

## Series 11:

Task 1: We have (with  $P(\alpha) = \mathcal{A}F + G$ )

$$\text{rank}_{\mathbb{C}(\alpha)}(\mathcal{H}) = \text{rank}_{\mathbb{C}(\alpha)} \begin{bmatrix} 0 & P \\ P^\sim & H \end{bmatrix}$$

$$= \text{rank}_{\mathbb{C}(\alpha)} \begin{bmatrix} I & -PH^{-1} \\ 0 & I \end{bmatrix} \begin{bmatrix} 0 & P \\ P^\sim & H \end{bmatrix}$$

$$= \text{rank}_{\mathbb{C}(\alpha)} \begin{bmatrix} -PH^{-1}P^\sim & 0 \\ P^\sim & H \end{bmatrix}$$

$$= \underbrace{\text{rank}_{\mathbb{C}(\alpha)}(PH^{-1}P^\sim)}_{=p} + \underbrace{\text{rank}_{\mathbb{C}(\alpha)} H}_{=q}$$

shown in the following

by Assumption 4.8

$P$  has full row rank, since  $F$  has full row rank. Thus there exists a  $\lambda_0 \in \mathbb{C}$  such that  $\text{rank } P(\lambda_0) = p = \text{rank } P^\sim(\lambda_0)$ .

$$\Rightarrow \text{rank } P(\lambda_0) H^{-1} P^\sim(\lambda_0) = p$$

$$\Rightarrow p \geq \text{rank}_{\mathbb{C}(\alpha)} PH^{-1}P^\sim \stackrel{\text{Lemma 4.9}}{\geq} \text{rank } P(\lambda_0) H^{-1} P^\sim(\lambda_0) = p.$$

## Task 2:

Compute the singular value decomposition of  $Y = U \Sigma V^*$  and

set

$$W := VU^* \text{ to obtain}$$

$$YW = (U\Sigma V^*)(VU^*) = U \underbrace{\Sigma}_{\text{diagonal and real}} U^*$$

Task 3:

$$\begin{pmatrix} I & 0 \\ Z^* & I \end{pmatrix} \begin{pmatrix} 0 & \lambda F + G \\ -\lambda F^* + G^* & H \end{pmatrix} \begin{pmatrix} I & Z \\ & I \end{pmatrix} \\ \parallel \\ = \begin{pmatrix} 0 & \lambda F + G \\ -\lambda F^* + G^* & C^* C \end{pmatrix}$$

$$\begin{pmatrix} I & \\ Z^* & I \end{pmatrix} \begin{pmatrix} 0 & \lambda F + G \\ -\lambda F^* + G^* & -\lambda F^* Z + G^* Z + H \end{pmatrix} \\ = \begin{pmatrix} 0 & \lambda F + G \\ -\lambda F^* + G^* & \lambda Z^* F + Z^* G - \lambda F^* Z + G^* Z + H \end{pmatrix}$$

$$\Leftrightarrow C^* C = \lambda(Z^* F - F^* Z) + H + G^* Z + Z^* G$$

$$\Leftrightarrow Z^* F - F^* Z = 0 \quad \text{and} \\ H + G^* Z + Z^* G = C^* C$$

## Task 4:

$$\begin{aligned} (N(\tau\omega_0))^* &= N^*(\tau\omega_0) = N^*(-\overline{\tau\omega_0}) \\ &= -\overline{\tau}\omega_0 = -\overline{\tau}\omega_0 \\ &= N^{\sim}(\tau\omega_0) = N(\tau\omega_0) \end{aligned}$$

Task 5: We have

$$\Theta(z(t_1)) - \Theta(z(t_0)) = \int_{t_0}^{t_1} \frac{d}{dt}(\Theta(z(t))) dt$$

$$= \int_{t_0}^{t_1} \frac{d}{dt} (z^*(t) F^* z(t)) dt$$

$$\begin{aligned} &= \int_{t_0}^{t_1} \underbrace{\dot{z}^*(t) F^* z(t)}_{z \in \mathcal{B}(F, H)} + z^*(t) F^* \dot{z}(t) dt \\ &= (F \dot{z}(t))^* = (-G z(t))^* \end{aligned}$$

$$= \int_{t_0}^{t_1} -z^*(t) G^* z(t) - z^*(t) z^* G z(t) dt$$

$$= - \int_{t_0}^{t_1} z^*(t) \underbrace{[G^* z + z^* G]}_{\substack{= C^* C - H \\ \text{Task 3}}} z(t) dt$$

