

# Structure Preserving Model Order Reduction by Parameter Optimization

## Scope

We illustrate the method proposed in P. SCHWERDTNER AND M. VOIGT, *Structure Preserving Model Order Reduction by Parameter Optimization* (<https://arxiv.org/abs/2011.07567>) on the basis of a port-Hamiltonian system. For further explanations, references, and all technical details we refer to our paper.

A brief talk on the method can be found at: <https://www.youtube.com/watch?v=kzUr1GdA5T0>

## Setup – Large Scale Port Hamiltonian Systems

We are given a dynamical system

$$\Sigma_{\text{pH}} : \begin{cases} \dot{x}(t) = (J - R)Qx(t) + Bu(t), \\ y(t) = B^\top Qx(t), \end{cases}$$

where  $J, R, Q \in \mathbb{R}^{n_x \times n_x}$  and  $B \in \mathbb{R}^{n_x \times n_u}$  with the structural constraints  $J = -J^\top$ ,  $R \geq 0$ , and  $Q \geq 0$ . Furthermore, we have that  $n_x \gg n_u \geq 1$ .

The system is described in *frequency domain* by its transfer function

$$G_{\text{pH}}(s) = B^\top Q(sI - (J - R)Q)^{-1}B.$$

$G_{\text{pH}}$  describes the frequency-wise amplification of the input, i.e.  $Y(s) = G_{\text{pH}}(s)U(s)$ , where  $U$  and  $Y$  are the Laplace transformed input and output, respectively.

## Task – Model Order Reduction

We must find an approximation to the system  $\Sigma_{\text{pH}}$  with the same structure, i.e.

$$\Sigma_{\text{pH},r} : \begin{cases} \dot{x}_r(t) = (J_r - R_r)Q_r x_r(t) + B_r u(t), \\ y_r(t) = B_r^\top Q_r x_r(t), \end{cases}$$

where  $J_r, R_r, Q_r \in \mathbb{R}^{r \times r}$ ,  $B_r \in \mathbb{R}^{r \times n_u}$  with  $J_r = -J_r^\top$ ,  $R_r \geq 0$ , and  $Q_r \geq 0$  and with  $r \ll n_x$ .

When approximating a dynamical system, we want to achieve that for any given input  $u$ , the outputs  $y$  and  $y_r$  are similar. The approximation quality of a *reduced order model* is often measured in terms of the  $\mathcal{H}_\infty$  norm of the error transfer function  $G_{\text{pH}} - G_{\text{pH},r}$ . It is defined as

$$\|G_{\text{pH}} - G_{\text{pH},r}\|_{\mathcal{H}_\infty} := \sup_{\omega \in \mathbb{R}} \|G_{\text{pH}}(i\omega) - G_{\text{pH},r}(i\omega)\|_2$$

## Our Approach – Approximation by Nonlinear Optimization

We define a parametrized port-Hamiltonian system

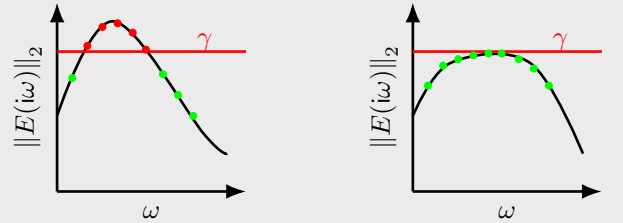
$$\Sigma_{\text{pH}}(\theta) : \begin{cases} \dot{x}(t) = (J(\theta) - R(\theta))Q(\theta)x(t) + B(\theta)u(t), \\ y(t) = B(\theta)^\top Q(\theta)x(t). \end{cases}$$

The matrices  $J(\theta), R(\theta), Q(\theta) \in \mathbb{R}^{r \times r}$  and  $B(\theta) \in \mathbb{R}^{r \times n_u}$  depend on the parameter vector  $\theta$  such that for all  $\theta \in \mathbb{R}^{n_\theta}$  the structural constraints are satisfied and  $\Sigma_{\text{pH}}(\theta)$  is a port-Hamiltonian system.

Therefore, we can use unconstrained optimization algorithms to minimize  $\|G_{\text{pH}} - G_{\text{pH}}(\cdot, \theta)\|_{\mathcal{H}_\infty}$ .

Note that we do not minimize  $\|G_{\text{pH}} - G_{\text{pH}}(\cdot, \theta)\|_{\mathcal{H}_\infty}$  directly since  $\|\cdot\|_{\mathcal{H}_\infty}$  is hard to compute.

We minimize the sum of the squared error at test frequencies  $s_i$ , if the error  $\|E(s_i)\|_2$  is larger than a prescribed value  $\gamma$ . This yields the following behavior.

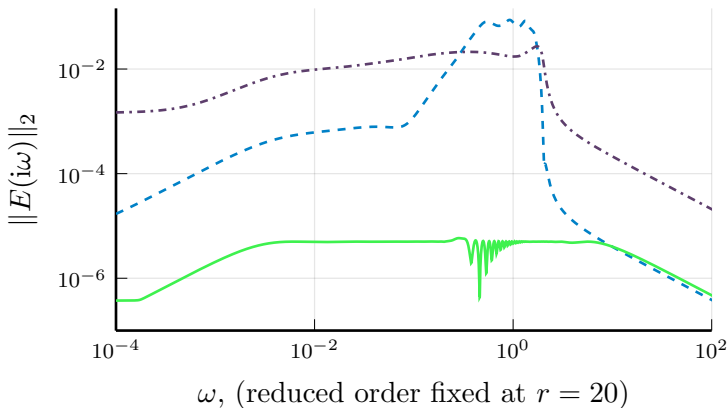


(a) Before

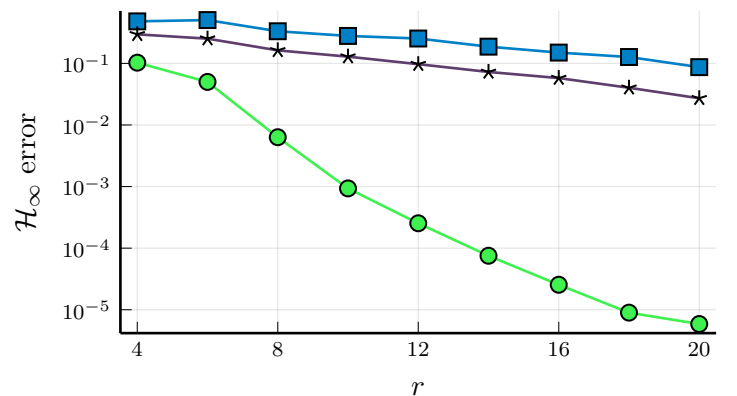
(b) After

We iteratively reduce  $\gamma$  until the accuracy of  $\Sigma_{\text{pH}}(\theta)$  is satisfactory.

## Results



$\omega$ , (reduced order fixed at  $r = 20$ )



— pH balanced truncation — pH IRKA — our method