

# Optimization of Time Delays in a Parabolic Delay Equation

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New trends in PDE constrained optimization

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## Joint work with

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**Mariano Mateos** (Gijón, Spain)

# Outline

- 1 Introduction
- 2 Control-to-state mapping
- 3 Optimization problem
- 4 Numerical Discretization
- 5 Numerical examples
- 6 Nonlocal Pyragas type feedback
- 7 The problem of stability

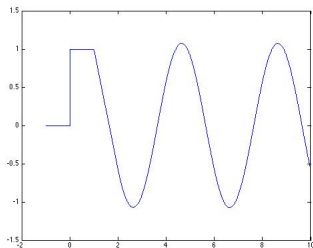
# A linear ODE with time delay

$$y'(t) = \kappa y(t-1), \quad t > 0$$
$$y(t) = y_0(t), \quad -1 \leq t \leq 0.$$

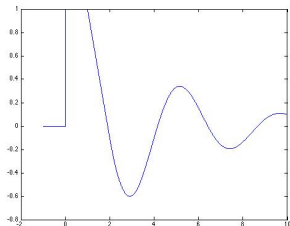


T. Erneux,

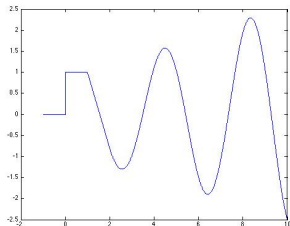
Applied delay differential equations,  
Springer, 2009



$$\kappa = -\pi/2$$



$$\kappa = -1.8, \quad y_0(0) = 1, y_0(t) = 0, t < 0$$



$$\kappa = -1.1$$

# Nonlinear ODE with delay

We consider nonlinear equations with cubic nonlinearity, e.g.

$$\begin{aligned}y'(t) + y^3(t) &= \kappa y(t - \tau), & t > 0 \\y(t) &= y_0(t), & -1 \leq t \leq 0.\end{aligned}$$

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Instead of  $R(y) = y^3$ , consider more general reaction terms like

$$R(y) = (y - y_1)(y - y_2)(y - y_3)$$

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$$y'(t) + R(y(t)) = \kappa y(t - \tau).$$

Pyragas feedback control:

$$y'(t) + R(y(t)) = \kappa (y(t - \tau) - y(t)).$$

So far, we had  $y : [0, T] \rightarrow \mathbb{R}$ . Let  $y$  also depend on a spatial variable  $x \in \Omega \subset \mathbb{R}^n$ ,  $y = y(x, t)$ , and consider

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$$\begin{aligned}(\partial_t y - \Delta_x y + R(y))(x, t) &= \kappa y(x, t - \tau) && \text{in } \Omega \times (0, T) \\ y &= y_0, && \text{in } \Omega \times [-\tau, 0] \\ \partial_n y &= 0 && \text{in } \partial\Omega \times (0, T).\end{aligned}$$

Reaction term:

$$R(y) = \rho (y - y_1)(y - y_2)(y - y_3),$$

$\rho > 0$ ,  $y_1 \leq y_2 \leq y_3$ .

Let

$$m_R := \min_y R'(y).$$

# Multiple time delays

More general are multiple time delays  $0 \leq \tau_1 < \tau_2 \dots < \tau_m \leq T$ ,

$$\begin{aligned}(\partial_t y - \Delta y + R(y))(x, t) &= \sum_{i=1}^m \kappa_i y(x, t - \tau_i) & (x, t) \in Q &= \Omega \times (0, T) \\ y &= y_0 & \text{in } Q_- &= \Omega \times [-T, 0] \\ \partial_n y &= 0 & \text{in } \Sigma &= \partial\Omega \times (0, T).\end{aligned}$$

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$$u(x, t) = \sum_{i=1}^m \kappa_i y(x, t - \tau_i).$$

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We will optimize the weights  $\kappa_i$  and the delays  $\tau_i$  for fixed  $m$ .

Set  $\tau := (\tau_1, \dots, \tau_m)$ ,  $\kappa := (\kappa_1, \dots, \kappa_m)$ ,  $U := (\tau, \kappa)$ .