$$= \int_{t_0}^{t_1} z^*(t) H z(t) dt - \underbrace{\int_{t_0}^{t_1} z^*(t) C^* C z(t) dt}_{= \|C z(t)\|_2^2 \geq 0} \geq 0$$

$$\leq \int_{t_0}^{t_1} z^*(t) H z(t) dt$$

Furthermore, if  $z \in \mathcal{L}_+(\mathcal{A}T + G)$  is with  $z(t) = 0, t \leq 0$ , then

$$\int_0^{\infty} z^*(t) H z(t) dt = \lim_{R \rightarrow \infty} \int_0^R z^*(t) H z(t) dt$$

$$\geq \lim_{R \rightarrow \infty} \theta(z(R)) - \theta(z(0))$$

$$= \theta(\underbrace{\lim_{R \rightarrow \infty} z(R)}_{=0, \text{ since } z \in \mathcal{L}_+(\dots)}) - \theta(\underbrace{z(0)}_{=0, \text{ since } z(t)=0, t \leq 0})$$

$\nearrow$   $\theta(y) = y^* z^* T y$  is continuous

$$= \theta(0) - \theta(0) = 0.$$

This implies non-negativity.

Task 6e If  $\hat{z}$  is optimal by Theorem 4.4 there exists a  $\hat{\mu} \in \mathcal{L}_+^*$  such that for  $t \geq 0$

$$\underbrace{(\hat{\mu}(t), \hat{z}(t))}_{:= \begin{bmatrix} \hat{\mu}(t) \\ \hat{z}(t) \end{bmatrix}} \in \mathcal{L}_+ \left( \begin{bmatrix} 0 & P \\ P^* & H \end{bmatrix} \right) \stackrel{\text{Theorem 4.10}}{=} \begin{matrix} \text{z.) b.)} \\ \begin{bmatrix} Z \\ I \end{bmatrix} \mathcal{L}_+ \begin{bmatrix} P \\ C \end{bmatrix} \end{matrix}$$

$$\Rightarrow \begin{matrix} \cancel{0} \\ 0 \end{matrix} = C \hat{z}(t) \quad \cancel{\text{and}} \\ (\text{and } \hat{\mu}(t) = Z \hat{z}(t)) \text{ for } t \geq 0.$$

Thus we obtain

$$\int_0^\infty \dot{\hat{z}}^*(t) H \dot{\hat{z}}(t) dt = \int_0^\infty \dot{\hat{z}}^* (C^* C - Z^* G - \cancel{G^* Z}) \dot{\hat{z}} dt$$

$$= \int_0^\infty \underbrace{\|C \dot{\hat{z}}(t)\|_2^2}_{=0} dt - \int_0^\infty \underbrace{\dot{\hat{z}}^* G^* Z \dot{\hat{z}} + \dot{\hat{z}}^* Z^* G Z \dot{\hat{z}}}_{=(G \dot{\hat{z}})^* = (Z^* \dot{\hat{z}})^*} dt$$

$$= \int_0^\infty \dot{\hat{z}}^* F^* Z \dot{\hat{z}} + \dot{\hat{z}}^* Z^* F \dot{\hat{z}} dt$$

$$= \int_0^\infty \frac{d}{dt} (\dot{\hat{z}}^* F^* Z \dot{\hat{z}}) dt$$

$$= \underbrace{\left( \lim_{R \rightarrow \infty} \dot{\hat{z}}^*(R) F^* Z \dot{\hat{z}}(R) \right)}_{=0, \text{ da } \dot{\hat{z}} \in \mathcal{L}_2(\cdot)} - \dot{\hat{z}}^*(0) F^* Z \dot{\hat{z}}(0)$$

$$= \Theta(\dot{\hat{z}}(0)) .$$

### Task 7:

(a) 1.)  $\Rightarrow$  2.)

$$J^* J = (J^* J)^* = J^* J^* = -J^* J$$

2.)  $\Rightarrow$  3.)

$$J^* J = J^* (J J^*)^{-1} = \underbrace{-J^* J^*}_{=I} J^* = J^* J^* = (J^* J)^*$$

3.)  $\Rightarrow$  4.) Denote the Hermitian matrix  $J^* J$  in the form

$$Jb =: \begin{bmatrix} +C & -B^* \\ -B & -A \end{bmatrix}, \text{ with } C = C^*, A = A^*.$$

