0$, so we can use Newton method to find out the value of G_1 and G_2 .

Remark: Equation (*) and (**) are verified in the Python Code.

3 Appendix

$$\bullet \frac{\partial D}{\partial G_1} = \begin{pmatrix} 0 & 0 & 0 \\ a_{11} & a_{12} & 0 \\ 2(a_{21} + G_1) & 2(a_{22} + G_2) & 0 \end{pmatrix},
\frac{\partial D}{\partial G_2} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & a_{11} & a_{12} \\ 0 & 2(a_{21} + G_1) & 2(a_{22} + G_2) \end{pmatrix},$$

•
$$D = \begin{pmatrix} M_{11}^2 & 2M_{11}M_{12} & M_{12}^2 \\ M_{11}M_{21} & M_{11}M_{22} + M_{12}M_{21} & M_{22}M_{12} \\ M_{21}^2 & 2M_{21}M_{22} & M_{22}^2 \end{pmatrix}$$

with $M_{11} = a_{11}, M_{12} = a_{12}, M_{21} = a_{21} + G_1, M_{22} = a_{22} + G_2$

$$\bullet \ \ \tfrac{\partial \xi_G}{\partial G_1} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}, \, \tfrac{\partial \xi_G}{\partial G_2} = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix},$$

References