

# PREDICTIVE CONTROL OF NONLINEAR SYSTEMS WITH GUARANTEED STABILITY REGIONS

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# INTRODUCTION

- **Input constraints:**
  - ◇ Finite capacity of control actuators
  - ◇ Influence stabilizability of an initial condition
- **Desired characteristics of an effective control policy:**
  - ◇ Synthesis of stabilizing feedback laws
  - ◇ Explicit characterization of set of admissible initial conditions
- **Direct methods for control with constraints:**
  - ◇ Bounded control
    - ▷ Constraint handling via explicit characterization of stability region
  - ◇ Model predictive control
    - ▷ Constraint handling within open-loop optimal control setting
    - ▷ Successful applications in industry

# NONLINEAR SYSTEMS WITH INPUT CONSTRAINTS

- State–space description:

$$\begin{aligned}\dot{x}(t) &= f(x(t)) + g(x)u(t) \\ u(t) &\in \mathcal{U}\end{aligned}$$

- ◇  $x(t) \in \mathbb{R}^n$  : state vector
- ◇  $u(t) \in \mathcal{U} \subset \mathbb{R}^m$  : control input
- ◇  $\mathcal{U} \subset \mathbb{R}^m$ : compact & convex
- ◇  $u = 0 \in$  interior of  $\mathcal{U}$
- ◇  $(0, 0)$  an equilibrium point

- Stabilization of origin under constraints

# MODEL PREDICTIVE CONTROL

- Control problem formulation

- ◇ Finite-horizon optimal control:

$$P(x, t) : \min\{J(x, t, u(\cdot)) \mid u(\cdot) \in U_{\Delta}\}$$

- ◇ Performance index:

$$J(x, t, u(\cdot)) = F(x(t+T)) + \int_t^{t+T} [\|x^u(s; x, t)\|_Q^2 + \|u(s)\|_R^2] ds$$

- ▷  $\|\cdot\|_Q$  : weighted norm

- ▷  $T$  : horizon length

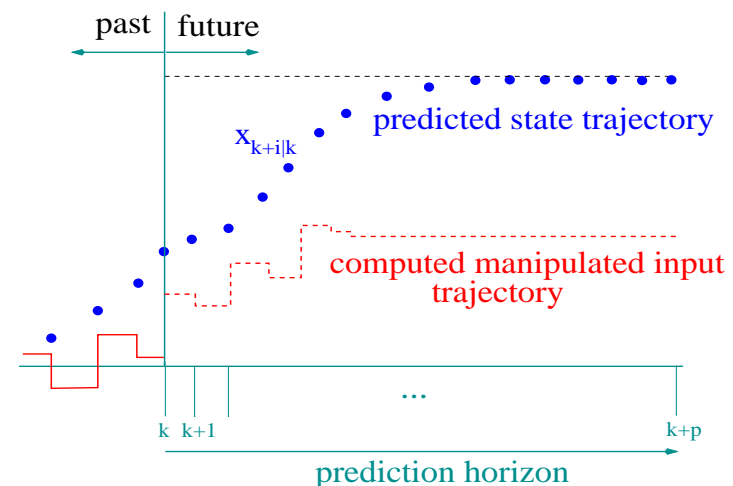
- ▷  $Q, R > 0$  : penalty weights

- ▷  $F(\cdot)$  : terminal penalty

- ◇ Implicit feedback law

$$M(x) = u^0(t; x, t)$$

“repeated on-line optimization”



# MODEL PREDICTIVE CONTROL

- Formulations for closed-loop stability:

(Mayne et al, Automatica, 2000)

- ◇ Adjusting horizon length, terminal penalty, weights, etc.

