

# Beyond Process Control

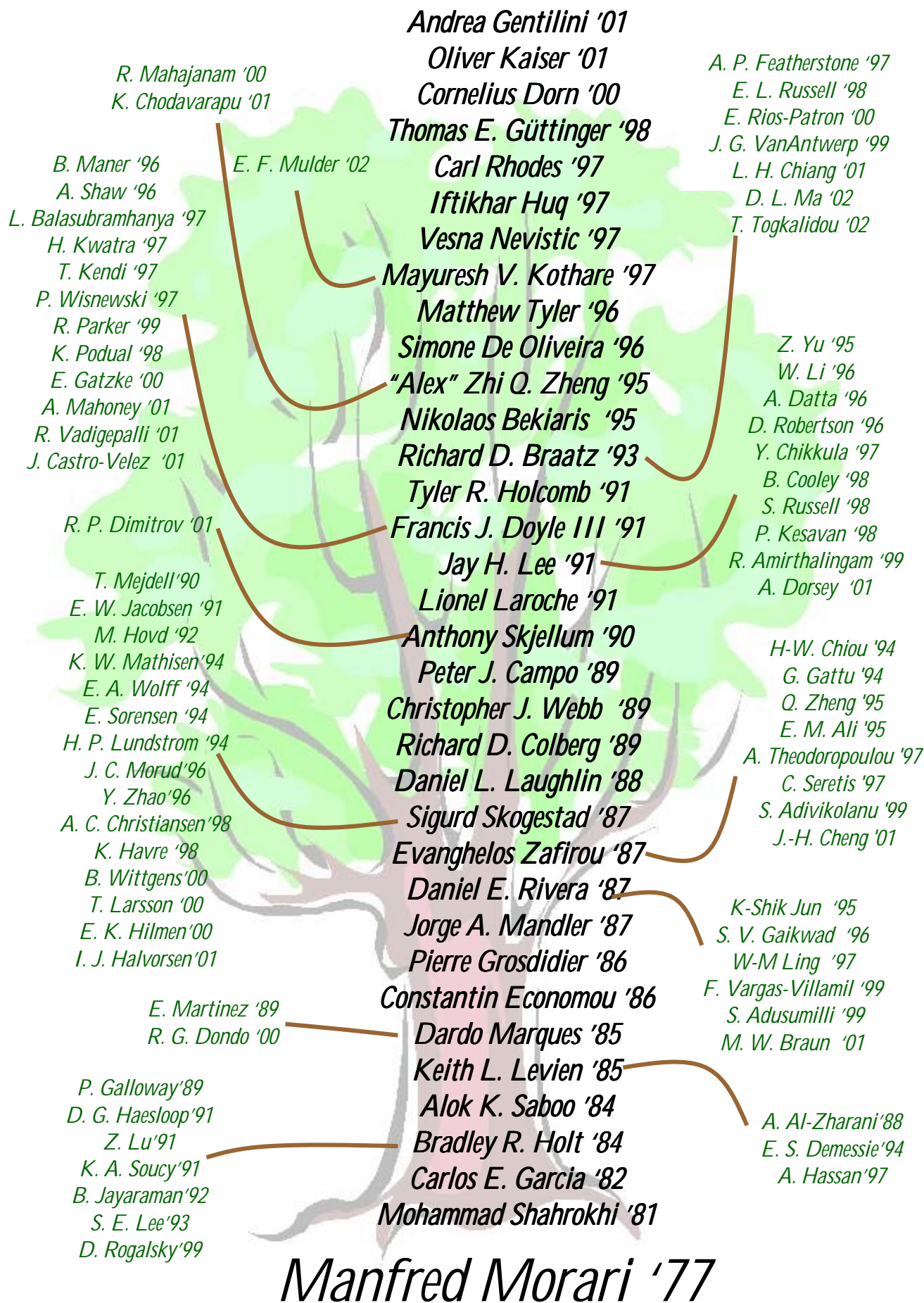
MANFRED MORARI



Automatic Control Laboratory, ETH Zürich

[WWW.CONTROL.ETHZ.CH](http://WWW.CONTROL.ETHZ.CH)





# Conclusions

- Process Control has been leading many important developments.
- Process Control tools can have significant impact in a wide range of other application areas.

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- Process Control tools can have significant impact in a wide range of other application areas.



## Nathaniel B. Nichols 1914-97

MS Physics, U. Mich..

Taylor Instruments (with Ziegler)

MIT with Draper & Brown (Nichols Chart)

Taylor Instruments

University of Minnesota

Raytheon

# Optimum temperature gradients in tubular reactors—I

## General theory and methods

OLEGH BILOUS\* and NEAL R. AMUNDSON

University of Minnesota, Minneapolis 14, Minnesota

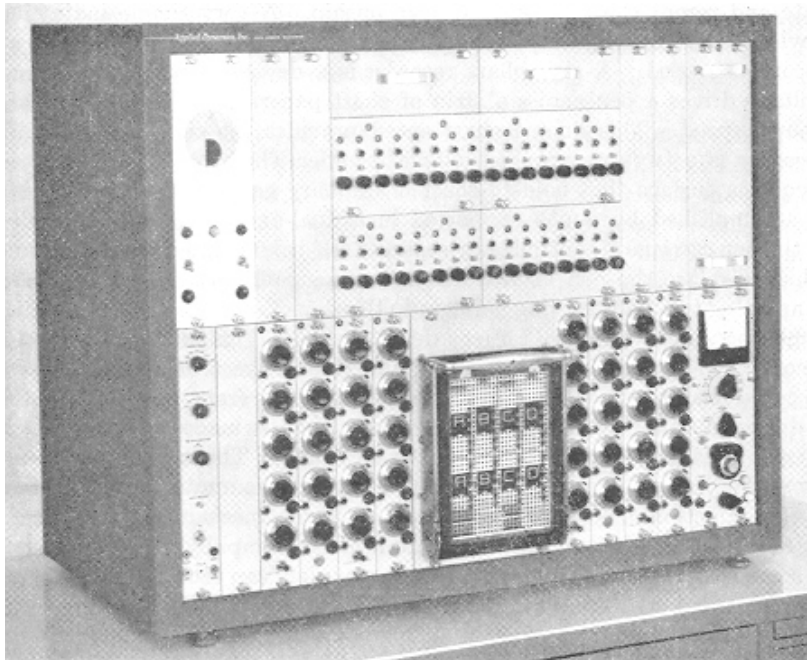
*(Received 28 May 1955)*

**Abstract**—In this paper the mathematical techniques necessary for the determination of the optimum temperatures profile in a tubular reactor to insure maximum yields or minimum contact times are developed, and applications are made to reversible and consecutive reaction systems. The problem is shown to be reducible to a system of ordinary non-linear differential equations. The solution of these differential equations can be made by conventional numerical methods, and will allow the specification of the temperatures in the reactor. In a succeeding paper numerical calculations obtained with an analogue computer (REAC) will be presented. The problem of two consecutive reactions  $A \rightarrow B \rightarrow C$ , in which the reactions are of first or second order, is discussed in detail. The method of attack on more complicated problems is sketched. It is shown in general that appreciable gains in the yield may be obtained if the optimum temperature distribution is used.

