

## Descriptor systems

Consider the **descriptor system**

$$\begin{aligned} E\dot{x}(t) &= Ax(t) + Bu(t), \\ y(t) &= Cx(t), \end{aligned}$$

with  $E, A \in \mathbb{R}^{n \times n}$ ,  $B \in \mathbb{R}^{n \times m}$ ,  $C \in \mathbb{R}^{p \times n}$ , and regular  $sE - A$ . The associated **transfer function** is given by

$$G(s) = C(sE - A)^{-1}B.$$

## $\mathcal{H}_\infty$ -spaces

$\mathcal{RH}_\infty^{p \times m}$  := Banach space of  $p \times m$  transfer functions  $G(s)$  that are:

- well-defined**, i.e.,  $sE - A$  is a regular matrix pencil;
- stable**, i.e., all poles of  $G(s)$  are in the open left half-plane;
- proper**, i.e.,  $G(\lambda)$  is bounded for  $|\lambda| \rightarrow \infty$ .

Induced norm:

$$\|G\|_{\mathcal{H}_\infty} := \sup_{\omega \in \mathbb{R}} \sigma_{\max}(G(i\omega)).$$

## Problem Setting

### Problem

Computation of the  $\mathcal{H}_\infty$ -norm for systems with **large and sparse  $E, A, B, C$**

### Applications:

- error measure for model order reduction;
- performance measure in robust control.

## Method 1: Structured Pseudospectra

### Structured stability radius

Consider the perturbed transfer function

$$G_\Delta(s) := C(sE - (A + B\Delta C))^{-1}B$$

with  $\Delta \in \mathbb{C}^{m \times p}$ . Define

$$\begin{aligned} r_{\mathbb{C}}^f(E, A, B, C) &:= \inf \{\|\Delta\|_2 : G_\Delta(s) \text{ is not stable}\}, \\ r_{\mathbb{C}}^\infty(E, A, B, C) &:= \inf \{\|\Delta\|_2 : G_\Delta(s) \text{ is not well-defined or improper}\}. \end{aligned}$$

Then

$$r_{\mathbb{C}}(E, A, B, C) := \min \{r_{\mathbb{C}}^f(E, A, B, C), r_{\mathbb{C}}^\infty(E, A, B, C)\}$$

is called the **structured stability radius**.

### Theorem

$$\|G\|_{\mathcal{H}_\infty} = \begin{cases} 1/r_{\mathbb{C}}(E, A, B, C) & \text{if } G \neq 0, \\ 0 & \text{if } G \equiv 0. \end{cases}$$