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# Existence and uniqueness

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**Nonlocal problem with Borel measure  $\mu \in \mathcal{M}[0, T]$**

$$\begin{aligned}\partial_t y(x, t) - \Delta y(x, t) + R(y(x, t)) &= \int_0^T y(x, t - s) d\mu(s) & (x, t) \in Q \\ y &= y_0 & \text{in } Q_- \\ \partial_n y &= 0 & \text{in } \Sigma.\end{aligned}$$

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Assume  $y_0 \in C(\overline{Q_-})$ .

# The nonlocal problem with measures

## Theorem (Casas, Mateos, Tr. 2017)

For all  $T > 0$  and every  $\mu \in \mathcal{M}[0, T]$ , the nonlocal problem has a unique solution  $y_\mu \in \mathcal{Y} = W(0, T) \cap C(\bar{Q})$ . We have

$$\begin{aligned}\|y_\mu\|_{L^2(0, T; H^1(\Omega))} &\leq C(\|y_0\|_{L^2(Q_-)} \|\mu\|_{\mathcal{M}[0, T]} + \|y_0(\cdot, 0)\|_{L^2(\Omega)} + |R(0)|) \\ \|y_\mu\|_{C(\bar{Q})} &\leq C(\|y_0\|_{C(\bar{Q}_-)} \|\mu\|_{\mathcal{M}[0, T]} + \|y_0(\cdot, 0)\|_{C(\bar{\Omega})} + |R(0)|),\end{aligned}$$

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$\implies$

## Theorem (Casas, Mateos, Tr. 2018)

To each weight  $\kappa \in \mathbb{R}^m$  and delay  $\tau \in \mathbb{R}^m$ , there exists a unique solution  $y_{\tau, \kappa} \in \mathcal{Y}$ . The mapping  $(\tau, \kappa) \mapsto y_{\tau, \kappa}$  is continuous from  $\mathbb{R}^{2m}$  to  $\mathcal{Y}$ .

# Differentiability of the mapping $(\tau, \kappa) \mapsto y_{\tau, \kappa}$

## Theorem (Partial derivatives)

*The mapping  $G : (\tau, \kappa) \mapsto y_{(\tau, \kappa)}$  from  $\mathbb{R}^{2m}$  to  $\mathcal{Y}$  has all partial derivatives  $\partial_{\kappa_i} G$  and  $\partial_{\tau_i} G$ ,  $i = 1, \dots, m$ .*

The result cannot be directly obtained as a particular case of the problem with measures  $\mu$ , where we proved the differentiability w.r. to  $\mu$ .

The partial derivatives can be obtained by solving linearized PDEs.

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**Formal computation of derivatives:** Consider e.g.  $\partial_{\kappa_i} G$ ; insert

$$y = (G(\tau, \kappa))(x, t)$$

$$\left( \partial_t G(\tau, \kappa) - \Delta G(\tau, \kappa) + R(G(\tau, \kappa)) \right)(x, t) = \sum_{j=1}^m \kappa_j G(\tau, \kappa)(x, t - \tau_j)$$



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## Theorem (Partial derivatives)

The partial derivatives of  $G : \mathbb{R}^{2m} \rightarrow \mathcal{Y}$ ,  $(\tau, \kappa) \mapsto y_{\tau, \kappa}$  are given as follows: For every  $(\tau, \kappa) = u$  and  $1 \leq i \leq m$ , we have  $\partial_{\tau_i} G(u) = z_i$ , where  $z_i$  satisfies the equation

$$\left\{ \begin{array}{l} \partial_t z - \Delta z + R'(y_u) z = \sum_{j=1}^m \kappa_j z(x, t - \tau_j) - \kappa_i \partial_t y_u(x, t - \tau_i) \text{ in } Q \\ \partial_n z = 0 \text{ on } \Sigma, z = 0 \text{ in } Q_-, \end{array} \right.$$

