

# Internal Model Control of Partial Differential Equations

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including joint work with

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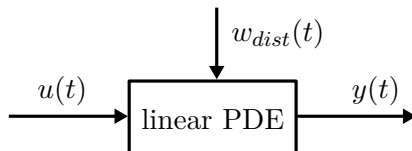
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# Introduction

## Problem

Study **robust output regulation** of linear PDE models.



**Output Regulation = Tracking + Disturbance Rejection:**

Design a controller such that the output  $y(t)$  of the system converges to a reference signal despite the disturbance  $w_{dist}(t)$ , i.e.,

$$\|y(t) - y_{ref}(t)\| \rightarrow 0, \quad \text{as } t \rightarrow \infty$$

**Robustness:** The controller is required to tolerate uncertainty in the parameters of the system.

# Applications

## Applications of regulation for PDEs:

- Temperature tracking control, e.g., in manufacturing processes
- Tracking control of flexible robotic manipulators
- Rejection of unwanted periodic noises or vibrations

## Robustness:

- Tolerance to the unavoidable **uncertainty** in models.
- Allows reliable use of **approximate** controller parameters.

# Goal of the Talk

- Highlight differences between internal model control for (linear) ODE and PDE systems
- Discuss selected approaches to controller design
- Examples in tutorial style

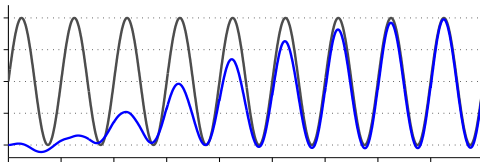
# The Reference and Disturbance Signals

The reference and disturbance signals are of the form

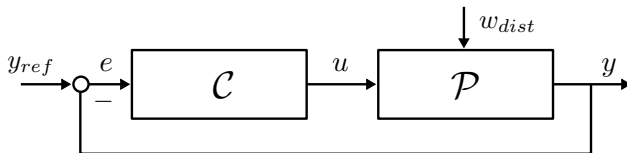
$$y_{ref}(t) = \sum_{k=0}^q a_k \cos(\omega_k t + \theta_k)$$

$$w_{dist}(t) = \sum_{k=0}^q b_k \cos(\omega_k t + \varphi_k)$$

with **known frequencies**  $0 = \omega_0 < \omega_1 < \dots < \omega_q$  and unknown amplitudes and phases.



# The Dynamic Error Feedback Controller



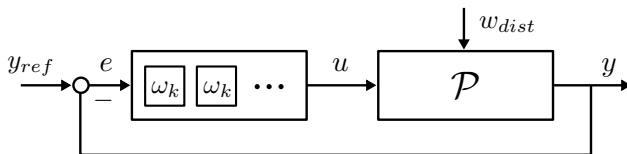
We consider a dynamic error feedback controller which is another linear system.

## Theorem

*The Robust Output Regulation Problem is solvable if the system*

- *is stabilizable and detectable*
- *does not have transmission zeros at the frequencies  $\pm i\omega_k$  of  $y_{ref}(t)$  and  $w_{dist}(t)$ .*

# The Internal Model Principle



## Theorem (Francis–Wonham, Davison 1970's, ...)

*The following are equivalent:*

- *The controller solves the robust output regulation problem.*
- *Closed-loop system is stable and the controller has **an internal model** of the frequencies  $\{\omega_k\}_k$  of  $w_{dist}(t)$  and  $y_{ref}(t)$ .*

**“Internal Model”:** For every  $k$ , the complex frequencies  $\pm i\omega_k$  must be eigenmodes of the controller dynamics with at least  $p = \dim Y$  independent eigenvectors.

