

### 3.9) BALANCED REALIZATIONS

**Lemma 3.20** Let  $\begin{bmatrix} A & B \\ C & D \end{bmatrix}$  be a state space realization of a transfer function  $G(s)$ . Suppose that there exists a symmetric matrix

$$P = P^T = \begin{bmatrix} P_1 & 0 \\ 0 & 0 \end{bmatrix}$$

with  $P_1$  nonsingular such that

$$AP + P A^T + B B^T = 0$$

Let the realization  $(A, B, C, D)$  be partitioned compatibly with  $P$  as

$$\left[ \begin{array}{cc|c} A_{11} & A_{12} & B_1 \\ A_{21} & A_{22} & B_2 \\ \hline C_1 & C_2 & D \end{array} \right]$$

then

$$\begin{bmatrix} A_{11} & B_1 \\ C_1 & D_1 \end{bmatrix}$$

is also a realization of  $G(s)$ . Even more,  $(A_{11}, B_1)$  is controllable if  $A_{11}$  is stable

Proof of Lemma 3.20

$$AP + PA^T + BB^T = \begin{bmatrix} A_{11}P_1 + P_1A_{11}^T + B_1B_1^T & P_1A_{21} + B_1B_2^T \\ A_{21}P_1 + B_2B_1^T & B_2B_2^T \end{bmatrix} = 0$$

$$(2,2) \Rightarrow B_2 = 0$$

$$(2,1) \Rightarrow A_{21} = 0 \quad (\text{since } P_1 \text{ is nonsingular})$$

$$\Rightarrow \left[ \begin{array}{cc|c} A_{11} & A_{12} & B_1 \\ \hline A_{21} & A_{22} & B_2 \\ \hline C_1 & C_2 & D \end{array} \right] = \left[ \begin{array}{cc|c} A_{11} & A_{12} & B_1 \\ \hline 0 & A_{22} & 0 \\ \hline C_1 & C_2 & D \end{array} \right] = \left[ \begin{array}{c|c} A_{11} & B_1 \\ \hline C_1 & D \end{array} \right] \quad \begin{array}{l} \text{since } A_{22} \\ \text{is uncontrollable} \\ \text{part} \end{array}$$

From the Lyapunov controllability theorem we have that  
 $A_{11}$  stable  $\Rightarrow (A_{11}, B_1)$  is controllable.

**Lemma 3.21** Let  $\begin{bmatrix} A & B \\ C & D \end{bmatrix}$  be a realization of  $G(s)$ .  
 Assume that there exists

$$Q = Q^T = \begin{bmatrix} Q_1 & 0 \\ 0 & 0 \end{bmatrix}$$

with  $Q_1$  nonsingular such that

$$QA + QA^T + C^TC = 0$$

Let us partition the realization  $(A, B, C, D)$  compatibly  
 with  $Q$  as

$$\left[ \begin{array}{cc|c} A_{11} & A_{12} & B_1 \\ \hline A_{21} & A_{22} & B_2 \\ \hline C_1 & C_2 & D \end{array} \right]$$

then  $\begin{bmatrix} A_{11} & B_1 \\ C_1 & D \end{bmatrix}$  is also a realization of  $G(s)$ .

Moreover,  $(C_1, A_{11})$  is observable if  $A_{11}$  is stable.

From Lemmas 3.20 and 3.21 we have the following algorithm for eliminating uncontrollable and unobservable modes:

1) Let  $G(s) = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$  be a stable realization

2) Compute the controllability Grammian  $P > 0$  from

$$AP + PA^T + BB^T = 0$$

3) Diagonalize  $P$  to get

$$P = \begin{bmatrix} v_1 & v_2 \end{bmatrix} \begin{bmatrix} \Lambda & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} u_1 & u_2 \end{bmatrix}^T, \quad \Lambda_1 > 0, \begin{bmatrix} u_1 & u_2 \end{bmatrix} \text{ unitary matrix}$$

4) then

$$G(s) = \begin{bmatrix} u_1^T A u_1 & u_1^T B \\ C u_1 & D \end{bmatrix}$$

is a controllable realization.

Similarly we can eliminate non-observable modes.