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and  $\partial_{\kappa_i} G(u) = \eta_i$ , where  $\eta_i$  satisfies

$$\begin{cases} \partial_t \eta - \Delta \eta + R'(y_u) \eta = \sum_{j=1}^m \kappa_j \eta(x, t - \tau_j) + y_u(x, t - \tau_i) & \text{in } Q \\ \partial_n \eta = 0 & \text{on } \Sigma, \quad \eta = 0 & \text{in } Q_-. \end{cases}$$

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# The optimization problem

Admissible set:

$$U_{ad} = \{u = (\tau, \kappa) \in \mathbb{R}^m \times \mathbb{R}^m : a_i \leq \tau_i \leq b_i, \alpha_i \leq \kappa_i \leq \beta_i, 1 \leq i \leq m\},$$

Desired state  $y_Q \in L^2(Q)$

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Optimization problem

$$(P) \quad \min_{u \in U_{ad}} J(u) = \frac{1}{2} \int_Q (y_u - y_Q)^2 dxdt + \frac{\nu}{2} |\kappa|^2,$$

where  $\nu \geq 0$  is fixed;  $y_u$  solves

$$\begin{cases} \partial_t y - \Delta y + R(y) = \sum_{i=1}^m \kappa_i y(x, t - \tau_i) & \text{in } Q \\ \partial_n y = 0 & \text{on } \Sigma \\ y = y_0 & \text{in } Q_- \end{cases}$$



## Theorem

*The partial derivatives of  $J$  are given by*

$$\partial_{\tau_i} J(u) = -\kappa_i \int_Q \varphi_u(x, t) \partial_t y_u(x, t - \tau_i) dx dt,$$

$$\partial_{\kappa_i} J(u) = \nu \kappa_i + \int_Q \varphi_u(x, t) y_u(x, t - \tau_i) dx dt,$$

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$1 \leq i \leq m$ , where *the adjoint state*  $\varphi_u \in \mathcal{Y}$  is the unique solution to the *advanced adjoint equation*

$$\begin{cases} -\partial_t \varphi - \Delta \varphi + R'(y_u) \varphi = y_u - y_Q + \sum_{i=1}^m \kappa_i \varphi(x, t + \tau_i) & \text{in } Q \\ \partial_n \varphi(x, t) = 0 & \text{on } \Sigma, \varphi(x, t) = 0 & \text{if } t \geq T. \end{cases}$$

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The theorem follows from the last one by the chain rule.

# Existence of optimal solutions

## Theorem (Existence of an optimal solution)

*If  $\nu > 0$  or  $-\infty < \alpha_i \leq \beta_i < \infty$  for all  $i \in \{1, \dots, m\}$ , then Problem (P) has a solution  $\bar{u} = (\bar{\tau}, \bar{\kappa})$ .*

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## Proof.

Let  $\kappa^k \rightarrow \kappa$  and  $\tau^k \rightarrow \tau$  in  $\mathbb{R}^m$  be infimal sequences, then

$$\sum_{i=1}^m \kappa_i^k \delta_{\tau_i^k} \xrightarrow{*} \sum_{i=1}^m \kappa_i \delta_{\tau_i} \quad \text{in } \mathcal{M}[0, T]$$

as  $k \rightarrow \infty$ .

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By a **weak\*  $\rightarrow$  strong continuity** result for the mapping  $\mu \mapsto y_\mu$  (Casas/Mateos/Tr. 2017), it follows that

$$y_{\kappa^k} \rightarrow y_\kappa \quad \text{in } L^2(0, T; H^1(\Omega)) \cap C(\bar{Q}).$$

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$$\sum_{i=1}^m \kappa_i^k \delta_{\tau_i^k} \xrightarrow{*} \sum_{i=1}^m \kappa_i \delta_{\tau_i} \quad \text{in } \mathcal{M}[0, T]$$

as  $k \rightarrow \infty$ .