# Internal Model Based Controller Design

The robust output regulation problem can be solved in two parts:

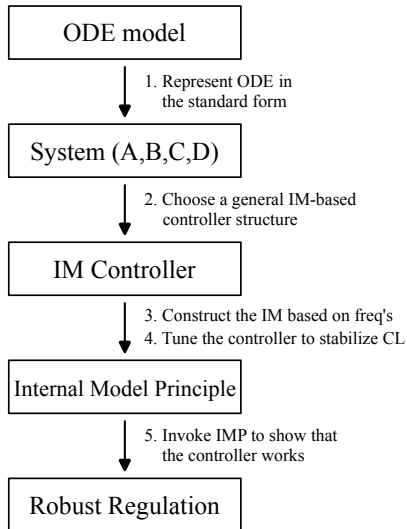
**Step 1°** Include a suitable internal model into the controller

**Step 2°** Use the rest of the controller's parameters to stabilize the closed-loop system.

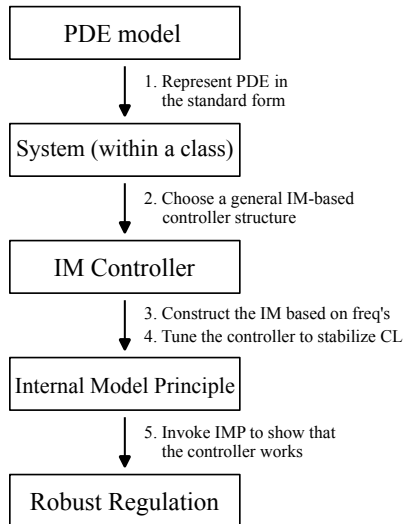
Internal model has fixed structure (easy), the closed-loop stability can be achieved in several ways (often the main challenge).



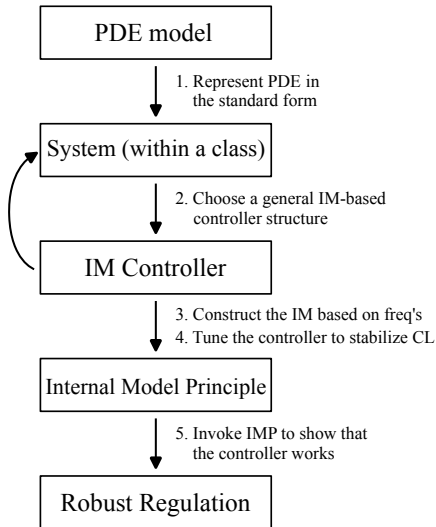
# Robust Regulation of Linear ODE Systems



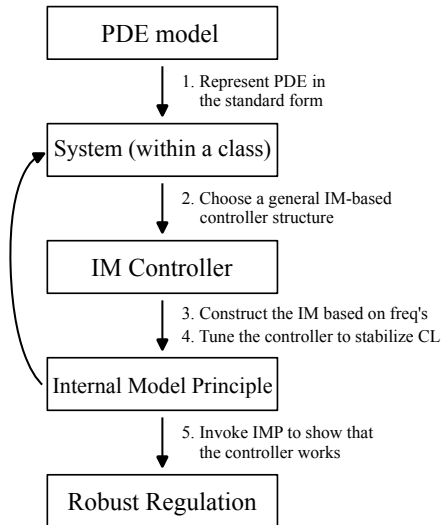
# Robust Regulation of linear PDE systems



# Robust Regulation of linear PDE systems



# Robust Regulation of linear PDE systems



# The Possible Types of “Standard Representations”

## 1. “A class of PDEs”

- A parametrised collection of PDEs of the same type (parabolic, hyperbolic, coupled)
- Allows various input/output configurations
- Examples: Distributed port-Hamiltonian systems,  $n \times m$  hyperbolic systems, reaction-convection-diffusion equations

# The Possible Types of “Standard Representations”

## 1. “**A class of PDEs**”

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- Examples: Distributed port-Hamiltonian systems,  $n \times m$  hyperbolic systems, reaction-convection-diffusion equations

## 2. “**Abstract linear systems**” (or “DPS”)

- Represents the PDE as a linear system  $(A, B, C, D)$  but on an *infinite-dimensional* space  $X$
- Different subclasses based on properties of the *operators*
- Examples: Regular Linear Systems, Well-Posed Systems

# Comparison of Internal Model Controllers

## 1. **A class of PDEs:**

- Currently there's no actual “IMP” for PDE classes
- **but** IM-based controllers (with tuning recipes) exist!