- ◇ Imposing stability constraints on optimization:

- ▷ Terminal equality constraints:  $x(t + T) = 0$

- Issues of practical implementation:

- ◇ Optimization problem non-convex

- ▷ Possibility of multiple, local optima

- ▷ Optimization problem hard to solve (e.g., algorithm failure)

- ▷ Difficult to obtain solution within “reasonable” time

- ◇ Lack of explicit characterization of stability region

- ▷ Extensive closed-loop simulations

- ▷ Restrict implementation to small neighborhoods

## BOUNDED LYAPUNOV-BASED CONTROL

- **Explicit bounded nonlinear control law:**

$$u = -k(x, u_{max})(L_G V)^T$$

- ◇ An example: (Lin & Sontag, 1991)

$$k(x, u_{max}) = \left( \frac{L_f V + \sqrt{(L_f V)^2 + (u_{max} \|(L_G V)^T\|)^4}}{\|(L_G V)^T\|^2 \left[ 1 + \sqrt{1 + (u_{max} \|(L_G V)^T\|)^2} \right]} \right)$$

- ◇ Nonlinear gain-shaping procedure:
  - ▷ **Accounts explicitly for constraints & closed-loop stability**

- **Constrained closed-loop properties:**

- ◇ Asymptotic stability
- ◇ Inverse optimality

# CHARACTERIZATION OF STABILITY PROPERTIES

$$D(u_{max}) = \{x \in \mathbb{R}^n : L_f V < u_{max} |(L_G V)^T|\}$$

- **Properties of inequality:**

- ◇ Describes open unbounded region where:

- ▷  $|u| \leq u_{max} \quad \forall x \in D$

- ▷  $\dot{V} < 0 \quad \forall 0 \neq x \in D$

- ◇ Captures constraint-dependence of stability region

- ◇  $D$  not necessarily invariant

- **Region of guaranteed closed-loop stability:**

$$\Omega(u_{max}) = \{x \in \mathbb{R}^n : V(x) \leq c_{max}\}$$

- ◇ Region of invariance:  $x(0) \in \Omega \implies x(t) \in \Omega \subset D \quad \forall t \geq 0$

- ◇ Larger estimates using a combination of several Lyapunov functions

- ◇ Other Lyapunov-based bounded control designs can be used

# UNITING BOUNDED CONTROL AND MPC

(El-Farra, Mhaskar & Christofides, Automatica, 2003; IJRNC, 2003)

- Objectives:

- ◇ Development of a framework for merging the two approaches:

- ▷ Reconcile tradeoffs in stability and optimality properties

- ▷ Explicit characterization of constrained stability region

- ▷ Possibility of improved performance

- ▷ Implement computationally inexpensive MPC formulations

- Central idea:

Decoupling “optimality” & “constrained stabilizability”

- ◇ Stability region provided by bounded controller

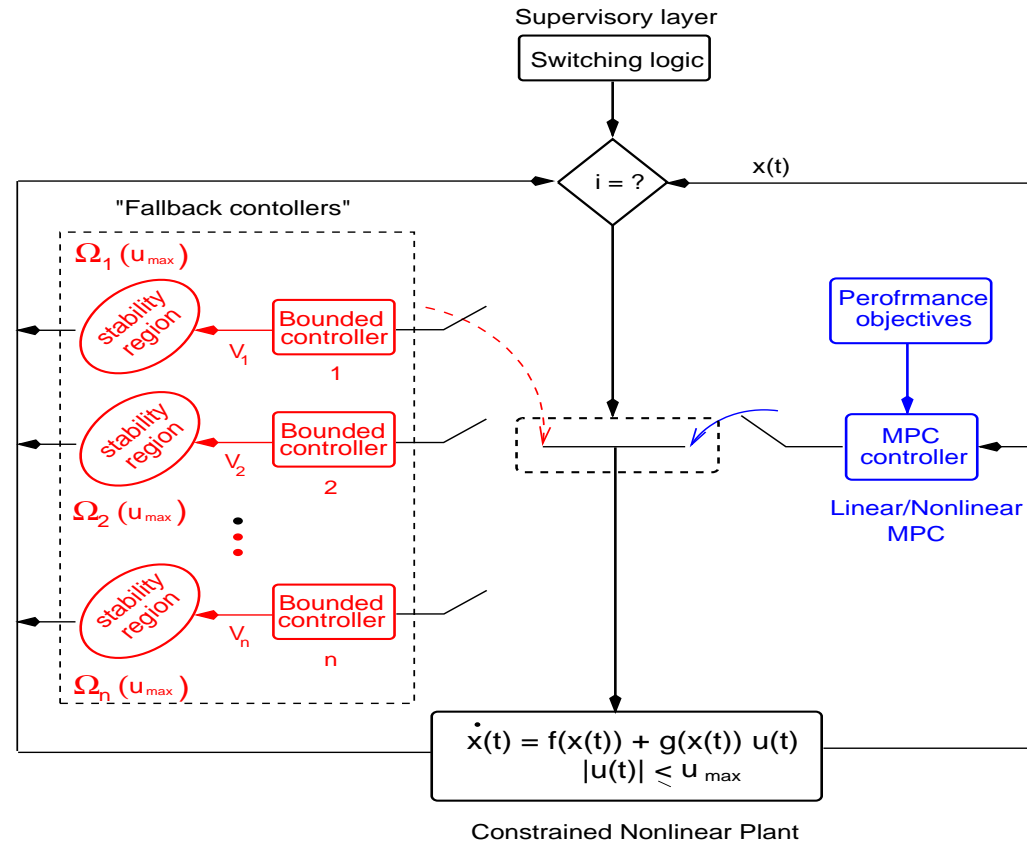
- ◇ Optimal performance supplied by MPC controller

