

# Fixed-Order H-infinity Optimization of Time-Delay Systems

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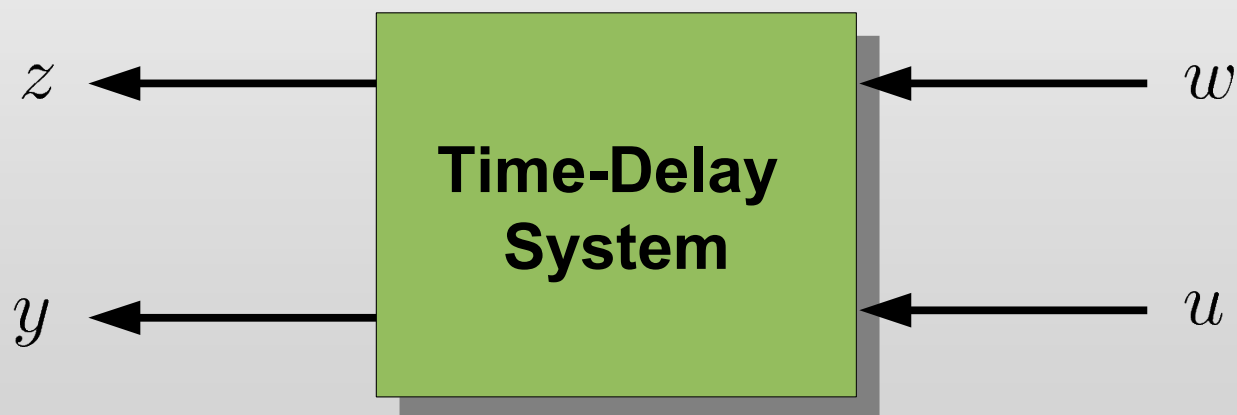
# Plant Definition

The state-space equations of the **generalized plant G** are

$$\dot{x}(t) = A_0 x(t) + \sum_{i=1}^m A_i x(t - \tau_i) + B_1 w(t) + B_2 u(t - \tau_{m+1})$$

$$z(t) = C_1 x(t) + D_{11} w(t) + D_{12} u(t)$$

$$y(t) = C_2 x(t) + D_{21} w(t) + D_{22} u(t - \tau_{m+1}), A_0 \in \mathbb{R}^{n \times n}$$



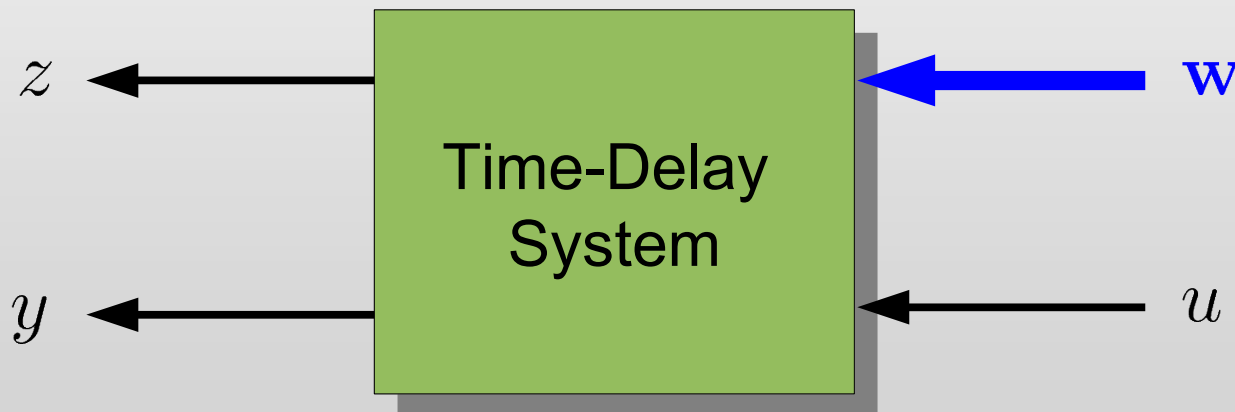
# Plant Definition

Input functions are **disturbances**

$$\dot{x}(t) = A_0x(t) + \sum_{i=1}^m A_i x(t - \tau_i) + B_1 \mathbf{w}(t) + B_2 u(t - \tau_{m+1})$$