## 2. **Abstract linear systems:**

- Internal Model Principle exists for several system classes.
- Several different types of IM controllers

# Comparison of Internal Model Controllers

## 1. A class of PDEs:

- Currently there's no actual "IMP" for PDE classes
- **but** IM-based controllers (with tuning recipes) exist!
- **Benefit:** Usually straightforward design!

## 2. Abstract linear systems:

- Internal Model Principle exists for several system classes.
- Several different types of IM controllers
- **Trade-off:** Sometimes technically demanding design.



## Part II: Controller Design for PDEs

- Illustrate how controller design works for concrete PDEs
- Keep the PDE models simple
- Swipe a lot of technical terms and details under the rug

## The “Simplest Example”

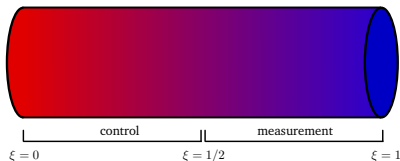
Consider a one-dimensional heat equation

$$\partial_t v(\xi, t) = \partial_{\xi\xi} v(\xi, t) + b(\xi)u(t) + b_d(\xi)w_{dist}(t), \quad \xi \in (0, 1)$$

$$\partial_{\xi} v(0, t) = \partial_{\xi} v(1, t) = 0, \quad v(\xi, 0) = v_0(\xi)$$

$$y(t) = \int_0^1 v(\xi, t) c(\xi) d\xi$$

with piecewise continuous functions  $b, b_d, c : [0, 1] \rightarrow \mathbb{R}$ .



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with piecewise continuous functions  $b, b_d, c : [0, 1] \rightarrow \mathbb{R}$ .

Choose  $x(t) = v(\cdot, t)$  and look for a representation

$$\dot{x}(t) = Ax(t) + Bu(t) + B_d w_{dist}(t), \quad x(0) = x_0 \in X$$

$$y(t) = Cx(t)$$

## Abstract Representation

The heat equation can be rewritten as a linear system

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + B_d w_{dist}(t), & x(0) &= x_0 \in X \\ y(t) &= Cx(t)\end{aligned}$$

if we choose

- State  $x(t) = v(\cdot, t)$  (heat profile at  $t \geq 0$ ),
- State space  $X = L^2(0, 1)$  (Hilbert),  $U = Y = U_d = \mathbb{C}$
- Operators  $A : \mathcal{D}(A) \subset X \rightarrow X$ ,  $B$ ,  $B_d$ , and  $C$

$$Af = \frac{d^2 f}{d\xi^2}, \quad \mathcal{D}(A) = \{ f \in H^2(0, 1) \mid f'(0) = f'(1) = 0 \}$$

$$Bu = b(\cdot)u, \quad B_d u = b_d(\cdot)u \quad Cf = \int_0^1 f(\xi)c(\xi)d\xi$$

# Internal Model Controller design

Simplifying assumptions:

- $y_{ref}(t) = a \sin(\omega t + \theta)$ , single freq  $\omega > 0$  and  $w_{dist}(t) \equiv 0$
- SISO systems only (generalisation to MIMO easy)
- Use one particular “observer-based” controller structure (others exist too!)