Then

$$\begin{aligned} b &= J^{-1} J b = \begin{bmatrix} & -I \\ I & \end{bmatrix} \begin{bmatrix} +C & -B^* \\ -B & -A \end{bmatrix} \\ &= \begin{bmatrix} B & A \\ C & -B^* \end{bmatrix}. \end{aligned}$$

(4.)  $\Rightarrow$  1.) We have that

$$Jb = \begin{bmatrix} & I \\ -I & \end{bmatrix} \begin{bmatrix} B & A \\ C & -B^* \end{bmatrix} = \begin{bmatrix} C & -B^* \\ -B^* & -A \end{bmatrix}$$

is Hermitian

$$(b) \text{ If } \sigma(b) = \mathfrak{J}(I - b) = \mathfrak{J}(AI - b) \mathfrak{J}$$

~~$$\mathfrak{J} \mathfrak{J} (AI - b) = \mathfrak{J} (b) \mathfrak{J}$$~~

$$= \mathfrak{J} \left( \underbrace{I}_{\text{skew-Hermitian}} - \underbrace{(b)}_{\text{Hermitian}} \right) =: \mathfrak{J}(N)$$

where  $N^* = N$  is para-Hermitian. Thus

$\mathfrak{J}(N) = \sigma(b)$  is symmetric w.r.t. the imaginary axis.

(c) Since  $N(\lambda) := \lambda J - gJ$  is para-Hermitian without imaginary zeros

$$\mathcal{Z}(N) \cap (i\mathbb{R}) \stackrel{(b)}{=} \sigma(g) \cap (i\mathbb{R}) \stackrel{\text{by Assumption}}{=} \emptyset$$

there exists (by Theorem 4.7) a unitary  $V = S \in \mathbb{C}^{2n \times 2n}$  such that  $V^* J V = j$  (i.e., it is symplectic) and

$$\begin{aligned} V^* N(\lambda) V &= \lambda j - \underbrace{V^* g J V}_{L} = \left[ \begin{array}{c|c} K & \lambda I - L \\ \hline -\lambda I - L^* & 0 \end{array} \right] \\ &= \left[ \begin{array}{c|c} K & L \\ \hline L^* & 0 \end{array} \right] \end{aligned}$$

where  $L = \begin{bmatrix} \Delta \end{bmatrix}$  and

$$\mathcal{C} \supseteq \mathcal{Z}(\lambda I - L) = \sigma(L).$$

This implies  $\begin{bmatrix} K & L \\ L^* & 0 \end{bmatrix} = V^* g V \underbrace{V^* J V}_{=j} = V^* g V j$

$$\Rightarrow V^* g V = \begin{bmatrix} K & L \\ L^* & 0 \end{bmatrix} \underbrace{\begin{bmatrix} 0 & -I \\ I & 0 \end{bmatrix}}_{=j^{-1} = -j} = \begin{bmatrix} L & L - K \\ 0 & -L^* \end{bmatrix}$$

Task 8: With  $P(\lambda) := \begin{bmatrix} \lambda & -1 \end{bmatrix}$

and  $H = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

the matrix polynomial for the optimality system is

$$\mathcal{M}(\lambda) := \begin{bmatrix} 0 & P \\ P^* & H \end{bmatrix} = \begin{bmatrix} 0 & \lambda & -1 \\ -\lambda & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}.$$

Thus we can choose  $X_1 = I \in \mathbb{C}^{3 \times 3}$

and

$$X_2^* \begin{bmatrix} 0 & \lambda & -1 \\ -\lambda & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} =: X_2$$

*should become zero*

$$= \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -1 & \lambda & -1 \\ -\lambda & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} -1 & \lambda & | \\ -\lambda & 1 & | \\ \hline & & 1 \end{bmatrix}$$

The zeros of  $N_1(\lambda) := \lambda \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} + \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$

$= \begin{bmatrix} M_{11} & M_{12} \\ M_{21} & M_{22} \end{bmatrix}$

are then



$$\mathcal{Z}(N_1) = \mathcal{Z}\left(\begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}\right) = \mathcal{Z}\left(\underbrace{\begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}}_{\text{unimodular}} \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}\right)$$

$$= \mathcal{Z}\left(\begin{bmatrix} -1 & 1 \\ 0 & -1^2+1 \end{bmatrix}\right) = \mathcal{Z}(-1^2+1) = \{\pm 1\}$$

Choosing the stable zero  $\lambda_1 = -1$  we have that (in the notation similar to Lemma 4.6)

$$N(\lambda_1)x_1 = \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix} \underbrace{\begin{bmatrix} 1 \\ -1 \end{bmatrix}}_{=: x_1} = 0$$