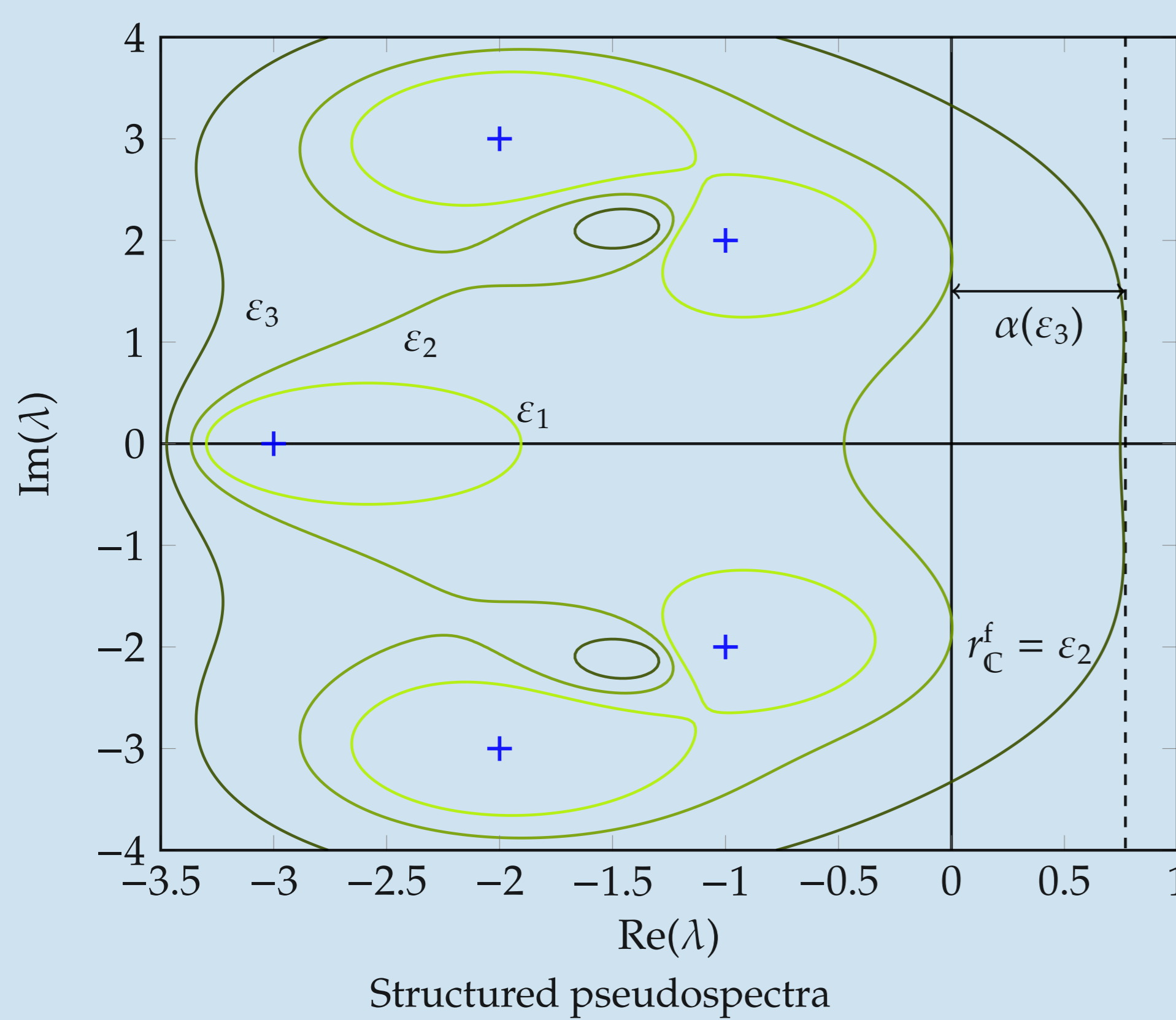
### Structured pseudospectra

- set of (finite) poles of  $G(s)$  denoted by  $\Pi_f(E, A, B, C)$ ;
- structured pseudospectrum**:

$$\Pi_\varepsilon(E, A, B, C) := \{\lambda \in \mathbb{C} : \lambda \in \Pi_f(E, A + B\Delta C, B, C) \text{ for } \Delta \in \mathbb{C}^{m \times p} \text{ with } \|\Delta\|_2 < \varepsilon\};$$

- structured pseudospectral abscissa**:

$$\alpha(\varepsilon) := \max \{\operatorname{Re}(\lambda) : \lambda \in \Pi_\varepsilon(E, A, B, C)\}.$$



### Algorithm outline

To compute the  $\mathcal{H}_\infty$ -norm we need to find the zero of  $\alpha(\cdot)$ .

#### Main Steps:

- Choose initial  $\varepsilon$ .
- Compute rightmost pseudopole to get  $\alpha(\varepsilon)$ .
- Update  $\varepsilon$  via Newton's method and repeat Steps 2 and 3.

### Software

MATLAB software is available under

<http://www2.mpi-magdeburg.mpg.de/mpcsc/voigtm/software/infnorm.php?lang=en>.

### References

- P. Benner and M. Voigt: *A structured pseudospectral method for  $\mathcal{H}_\infty$ -norm computation of large-scale descriptor systems*, Math. Control Signals Systems, 26(1):303–338, 2014.
- R. Lowe and M. Voigt:  *$\mathcal{L}_\infty$ -norm computation for large-scale descriptor systems using structured iterative eigensolvers*, Preprint MPIMD/13-20, MPI Magdeburg, 2013.
- J. Rommes and N. Martins: *Efficient computation of transfer function dominant poles using subspace acceleration*, IEEE Trans. Power Syst., 21(3):1218–1226, 2006.