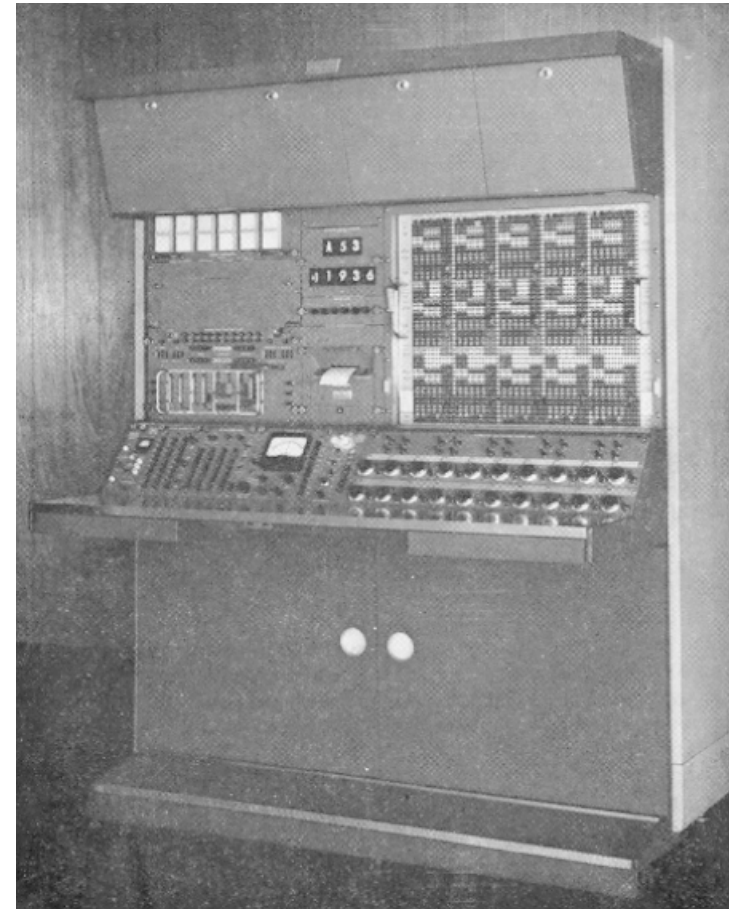


# Analog Computers ~1960

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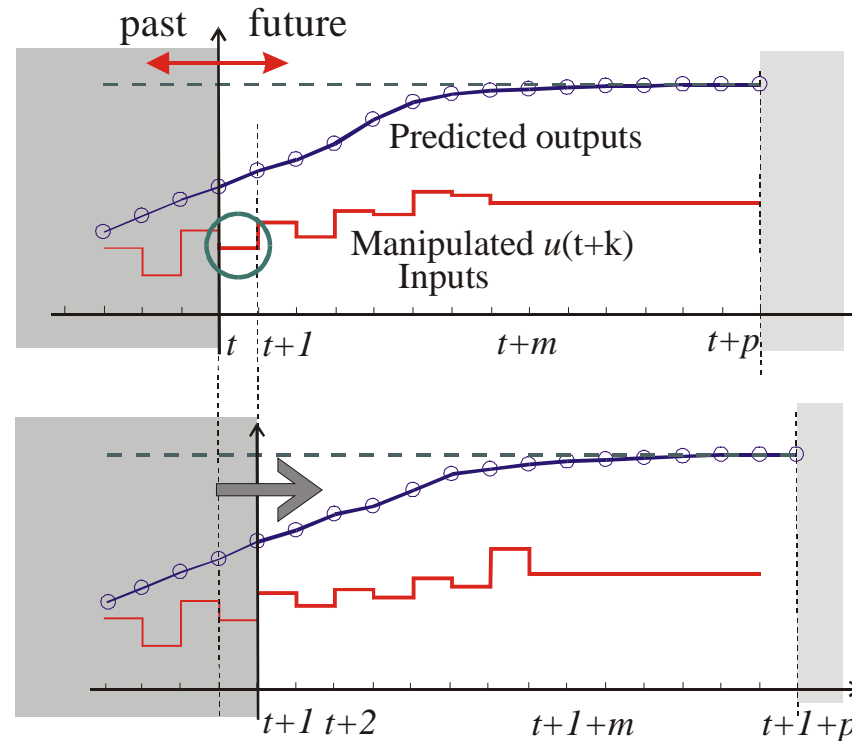


32-amplifier desk-top computer.  
(Courtesy of Applied Dynamics, Inc.)



Large-size (100-amplifier) computer.  
(Courtesy of Electronic Associates, Inc.)

# Model Predictive Control



- Optimize at time  $t$  (new measurements)
- Only apply the first optimal move  $u(t)$
- Repeat the whole optimization at time  $t + 1$
- Advantage of on-line optimization  $\Rightarrow$  **FEEDBACK**



# Model Predictive Control: A Singular Success Story

- Impact on industrial automation
- Impact on academic research

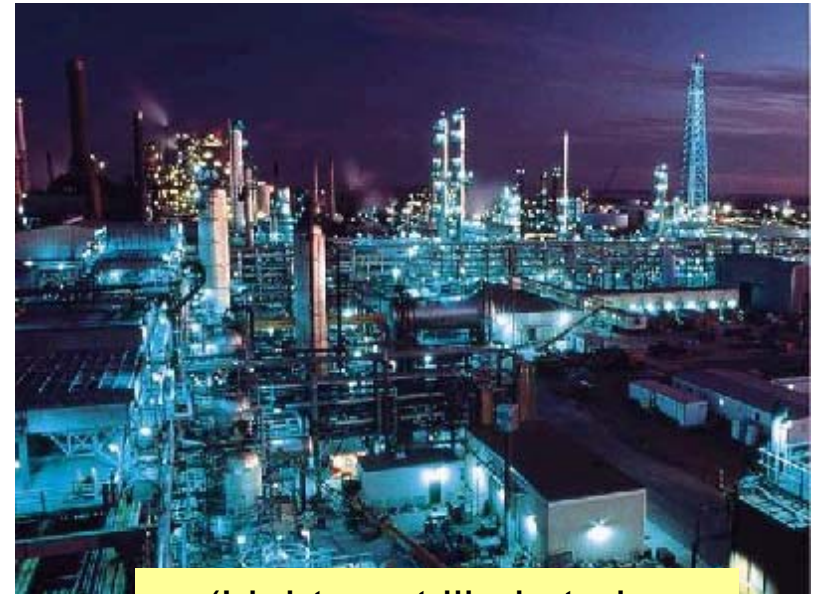
# MPC Vendor Applications by Areas

Area	Aspen Technology	Honeywell Hi-Spec	Adersa <sup>b</sup>	Invensys	SGS <sup>c</sup>	Total
Refining	1200	480	280	25		1985
Petrochemicals	450	80	—	20		550
Chemicals	100	20	3	21		144
Pulp and paper	18	50	—	—		68
Air & Gas	—	10	—	—		10
Utility	—	10	—	4		14
Mining/Metallurgy	8	6	7	16		37
Food Processing	—	—	41	10		51
Polymer	17	—	—	—		17
Furnaces	—	—	42	3		45
Aerospace/Defense	—	—	13	—		13
Automotive	—	—	7	—		7
Unclassified	40	40	1045	26	450	1601
Total	1833	696	1438	125	450	4542
First App.	DMC:1985 IDCOM-M:1987 OPC:1987	PCT:1984 RMPCT:1991	IDCOM:1973 HIECON:1986	1984	1985	
Largest App.	603 × 283	225 × 85	—	31 × 12	—	

Qin & Badgwell, Control Engineering Practice, 2003

# Increasing Autonomy in Industrial Processes

- An emphasis on reducing operators in process plants
- A telling metric: “loops per operator”
- United States refining industry data:
  - 1980: 93,000 operators, 5.3 bbl production
  - 1998: 60,000 operators, 6.2 bbl production(U.S. Bureau of the Census, 1999)



(Lights not likely to be turned off anytime soon)

ries



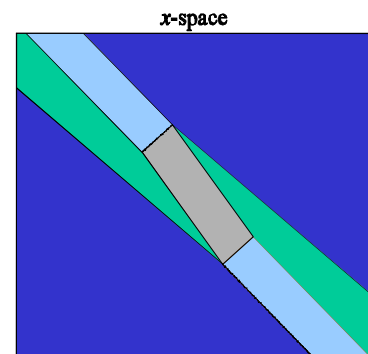
# Conclusions

- Process Control has been leading many important developments.
- Process Control tools can have significant impact in a wide range of other application areas.

# Developments extending the reach of MPC beyond PC

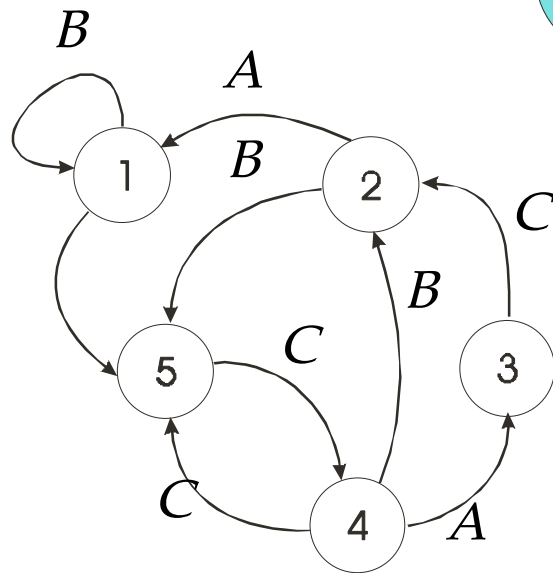
- Faster & cheaper computers
- Extension of MPC to switched/ hybrid systems
- On line optimization  $\Rightarrow$  look up table

**Different linear controller for each region of the state space**



# Hybrid Systems

$X = \{1, 2, 3, 4, 5\}$   
 $U = \{A, B, C\}$



Computer Science

Finite state machines

Control Theory

Continuous dynamical systems

Hybrid systems



$x \in \mathbb{R}^n$   
 $u \in \mathbb{R}^m$   
 $y \in \mathbb{R}^p$

$$\begin{cases} x(k+1) = f(x(k), u(k)) \\ y(k) = g(x(k), u(k)) \end{cases}$$

# Optimal Control for Constrained PWA Systems

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## System

- Discrete *PWA* Dynamics  $x(k+1) = f_{\text{PWA}}(x(k), u(k))$
  - Constraints on the state  $x(k) \in \mathcal{X}$
  - Constraints on the input  $u(k) \in \mathcal{U}$
- $$\left. \begin{array}{l} x(k) \in \mathcal{X} \\ u(k) \in \mathcal{U} \end{array} \right\} C^x x(k) + C^u u(k) \leq C^0$$

## Objectives

- **Stability** (feedback is stabilizing)
- **Feasibility** (feedback exists for all time)
- **Optimal Performance**



# Constrained Finite Time Optimal Control of PWA Systems

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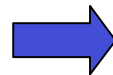
## Linear Performance Index ( $p=1,\infty$ )

$$J^*(x) := \min_U \|Px_T\|_p + \sum_{k=0}^{T-1} \|Qx_k\|_p + \|Ru_k\|_p$$

## Constraints

$$\begin{cases} x_0 = x, \\ x_{k+1} = f_{\text{PWA}}(x_k, u_k), \\ C^x x_k + C^u u_k \leq C^0 \end{cases}$$