$$z(t) = C_1 x(t) + D_{11} \mathbf{w}(t) + D_{12} u(t)$$

$$y(t) = C_2 x(t) + D_{21} \mathbf{w}(t) + D_{22} u(t - \tau_{m+1})$$



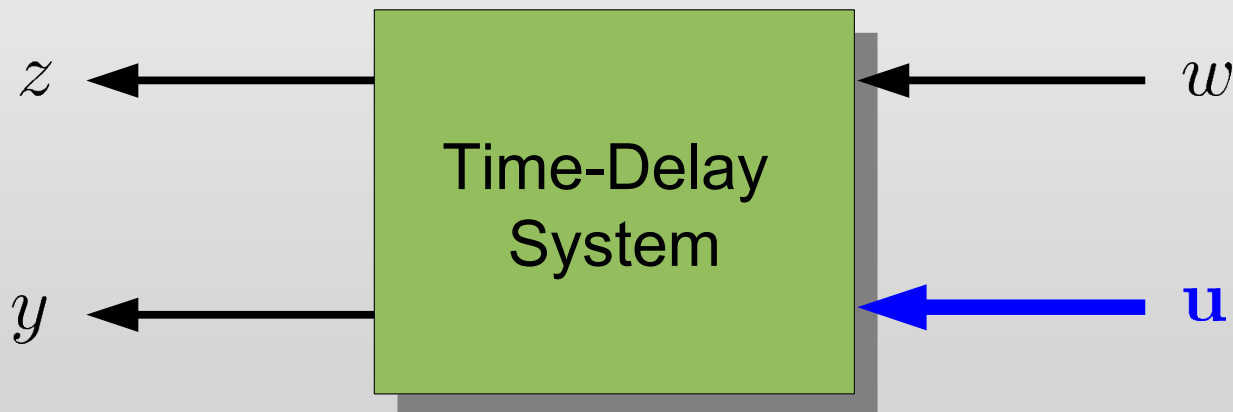
# Plant Definition

Input functions are **control signals**

$$\dot{x}(t) = A_0 x(t) + \sum_{i=1}^m A_i x(t - \tau_i) + B_1 w(t) + B_2 \mathbf{u}(t - \tau_{m+1})$$

$$z(t) = C_1 x(t) + D_{11} w(t) + D_{12} \mathbf{u}(t)$$

$$y(t) = C_2 x(t) + D_{21} w(t) + D_{22} \mathbf{u}(t - \tau_{m+1})$$



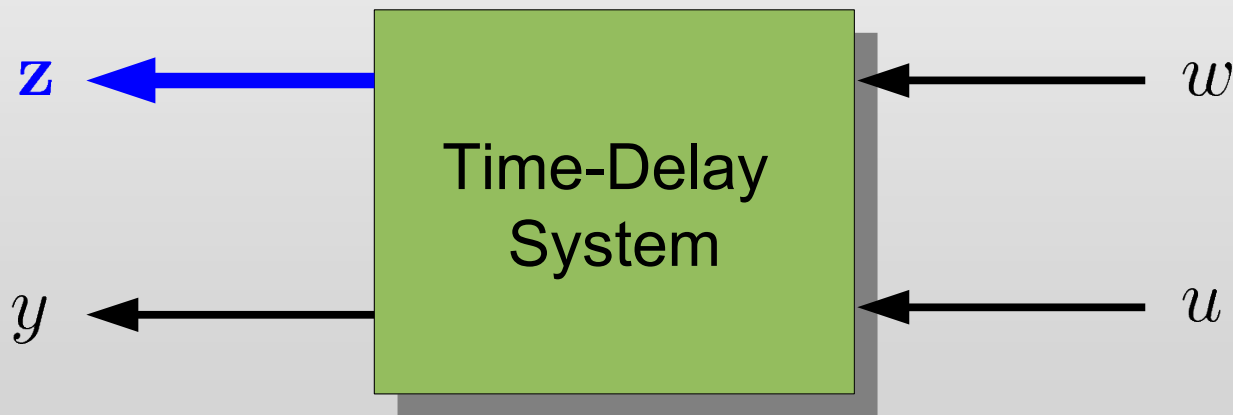
# Plant Definition

Output functions are **controlled signals**

$$\dot{x}(t) = A_0x(t) + \sum_{i=1}^m A_i x(t - \tau_i) + B_1w(t) + B_2u(t - \tau_{m+1})$$