By a **weak\*  $\rightarrow$  strong continuity** result for the mapping  $\mu \mapsto y_\mu$  (Casas/Mateos/Tr. 2017), it follows that

$$y_{\kappa^k} \rightarrow y_\kappa \quad \text{in } L^2(0, T; H^1(\Omega)) \cap C(\bar{Q}).$$

Therefore,  $J$  is continuous in  $U$ . Either the objective functional is coercive w.r. to  $\kappa$  or  $U_{ad}$  is compact in  $\mathbb{R}^{2m}$ . Therefore, (P) has a global solution.  $\square$

## Theorem (Necessary optimality conditions)

Let  $\bar{u} \in U_{ad}$  be a local solution of (P) and let  $\bar{y} := y_{\bar{u}}$  be the associated state. Then there exists a unique associated adjoint state  $\bar{\varphi} := \varphi_{\bar{u}} \in \mathcal{Y}$  such that the variational inequalities

$$-\bar{\kappa}_i \int_Q \partial_t \bar{y}(x, t - \bar{\tau}_i) \bar{\varphi}(x, t) \, dx dt (\tau_i - \bar{\tau}_i) \geq 0 \quad \forall \tau_i \in [a_i, b_i],$$

and

$$\left( \nu \bar{\kappa}_i + \int_Q \bar{y}(x, t - \bar{\tau}_i) \bar{\varphi}(x, t) \, dx dt \right) (\kappa_i - \bar{\kappa}_i) \geq 0 \quad \forall \kappa_i \in [\alpha_i, \beta_i] \cap \mathbb{R},$$

are satisfied for  $i = 1, \dots, m$ .



# Outline

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- 2 Control-to-state mapping
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# Discretization of the state equation

- $\Omega$ : polygonal or polyhedral domain of  $\mathbb{R}^d$ ,  $d \leq 3$ ,
- $\{\mathcal{K}_h\}_{h>0}$ : quasi-uniform family of triangulations of  $\bar{\Omega}$ , size  $h$ ,
- $0 = t_0 < t_1 < \dots < t_{N_\delta} = T$ : quasi-uniform partition of  $[0, T]$ ,
- $I_k = (t_{k-1}, t_k]$ ,  $\delta_k = t_k - t_{k-1}$ , size  $\delta = \max\{\delta_k\}$ ,  $\sigma = (h, \delta)$ .

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## Finite dimensional spaces

$$Y_h = \{z_h \in C(\bar{\Omega}) : z_{h|K} \in \mathcal{P}^1(K) \forall K \in \mathcal{K}_h\},$$

$$\mathcal{Y}_\sigma^0 = \{\phi_\sigma \in L^2(0, T; Y_h) : \phi_\sigma|_{I_k} \in \mathcal{P}^0(I_k; Y_h) \forall k = 1, \dots, N_\delta\},$$

$$\mathcal{Y}_\sigma^1 = \{y_\sigma \in C([0, T]; Y_h) : y_\sigma|_{I_k} \in \mathcal{P}^1(I_k; Y_h) \forall k = 1, \dots, N_\delta\}$$

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- $\mathcal{P}^1(K)$ : set of polynomials of degree 1 in  $K$
- $\mathcal{P}^i(I_k; Y_h)$ : set of polynomials of degree  $i$  defined in  $I_k$  with values in  $Y_h$ ,  $i = 0, 1$

For  $\phi_\sigma \in \mathcal{Y}_\sigma^0$ , we denote by  $\phi_\sigma^k \in Y_h$  the value of  $\phi_\sigma$  in  $I_k$ .

# Discretization of the state equation

## Discrete state equation in variational form:

The discrete state  $y_\sigma(u) \in \mathcal{Y}_\sigma^1$  is the unique solution to (cf. Becker-Meidner-Vexler (2007))

$$y_{\sigma,0}(x, 0) = \Pi_h y_0(x, 0),$$

$$\begin{aligned} & \int_Q \partial_t y_\sigma \phi_\sigma \, dxdt + \int_Q \nabla_x y_\sigma \cdot \nabla_x \phi_\sigma \, dxdt + \int_Q R(y_\sigma) \phi_\sigma \, dxdt \\ &= \sum_{i=1}^m \kappa_i \left[ \int_\Omega \int_0^{\tau_i} y_0(x, t - \tau_i) \phi_\sigma \, dxdt + \int_\Omega \int_{\tau_i}^T y_\sigma(x, t - \tau_i) \phi_\sigma \, dxdt \right] \quad \forall \phi_\sigma \in \mathcal{Y}_\sigma^0, \end{aligned}$$

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**Discretized optimization problem:** Defined upon the discretized state.