## Method 2: Even Matrix Pencils

### Theorem

Consider the **even matrix pencil**

$$\mathcal{H}_\gamma(s) := \begin{bmatrix} 0 & -sE^T - A^T & -C^T & 0 \\ sE - A & 0 & 0 & -B \\ -C & 0 & \gamma I_p & 0 \\ 0 & -B^T & 0 & \gamma I_m \end{bmatrix}.$$

Assume that  $sE - A$  has no purely imaginary eigenvalues and let  $\gamma > \inf_{\omega \in \mathbb{R}} \sigma_{\max}(G(i\omega))$ . Then

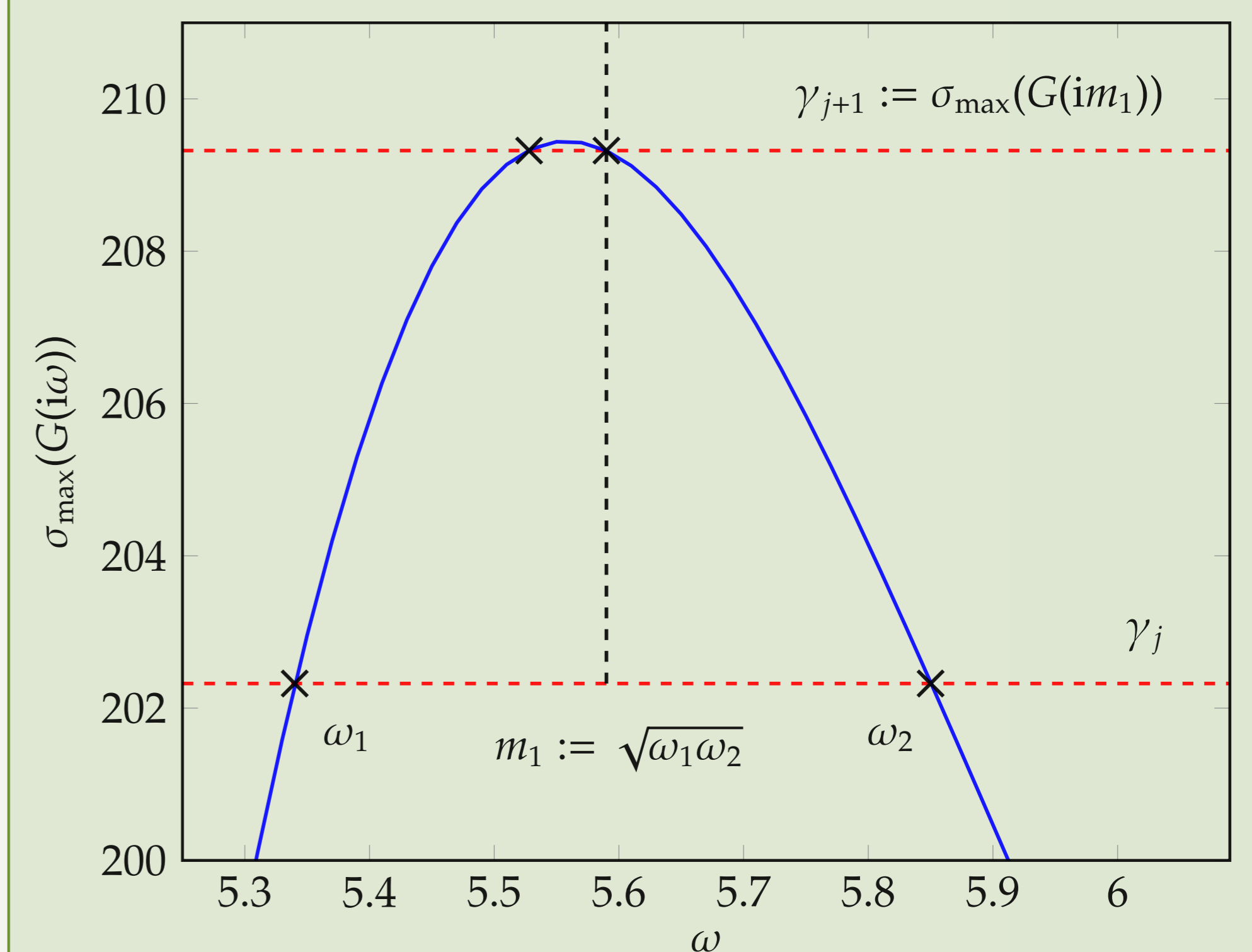
- $\gamma$  is a singular value of  $G(i\omega_0) \iff \mathcal{H}_\gamma(s)$  has the eigenvalue  $i\omega_0$ ;
- $\|G\|_{\mathcal{H}_\infty} \geq \gamma \iff \mathcal{H}_\gamma(s)$  has purely imaginary eigenvalues.