$$z(t) = C_1x(t) + D_{11}w(t) + D_{12}u(t)$$

$$y(t) = C_2x(t) + D_{21}w(t) + D_{22}u(t - \tau_{m+1})$$



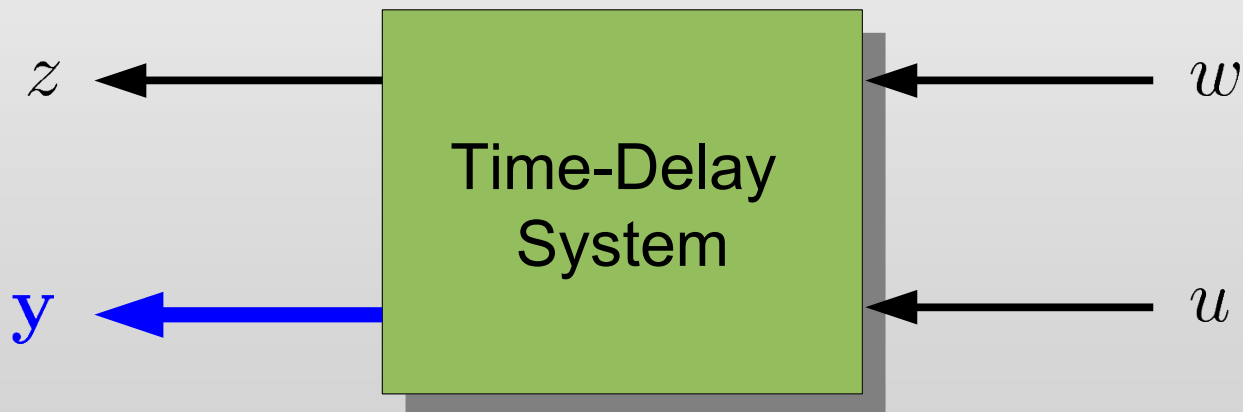
# Plant Definition

Output functions are **measured signals**

$$\dot{x}(t) = A_0x(t) + \sum_{i=1}^m A_i x(t - \tau_i) + B_1w(t) + B_2u(t - \tau_{m+1})$$

$$z(t) = C_1x(t) + D_{11}w(t) + D_{12}u(t)$$

$$y(t) = C_2x(t) + D_{21}w(t) + D_{22}u(t - \tau_{m+1})$$



# Controller structure

The **controller** has the following structure

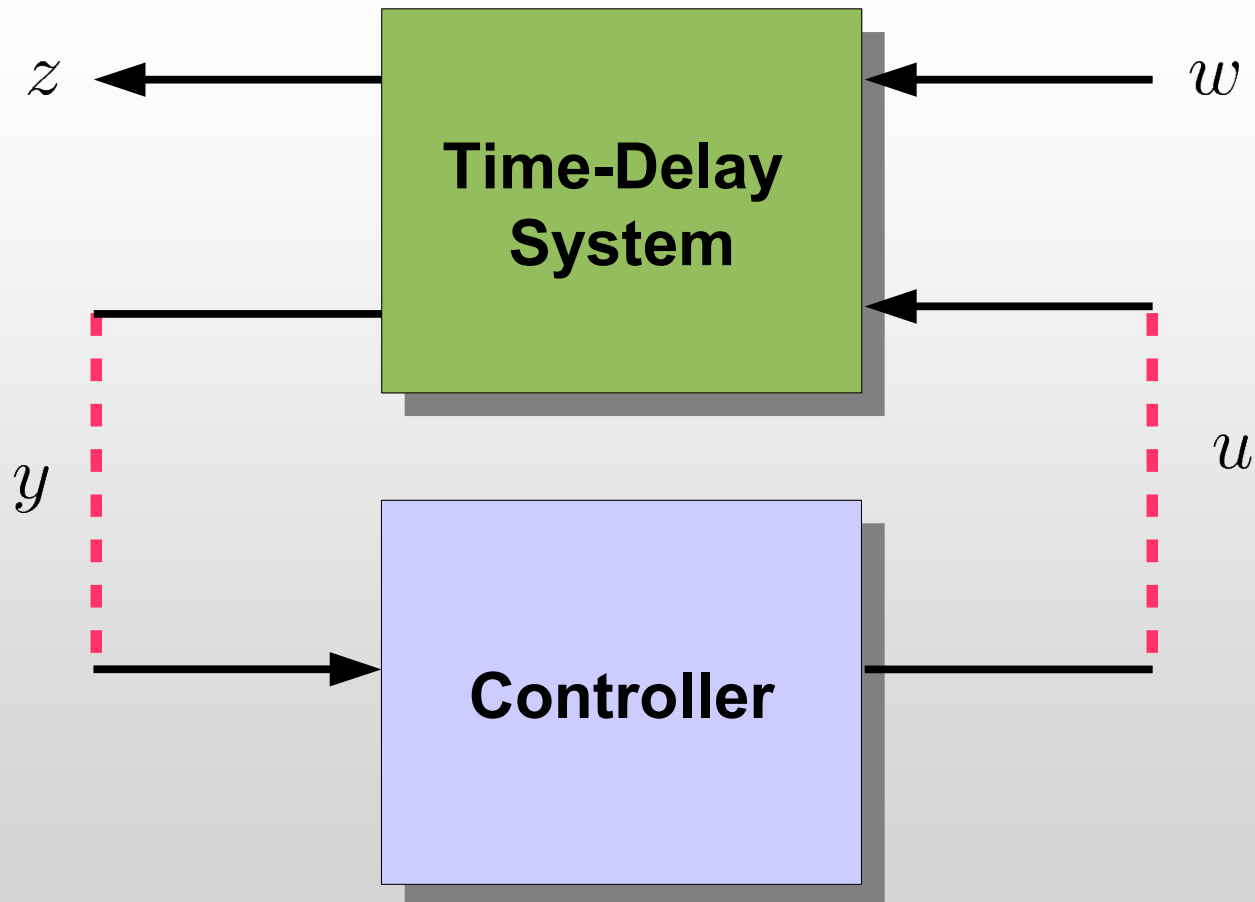
$$\begin{aligned}\dot{x}_K(t) &= A_K x_K(t) + B_K y(t) \\ u(t) &= C_K x_K(t)\end{aligned}$$