## Controller Design for the Heat System

Theorem (Immonen 2007, Hämäläinen–Pohjolainen 2010)

*The Robust Output Regulation Problem can be solved with the internal model controller*

$$\dot{z}_1(t) = G_1 z_1(t) + G_2(y(t) - y_{\text{ref}}(t))$$

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L(\hat{y}(t) - y(t) + y_{\text{ref}}(t))$$

$$\hat{y}(t) = C\hat{x}(t)$$

$$u(t) = K_1 z_1(t) + K_2 \hat{x}(t)$$

*with matrices  $G_1$ ,  $G_2$ ,  $K_1$  and bounded operators  $L$  and  $K_2$ .*

The matrices  $(G_1, G_2)$  contain the internal model,

$$G_1 = \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix}, \quad G_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

## Controller Parameters $L$ , $K_1$ and $K_2$

**The operators  $L$ ,  $K_1$  and  $K_2$ :** Chosen so that the systems

$$\dot{x}(t) = (A + LC)x(t) \quad \leadsto \quad \text{output injection}$$

and

$$\dot{x}(t) = \left( \begin{bmatrix} G_1 & G_2 C \\ 0 & A \end{bmatrix} + \begin{bmatrix} 0 \\ B \end{bmatrix} [K_1, K_2] \right) x(t)$$

are exponentially stable.

## Controller Parameters $L$ , $K_1$ and $K_2$

Stabilize

$$\dot{x}(t) = \left( \begin{bmatrix} G_1 & G_2 C \\ 0 & A \end{bmatrix} + \begin{bmatrix} 0 \\ B \end{bmatrix} \begin{bmatrix} K_1 & K_2 \end{bmatrix} \right) x(t)$$

Can be rewritten as feedback stabilization of a PDE-ODE cascade:

$$\dot{z}_1(t) = G_1 z_1(t) + G_2 \int_0^1 v(\xi, t) c(\xi) d\xi$$

$$\partial_t v(\xi, t) = \partial_{\xi\xi} v(\xi, t) + b(\xi) K_1 z_1(t) + b(\xi) \int_0^1 v(\xi, t) k_2(\xi) d\xi$$

$$\partial_{\xi} v(0, t) = \partial_{\xi} v(1, t) = 0,$$

Possible approaches:

- Numerical approximations and LQR (Banks–Kunisch '84)
- “**Forwarding**” (+ numerical approx)



# The PDE Controller

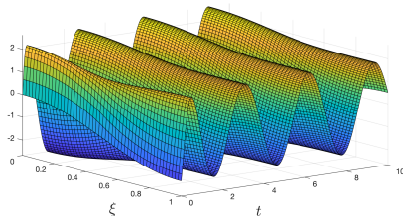
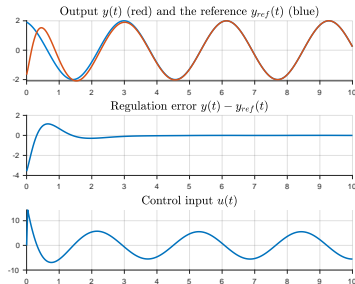
Finally, rewrite the “abstract controller” as a PDE-ODE system:

$$\dot{z}_1(t) = \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix} z_1(t) + \begin{bmatrix} 1 \\ 1 \end{bmatrix} (y(t) - y_{ref}(t))$$

$$\begin{aligned} \partial_t \hat{v}(\xi, t) &= \partial_{\xi\xi} \hat{v}(\xi, t) + b(\xi) K_1 z_1(t) + b(\xi) \int_0^1 k_2(\xi) \hat{v}(\xi, t) d\xi \\ &\quad + \ell(\xi) \left( \int_0^1 c(\xi) \hat{v}(\xi, t) d\xi - y(t) + y_{ref}(t) \right) \\ u(t) &= K_1 z_1(t) + \int_0^1 k_2(\xi) \hat{v}(\xi, t) d\xi, \quad \partial_{\xi} \hat{v}(\xi, t) = \partial_{\xi} \hat{v}(\xi, t) = 0. \end{aligned}$$

(Here we use the knowledge of  $X$ ,  $A$ ,  $B$ ,  $C$  etc.)