Algebraic  
manipulation



Mixed Integer  
Linear Program (MILP)

$$U^*(x) = \{u_0^*, u_1^*, \dots, u_{T-1}^*\}$$

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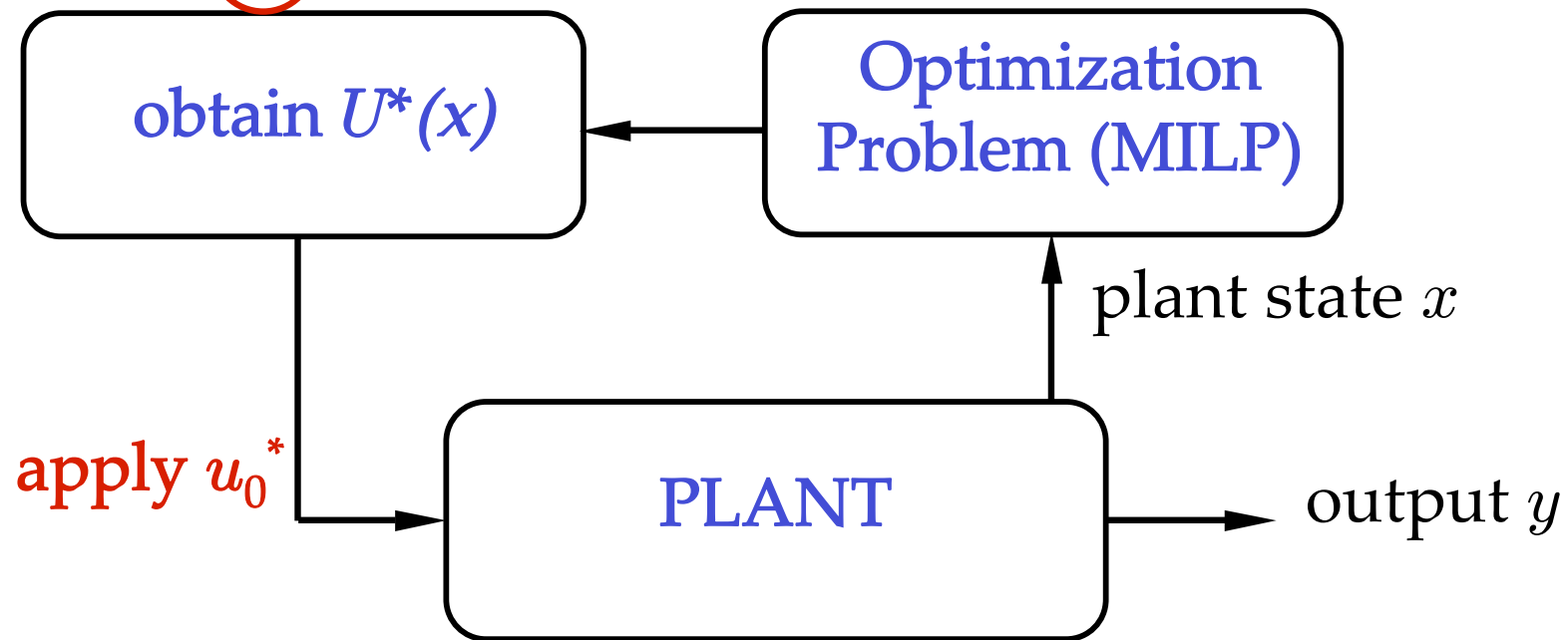
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*Receding Horizon Control*

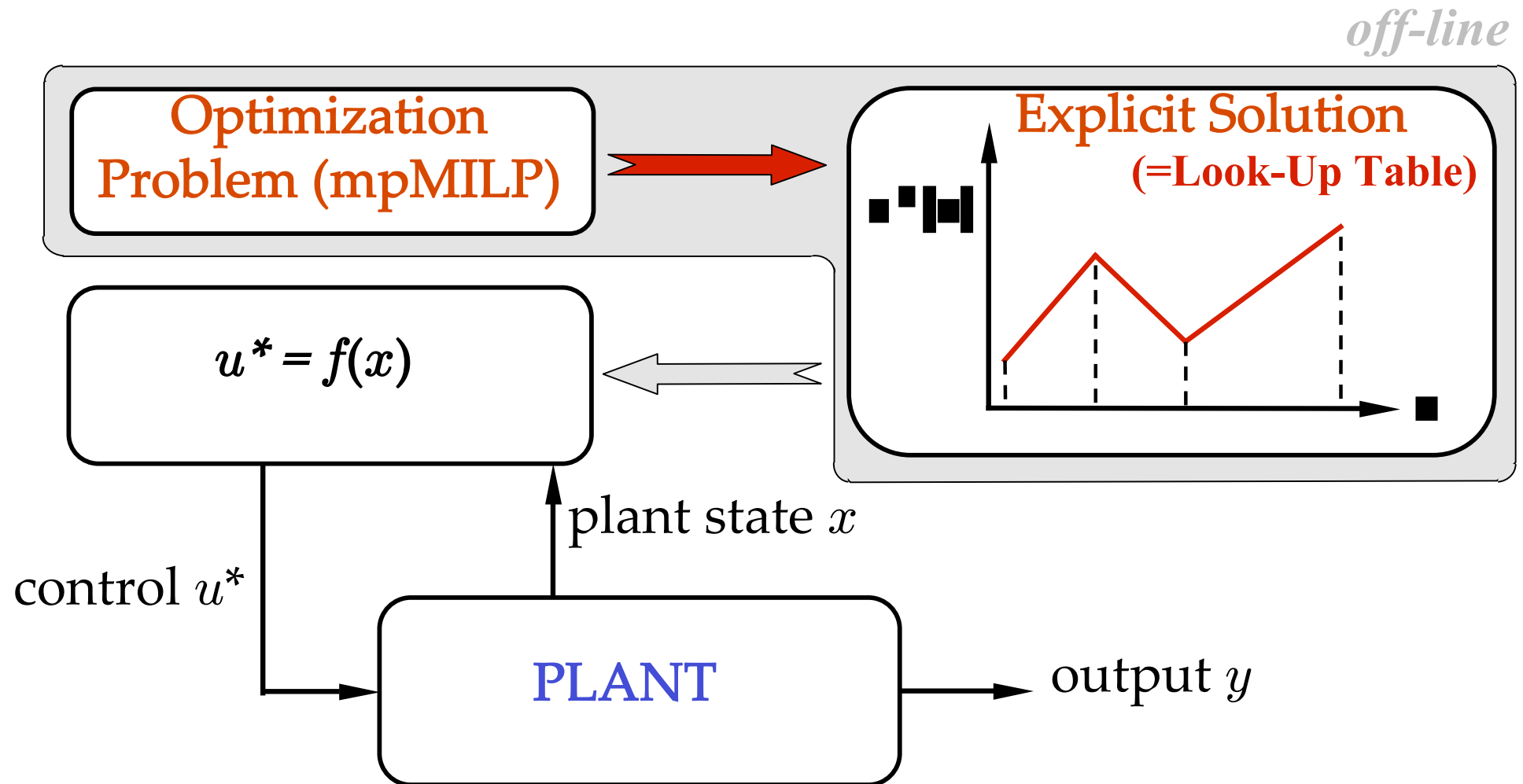
# Receding Horizon Control *On-Line* Optimization

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$$U^*(x) = \{u_0^*, u_1^*, \dots, u_{T-1}^*\}$$



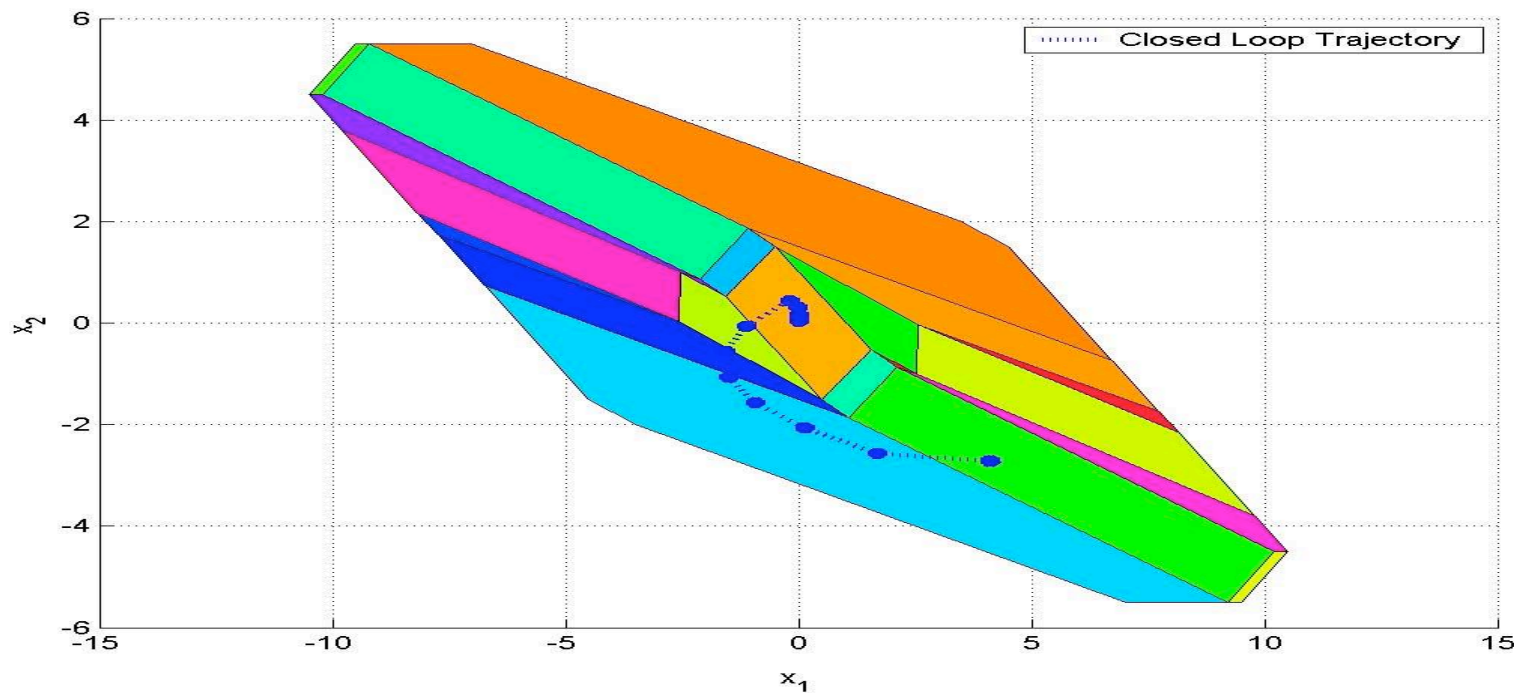
# Receding Horizon Policy *Off-Line* Optimization