The controller order is  $n_K$  where  $A_K \in \mathbb{R}^{n_K \times n_K}$ . Note that the order of the controller is a design parameter.



# Closing the loop

Connect the generalized plant and the controller

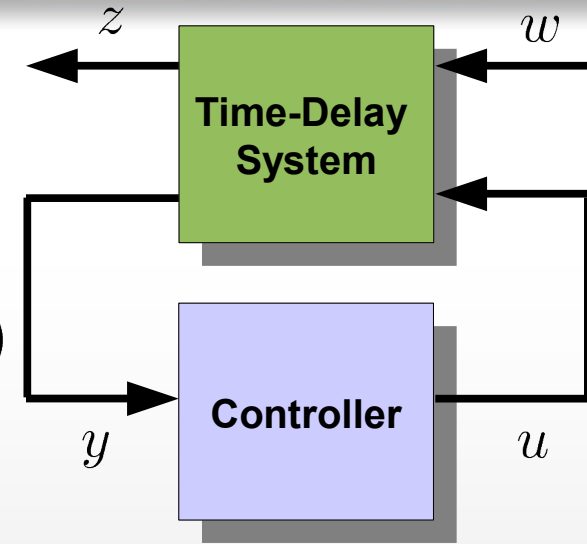




# Closed-loop equations

We can write the closed-loop equations as

$$\dot{x}_{cl}(t) = A_{cl,0}x_{cl}(t) + \sum_{i=1}^{m+1} A_{cl,i}x_{cl}(t - \tau_i) + B_{cl}w(t)$$
$$z(t) = C_{cl}x_{cl}(t) + D_{cl}w(t)$$



and the transfer function from  $w$  to  $z$  as

$$T_{zw}(s) = C_{cl} \left( sI - A_{cl,0} - \sum_{i=1}^{m+1} A_{cl,i}e^{-\tau_i s} \right)^{-1} B_{cl} + D_{cl}$$

The **controller matrices** can be tuned to achieve certain objectives.