NOTE THAT WE MAY REMOVE MODES THAT ARE BOTH WEAKLY CONTROLLABLE AND WEAKLY OBSERVABLE  
 $\Rightarrow$  order reduction of the original system

(Example, p. 74)

$$G(s) = \frac{3s+18}{s^2+3s+18} \Rightarrow P = \begin{bmatrix} 0.5 & 0 \\ 0 & \alpha^2 \end{bmatrix}$$

$$Q = \begin{bmatrix} 0.5 & 0 \\ 0 & 1/2 \end{bmatrix}$$

for  $\alpha$  very small we have one weakly controllable mode.  
 If we remove it  $\Rightarrow G_r(s) = \frac{-1}{s+1} \Rightarrow$  very different response from the original system.  
 The reason for this error is that the second mode is also strongly observable.

In MAITLANDS balancing transformations

(10)

$$[c, b, b, c, g, T] = \text{balreal}(a, b, c)$$

diagonal contr & obs. grammians

Also diagonal real for discrete time system

Goal to Balance controllability and observability

Grammians

$$AP + PA^T + BB^T = 0 \Rightarrow P > 0 \quad \begin{matrix} \text{Asymptotically} \\ \text{stable and} \\ (A, B) \text{ controllable} \end{matrix}$$

$$A^T Q + QA + C^T C = 0 \Rightarrow Q > 0 \quad \begin{matrix} \text{Asymptotically} \\ \text{stable and} \\ (A, C) \text{ observable} \end{matrix}$$

In general  $P \geq 0, Q \geq 0$ .

By using a state transformation  $\hat{x} = Tx$  we get

$$\left. \begin{matrix} \hat{P} = TPT^T \\ \hat{Q} = T^{-T}QT^{-1} \end{matrix} \right\} \Rightarrow \hat{P}\hat{Q} = TPT^{-1}T^{-T}Q$$

Theorem 7.5 in Chen's book 1999, 193

$\Rightarrow$  The eigenvalues of the product of the Grammians are invariant under state transformation.

Take for the columns of  $T^{-1}$  the eigenvectors of  $PQ$  then

$$PQ = T^{-1}\Lambda T \text{ with } \begin{cases} \lambda_i = \lambda(PQ) \\ \Lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n) \geq 0 \\ \text{note that } P \geq 0, Q \geq 0 \end{cases}$$

The eigenvectors can be chosen such that (not unique in general)

THE CONTROLLABILITY AND OBSERVABILITY GRAMMIANS ARE EQUAL, that is

$$\begin{matrix} \hat{P} = TPT^T = \Sigma \\ \hat{Q} = T^{-T}QT^{-1} = \Sigma \end{matrix} \Rightarrow \boxed{\begin{matrix} \text{BALANCED} \\ \text{REALIZATION} \\ \hat{P} = \hat{Q} = \Sigma \end{matrix}}$$

with

$$\Sigma = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n) \quad \text{and} \quad \Sigma^2 = \Lambda$$

$\sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_n \geq 0$  are known as Hankel singular values





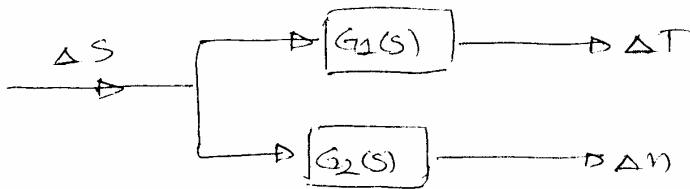
Example: Fusion Reactor Control  
(Plummer, 1995 paper)

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Linearized model

$$\frac{\Delta T}{\Delta S} = \frac{-3.81 \cdot 10^{-2} s + 24.28 \cdot 10^{-14}}{11.688 s^2 + 5.735 s + 1} \Rightarrow \frac{\Delta T}{\Delta S} = G_1(s)$$

$$\frac{\Delta n}{\Delta S} = \frac{s + 0.258}{s^2 + 0.4906 s + 0.085} \Rightarrow \frac{\Delta n}{\Delta S} = G_2(s)$$



two uncoupled systems with a common input

Let us apply the balancing order reduction technique to  $G_2(s)$ . We use MATLAB to solve this problem

Problem is solved in MATLAB v2. Upper numbers are corrected values obtained by MATLAB

>> num = [1 0.258]

>> den = [1 0.4906 0.085]

>> [A, B, C, D] = tf2ss(num, den) =>

$$A = \begin{bmatrix} -0.49069 & -0.8855 \\ 1 & 0 \end{bmatrix}, B = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$C = [1 \ 0.258], D = 0$$

% controllability and observability gramms

>> P = eypap(A, B\*B')