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# Example 1 (ODE)

In all examples:

$$R(y) = y(y - 0.25)(y - 1), \quad T = 80, \quad \nu = 0.$$

$$0 \leq \tau_i \leq T, \quad |\kappa_i| \leq 1000, \quad i = 1, \dots, m.$$

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$$y'(t) + R(y(t)) = \sum_{i=1}^m \kappa_i y(t - \tau_i), \quad t \in (0, T], \quad y(t) = y_0(t), \quad t \leq 0,$$

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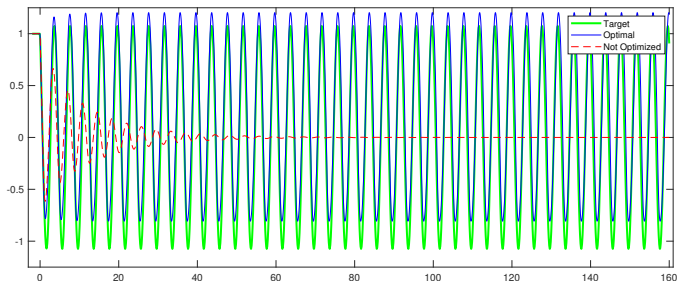
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Take  $m = 1$ . **Aim:** Find  $u = (\tau, \kappa)$ , to best approximate  $y_Q$  by the solution of the **nonlinear** delay equation with **initial function**  $y_0(t) = 1$ .

# Example 1, result

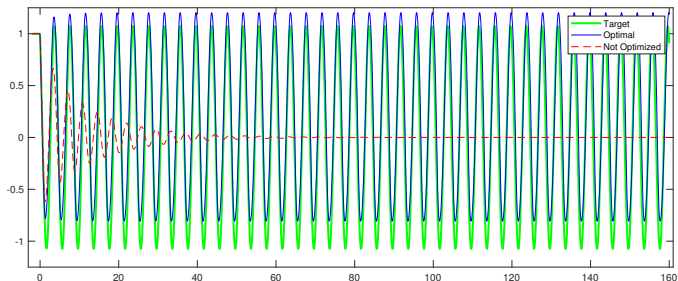
Computed optimal solution for  $m = 1$ :  $\bar{\tau} = 1.2409$ ,  $\bar{\kappa} = -1.7668$ ,  
 $J(\bar{u}) = 1.8701$ ,  $|\nabla J(\bar{u})| = 3.8 \cdot 10^{-7}$ .



Uncontrolled state: **red**, Target state: **green**, Optimal state: **blue**

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Optimization code: MATLAB code `fmincon`, derivatives by adjoint calculus

# Example 2

## Data of Nestler/Schöll/Tr. (2016)

- $\Omega = (-20, 20)$
- $y_0(x, t) = \frac{1}{2} \left[ 1 - \tanh \left( \frac{x-vt}{2} \right) \right]$  with  $v = 0.25\sqrt{2}$
- $y_Q(x, t) = 3 \sin \left( t - \cos \left( \frac{\pi}{20}(x + 20) \right) \right)$
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Desired state  $y_Q$

$i$	$\bar{\tau}_i$	$\bar{\kappa}_i$
1	0.0000	0.9846
2	0.9367	-1.5039
3	6.7481	0.4542
4	28.3843	-2.2799
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$$J(\bar{u}) = 4209.3, \quad \partial_{\tau_1} J(\bar{u}) = 486,$$
$$|\partial_{\tau_i} J(\bar{u})| \leq 2 \cdot 10^{-4}, \quad (i > 1)$$