# Problem definition

We would like to minimize H-infinity norm from  $w$  to  $z$

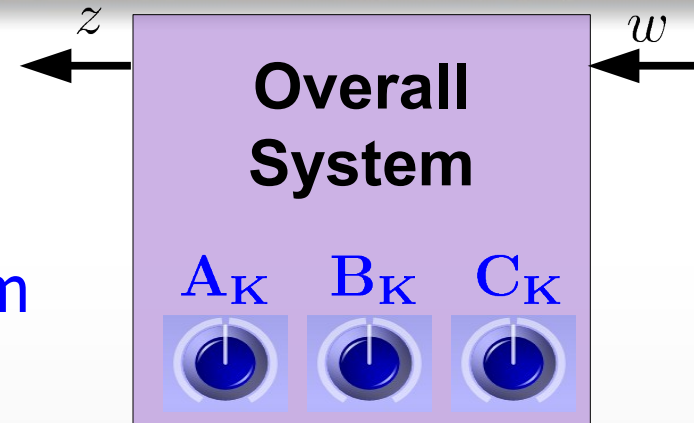
$$\|T_{zw}\|_{\infty} = \sup_{\omega \in \mathbb{R}} \sigma \left( C_{cl} \left( j\omega I - A_{cl,0} - \sum_{i=1}^{m+1} A_{cl,i} e^{-j\tau_i \omega} \right)^{-1} B_{cl} + D_{cl} \right)$$

by a fixed-order controller

while keeping the overall system stable

$$\alpha \left( \sum_{i=0}^m A_{cl,i} e^{-\tau_i s} \right) < 0$$

$\alpha$  denotes the spectral abscissa, max real parts of eigenvalues



# Why designing a fixed-order H-infinity controller?

- **Fixed-order** controllers are
  - cheap and easy to implement in hardware
  - non-restrictive in sampling rate and bandwidth
  - good alternative to controller order reduction and performance degradation check in design
- **H-infinity** controllers
  - stabilize the closed-loop under model uncertainties
  - achieve design performance for a predefined set of input signals
  - reject a set of disturbances (input disturbances, measurement noise)

# Literature tells us

- For MIMO finite dimensional plants, H-infinity controller design
  - Optimal: solvable by standard methods when  $n=n_k$  (Riccati-DGKF, LMI-Gahinet/ Apkarian). This is restrictive for high-order plants.
  - Optimal, fixed-order: only for certain types of closed-loop functions (Nagamune)
  - Fixed-order: via non-smooth, non-convex optimization methods ([HIFOO-Overton et.al.](#), Apkarian/Noll)
- For time-delay plants, there is no known method for fixed-order H-infinity controller design

# Objective function

We would like to find a local minimizer of  $\|T_{zw}\|_{\infty}$  where

$$\|T_{zw}\|_{\infty} = \sup_{\omega \in \mathbb{R}} \sigma \left( C_{cl} \left( j\omega I - A_{cl,0} - \sum_{i=1}^{m+1} A_{cl,i} e^{-j\tau_i \omega} \right)^{-1} B_{cl} + D_{cl} \right)$$

Note the objective function is

- not everywhere differentiable, may even be not Lipschitz continuous
- smooth almost everywhere

# Optimization problem

Optimization method looks for a local minimizer of  $\|T_{zw}\|_\infty$

where

$$\|T_{zw}\|_\infty = \sup_{\omega \in \mathbb{R}} \sigma \left( C_{cl} \left( j\omega I - A_{cl,0} - \sum_{i=1}^{m+1} A_{cl,i} e^{-j\tau_i \omega} \right)^{-1} B_{cl} + D_{cl} \right)$$

Similarly to HIFOO, the method implements a hybrid algorithm for nonsmooth, nonconvex optimization, based on the following elements:

- A quasi-Newton algorithm (BFGS) provides a fast way to approximate a local minimizer
- A local bundle method attempts to verify local optimality for the best point found by BFGS, and if this does not succeed,
- Gradient sampling attempts to refine the approximation of the local minimizer, returning a rough optimality measure.

## More information on optimization method

- J. V. Burke, A. S. Lewis and M. L. Overton, "A robust gradient sampling algorithm for nonsmooth nonconvex optimization," *SIAM Journal on Optimization*, vol.15, pp. 751–779, 2005.
- J.V. Burke, D. Henrion, A.S. Lewis and M.L. Overton, "HIFOO - A MATLAB Package for Fixed-Order Controller Design and H-infinity Optimization," Proc. of ROCOND 2006, Toulouse, July 2006.
- S. Gumussoy and M.L. Overton, "Fixed-Order H-infinity Controller Design via HIFOO, a Specialized Nonsmooth Optimization Package," Proc. of ACC, Seattle, 2008.
- S. Gumussoy, D. Henrion, M. Millstone and M.L. Overton, "Multiobjective Robust Control with HIFOO 2.0," IFAC Symposium on Robust Control Design, Haifa, 2009.
- <http://www.cs.nyu.edu/overton/software/hifoo/>

