





$$\Rightarrow \underbrace{\begin{matrix} \underbrace{\mathbb{1}^0}_{r} \\ \underbrace{\mathbb{1}^1}_{n-r} \end{matrix}}_{\mathbb{1}^0} \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \cdot \underbrace{\begin{matrix} \underbrace{\mathbb{1}^1}_{r} \\ \underbrace{\mathbb{1}^0}_{n-r} \end{matrix}}_{\mathbb{1}^1} + \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} \underline{\mu}$$

$\underbrace{\quad}_{r} \quad \underbrace{\quad}_{n-r}$   
 $C_{bal}$

$$\underline{y} = \underbrace{\begin{bmatrix} C_1 & C_2 \end{bmatrix}}_{C_{bal}} \underbrace{\mathbb{1}^0}_{\begin{matrix} r \\ n-r \end{matrix}} + \begin{bmatrix} D \end{bmatrix} \underline{\mu}$$

has very similar behavior to:

$$\left| \begin{array}{l} \mathbb{1}^0_{red} = A_{11} \cdot \mathbb{1}^0_{red} + B_1 \cdot \underline{\mu} \\ \underline{y} = C_1 \cdot \mathbb{1}^0_{red} + D \cdot \underline{\mu} \end{array} \right|$$

$$\mathbb{1}^0 = \begin{bmatrix} \mathbb{1}^0_{red} \\ \mathbb{1}^0_{elim} \end{bmatrix} \begin{matrix} r \\ n-r \end{matrix}$$

Assume :

$$\dot{\underline{x}} = A \underline{x} + B \underline{u}$$

$$\text{Rank}(G_c) < n$$

⇒ The system is truly uncontrollable!

⇒ Need "input decoupling."

The algorithm of ECE 501 requires  $Q_c$  and  $Q_c^{-1}$

⇒ FOR THE BIRDS !!!

Recipe:

Look at the system:

$$\dot{\underline{x}} = A \underline{x} + [B \quad B_{aux}] \cdot \begin{bmatrix} \underline{u} \\ \underline{u}_{aux} \end{bmatrix}$$

e.g.  $B_{aux} = 100 * \Sigma * \text{rand}(n, 1) * \|B\|$

We calculate  $G_c$  again for the system with the augmented input vector. If

$$\text{Rank}(G_c) < n$$

We increase the gain of  $B_{aux}$ , until

$$\text{Rank}(G_c) = n.$$

Yet, since the elements of  $B_{aux}$  are still small, the system is still almost uncontrollable.

Now, we can use Balreal.

We eliminate the almost uncontrollable state and get:

$$\dot{\underline{x}}_{red} = A_{11} \cdot \underline{x}_{red} + [B_1 \quad B_2] \cdot \begin{bmatrix} \underline{u} \\ \underline{u}_{aux} \end{bmatrix}$$

Because the elements of  $B_{aux}$  were small, the system output does not change much for any power-limited input signal  $u_{aux}$ . Since the I/O behavior is invariant to similarity transformations:

$\Rightarrow$  The elements of  $B_2$  must still be small, in spite of the similarity transformation. Thus, we simply throw  $B_2$  away:

$$\dot{x}_{red} = A_{11} \cdot x_{red} + B_1 \cdot u$$

Similarly, we augment the C matrix by an additional row of small elements if  $\text{Rank}(G_0) < n$ .

Assume that the original system was unstable.

If any of the unstable modes is unobservable or uncontrollable (or poorly observable / controllable), the system is garbage.

Otherwise, I create a linear output feedback around the original system (using either the stable pole placement or LQR algorithms) without worrying about performance too much.

The purpose is simply stabilization.

Once this has happened, we can use balancing to throw out uncontrollable and/or unobservable modes of the already controlled system. Then, we build a controller around this for performance improvement.