- Approach:

- ◇ Switching between MPC & a family of bounded controllers



# OVERVIEW OF HYBRID CONTROL STRATEGY



- **Hierarchical control structure**

- ◇ Plant level
- ◇ Control level
- ◇ Supervisory level

- Overall structure **independent** of specific MPC algorithm used

- ◇ Could use linear/nonlinear MPC with or without stability constraints

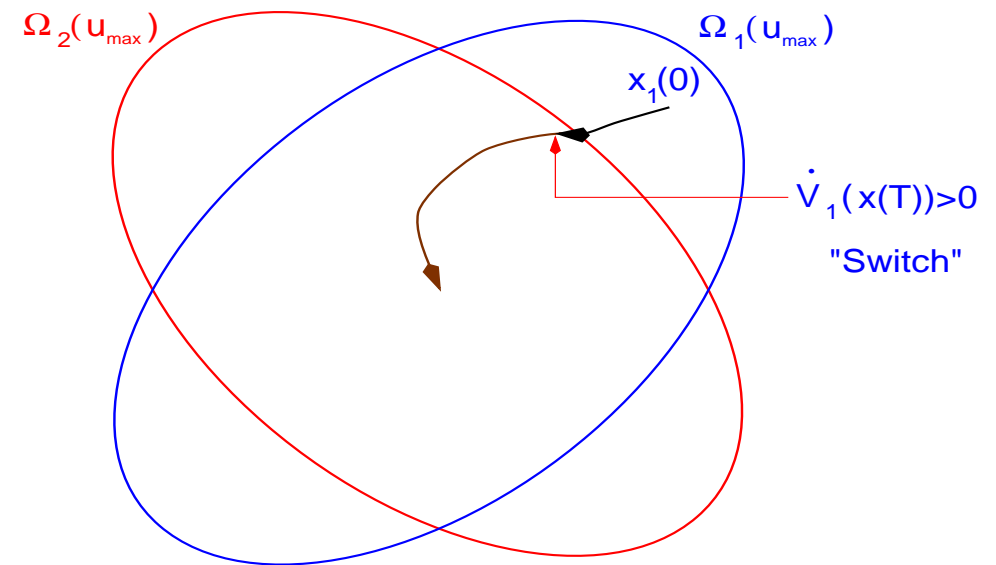
# STABILITY-BASED CONTROLLER SWITCHING

- Switching logic:

$$u_{\sigma}(x(t)) = \begin{cases} M(x(t)), & 0 \leq t < T^* \\ b(x(t)), & t \geq T^* \end{cases}$$

$$L_f V_k(x) + L_g V_k(x) M(x(T^*)) \geq 0$$

- ◇ Initially implement MPC,  $x(0) \in \Omega_k(u_{max})$
- ◇ Monitor temporal evolution of  $V_k(x^M(t))$
- ◇ Switch to bounded controller only if  $V_k(x^M(t))$  starts to increase