>> P =  $\begin{bmatrix} 1.0192 & 0 \\ 0 & 11.92 \end{bmatrix}$ , K =  $\begin{bmatrix} 1.8126 & 0.3893 \\ 0.3893 & 0.088 \end{bmatrix}$

>> K = eypap(A', C'+C)

note that the first mode is weakly controllable and strongly observable. The second mode is the other way around

( $\sigma(P)$  and  $\sigma(K)$  are controllability and observability margins)

>> [Ab, Bb, Cb, sigma, Trans] = balred(A, B, C)

>>  $A_b = \begin{bmatrix} -0.3146 & -0.1936 \\ 0.1736 & -0.176 \end{bmatrix}$ ,  $B_b = \begin{bmatrix} -1.0317 & 0.3117 \\ 0.2538 & 0.2538 \end{bmatrix}$ ,  $C_b = [-1.0317 \ -0.2538]$

$T = \begin{bmatrix} -0.6215 & 1.4134 \\ -1.5899 & -6.4622 \end{bmatrix}$ ,  $\Sigma = \begin{bmatrix} 1.6918 & 0 \\ 0 & 0.1830 \end{bmatrix}$

nice separation of Hankel's singular

Due to the fact that

$$\Sigma = \begin{bmatrix} \sigma_1 & 0 \\ 0 & \sigma_2 \end{bmatrix} = \begin{bmatrix} 1.6918^{75} & 0 \\ 0 & 0.1830^{733} \end{bmatrix} \quad \sigma_1 \gg \sigma_2$$

We can approximate this system by

$$\begin{aligned} \Rightarrow A_r &= A_b(1,1) & \Rightarrow A_r &= -0.3146^{33} \\ \Rightarrow B_r &= B_b(1,1) & \Rightarrow B_r &= -1.0317^{+1.0317} \\ \Rightarrow C_r &= C_b(1,1) & \Rightarrow C_r &= -1.0317^{+1.0317} \end{aligned} \Rightarrow \begin{cases} \dot{x}_r = -0.3146 x_r - 1.0317 u \\ y_r = -1.0317 x_r \end{cases}$$

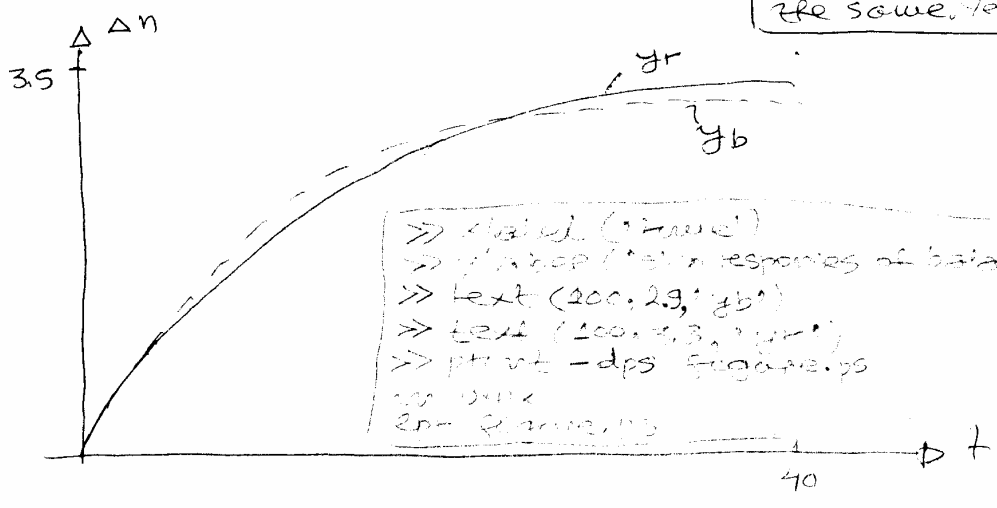
$u \approx \Delta S$   
 $y_r \approx \Delta n$

If we plot the step responses of the original second order system

and the reduced first order system we get

- >>  $D_b = 0$
- >>  $y_b = \text{step}(A_b, B_b, C_b, D_b)$
- >>  $y_r = \text{step}(A_r, B_r, C_r, D_r)$
- >>  $\text{plot}(y_b)$
- >>  $\text{hold on}$
- >>  $\text{plot}(y_r, '--')$

Note that MATLAB's error  
 $\lambda(A) = -0.2453 \pm j0.1596$   
 $\lambda(A_b) = -0.2453 \pm j0.1592$   
 they should be exactly the same, yes, in MATLAB