# Why Compute an Explicit Solution?

## 1. Understand the Controller

- Powerful
- Nice tool
- Visualization: e.g. saturation of the controller



# Why Compute an Explicit Solution?

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## 2. Fast Implementation

Parallelization Possible  $u(x) = F_r x + G_r, \quad \text{if } H_r x \leq K_r$

*versus*

Interior-Point Methods  
⇒ Sequential  $J^*(x) := \min_U \|Px_T\|_p + \sum_{k=0}^{T-1} \|Qx_k\|_p + \|Ru_k\|_p$

$$\text{subj. to } \begin{cases} x_0 = x, \\ x_{k+1} = Ax_k + Bu_k, \\ C^x x_k + C^u u_k \leq C^0 \end{cases}$$

# Why Compute an Explicit Solution?

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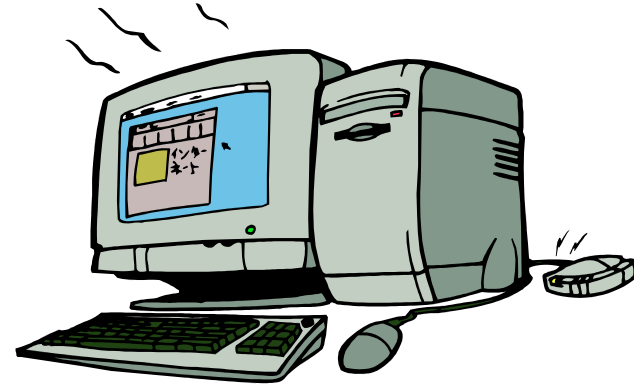
## 3. Cheap Implementation



~\$10

(Look-up-Table &  $\mu$ P)

*versus*



~\$10000

(PC & CPLEX)



# Multi-parametric controllers

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Algorithms have been developed for over 5 years:

...Minimization of linear and quadratic objectives

*(Baotic, Baric, Bemporad, Borrelli, De Dona, Dua, Goodwin, Grieder, Johansen, Mayne, Morari, Pistikopoulos, Rakovic, Seron, Toendel)*

...Minimum-Time controller computation

*(Baotic, Grieder, Kvasnica, Mayne, Morari, Schroeder)*

...Infinite horizon controller computation

*(Baotic, Borrelli, Christophersen, Grieder, Morari, Torrisi)*

...Computation of robust controllers

*(Borrelli, Bemporad, Kerrigan, Grieder, Maciejowski, Mayne, Morari, Parrilo, Sakizlis)*

**⇒ Computation schemes are mature !**

# Multi-parametric controllers

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## PROs:

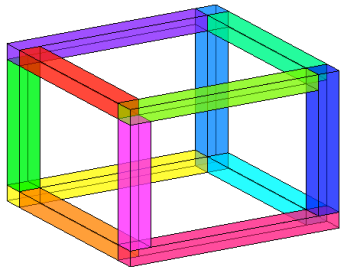
- Easy to implement
- Fast on-line evaluation
- Analysis of closed-loop system possible

## CONs:

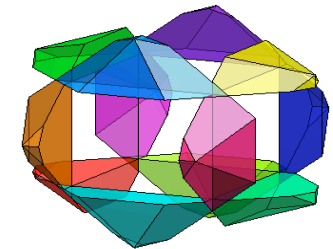
- Number of controller regions can be large
- Off-line computation time may be prohibitive
- Computation scales badly.

⇒ controller complexity is the crucial issue

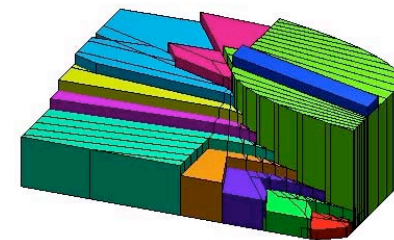
- All results and plots were obtained with the MPT toolbox



<http://control.ethz.ch/~mpt>



- MPT is a MATLAB toolbox that provides efficient code for
  - (Non)-Convex Polytope Manipulation
  - Multi-Parametric Programming
  - Control of PWA and LTI systems



# MULTI PARAMETRIC TOOLBOX



4000+ downloads



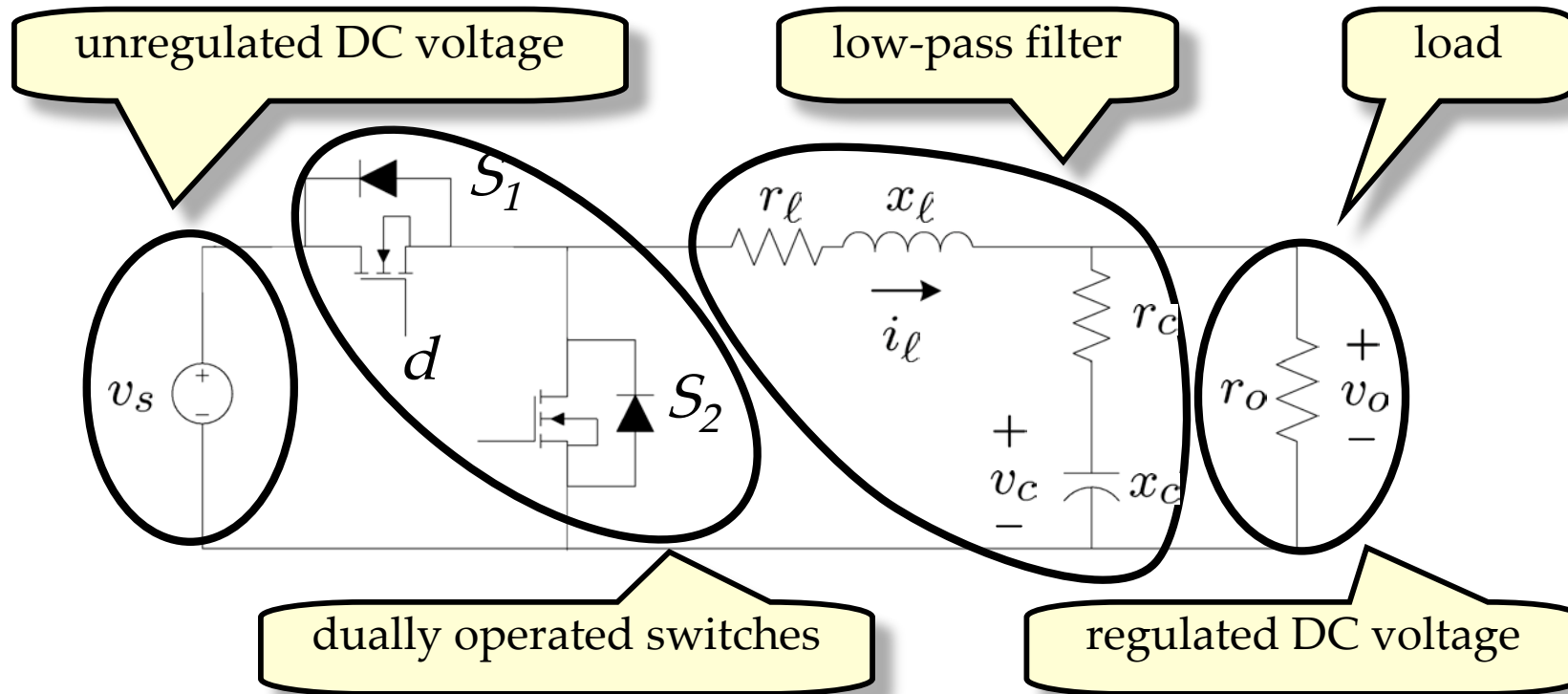
Rated 4.5 / 5 on mathworks.com



# Switch-mode DC-DC Converter

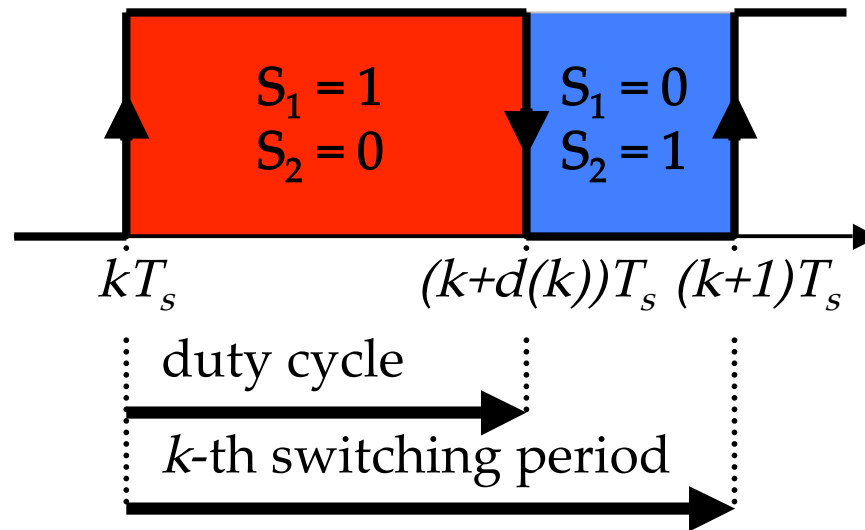
**Switched circuit:** supplies power to load with constant DC voltage