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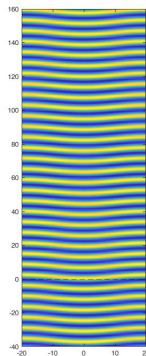
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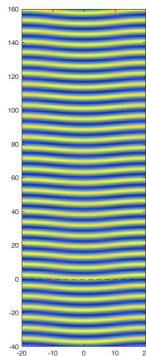
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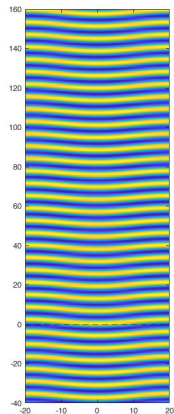
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Notice:  $Q = (-20, 20) \times (0, 80)$ ,  $|Q| = 3200$ ,  $\|1\|_{L^2(Q)}^2 = 3200$ .

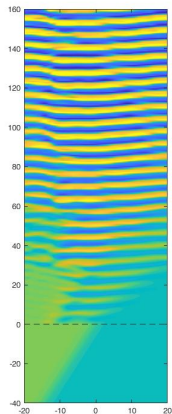
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# Example 2



Desired state



Computed optimal state

# References



Nestler, P., Schöll, E., T. F.

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Take  $u(t) = \kappa (y(t - \tau) - y(t))$ .

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**Pyragas feedback:** Let  $\tau > 0$  and  $y_0 : \mathbb{Q}_- \rightarrow \mathbb{R}$  be given,

$$\begin{aligned}\partial_t y(t) - \Delta y(t) + R(y(t)) &= \kappa (y(t - \tau) - y(t)), & t > 0, \\ y(t) &= y_0(t), & t \in [-\tau, 0].\end{aligned}$$

If  $y$  has period  $\tau$ , then the feedback vanishes. This type of feedback is very popular in Theoretical Physics.

**Pyragas feedback with multiple delays:**

$$\partial_t y(t) - \Delta y(t) + R(y(t)) = \sum_{i=1}^k \kappa_i y(t - \tau_i) - \kappa_0 y(t)$$



# Nonlocal Pyragas feedback

Nonlocal version with "distributed time delay"

$$\begin{aligned}\partial_t y(t) - \Delta y(t) + R(y(t)) &= \kappa \left( \int_0^T g(\tau) y(t - \tau) d\tau - y(t) \right), & t > 0, \\ y &= y_0, & \text{in } [-\tau, 0],\end{aligned}$$

with given **feedback gain**  $\kappa$  and a **kernel**  $g \in L^\infty(0, T)$ .

## Forward problem: $g \mapsto y$

Depending on the chosen **feedback kernel**  $g$ , different solutions  $y$  are generated.

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$$\Omega = (0, 200), \quad T = 400, \quad y_0(x, t) := \frac{1}{2} \left( 1 - \tanh \left( \frac{x-vt}{2\sqrt{2}} \right) \right)$$

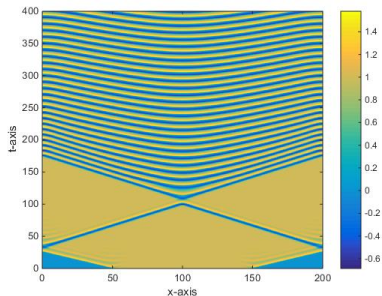
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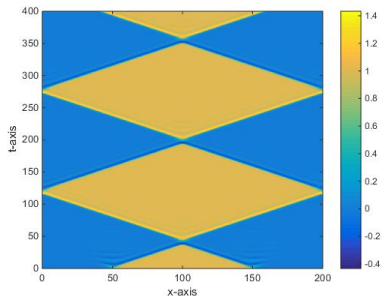
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$$y_2 = 0.25, \quad \kappa = -1.65$$



$$y_2 = 0.5, \quad \kappa = -1.4$$

# Pyragas feedback with Borel measures

**More general idea:** Cover standard and nonlocal Pyragas type feedback in a unified way by using a Borel measure.

$$\left\{ \begin{array}{ll} \frac{\partial y}{\partial t} - \Delta y + R(y) = \int_0^T y(t-s) d\mu(s) & \text{in } Q, \\ \partial_n y = 0 & \text{on } \Sigma, \\ y = y_0 & \text{in } Q_- \end{array} \right.$$

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Here, the regular Borel measure  $\mu$  plays the role of the "control".