Given the discrete system:

$$\underline{x}_{k+1} = F \cdot \underline{x}_k + G \cdot u_k; \quad \underline{x}(0) = \underline{x}_0$$

This system is controllable if the controllability Gramian:

$$G_c = \sum_{i=0}^{\infty} F^i \cdot G \cdot G^* \cdot (F^*)^i$$

is non-singular.

$$G_c = GG^* + FGG^*F^* + F^2GG^*(F^*)^2 + \dots$$

$$\Rightarrow F \cdot G_c \cdot F^* = FGG^*F^* + F^2GG^*(F^*)^2 + \dots$$

$$\Rightarrow F \cdot G_c \cdot F^* - G_c = -GG^*$$

$$\Rightarrow \boxed{F \cdot G_c \cdot F^* - G_c + GG^* = 0}$$

is a discrete Lyapunov equation. The above derivation holds as long as the infinite series converge, i.e., as long as  $\|F\| < 1$ , i.e., the system

is stable.

$$B = U \cdot \Sigma \cdot V^*$$

(Singular  
value  
decomposition)

$$\Rightarrow F \cdot G_c \cdot F^* - G_c + U \cdot \Sigma^2 \cdot U^* = \Phi$$

$$\Rightarrow U^* F G_c F^* U - U^* G_c U + \Sigma^2 = \Phi$$

$$\Rightarrow \underbrace{U^* F U}_{\hat{F}} \cdot \underbrace{U^* G_c U}_{\hat{G}_c} \cdot \underbrace{U^* F^* U}_{\hat{F}^*} - \underbrace{U^* G_c U}_{\hat{G}_c} + \Sigma^2 = \Phi$$

$$\Rightarrow \boxed{\hat{F} \cdot \hat{G}_c \cdot \hat{F}^* - \hat{G}_c + \Sigma^2 = \Phi}$$

is again a discrete Lyapunov equation.

function  $[G_c] = \text{dgram}(F, G)$

$[u, s, v] = \text{svd}(G);$

$F_{\text{hat}} = u' * F * u;$

$G_{\text{chat}} = \text{dlyap}(F_{\text{hat}}, s * s');$

$G_c = u * G_{\text{chat}} * u';$

return

is the Matlab implementation.



$G_c$  is uniquely determined by the discrete Lyapunov equation.  
 $G_c$  is always Hermitian:

$$\begin{aligned} (F \cdot G_c \cdot F^* - G_c + G \cdot G^*)^* &= \Phi \\ \Rightarrow F \cdot G_c^* \cdot F^* - G_c^* + G \cdot G^* &= \Phi \\ \Rightarrow \boxed{G_c^* = G_c} \end{aligned}$$

The observability Gramian:

$$G_o = \sum_{i=0}^{\infty} (F^*)^i \cdot H^* \cdot H \cdot F^i$$

can be computed using the Lyapunov equation:

$$\underline{\underline{F^* \cdot G_o \cdot F - G_o + H^* \cdot H = \Phi}}$$

In Matlab:

$$\begin{aligned} // G_c &= \text{dgram}(F, G) \\ // G_o &= \text{dgram}(F', H') \end{aligned}$$

# Similarity transformations:

$$\underline{M} = T \cdot X$$

$$\left| \begin{array}{l} \underline{H} = T \cdot F \cdot T \\ \underline{G} = T \cdot G \\ \underline{F} = F \cdot T \\ \underline{H} = H \end{array} \right|$$

$$\underline{H} \cdot \underline{G}_c \cdot \underline{F}^* - \underline{G}_c + \underline{G} \cdot \underline{G}^* = \Phi$$

$$\Rightarrow T \cdot F \cdot T^{-1} \cdot \underline{G}_c \cdot (T^{-1})^* \cdot F^* \cdot T^* - \underline{G}_c + T \cdot G \cdot G^* \cdot T^* = \Phi$$

$$\Rightarrow \underbrace{F \cdot T^{-1} \cdot \underline{G}_c \cdot (T^{-1})^* \cdot F^*}_{\underline{G}_c} - \underbrace{T \cdot \underline{G}_c \cdot (T^{-1})^*}_{\underline{G}_c} + G \cdot G^* = \Phi$$