Very close to each other

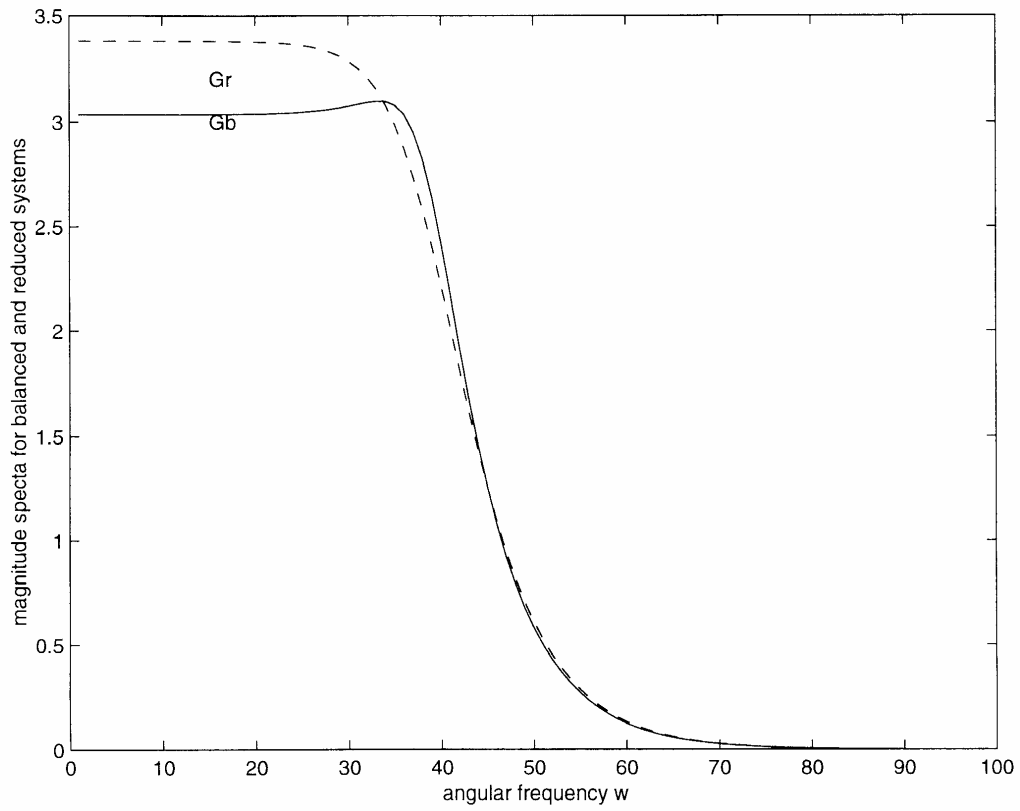
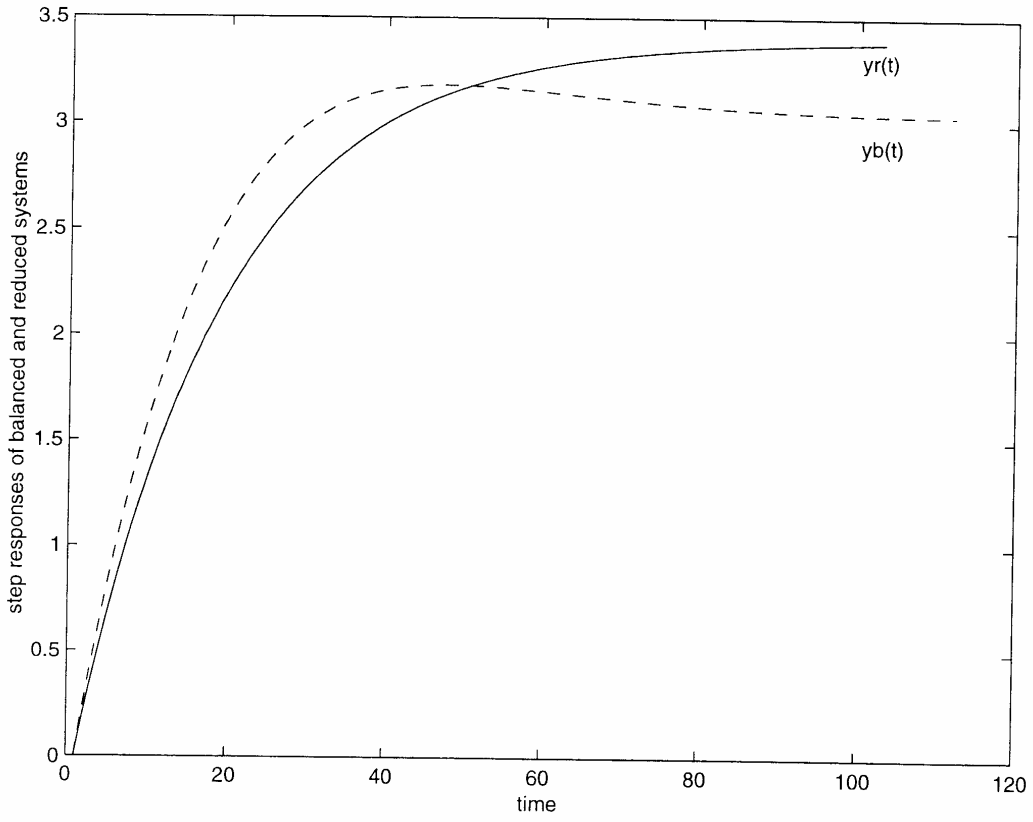
- >>  $\text{title}('Time')$
- >>  $\text{axis}([0 40 0 3.5])$  (step responses of balanced and ...)
- >>  $\text{text}(20, 2.9, 'y_b')$
- >>  $\text{text}(20, 2.9, 'y_r')$
- >>  $\text{print}(-dps 5 \text{ figure}.ps)$
- or  $\text{print}(-dps 5 \text{ figure}.ps)$
- $\text{end}$