Find the measure  $\bar{\mu}$  such that the solution of this equation best approximates a desired solution.

# Optimal feedback design problem

*Design problem:*



# Optimal feedback design problem

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$$\min \frac{1}{2} \|y - y_Q\|_{L^2(Q)}^2 + \frac{\nu}{2} \|\mu\|_{\mathcal{M}(0,T)}$$

where  $y$  solves

$$\begin{cases} \partial_t y(t) - \Delta y(t) + R(y(t)) &= \int_0^T y(t-\tau) d\mu(\tau) & \text{in } Q, \\ y &= y_0 & \text{in } Q_-, \\ \partial_n y &= 0 & \text{in } \Sigma. \end{cases}$$



Casas, E., Mateos, M., T. F.

Measure control of a semilinear parabolic equation with a nonlocal time delay,  
Accepted by SICON

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# A stability theorem

**Desired state:**  $y^{desi} \in L^\infty(Q_{2\tau})$ ,  $\tau$ -periodic,  $y^{desi} = y^{desi}(t, x)$

**Assume:**  $y^{desi}$  has Neumann traces  $y_x^{desi}(\cdot, 0) \in L^p(0, \tau)$  and  $y_x^{desi}(\cdot, L) \in L^p(0, \tau)$ ,  $p > 2$ .

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- Extend  $y^{desi}$  to a  $\tau$ -periodic function on  $[0, \infty) \times [0, L]$ .
- Let  $y_0 \in C([-\tau, 0], L^\infty(0, L))$ ; consider

# A stability theorem

**Desired state:**  $y^{desi} \in L^\infty(Q_{2\tau})$ ,  $\tau$ -periodic,  $y^{desi} = y^{desi}(t, x)$

**Assume:**  $y^{desi}$  has Neumann traces  $y_x^{desi}(\cdot, 0) \in L^p(0, \tau)$  and  $y_x^{desi}(\cdot, L) \in L^p(0, \tau)$ ,  $p > 2$ .

– Extend  $y^{desi}$  to a  $\tau$ -periodic function on  $[0, \infty) \times [0, L]$ .

– Let  $y_0 \in C([-\tau, 0], L^\infty(0, L))$ ; consider

(S)

$$y_t = y_{xx} - y_{xx}(\cdot - \tau, \cdot) - \rho [R(y) - R(y(\cdot - \tau, \cdot))] \\ + y_t(\cdot - \tau, \cdot) - \kappa [y - y(\cdot - \tau, \cdot)], \quad \text{in } (0, \infty) \times (0, L),$$

$$y_x(t, 0) = y_x^{desi}(t, 0), \quad t > 0,$$

$$y_x(t, L) = y_x^{desi}(t, L), \quad t > 0,$$

$$y(t, x) = y_0(t, x) \quad (t, x) \in [-\tau, 0] \times (0, L).$$

## Theorem (Exponential Stability)

Let  $\tau > 0$  and a  $\tau$ -periodic state  $y^{desi} \in H^2(Q_{2\tau})$  be given. Define  $\mu = 2\kappa - 2\rho|m_R|$ . Assume that  $\mu > 0$ . Then

$$V(t) = \frac{1}{2} \int_0^L (y(t, x) - y(t - \tau, x))^2 dx$$

is a strict Lyapunov function for **(S)**, i.e. for  $t \geq \tau$  we have

$$V(t) \leq \exp(-\mu(t - \tau)) V(\tau).$$

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Moreover, there are a  $\tau$ -periodic  $y^* \in L^2(Q_{2\tau})$  and  $c_0 > 0$  such that

$$\int_0^L (y(t, x) - y^*(t, x))^2 dx \leq c_0 \exp(-\mu t)$$

for all  $t \geq \tau$ .

## Example 1 continued

Let  $R(y) = y(y - 0.25)(y - 1)$ .