# Computation of H-infinity norm

Optimization method requires the computation of  $\|T_{zw}\|_\infty$  and gradients with respect to controller parameters

where

$$\|T_{zw}\|_\infty = \sup_{\omega \in \mathbb{R}} \sigma \left( C_{cl} \left( j\omega I - A_{cl,0} - \sum_{i=1}^{m+1} A_{cl,i} e^{-j\tau_i \omega} \right)^{-1} B_{cl} + D_{cl} \right)$$

We implemented a predictor-corrector type method to evaluate H-infinity norm of  $T_{zw}$

- Prediction step: calculate the approximate H-infinity norm and corresponding frequencies
- Correction step: correct the approximate results from the predicted step

Note that derivatives with respect to controller parameters can be computed from the derivative of an individual singular value



## Predicton step

**[Thm]** Let  $\zeta > 0$  be such that the matrix  $D_\xi := D_{cl}^T D_{cl} - \xi^2 I$  is non-singular. Singular values of  $T_{zw}$  and eigenvalues of the Hamiltonian-like operator  $L_\zeta$  of  $T_{zw}$  have the relation:

$$\sigma_i(T_{zw}(j\omega_0)) = \xi \iff L_\xi u = j\omega_0 u$$

### [Corollary]

$\|T_{zw}\|_\infty = \sup\{\xi \in \mathbb{R}_+ : \text{operator } \mathcal{L}_\xi \text{ has an eigenvalue on the imaginary axis}\}$

## Prediction step: Properties of $L_\zeta$

- infinite dimensional linear operator
- has infinitely many eigenvalues, finite on imaginary axis
- its spectrum symmetric wrt imaginary axis
- eigenvalues of  $L_\zeta$  can be computed approximately by computing eigenvalues of the matrix  $L_\zeta^N$  obtained by spectral methods (Breda et.al.)

$$\|T_{zw}\|_\infty \approx \sup\{\xi \in \mathbb{R}_+ : \text{matrix } \mathcal{L}_\xi^N \text{ has an eigenvalue on the imaginary axis}\}$$

based on

$$\sigma_i(T_{zw}(j\omega_0)) = \xi \iff \det(j\hat{\omega}_0 I - L_\xi^N) = 0$$

## Correction step

We want to correct the approximate results from prediction step.

**[Thm]** Let  $\zeta > 0$  be such that the matrix  $D_\xi := D_{cl}^T D_{cl} - \xi^2 I$  is non-singular.  $\lambda$  is an eigenvalue of  $L_\zeta$  if and only if

$$h_\xi(\lambda) := \det H_\xi(\lambda) = 0$$

where

$$H_\xi(\lambda) := \lambda I - M_0 - \sum_{i=1}^{m+1} (M_i e^{-\lambda \tau_i} + M_{-i} e^{\lambda \tau_i})$$

$$\sigma_i(T_{zw}(j\omega_0)) = \xi \iff L_\xi u = j\omega_0 u \iff h_\xi(j\omega_0) = 0$$

## Correction step

Using the properties of  $h_\xi(j\omega)$ ,  $h_\xi(j\omega) = 0$ ,  $h'_\xi(j\omega) = 0$

$$\begin{cases} H(j\omega, \xi) \begin{bmatrix} u \\ v \end{bmatrix} = 0, & n(u, v) = 0 \\ \Im \left\{ v^* \left( I + \sum_{i=1}^{m+1} A_{cl,i} \tau_i e^{-j\omega \tau_i} \right) u \right\} = 0 \end{cases}$$

- Overdetermined system (4n+3 equations, 4n+2 unknowns)
- Exactly solvable, overdetermined due to symmetry of the spectrum
- Solved using Gauss-Newton algorithm (quadratically converging because residual in the solution is zero)

# Computation of H-infinity Norm

## Prediction step

For fixed  $N$ , determine

$$\sup\{\xi \in \mathbb{R}_+ : \text{matrix } \mathcal{L}_\xi^N \text{ has an eigenvalue on the imaginary axis}\}$$

and determine the corresponding eigenvalues on the imaginary axis

(solved by a criss-cross search in two directions)

## Correction step

Correct the results by solving the equations

$$\begin{cases} H(j\omega, \xi) \begin{bmatrix} u \\ v \end{bmatrix} = 0, & n(u, v) = 0 \\ \Im \left\{ v^* \left( I + \sum_{i=1}^{m+1} A_{cl,i} \tau_i e^{-j\omega \tau_i} \right) u \right\} = 0 \end{cases}$$