It is interesting that balancing realization works very well on the second order system.

Imagine its importance in reducing high-order systems.

Note that balancing changes a lot of the original matrix A.





Some Important Results from Chapter 7

Let  $G(s)$  be stable and balanced

$G(s) = \begin{bmatrix} A_b & B_b \\ C_b & D_b \end{bmatrix}$   $A$  stable,  $(A, B)$  controllable,  $(A, C)$  observable

$A_b \Sigma + \Sigma A_b^T + B_b B_b^T = 0$

$A_b^T \Sigma + \Sigma A_b + C_b^T C_b = 0$

Partition  $\Sigma = \begin{bmatrix} \Sigma_1 & 0 \\ 0 & \Sigma_2 \end{bmatrix}$  and  $G = \begin{bmatrix} A_{11} & A_{12} & B_1 \\ A_{21} & A_{22} & B_2 \\ C_1 & C_2 & D \end{bmatrix}$

**Theorem 7.1** If  $\Sigma_1$  and  $\Sigma_2$  have no diagonal entries in common then both subsystems  $(A_{ii}, B_i, C_i)$  are asymptotically stable.

No proof. In the fusion reactor example  $A_{11} = -0.3146$   
 $A_{22} = -0.176$

**Theorem 7.2** Let  $\Sigma_1 = \text{diag}(\sigma_1 I_{s_1}, \sigma_2 I_{s_2}, \dots, \sigma_r I_{s_r})$

and  $\Sigma_2 = \text{diag}(\sigma_{r+1} I_{s_{r+1}}, \sigma_{r+2} I_{s_{r+2}}, \dots, \sigma_n I_{s_n})$

and  $\sigma_1 > \sigma_2 > \dots > \sigma_r > \sigma_{r+1} > \dots > \sigma_n$   
 $s_1 + s_2 + \dots + s_n = n$

then the reduced system

$G_r(s) = \begin{bmatrix} A_{11} & B_1 \\ C_1 & D \end{bmatrix}$

is balanced and asymptotically stable. Furthermore

$\|G(s) - G_r(s)\|_{\infty} \leq 2(\sigma_{r+1} + \sigma_{r+2} + \dots + \sigma_n)$

FR-Example  $\|G(s) - G_r(s)\|_{\infty} \leq 2 \cdot \sigma_2 = 2 \cdot 0.1830 = 0.366$