**Illustrating example:** synchronous step-down DC-DC converter



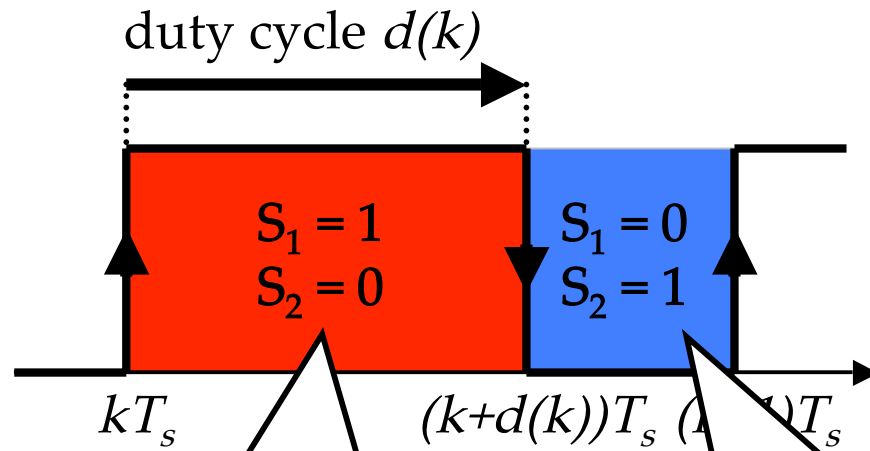
# Operation Principle

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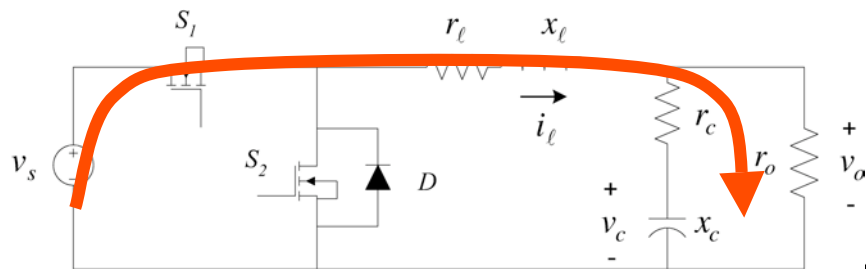


- Length of switching period  $T_s$  constant (*fixed switching frequency*)
- Switch-on transition at  $kT_s$ ,  $k \in \mathbb{N}$
- Switch-off transition at  $(k+d(k))T_s$  (*variable pulse width*)
- Duty cycle  $d(k)$  is real variable bounded by 0 and 1

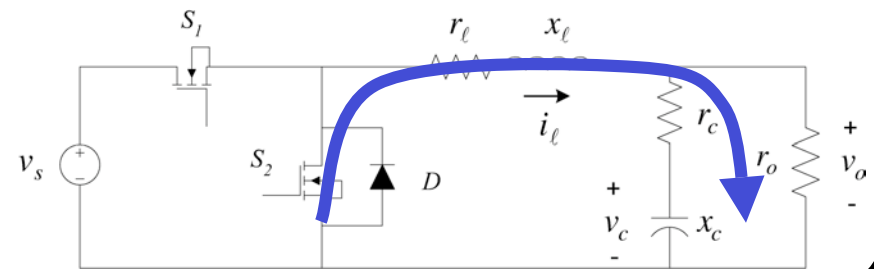
# Mode 1 and 2



mode 1:



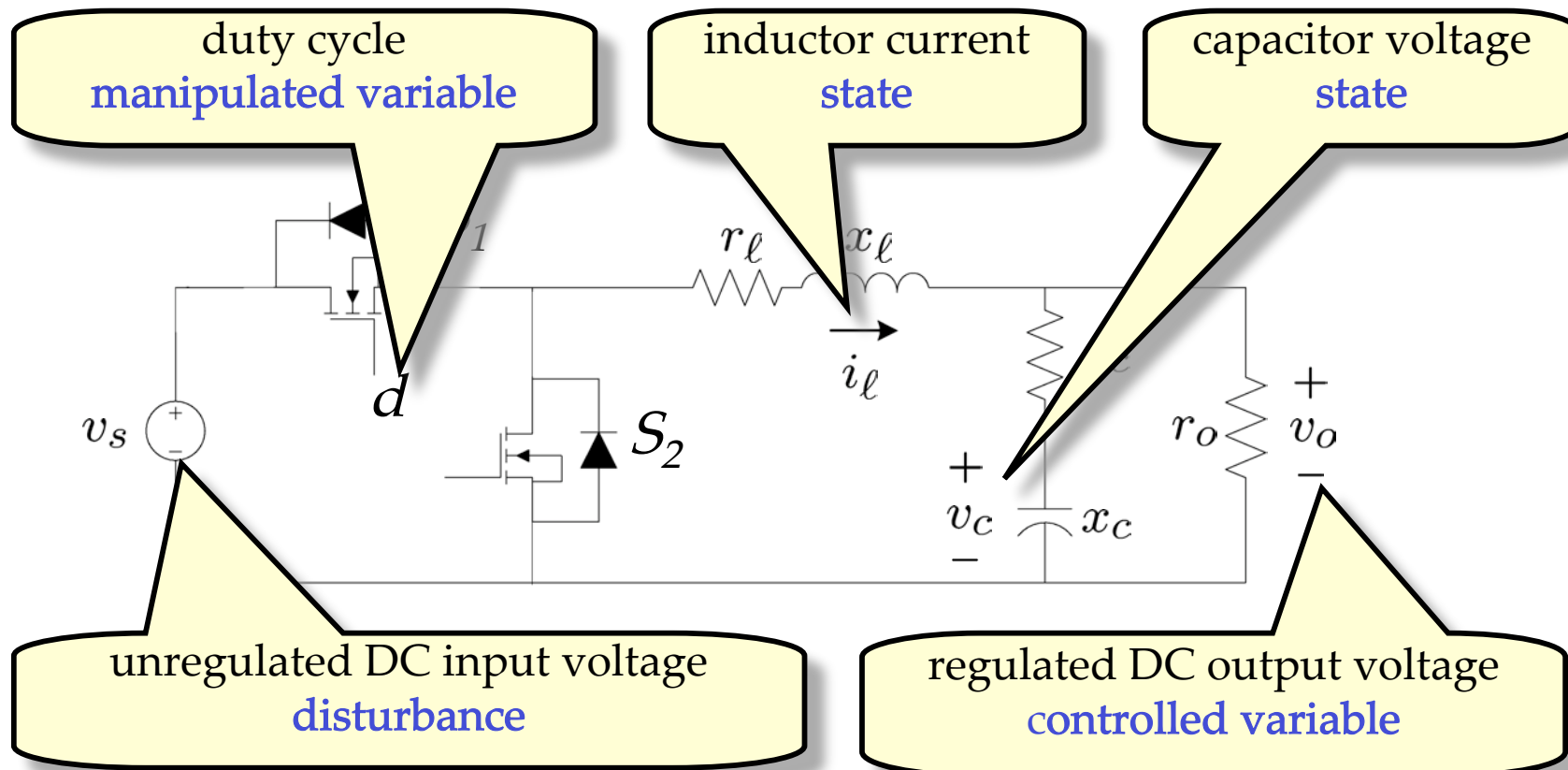
mode 2:





# Control Objective

Regulate DC output voltage by appropriate choice of duty cycle



# Control Objectives

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**Minimize** (average) output voltage error and changes in duty cycle

**Respect** constraint on current limit

Translate in Receding Horizon Control (RHC) problem

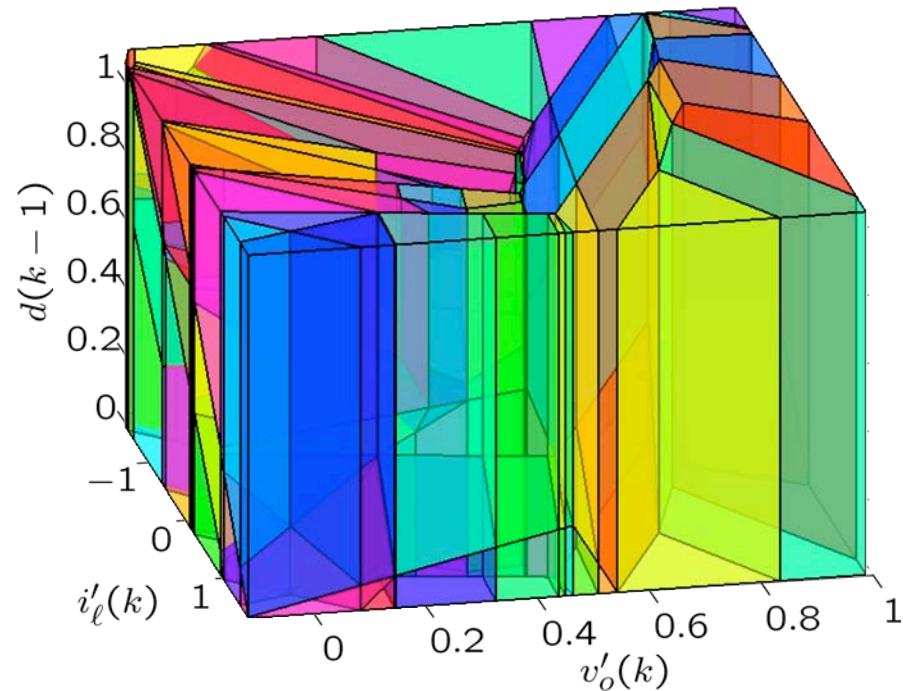
# State-feedback Controller: Polyhedral Partition

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PWA state-feedback  
control law:

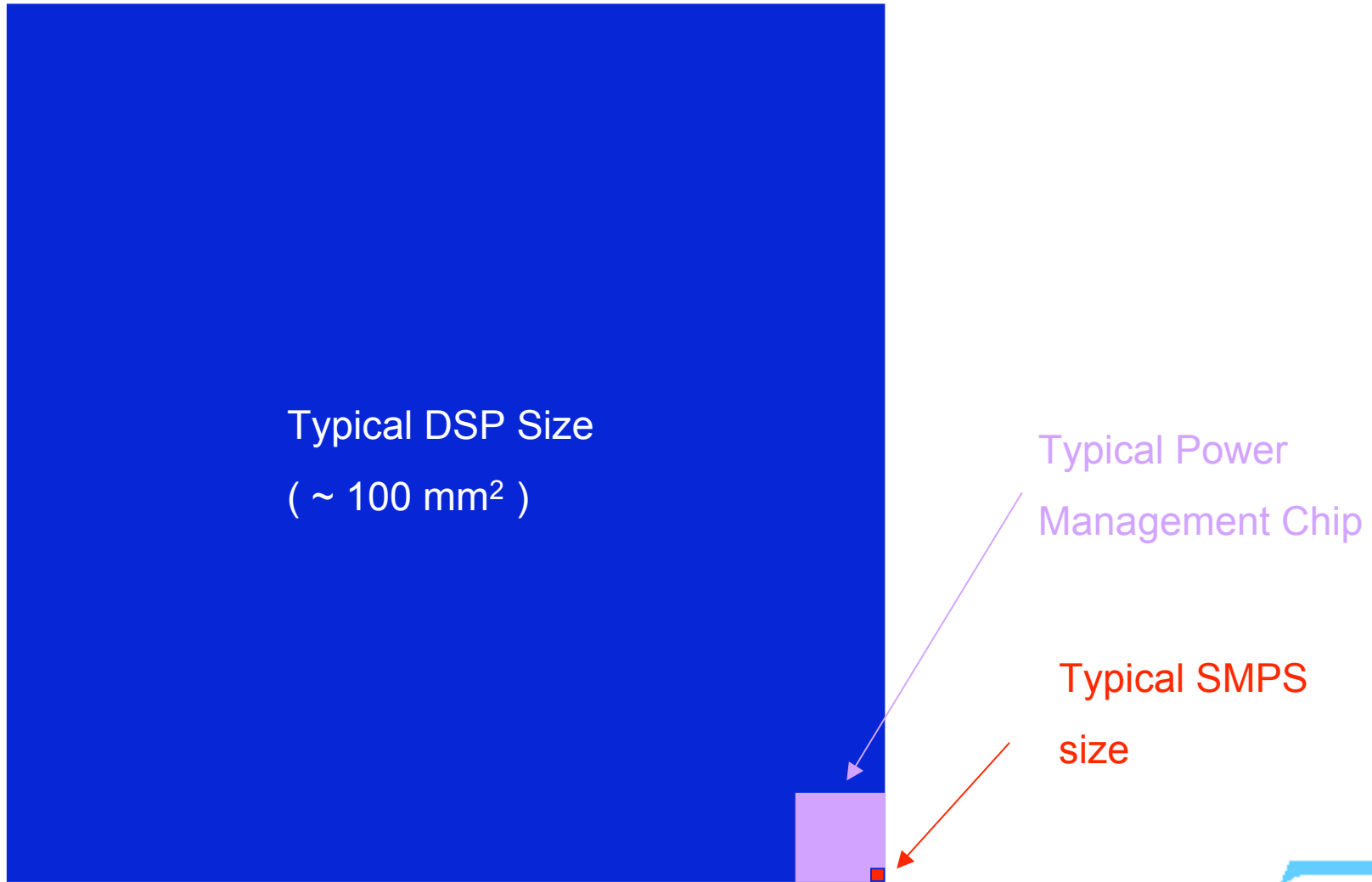
computed in 100s using the  
MPT toolbox

121 polyhedra after simplification  
with optimal merging



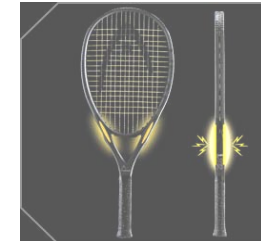
Colors correspond to the 121 polyhedra

# DSP vs. PM IC



# Smart Damping Materials

Niederberger, Moheimani



## • Demands

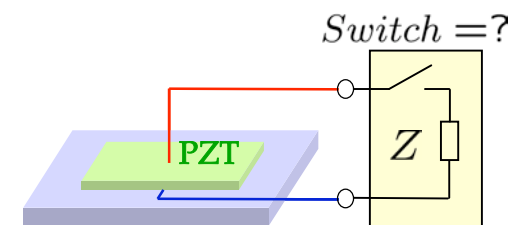
- Device *suppresses vibration*
- *External power* source for operation is not required
- *Weight* and *size* of the device have to be kept to a minimum

## • Idea

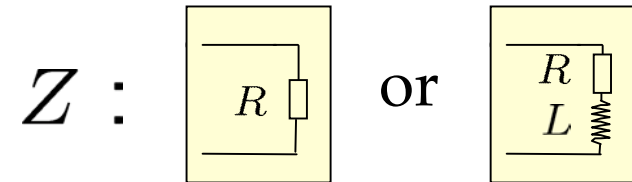
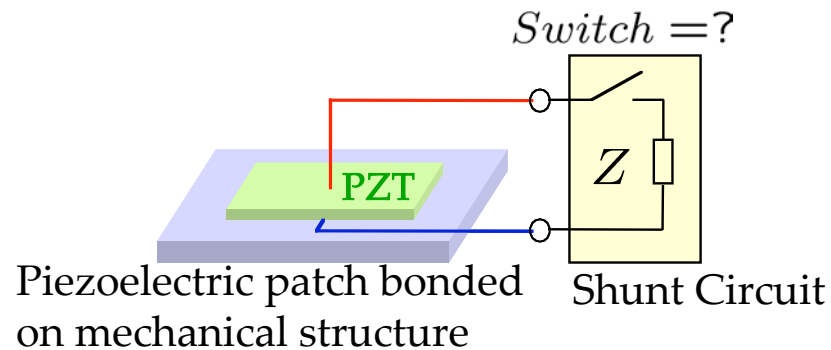
- *Switched Piezoelectric (PZT) Patches*

## • Problem

- *What is the optimal switching law for optimal vibration suppression?*



# Idea of Shunt Damping



- **How does it work?**

- Piezoelectric device converts mechanical energy into electrical energy.
- Shunt Circuit dissipates and stores electrical energy.
- Stored energy is supplied back to the mechanical system at the right time.

- **Problem**

- How to switch *optimally*?

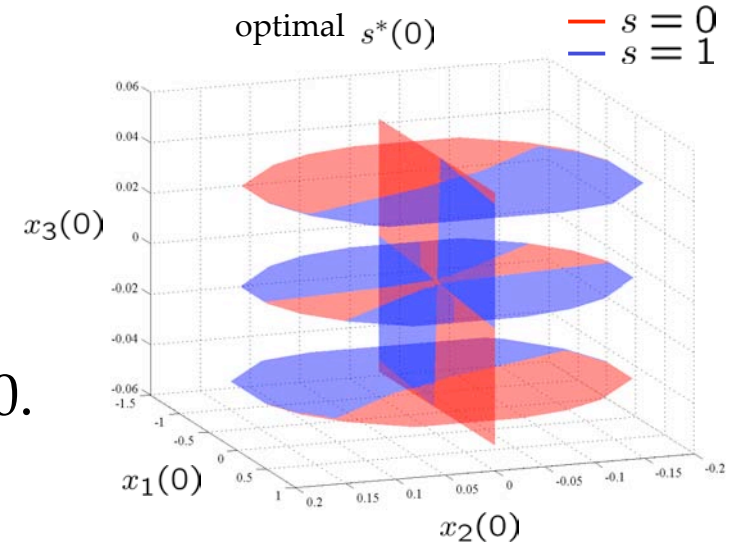
# Optimal Feedback using Multi-Parametric Programming

- **Optimal  $s^*(0)$  as a function of state  $x_0$** 
  - Multi-parametric programming

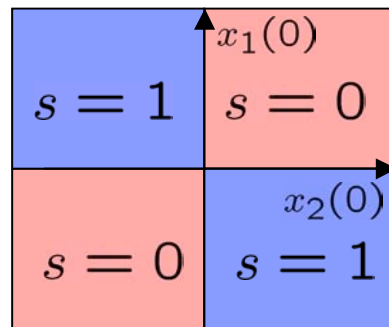
$$J_N^*(x(0)) = \min_{s_0, \dots, s_{N-1}} \sum_{k=0}^{N-1} \|v(k)\|_2^2$$

subj. to  $GS_N \leq W + Ex_0$

- State-space is partitioned into regions where  $s^*(0)$  is either 1 or 0.



- **After some simplifications**



$$x_2 = x$$

$$x_1 = -c \frac{dx}{dt}$$



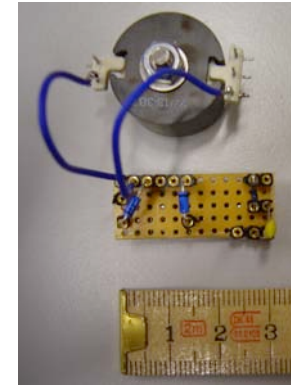
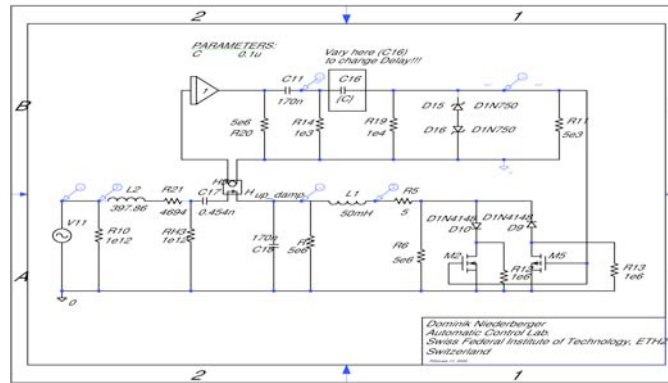
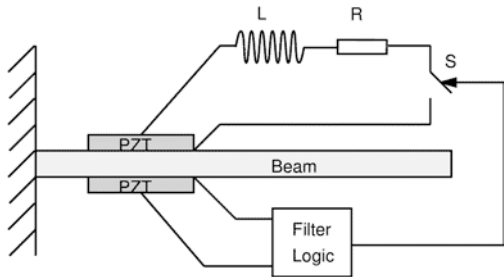
$$x \cdot \dot{x} \geq 0 \Rightarrow s = 1$$

$$\text{else } s = 0$$

Former Heuristic Controller  
*[Clark et al., J.Int.Mat.S.S. 2000]*

# Experimental Results

- Implementation as Autonomous Circuit
  - Switching circuit without external power source