For delay  $\tau = 1.240683838477202$  and weight  $\kappa = -1.766552137106608$ , let  $y : [-\tau, \infty) \rightarrow \mathbb{R}$  be the solution  $y$  of the **nonlinear** delay equation

$$\begin{aligned}y'(t) + R(y(t)) &= \kappa y(t - \tau) \quad \text{for } t \in (0, \infty), \\y(t) &= 1 \quad \text{if } t \leq 0\end{aligned}$$

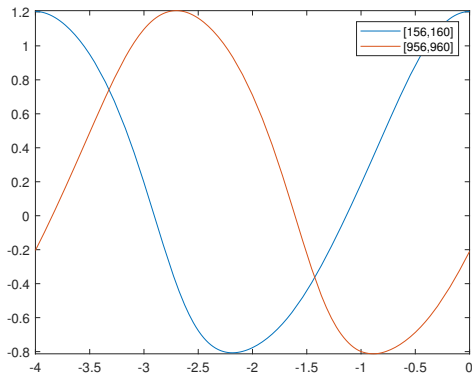
that in  $[0, 160]$  minimizes the  $L^2$ -distance to the solution  $v$  of

$$\begin{aligned}v'(t) &= -\frac{\pi}{2}v(t - 1) \quad \text{for } t \in [0, \infty), \\v(t) &= 1 \quad \text{if } t \leq 0.\end{aligned}$$

This  $v$  has period  $\tau = 4$ .



The long-term behaviour in  $[0, 960]$  shows that  $y$  is possibly not periodic



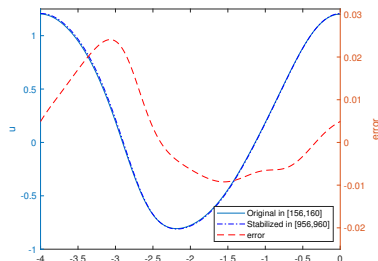
We stabilize the solution by Pyragas feedback with  $\tau = 4$ .

# Stabilization by Pyragas feedback

We solve the equation

$$u'(t) + R(u(t)) = \kappa u(t - \tau) + 100(u(t - 4) - u(t)), \quad t \in [160, 960],$$
$$u(t) = y(t), \quad t \in [156, 160)$$

and obtain a very good numerical adjustment to 4-periodicity;



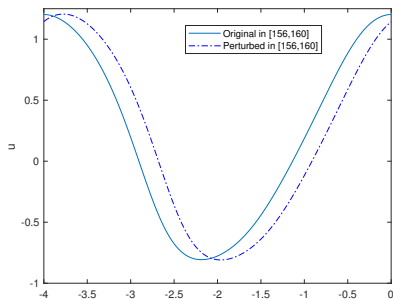
(error graph is magnified 40 times.)

# Example 1 under perturbation

The solution is very sensitive w.r. to perturbations of the data: For the tiny modification  $\hat{\tau} = 1.24$ ,  $\hat{\kappa} = -1.77$ , the solution of

$$\begin{aligned}\hat{y}'(t) + R(\hat{y}(t))_+ &= \hat{\kappa}\hat{y}(t - \hat{\tau}), & t \in (0, \infty), \\ \hat{y}(t) &= 1, & t \leq 0\end{aligned}$$

considerably differs from the unperturbed solution.

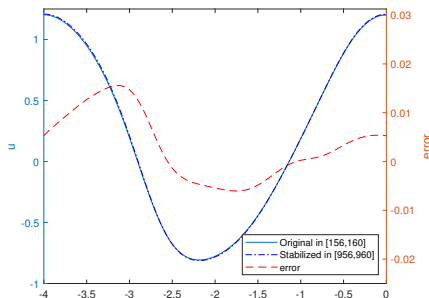


# Example 1, perturbation and Pyragas feedback

Again, Pyragas feedback stabilizes the solution: The solution of

$$y'(t) + R(y(t)) = \hat{\kappa}y(t - \hat{\tau}) + \kappa(y(t - 4) - y(t)), \quad t \in (160, 960],$$
$$y(t) = 1.01v(t), \quad t \in [156, 160]$$

not only appears to be periodic, but also is very close to the solution of the original unperturbed problem.





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**Thank you**