### Eigenvalue computation

- Eigenvalues computed using the **structure-preserving** even IRA method;
- even IRA computes some eigenvalues close to a **pre-specified shift**;
- heuristically, the  $\mathcal{H}_\infty$ -norm is attained close to a dominant pole!  $\implies$  use the imaginary parts of the dominant poles as initial shifts!

### Algorithm outline

- Choose an initial value for  $\gamma$ .
- Determine purely imaginary eigenvalues of  $\mathcal{H}_\gamma(s)$ .
- If imaginary eigenvalues exist, increase  $\gamma$  and repeat Step 2.
- If no imaginary eigenvalues exist,  $\gamma$  is an upper bound for  $\|G\|_{\mathcal{H}_\infty}$ .



### Dominant poles

Assume that  $\lambda E - A$  has only simple eigenvalues  $\lambda_k$  with left and right eigenvectors  $y_k$  and  $x_k$  such that  $y_k^H E x_k = 1$ . If  $G(s)$  is proper then

$$G(s) = \sum_{k=1}^n \frac{R_k}{s - \lambda_k} + R_\infty$$

with residues

$$R_k = C x_k y_k^H B, \quad R_\infty = \lim_{\omega \rightarrow \infty} G(i\omega).$$

**Observation:** If  $\lambda_j$  is close to the imaginary axis and  $\|R_j\|_2$  is large, then

$$G(i\omega) \approx \frac{R_j}{-i\operatorname{Re}(\lambda_j)} + \sum_{k=1, k \neq j}^n \frac{R_k}{i\omega - \lambda_k} + R_\infty$$

for  $\omega \approx \operatorname{Im}(\lambda_j)$  and therefore  $\|G(i\omega)\|_2$  is large, too.

#### Dominant poles:

- = the  $\lambda_j$  with largest  $\|R_j\|_2 / \operatorname{Re}(\lambda_j)$ ;
- computation by the dominant pole algorithm.

### Numerical Results and Comparison

#### Comparison of runtimes

example	n	m	p	time in s	
				Method 1	Method 2
build	48	1	1	1.54	0.54
beam	348	1	1	52.22	38.15
M80PI_n	4182	3	3	10.41	4.51
bips98_1142	9735	4	4	69.65	16.37
xingo_afonso_itaipu	13250	1	1	39.24	16.80
mimo8x8_system	13309	8	8	78.47	23.25
mimo28x28_system	13251	28	28	85.36	35.45
mimo46x46_system	13250	46	46	115.91	49.13

#### Summary of Method 1:

- + smaller number of user-defined parameters, easier to use;
- slower than Method 2;
- perturbed pencils are complex  $\rightsquigarrow$  complex arithmetics;
- no guaranteed global optimization.

#### Summary of Method 2:

- + faster than Method 1;
- + works under less restrictive conditions than Method 1;
- larger number of user-defined parameters;
- no guaranteed global optimization.

### Computation of the rightmost pseudopole

- Useful fact:** The whole structured pseudospectrum can be obtained by using only rank-1 perturbations, i.e.,  $\Delta = \varepsilon u v^H$  with vectors  $u, v$  with  $\|u\|_2 = \|v\|_2 = 1$ .
- Strategy:** Compute a sequence of suitable structured rank-1 perturbed transfer functions

$$G_{\varepsilon u v^H}(s) := C(sE - (A + \varepsilon B u v^H C))^{-1}B$$

such that one of the perturbed poles converges to the rightmost pseudopole of  $G(s)$ .

- Optimal perturbation:** Let  $\lambda$  be a simple eigenvalue of  $sE - A$  with right/left eigenvectors  $x, y$ . Then maximal local growth of the real part of  $\lambda$  for

$$u = B^T y / \|B^T y\|_2, \quad v = C x / \|C x\|_2.$$