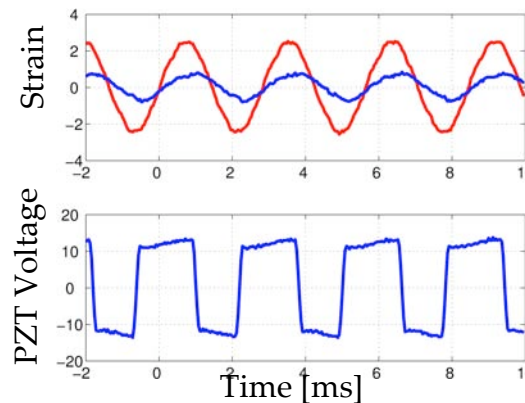


- Experiment with a Beam



(One-side clamped beam)

## Results



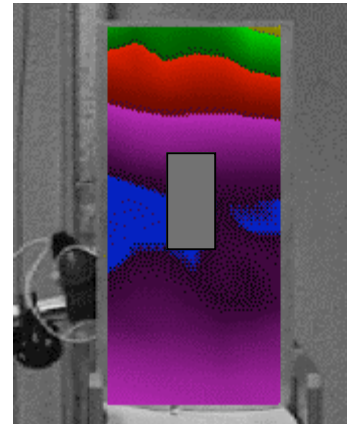
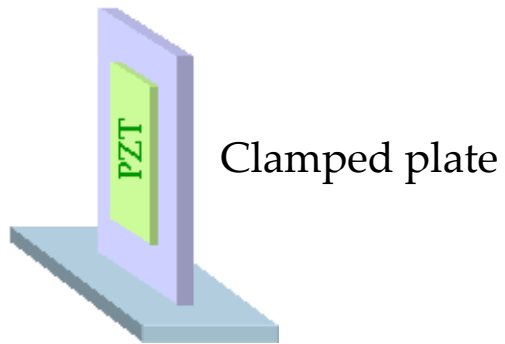
– 60% vibration suppression

- uncontrolled
- Switching Circuit (Autonomous)

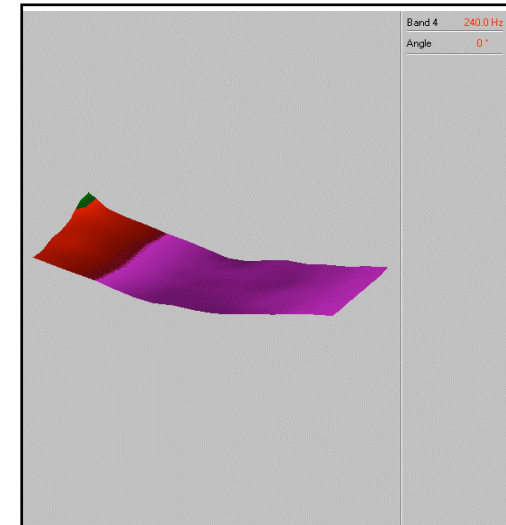


# Experiments with a Plate

- No Shunt Damping  
(Open system)

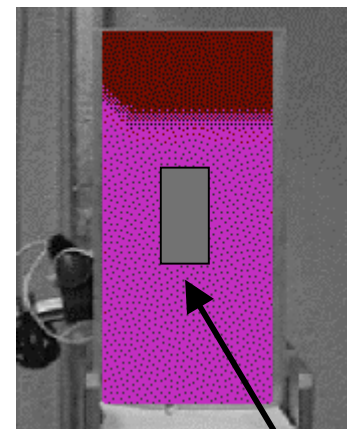
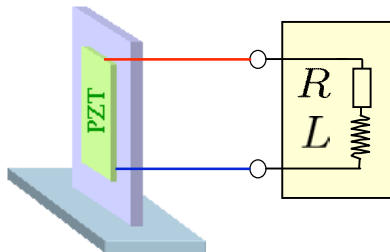


Measured Velocity

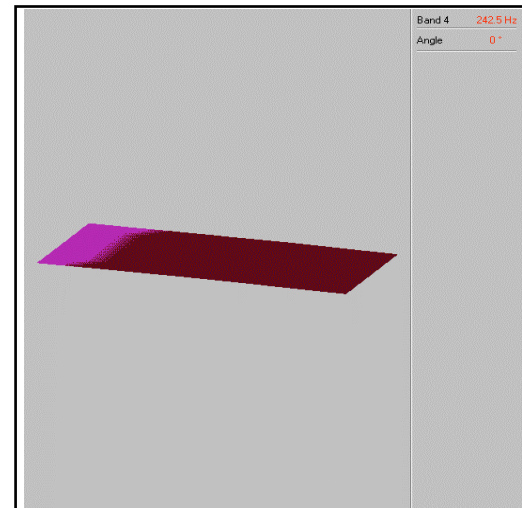


Measured Mode Shape

- With Shunt Circuit



Piezoelectric Patch



# Application: Brake Squeal Reduction

- Friction induced vibration in brakes
  - Strong vibration radiates unwanted noise
  - One frequency, small bandwidth
  - Frequency can vary



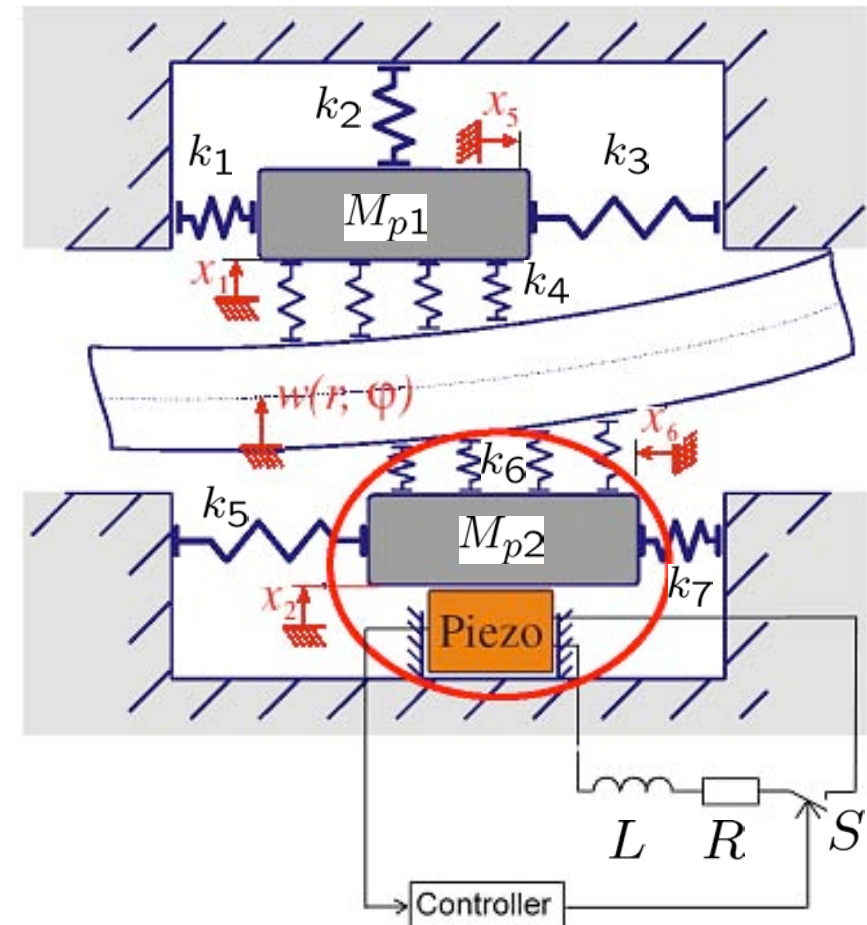
# Brake Squeal Reduction using Shunt Control

- **Vibration reduction**

- Piezoelectric actuator between brake pad and calliper
- Switching shunt control

- **Advantages**

- Tracks resonance frequency
- Cheap solution
- No electrical power required

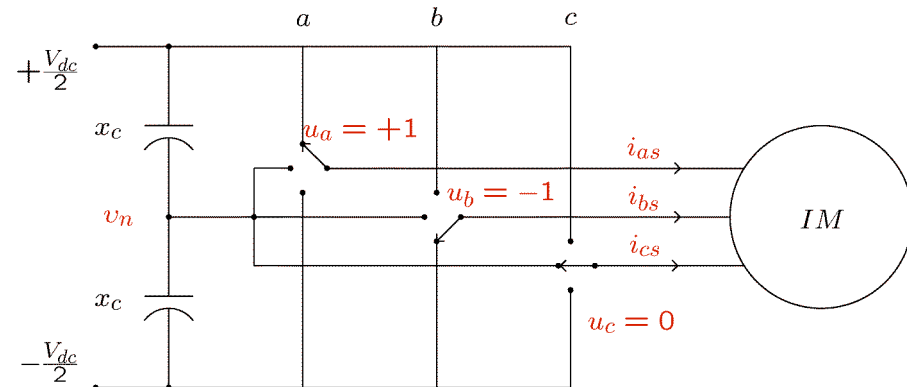


# Direct Torque Control



## Physical Setup:

- Three-level DC link inverter driving a three-phase symmetric induction motor
- Binary control inputs



## Control Objectives:

- Keep torque, stator flux and neutral point potential within given bounds
- Minimize average switching frequency (losses)