# Discussion on computational time

- The main step is the computing eigenvalues of the Hamiltonian matrix  $L_{\zeta}^N$  (N: num of discretization points).
- The computational time proportional to
  - ODE  $\sim (2n_1)^3$  ( $n_1$ : dim of ODE)
  - DDE  $\sim (2n_2 \times 2N)^3$  ( $n_2$ : dim of DDE)

# Benchmarking – Example 1

$$\dot{x}(t) = -x(t) - 0.5x(t-1) + w(t) + u(t)$$

$$z(t) = x(t) + u(t)$$

$$y(t) = x(t) + w(t)$$

- The time-delay system is stable and H-infinity norm is **0.72**
- When  $n_K=1$ , our method finds the controller

$$\dot{x}_K(t) = 3.61x_K(t) + 1.39y(t)$$

$$u(t) = -0.83x_K(t)$$

achieving the closed-loop H-infinity norm **0.064**

- For  $n_K=2$ , the norm is **0.021**. For  $n_K=3$ , **0.020**.

# Benchmarking – Example 2

$$\begin{aligned} \dot{x}(t) = & \begin{pmatrix} -4.4656 & -0.4271 & 0.4427 & -0.1854 \\ -0.8601 & -5.6257 & 0.8577 & -0.5210 \\ 0.9001 & -0.7177 & -6.5358 & 0.0417 \\ -0.6836 & 0.0242 & 0.4997 & -3.5618 \end{pmatrix} x(t) + \begin{pmatrix} 0.6848 & -0.0618 & 0.5399 & 0.5057 \\ 0.3259 & -0.3810 & 0.6592 & -0.0066 \\ 0.6325 & 0.3752 & 0.4122 & 0.7303 \\ 0.5878 & 0.9737 & 0.1907 & -0.8639 \end{pmatrix} x(t - 3.2) \\ & + \begin{pmatrix} 0.9371 & -0.7859 & 0.1332 & 0.7429 \\ -0.8025 & 0.4483 & 0.6226 & 0.0152 \\ 0.0940 & 0.2274 & 0.1536 & 0.5776 \\ -0.1941 & 0.5659 & 0.8881 & -0.0539 \end{pmatrix} x(t - 3.4) + \begin{pmatrix} 0.6576 & -0.8543 & -0.3460 & 0.6415 \\ -0.3550 & 0.5024 & 0.6081 & 0.9038 \\ 0.9523 & 0.6624 & 0.0765 & -0.8475 \\ -0.4436 & 0.8447 & -0.0734 & 0.4173 \end{pmatrix} x(t - 3.9) \\ & + \begin{pmatrix} 1 & 0 \\ -1.6 & 1 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} w(t) + \begin{pmatrix} 0.2 \\ -1 \\ 0.1 \\ -0.4 \end{pmatrix} u(t - 0.2) \end{aligned}$$

$$z(t) = \begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & -1 & 1 & 0 \end{pmatrix} x(t) + \begin{pmatrix} 0.1 & 1 \\ -1 & 0.2 \end{pmatrix} w(t) + \begin{pmatrix} 1 \\ -1 \end{pmatrix} u(t)$$

$$y(t) = \begin{pmatrix} 1 & 0 & -1 & 0 \end{pmatrix} x(t) + \begin{pmatrix} -2 & 0.1 \end{pmatrix} w(t) + 0.4u(t - 0.2)$$



## Benchmarking – Example 2

- The time-delay system is stable and H-infinity norm is **1.3907**
- When  $n_K=1$ , our method finds the controller

$$\begin{aligned}\dot{x}_K(t) &= -0.712x_K(t) - 0.1639y(t) \\ u(t) &= -0.2858x_K(t)\end{aligned}$$

achieving the closed-loop H-infinity norm **1.2606**.

- For  $n_K=2$ , the norm is **1.2573**. For  $n_K=3$ , **1.2505**.

# Concluding Remarks

- Fixed-order H-infinity controller design via non-smooth, non-convex optimization method is successfully implemented for a class of time-delay systems
- Our method allows to choose the controller order as desired
- This method is a promising alternative to controller order reduction methods
- Extension to general time-delay systems and allowing the structure in the controller are future research directions.
- Optimization based approach is essentially an approach for parameter tuning, thus applicable to a very broad class of problems, e.g. also controllers featuring delays  $\sum_{k_i} y(t - \tau_i)$