

## Model Reduction of Dynamical Systems - 3

### Exercise: 3.1 (Balancing-free square root (BFSR) method)

We have already discussed the square-root balanced truncation technique for model reduction in Exercise 2.1. Another way of model reduction is to use the balancing-free square root (BFSR) algorithm. Analogue to the standard square-root balanced truncation approach, one has to compute the Cholesky factors  $S$  and  $R$  of the solutions of the Lyapunov equations and the corresponding SVD of those factors, i.e.,

$$AP + PA^T + BB^T = 0, \quad A^T Q + QA + C^T C = 0, \quad SR^T = [U_1 \quad U_2] \begin{bmatrix} \Sigma_1 & 0 \\ 0 & \Sigma_2 \end{bmatrix} \begin{bmatrix} V_1^T \\ V_2^T \end{bmatrix}.$$

The left and right projection matrices for the computation of a reduced-order model of dimension  $r$  now are given as  $T_l = (Q_1^T P_1)^{-1} Q_1^T$  and  $T_r = P_1$ , with

$$S^T U_1 = [P_1 \quad P_2] \begin{bmatrix} \hat{R} \\ 0 \end{bmatrix}, \quad R^T V_1 = [Q_1 \quad Q_2] \begin{bmatrix} \bar{R} \\ 0 \end{bmatrix},$$

and  $P_1, Q_1 \in \mathbb{R}^{n \times r}$  have orthonormal columns and  $\hat{R}, \bar{R} \in \mathbb{R}^{r \times r}$  are upper triangular.

Show that the reduced-order system is equivalent to a balanced system and that it satisfies the same error bound as the one obtained by the standard square root BT method.

### Exercise: 3.2 (Balanced model reduction for non-minimal systems)

Consider a system which is neither controllable nor observable, i.e.,

$$\mathcal{K} = \text{rank} \left( \begin{bmatrix} B & AB & \dots & A^{n-1}B \end{bmatrix} \right) = k_1 < n \quad \text{and} \quad \mathcal{O} = \text{rank} \left( \begin{bmatrix} C \\ CA^T \\ \vdots \\ C(A^T)^{n-1} \end{bmatrix} \right) = k_2 < n.$$

Show that if  $Y \in \mathbb{R}^{n \times k_1}$  and  $Z \in \mathbb{R}^{n \times k_2}$  are low rank factors that satisfy  $P = YY^T$  and  $Q = ZZ^T$ , a balanced reduced-order model can be obtained by projection matrices  $T_l = \Sigma_1^{-\frac{1}{2}} V_1 Z^T$  and  $T_r = Y U_1 \Sigma_1^{-\frac{1}{2}}$ , where  $Y^T Z = [U_1 \quad U_2] \begin{bmatrix} \Sigma_1 & 0 \\ 0 & \Sigma_2 \end{bmatrix} \begin{bmatrix} V_1^T \\ V_2^T \end{bmatrix}$ .