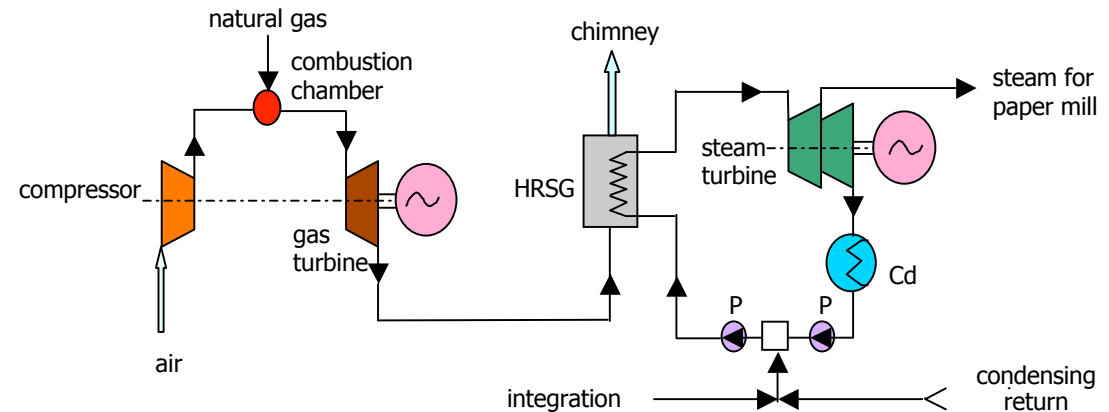
Reduction of switching frequency by up to 45 % (in average 25 %) with respect to ABB's commercial DTC scheme (ACS 6000)

# Control of Cogeneration Power Plants



## Physical Setup:

- Gas and steam turbines
- Different start-ups
- Logic implications
- Operating constraints



## Control Objective:

- Maximize profit (based on predicted profile of electricity price)

# Emergency Voltage Control in Power Systems

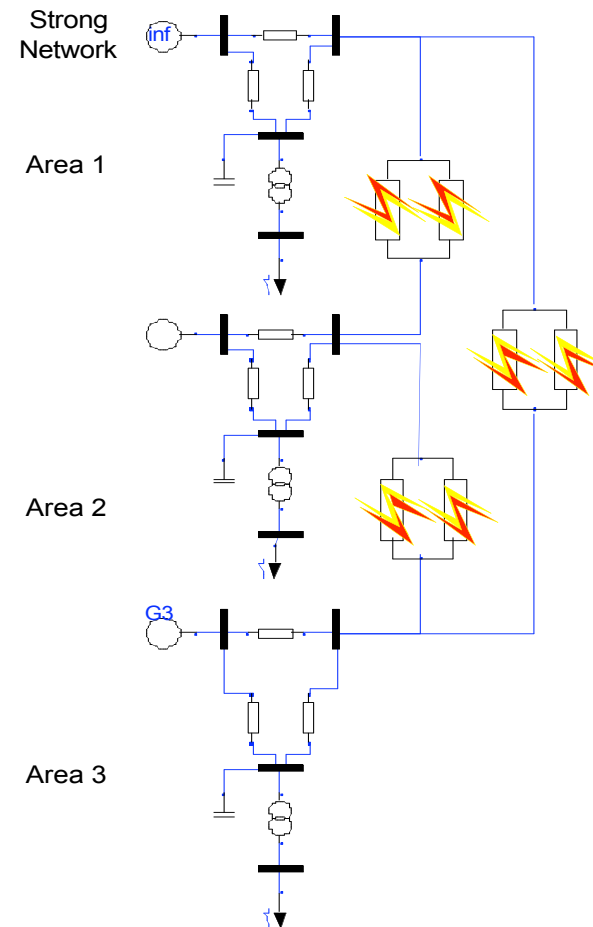


## Physical Setup:

- 3 area transmission system
- Integer control inputs
- Line outages trigger nonlinear network dynamics

## Control Objectives:

- Stabilize voltages
- Minimize disruptive control actions (load shedding)



Voltages effectively maintained within security limits



# Control of Anaesthesia

## Physical Setup:

- Patient undergoing surgery
- Analgesic infusion pump

## Control Objectives:

- Minimize stress reaction to surgical stimulation  
(by controlling mean arterial pressure)
- Minimize drug consumption



Excellent performance of administration scheme,  
mean arterial pressure variations kept within bounds

# Control of Thermal Print-Heads



## Physical Setup:

- Thermal print-head with ~ 1400 heat elements
- Binary control inputs
- Printing on a wide range of materials



## Control Objectives:

- Maximize printout quality
- Achieve robustness to parameter variations

90% quality gain over traditional controllers [ANSI X3.182-1990];  
Straight-forward design method for print-head controller

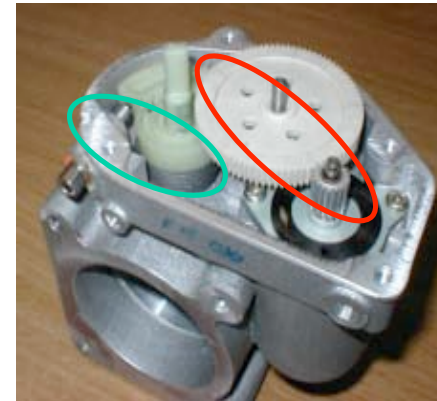


# Electronic Throttle Control



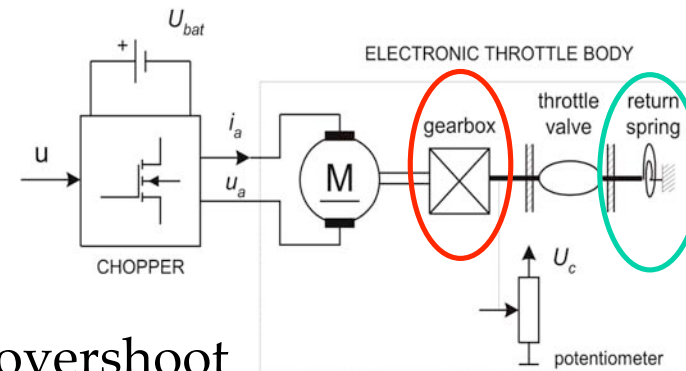
## Physical Setup:

- Valve (driven by DC motor) regulates air inflow to the car engine
- Friction nonlinearity
- Limp-Home nonlinearity
- Physical constraints



## Control Objectives:

- Minimize steady-state regulation error
- Achieve fast transient behavior without overshoot



Systematic controller synthesis procedure. On average twice as fast transient behavior compared to state-of-the-art PID controller with ad-hoc precompensation of nonlinearities.

# Traction Control



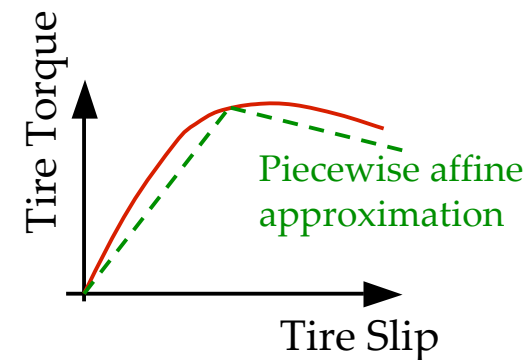
## Physical Setup:

- Improve driver's ability to control vehicle under adverse external conditions (wet or icy roads)
- Tire torque is nonlinear function of slip
- Uncertainties and constraints



## Control Objectives:

- Maximize tire torque by keeping tire slip close to the desired value

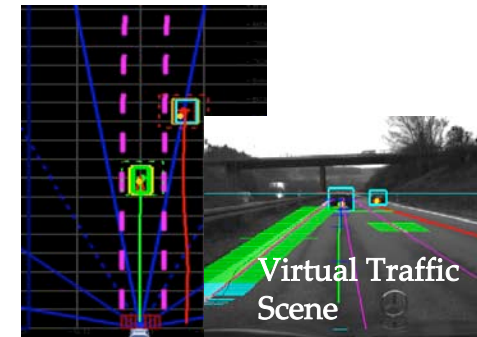
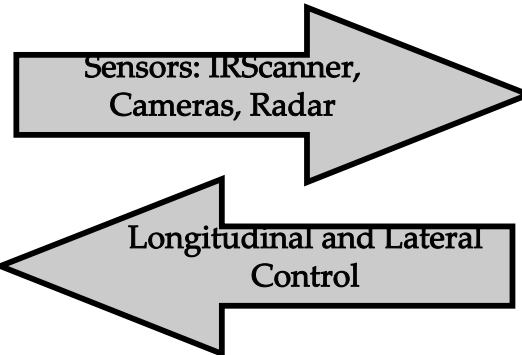


Experimental results: 2000 Ford Focus on a Polished Ice Surface;  
Receding Horizon controller with 20 ms sampling time

# Adaptive Cruise Control

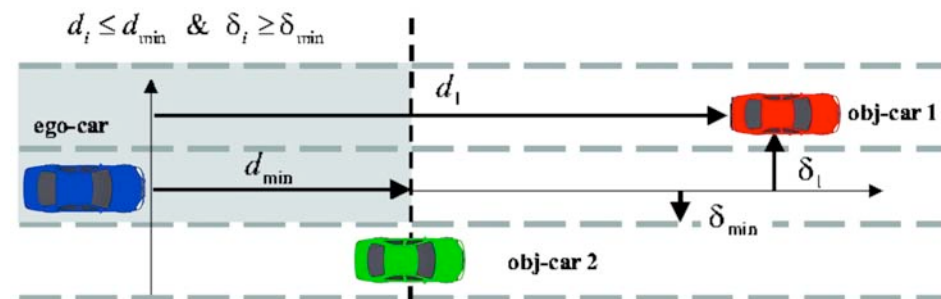
DAIMLERCHRYSLER

## Physical Setup:



## Control Objectives:

- Track reference speed
- Respect traffic rules
- Consider all objects on all lanes



Optimal state-feedback control law successfully implemented and tested on a research car Mercedes E430 with 80ms sampling time

# Conclusions

- Process Control has been leading many important developments.
- Process Control tools can have significant impact in a wide range of other application areas.