—▶ MPC  
—▶ Bounded control

# ENHANCING CLOSED-LOOP PERFORMANCE

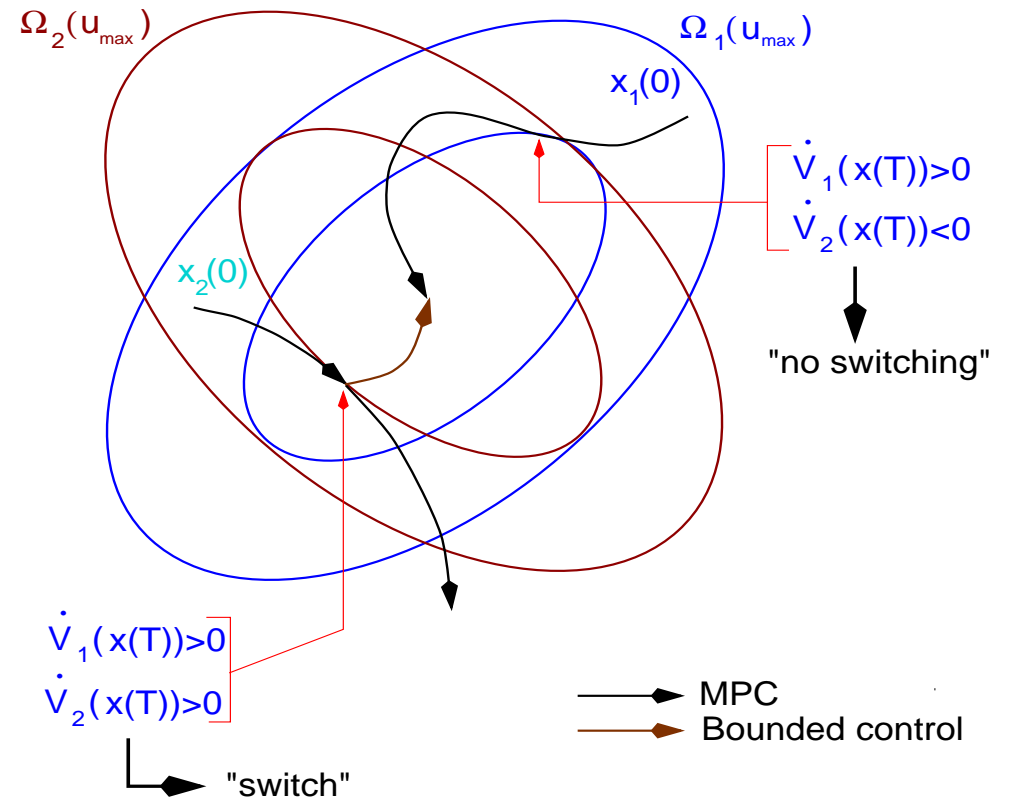
- **Switching policy:**

- ◇ Initialize the system in  $\Omega_k$
- ◇ Monitor  $V_k(x)$  for which  $x \in \Omega_k$
- ◇ Discard any  $V_k$  whose value ceases to decay

$$V_k(x(T_k)) \leq c_k^{max}$$

$$L_f V_k(x) + L_g V_k(x)M(x(t)) \geq 0$$

- ◇ Monitor all active  $V_j$ 's for which  $x \in \Omega_j$
- ◇ Continue MPC if  $\dot{V}_j < 0$  for some active  $V_j$ . Else switch to the appropriate bounded controller



## IMPLICATIONS OF SWITCHING SCHEME

- Switched closed-loop inherits bounded controller's stability region
  - ◇ A priori guarantees for all  $x(0) \in \Omega(u_{max})$
- Lyapunov stability condition checked & enforced by “supervisor”
  - ◇ Reduce computational complexity of optimization
  - ◇ Scheme does not require stability of MPC within  $\Omega(u_{max})$
  - ◇ Provides a safety net for implementing MPC
  - ◇ Stability independent of horizon length
- Conceptual differences from other schemes:
  - ◇ Switching does not occur locally
  - ◇ Provides stability region explicitly
  - ◇ No switching occurs if  $V(x^M(t))$  decays continuously
    - ▷ Only MPC is implemented  $\implies$  optimal performance recovered