# Simulation Results



**RORPack** – Matlab/Python libraries for Robust Output Regulation  
Available at <https://github.com/lassipau/rorpack-matlab/>

## Example: What did we learn?

- ① Representation of the PDE as an abstract system was “easy”
- ② Choices of  $L$ ,  $K_1$  and  $K_2$  could be completed by solving stabilization problems for a PDE and a PDE-ODE cascade
- ③ The abstract controller that we constructed could be reinterpreted as a PDE-ODE system

## Example: What did we learn?

- 1 Representation of the PDE as an abstract system was “easy”
- 2 Choices of  $L$ ,  $K_1$  and  $K_2$  could be completed by solving stabilization problems for a PDE and a PDE-ODE cascade
- 3 The abstract controller that we constructed could be reinterpreted as a PDE-ODE system

Why was this a “simple” example?

- The control and observation were **inside the spatial domain**

## A More Challenging Case

Consider the  $2 \times 2$  hyperbolic system

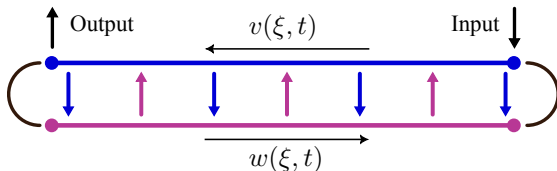
$$\partial_t v(\xi, t) - \mu_1 \partial_\xi v(\xi, t) = a_1(\xi) w(\xi, t)$$

$$\partial_t w(\xi, t) + \mu_2 \partial_\xi w(\xi, t) = a_2(\xi) v(\xi, t)$$

$$v(0, t) = q_0 w(0, t), \quad v(1, t) = q_1 w(1, t) + u(t)$$

$$y(t) = v(0, t)$$

with  $\mu_1, \mu_2 > 0$ ,  $q_0, q_1 \in \mathbb{R}$ .



- The model has **boundary control** and **boundary observation**

## Representation as an Abstract System

The hyperbolic system

$$\begin{aligned}\partial_t v - \mu_1 \partial_\xi v &= a_1 w, & v(0, t) &= q_0 w(0, t) + w_{dist}(t), \\ \partial_t w + \mu_2 \partial_\xi w &= a_2 v, & v(1, t) &= q_1 w(1, t) + u(t), \\ y(t) &= v(0, t)\end{aligned}$$

Choose  $x(t) = (v(\cdot, t), w(\cdot, t))^T \rightsquigarrow$  **A regular linear system**

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) + B_d w_{dist}(t), & x(0) &= x_0 \in X \\ y(t) &= C_\Lambda x(t)\end{aligned}$$

on  $X = L^2(0, 1) \times L^2(0, 1)$  with  $u(t), y(t) \in \mathbb{C}$ .

### Question

**“What’s so difficult?”**

## Representation as an Abstract System

$$\begin{aligned}\partial_t v - \mu_1 \partial_\xi v &= a_1 w, & v(0, t) &= q_0 w(0, t), \\ \partial_t w + \mu_2 \partial_\xi w &= a_2 v, & w(1, t) &= q_1 v(1, t) + u(t), \\ y(t) &= v(0, t)\end{aligned}$$

**regular linear system** on  $X = L^2(0, 1)^2$

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) & x(0) &= x_0 \in X \\ y(t) &= C_\Lambda x(t)\end{aligned}$$

Question (“What’s so difficult?”)

- (1) **Boundary input and output**  $\leadsto$  the abstract class is “**large**”
- (2) Stabilization requires **boundary feedback**.

Same applies to undamped wave, plate and beam equations.

# Overview

## Question (“What’s so difficult?”)

- (1) **Boundary input and output**  $\leadsto$  the abstract class is “**large**”
- (2) Stabilization requires **boundary feedback**.