*should be determined*

and thus we choose

$$\underbrace{\frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}}_{=: V_3^*} \begin{bmatrix} 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ -2 \end{bmatrix} \frac{1}{\sqrt{2}} = \begin{bmatrix} 0 \\ -\sqrt{2} \end{bmatrix}$$

and obtain the para-Hermitian Schur form of Theorem 4.7

$$V_3^* N_1(\lambda) V_3 = V_3^* \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \frac{1}{\sqrt{2}}$$

$$= \frac{1}{2} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 1-1 & 1+1 \\ -(1-1) & 1+1 \end{bmatrix} = \frac{1}{2} \left[ \begin{array}{c|c} 0 & 2(1+1) \\ \hline -2(1-1) & 0 \end{array} \right]$$

$$= \left[ \begin{array}{c|c} 0 & 1+1 \\ \hline -1+1 & 0 \end{array} \right].$$

*all stable zeros are here*

This means that (in the proof of Theorem 4.10)

$$X_3 := \begin{bmatrix} V_3 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

and thus

$$Y := X_1 X_2 X_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & \\ & & 1 \end{bmatrix}$$

$$= \left[ \begin{array}{ccc|c} 1 & -1 & 0 & \\ \hline 1 & 1 & 0 & \\ \hline 1 & -1 & \sqrt{2} & \end{array} \right] \frac{1}{\sqrt{2}}$$

$$=: \left[ \begin{array}{c|c} Y_1 & Y_2 \\ \hline Y_3 & Y_4 \end{array} \right]$$

$$\Rightarrow Y_4^{-1} = \begin{bmatrix} 1 & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \sqrt{2} = \begin{bmatrix} \sqrt{2} & 0 \\ 1 & 1 \end{bmatrix}$$

$$\Rightarrow Y_4^{-1} Y_3 = \begin{bmatrix} \sqrt{2} & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \frac{1}{\sqrt{2}} = \begin{bmatrix} \sqrt{2} \\ 2 \end{bmatrix} \frac{1}{\sqrt{2}} = \begin{bmatrix} 1 \\ \sqrt{2} \end{bmatrix}$$

$$\Rightarrow Y_2 Y_4^{-1} Y_3 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ \sqrt{2} \end{bmatrix} = -\frac{1}{\sqrt{2}}$$

$$\Rightarrow Y_1 - Y_2 Y_4^{-1} Y_3 = \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} = \frac{2}{\sqrt{2}} = \sqrt{2}$$

which means that

$$Y^{-1} = \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{1}{\sqrt{2}} & & \\ \hline -1 & \sqrt{2} & 0 & & 1 & \\ -\sqrt{2} & 1 & 1 & & & 1 \end{array} \right]$$
$$= \begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & 0 \\ -\frac{1}{\sqrt{2}} & \sqrt{2} & 0 \\ -1 & 1 & 1 \end{bmatrix}$$

and

$$Y Y^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 & 0 \\ 1 & 1 & 0 \\ 1 & -1 & \sqrt{2} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} & 0 & 0 \\ -\frac{1}{\sqrt{2}} & \sqrt{2} & 0 \\ -1 & 1 & 1 \end{bmatrix}$$
$$= \frac{1}{\sqrt{2}} \begin{bmatrix} \frac{2}{\sqrt{2}} & -\sqrt{2} & 0 \\ 0 & \sqrt{2} & 0 \\ \frac{2}{\sqrt{2}} - \sqrt{2} & 0 & \sqrt{2} \end{bmatrix} = \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = Z$$

$$\Rightarrow Z := [-1, 0]$$

$$\Rightarrow H + G^* Z + Z^* G = \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} [-1, 0]$$

$$\dots + \begin{bmatrix} -1 \\ 0 \end{bmatrix} [0, -1] = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ -1 & 0 \end{bmatrix} + \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} [1, 1]$$

$$\Rightarrow C := [1, 1]$$

The necessary equation for optimality is thus

$$z = \begin{bmatrix} x \\ u \end{bmatrix}$$

$$0 = C z(t) = \cancel{[1, 1]} \begin{bmatrix} x(t) \\ u(t) \end{bmatrix} = x(t) + u(t)$$

$$\Rightarrow u(t) = -x(t)$$

$$\Rightarrow \text{choose } c_1 = c_2 = 1$$

The closed-loop dynamics are

$$\dot{x}(t) = u(t) = -x(t) \quad \text{which means}$$

$$\underline{\underline{\dot{x}(t) = -x(t)}}$$