$\Rightarrow \text{numr} = Br + Cr \Rightarrow 2.0635$

$\Rightarrow \text{denr} = [1 \quad 0.3146]$

$\Rightarrow [Re, Im] = \text{nyquist}(\text{numr}, \text{denr})$

$\Rightarrow [Per, Imr] = \text{nyquist}(\text{numr}, \text{denr})$

$\Rightarrow G = Re + j * Im \Rightarrow |G| = G_0$

$\Rightarrow G_1 = \text{abs}(G)$

$\Rightarrow Gr = Per + j * Imr \Rightarrow |Gr| = G_{r0}$

$\Rightarrow G_{r1} = \text{abs}(Gr)$

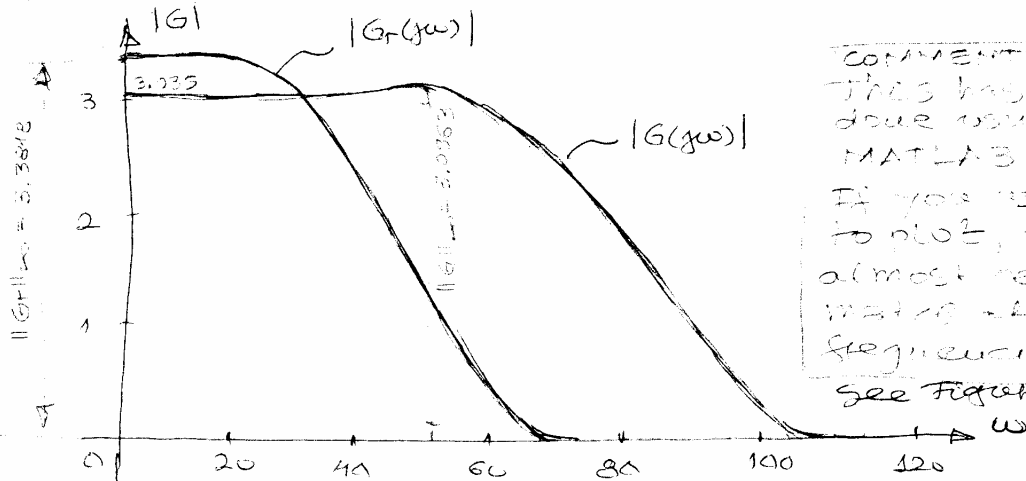
$\Rightarrow \text{RunG} = \max(G_1) \Rightarrow \|G\|_{\infty} = 3.0963$

$\Rightarrow \text{RunGr} = \max(G_{r1}) \Rightarrow \|Gr\|_{\infty} = 3.3816$

$\Rightarrow \text{plot}(G_0)$

$\Rightarrow \text{hold}$

$\Rightarrow \text{plot}(Gr, '---')$



COMMENT:  
 This has been done using MATLAB 4.2  
 If you use MATLAB to plot, you get almost perfect match at high frequencies  
 See Figure on page 10

According to theorem 7.2 we have

$$\left\| \frac{3 + 0.258}{s^2 + 0.49065s + 0.085} - \frac{Br + Cr}{s + Ar} \right\|_{\infty} \leq 2G_2 = 2 \cdot 0.1830 = 0.366$$

In order to capture the approximation in frequency bands of interest we can use the frequency weighted balanced model reduction

$$\| W_0(s) [G(s) - G_H(s)] W_1(s) \|_\infty$$

$W_0, W_1 \in \mathbb{R}^{H_\infty}$  (stable matrices) are known as the output and input weighting matrices. (Section 7.2)

Another interesting theorem related to the Hankel singular values and  $H_\infty$  norm is stated in Chapter 4.

**THEOREM 4.5**

$$\sigma_1 \leq \|G(s)\|_\infty \leq 2 \sum_{i=1}^n \sigma_i$$

Also

$$\|G(s)\|_\infty \leq \int_0^\infty \|g(t)\|_2 dt \leq 2 \sum_{i=1}^n \sigma_i$$

**FR Example**

$$\|G(s)\|_\infty \leq 2(1.6918 + 0.1836) = 3.7496$$

for the reduced system

$$A_H(\sigma_{1r}) + A_H(\sigma_{1r}) + B_1 \cdot B_1 = 0 \Rightarrow \sigma_{1r} = \frac{B_1^2}{2A_H} = \frac{(1.0317)^2}{2 \cdot (0.3146)} = 1.692$$

$\Rightarrow$

$$\|G_r(s)\|_\infty \leq 2 \cdot \sigma_{1r} = 2 \cdot 1.692 = 3.384 \quad (\text{pretty good upper bound the actual value is } \|G_r(s)\|_\infty = 3.3818)$$

Note that MATLAB has the ~~control~~ toolbox for ROBUST CONTROL

$\gg [G_b, \text{sig}] = \text{sybal}(G) \Rightarrow G_b = \text{balanced}$

$\gg G_r = \text{strunc}(G_b, 2)$   $\text{sig} = \text{Hankel singular values}$

does truncation to the second order system