**The good news:** An IMP for regular linear systems exists [P. '16]

**The bad news:** The existing controller constructions do not manage boundary feedbacks well.



# Overview

## Question (“What’s so difficult?”)

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**The good news:** An IMP for regular linear systems exists [P. '16]

**The bad news:** The existing controller constructions do not manage boundary feedbacks well.

## Result (P. 2023+)

*New abstract controller design techniques for regular linear systems allowing boundary feedback.*

# Controller Design for the Hyperbolic System

## Theorem (P. 2023+)

*The Robust Output Regulation Problem can be solved with the internal model controller **which formally resembles***

$$\dot{z}_1(t) = G_1 z_1(t) + G_2 (y(t) - y_{\text{ref}}(t))$$

$$\dot{\hat{x}}(t) = A\hat{x}(t) + Bu(t) + L(\hat{y}(t) - y(t) + y_{\text{ref}}(t))$$

$$\hat{y}(t) = C\hat{x}(t)$$

$$u(t) = K_1 z_1(t) + K_2 \hat{x}(t)$$

*with matrices  $G_1$ ,  $G_2$ ,  $K_1$  and **unbounded** operators  $L$  and  $K_2$ .*

- Matrices  $(G_1, G_2)$  contain the internal model
- $L$  contains output injections for the hyperbolic system
- $K_1$  and  $K_2$  stabilize a cascade of the system and the IM

## Choosing output injection $L$

$L$  can be constructed by choosing  $\ell_b \in \mathbb{R}$ ,  $\ell_1, \ell_2 : [0, 1] \rightarrow \mathbb{R}$  so that

$$\begin{aligned}\partial_t v - \mu_1 \partial_\xi v &= a_1 w + \ell_1(\xi) v(0, t), \\ \partial_t w + \mu_2 \partial_\xi w &= a_2 v + \ell_2(\xi) v(0, t), \\ v(0, t) &= q_0 w(0, t) + \ell_b v(0, t), \quad w(1, t) = q_1 v(1, t)\end{aligned}$$

is exponentially stable.

This problem can be solved with **backstepping**.

## Choosing Feedback Operators $K_1$ and $K_2$

Choose  $K_1 \in \mathbb{R}^{2 \times 2}$ ,  $k_b \in \mathbb{R}$ ,  $k_1, k_2 : [0, 1] \rightarrow \mathbb{R}$  that the cascade

$$\dot{z}_1(t) = G_1 z_1(t) + G_2 v(0, t)$$

$$\partial_t v - \mu_1 \partial_\xi v = a_1 w,$$

$$\partial_t w + \mu_2 \partial_\xi w = a_2 v, \quad v(0, t) = q_0 w(0, t)$$

$$\begin{aligned} w(1, t) = & q_1 v(1, t) + K_1 z_1(t) + k_b v(1, t) \\ & + \int_0^1 k_1(\xi) v(\xi, t) + k_2(\xi) w(\xi, t) d\xi \end{aligned}$$

is exponentially stable.

This problem can either be solved directly with **backstepping** cascade techniques, or reduced using **forwarding** to stabilization of the hyperbolic system.

## The PDE Controller

Finally, rewrite the abstract controller as a PDE-ODE system:

$$\begin{aligned}\dot{z}_1(t) &= \begin{bmatrix} 0 & \omega \\ -\omega & 0 \end{bmatrix} z_1(t) + \begin{bmatrix} 1 \\ 1 \end{bmatrix} (v(0,t) - y_{ref}(t)) \\ \partial_t \hat{v} - \mu_1 \partial_\xi \hat{v} &= a_1 \hat{w} + \ell_1(\xi)(\hat{v}(0,t) - v(0,t) + y_{ref}(t)), \\ \partial_t \hat{w} + \mu_2 \partial_\xi \hat{w} &= a_2 \hat{v} + \ell_2(\xi)(\hat{v}(0,t) - v(0,t) + y_{ref}(t)), \\ \hat{v}(0,t) &= q_0 \hat{w}(0,t) + \ell_b(\hat{v}(0,t) - v(0,t) + y_{ref}(t)) \\ \hat{w}(1,t) &= q_1 \hat{v}(1,t) + K_1 z_1(t) + k_b \hat{v}(1,t) \\ &\quad + \int_0^1 k_1(\xi) \hat{v}(\xi,t) + k_2(\xi) \hat{w}(\xi,t) d\xi\end{aligned}$$

(Here we use the knowledge of  $X$ ,  $A$ ,  $B$ ,  $C$  etc.)

