

Controllability Gramian

$$\begin{aligned}
x_{t+1} &= Ax_t + Bu_t & x_0 \\
x_1 &= Ax_0 + Bu_0 \\
x_2 &= A(Ax_0 + Bu_0) + Bu_1 \\
x_3 &= A^3x_0 + A^2Bu_0 + ABu_1 + Bu_2 \\
&\vdots \\
x_t &= A^tx_0 + A^{t-1}Bu_0 + A^{t-2}Bu_1 + \dots + ABu_{t-2} + Bu_{t-1} \\
x_t &= A^tx_0 + \begin{bmatrix} A^{t-1}B & \dots & \dots & AB & B \end{bmatrix} \begin{bmatrix} u_0 \\ \vdots \\ u_{t-2} \\ u_{t-1} \end{bmatrix} \\
U_{t-1} &= (u_0, u_1, \dots, u_{t-2}, u_{t-1})^T \\
\mathcal{W}_c(t) &= (A^{t-1}B, A^{t-2}B, \dots, AB, B) \\
x_t &= A^tx_0 + \sum_{s=0}^{t-1} A^{t-s-1}Bu_s = A^tx_0 + \mathcal{W}_c(t)U_{t-1}
\end{aligned}$$

define,

$$\Sigma_t^c = \mathcal{W}_c(t)\mathcal{W}_c^T(t) \quad n \times n$$

Controllable if Σ^c is non singular. Discrete controllability Gramian:

$$\begin{aligned}
\Sigma_\tau^c &= A^{\tau-1}BB^T(A^{\tau-1})^T + \dots + ABB^TA^T + BB^T \\
&= \sum_{\tau=0}^{\tau-1} A^\tau BB^T(A^\tau)^T
\end{aligned} \tag{1}$$

upon expansion, satisfies the discrete time Lyapunov equation:

$$\Sigma_{\tau+1}^c = A\Sigma_\tau^c A^T + BB^T \quad \Sigma_0^c = 0 \tag{2}$$

consider the difference equation:

$$u_{n+1} - a^2u_n - b^2 = 0$$

in continuous time form:

$$\dot{u} - a^2u - b^2 = 0$$

homogeneous sol'n:

$$\dot{u} - a^2u = 0$$

$$\dot{u} = a^2u$$

$$u = Ce^{a^2t}$$

particular sol'n:
let $u=d=\text{constant}$

$$0 - a^2 d - b^2 = 0$$

$$d = -\frac{b^2}{a^2}$$

general sol'n:

$$u = Ce^{a^2 t} - \frac{b^2}{a^2}$$

$$u_0 = c - \frac{b^2}{a^2} = 0 \Rightarrow c = \frac{b^2}{a^2}$$

$$u = \frac{b^2}{a^2} e^{a^2 t} - \frac{b^2}{a^2}$$

$$\dot{u} = b^2 e^{a^2 t} \Rightarrow u = \int b^2 e^{a^2 t} dt$$

hence, analog to the scalar case (this is admittedly a hand-wavy argument), the continuous controllability Gramian is

$$\Sigma_\tau^c = \int_0^\tau e^{A\tau} B B^T (e^{A\tau})^T d\tau$$

it turns out that the continuous time version of Lyapunov equation (eq. (2)), is:

$$\dot{\Sigma}_t^c = A \Sigma_t^c + \Sigma_t^c A^T + B B^T$$

Furthermore, suppose we want to drive x_t to a desirable state at time $t = t_1$. Given, $x_t = A^t x_0 + \mathcal{W}_c(t) U_{t-1}$ for $t \geq n$:

$$\text{error : } \tilde{x}_t = x_t - A^t x_0 = (W)_c(t) U_{t-1}$$

define the cost function,

$$J = \frac{1}{2} U_{t-1}^T U_{t-1} \quad \text{s.t.} \quad \mathcal{W}_c(t) U_{t-1} = \tilde{x}_t = x_r - A^t x_0$$

Using the method of Lagrange multipliers,

$$J = \frac{1}{2} U_{t-1}^T U_{t-1} - \lambda^T (\mathcal{W}_c(t) U_{t-1} - (x_r - A^t x_0))$$

set $\frac{\partial J}{\partial U_{t-1}} = 0$ and $\frac{\partial J}{\partial \lambda} = 0$.

$$\frac{\partial J}{\partial U_{t-1}} = U_{t-1} - \mathcal{W}_c(t)^T \lambda = 0 \tag{3}$$

$$\frac{\partial J}{\partial \lambda} = \mathcal{W}_c(t) U_{t-1} - (x_r - A^t x_0) = 0 \tag{4}$$

from (3),

$$U_{t-1} = \mathcal{W}_c(t)^T \lambda \quad (5)$$

substituting (5) to (4),

$$\mathcal{W}_c(t) \mathcal{W}_c(t)^T \lambda - (x_r - A^t x_0) = 0$$

$$\lambda = (\mathcal{W}_c(t) \mathcal{W}_c(t)^T)^{-1} (x_r - A^t x_0)$$

substituting back λ to (5) yields

$$\begin{aligned} U &= \mathcal{W}_c(t)^T (\mathcal{W}_c(t) \mathcal{W}_c(t)^T)^{-1} (x_r - A^t x_0) \\ &= \mathcal{W}_c(t_1)^T (\Sigma_{t_1}^c)^{-1} (x_r - A^{t_1} x_0) \end{aligned}$$

note: this is one of the many ways to reach x_1 since the cost function could be formulated differently.

references:

1. https://en.wikipedia.org/wiki/Controllability_Gramian
2. <http://www2.imm.dtu.dk/courses/02421/tfoils.pdf>
3. solving difference equation, <https://www.cl.cam.ac.uk/teaching/2003/Probability/prob07.pdf>