$$\Rightarrow \underline{G}_c = T^{-1} \cdot \underline{G}_c \cdot (T^{-1})^*$$

$$\Rightarrow \underline{\underline{G_c = T \cdot G_c \cdot T^*}}$$

$$F^* G_0 F - G_0 + F^* F = \Phi$$

$$\Rightarrow (T^{-1})^* F^* T^* \cdot G_0 \cdot T F T^{-1} - G_0 + (T^{-1})^* F^* F T^{-1} = \Phi$$

$$\Rightarrow \underbrace{F^* T^* \cdot G_0 \cdot T F}_{G_0} - \underbrace{T^* G_0 T}_{G_0} + F^* F = \Phi$$

$$\Rightarrow G_0 = T^* G_0 T$$

$$\Rightarrow \underline{\underline{G_0 = (T^{-1})^* \cdot G_0 \cdot (T^{-1})}}$$

Let :  $G_c = U_c \cdot \Sigma_c^2 \cdot U_c^*$

$$\Rightarrow F \cdot G_c \cdot F^* - G_c + G \cdot G^* = \Phi$$

$$\Rightarrow F \cdot U_c \cdot \Sigma_c^2 \cdot U_c^* \cdot F^* - U_c \Sigma_c^2 U_c^* + G G^* = \Phi$$

$$\Rightarrow (U_c^* F U_c) \Sigma_c^2 (U_c^* F^* U_c) - \Sigma_c^2 + U_c^* G G^* U_c = \Phi$$

$$\Rightarrow \boxed{T = U_c^*}$$

$$\Rightarrow \underline{\underline{G_c = \Sigma_c^2}}$$

Similarly:

$$G_0 = U_0 \cdot \Sigma_0^2 \cdot U_0^*$$

$$\Rightarrow \boxed{T = U_0^*}$$

$$\Rightarrow \underline{\underline{G_0 = \Sigma_0^2}}$$

$$G_c = U_c \cdot \Sigma_c^2 \cdot U_c^*$$

$$G_0 = U_0 \cdot \Sigma_0^2 \cdot U_0^*$$

$$\Rightarrow G_0 G_c = (U_0 \cdot \Sigma_0) \underbrace{(\Sigma_0 \cdot U_0^* \cdot U_c \cdot \Sigma_c)}_H \cdot (\Sigma_c \cdot U_c^*)$$

$$H = U_H \cdot \Sigma_H^2 \cdot V_H^*$$

$$T = \Sigma_H^{-1} \cdot U_H^* \cdot \Sigma_0 \cdot U_0^*$$

Since:  $\hat{G}_c = T \cdot G_c \cdot T^*$

$$\hat{G}_0 = (T^{-1})^* \cdot G_0 \cdot (T^{-1})$$

is still true

$$\Rightarrow \underline{\hat{G}_c = \hat{G}_o = \sum_H^2}$$

In Matlab:

function [Fbal, Gbal, Hbal, Mode, T] =  
dbalreal (F, G, H)

$$G_c = \text{dgram}(F, G);$$

$$G_o = \text{dgram}(F', H');$$

$$[U_c, S_c, V_c] = \text{svd}(G_c);$$

$$[U_o, S_o, V_o] = \text{svd}(G_o);$$

$$S_c = \text{diag}(\text{diag}(S_c)^{\wedge} \phi.s);$$

$$S_o = \text{diag}(\text{diag}(S_o)^{\wedge} \phi.s);$$

$$H = S_o * U_o' * U_c * S_c;$$

$$[U_H, S_H, V_H] = \text{svd}(H);$$

$$T = \text{diag}(\text{diag}(S_H)^{\wedge} \phi.s) \setminus U_H' * S_o * U_o';$$

$$F_{bal} = T * F / T;$$

$$G_{bal} = T * G;$$

$$H_{bal} = H / T;$$

$$\text{Mode} = \text{diag}(S_H);$$

return