## Discussion on the Examples

- We focused on two 1D examples with differing properties
- Note that **boundary control** is not always difficult!

Easy cases:

- If the system is **already stable**, can use a very easy **finite-dimensional controller** by Rebarber–Weiss '03 (for a large class of systems)
- For parabolic PDEs, you can often design a **finite-dim. controller** using Galerkin approximations even in the unstable case [P.–Phan '20, '21, Huhtala–P.–Hu '22]
- In PDEs with boundary control, **actuator and sensor dynamics** make things easier\*!

# Historical Highlights Related to PDEs

Internal Model Principle (*characterization* of controllers) for PDEs:

- **Starting point:** IMP for linear finite-dimensional systems
  - Francis–Wonham '75, Davison '76:
- **Extension to PDEs** with distributed control and observation
  - Bhat '76: Geometric approach, PhD with Koivo and Wonham
  - Immonen '05–'07: Approach using Sylvester equations
  - P.–Pohjolainen '10: The “classical” form of the IMP
- **Extension to PDEs** with boundary control and observation
  - P. '14, '16: The class of Regular Linear Systems
  - Humaloja–Kurula–P. '19: Boundary Control Systems
- **Frequency domain extensions** of the IMP
  - Nett '84, Yamamoto–Hara '88, Vidyasagar '88, Laakkonen '13

## Developments: Internal Model based control design for PDEs

- **Classes of linear PDEs** with distributed/boundary control
  - Pohjolainen '83, Logemann–Townley '97, Hämäläinen–Pohjolainen '00, '06, '10, Rebarber–Weiss '03, Boulite et. al. '09, Harkort–Deutscher '11, '17, P. '16, '17
- **Parabolic PDEs**
  - Chentouf et. al. '08, '10, Deutscher '13, '15, '16, Guo-Meng '20, Huhtala–P.–Hu '22, ...
- **Hyperbolic PDEs**
  - Guo–Guo '13, '16, Guo–Krstic '17, '18, Humaloja–Kurula–P. '18, '19, Deutscher–Gabriel '18–21, Wang et. al. '18, '21, Guo-Meng '21, '22, ...
- **PIDEs, Coupled systems, Networks, ...**

In addition: Controllers for regulation without robustness

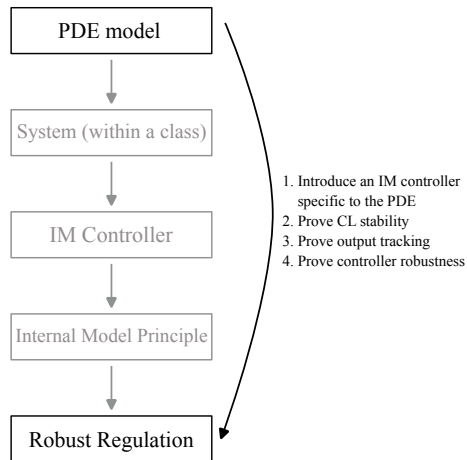
- Schumacher '83, Byrnes et. al. '00, Boulite–Saij et. al. '13, '18, Natarajan et. al. '14, Xu–Dubljevic '16, '17, ...



# Our “Abstract Approach” vs. “PDE-Based Approach”

In the “*PDE-based approach*”, an IM controller is designed for each PDE separately, followed by proofs for closed-loop stability, tracking and robustness.

In comparison, our “abstract approach” **avoids repetition** since regulation and robustness are guaranteed by the IMP, and do not need to be **proved separately** for each PDE!



## Conclusions

- An overview of the “abstract approach” to internal model control of linear PDEs.
- Examples of the control design process.
- An invitation to use the techniques in regulation of PDEs!
- Here the examples were chosen to be simple, but in principle, *if you can stabilize, you can do internal model control.*

## Things to Discuss

- How could the abstract results be made more user-friendly?
- How could we popularize internal model control?
- Interesting applications?
- Non-linear or semi-linear cases?