# PREDICTIVE CONTROL IN INDUSTRIAL PRACTICE

- A “typical” predictive control design:

- ◇ Nonlinear process model:

$$\dot{x} = f(x) + g(x)u$$
$$u_{min}^i \leq u_i \leq u_{max}^i$$

- ◇ Linear representation:

$$\dot{x} = Ax + Bu$$
$$u_{min}^i \leq u_i \leq u_{max}^i$$

★ Linearization

(around desired steady-state)

★ Model identification

(e.g., through step tests)

- ◇ Use of computationally efficient linear MPC (QP) algorithms
  - ◇ No closed-loop stability guarantees for nonlinear system
- Practical value of the hybrid control structure:
    - ◇ Provides stability guarantees through fall-back controllers
    - ◇ Entails no modifications in existing predictive controller design

## APPLICATION TO A CHEMICAL REACTOR

- State–space description:

$$\begin{aligned}\dot{C}_A &= \frac{F}{V}(C_{A0} - C_A) - k_0 e^{\frac{-E}{RT_R}} C_A \\ \dot{T}_R &= \frac{F}{V}(T_{A0} - T_R) + \frac{(-\Delta H)}{\rho c_p} k_0 e^{\frac{-E}{RT_R}} C_A \\ &\quad + \frac{UA}{\rho V c_p} (T_c - T_R)\end{aligned}$$

- ◇ Multiple steady states
- ◇ Control objective: stabilization at the open–loop unstable equilibrium point,  $(C_{As}, T_s) = (.52, 398)$
- ◇ Manipulated input:  $u = T_c \in [250, 500]$

# APPLICATION TO A CHEMICAL REACTOR

- **Model predictive controller:**

- ◇ Performance index:

$$J = \int_t^{t+T} [\|x(\tau)\|_Q^2 + \|u(\tau)\|_R^2 + \|\dot{u}(\tau)\|_S^2] d\tau$$

- ◇  $Q = qI > 0, R = rI > 0, S = sI > 0$

- ◇ Prediction model:

$$\dot{x} = Ax + Bu$$

- ▷  $A, B$  obtained by linearizing the nonlinear model around  $(C_{As}, T_s)$

- ◇ Terminal equality constraint:  $x(t + T) = 0$

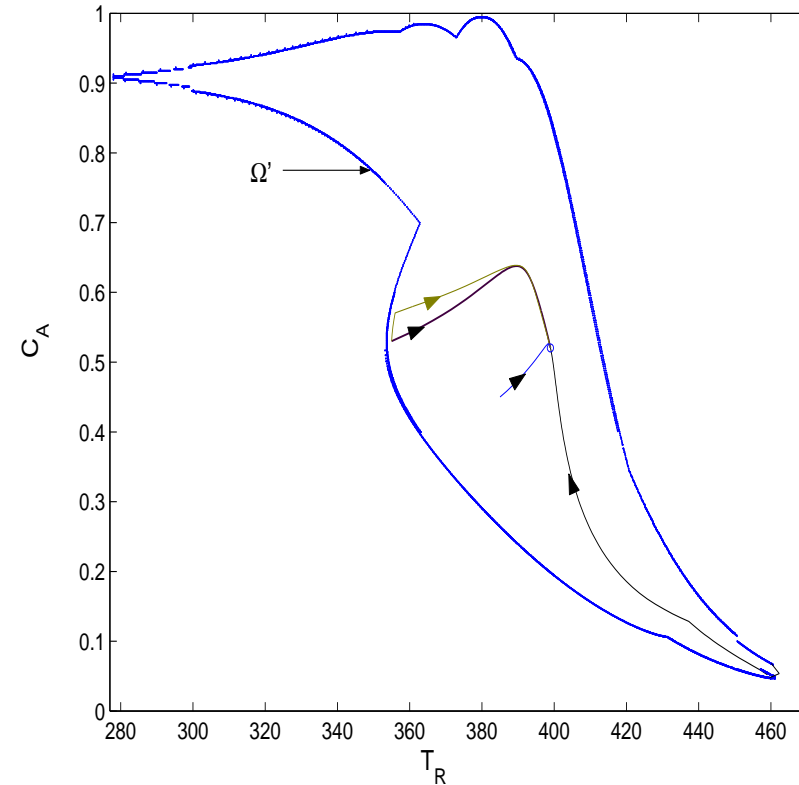
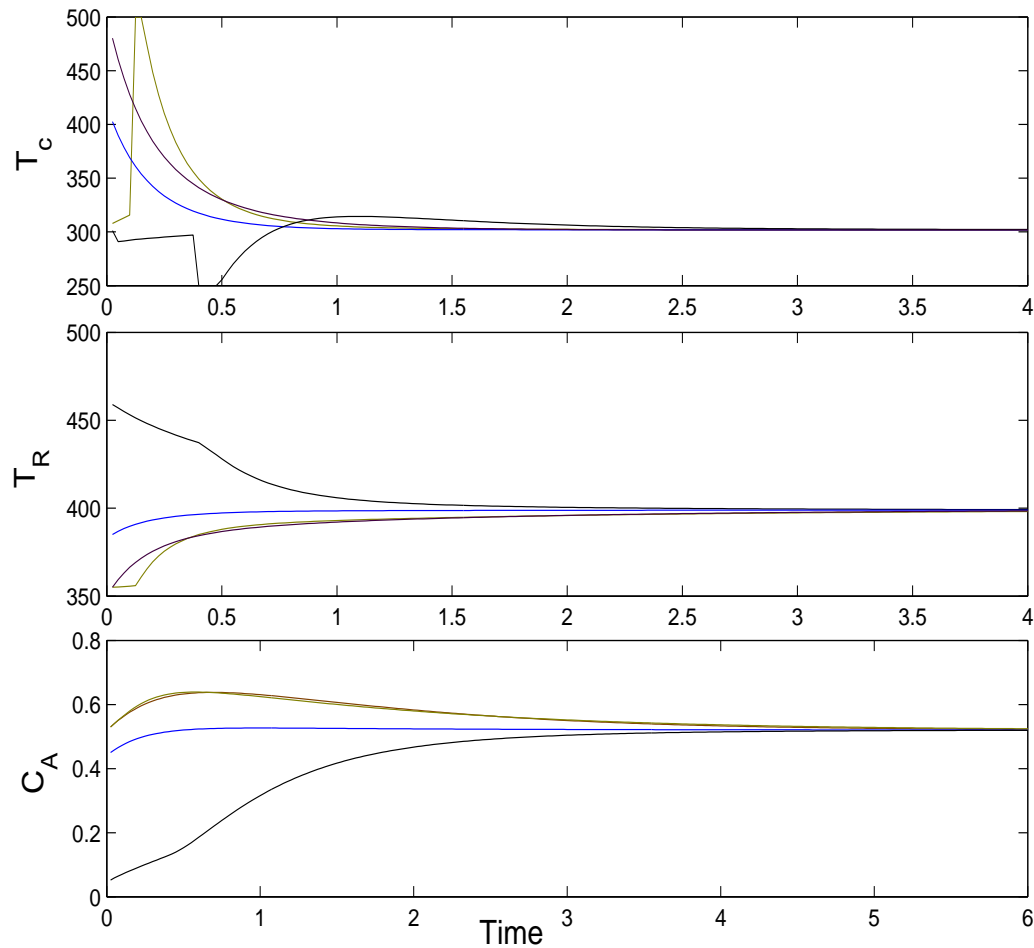
- **Bounded controller:**

- ◇ Bounded controller designed using a normal form representation

- ◇ Use  $V_k = \xi^T P_k \xi$

# CLOSED-LOOP SIMULATION RESULTS

“Stability-based switching”



◇ Closed-loop trajectories

◇ Input & state profiles

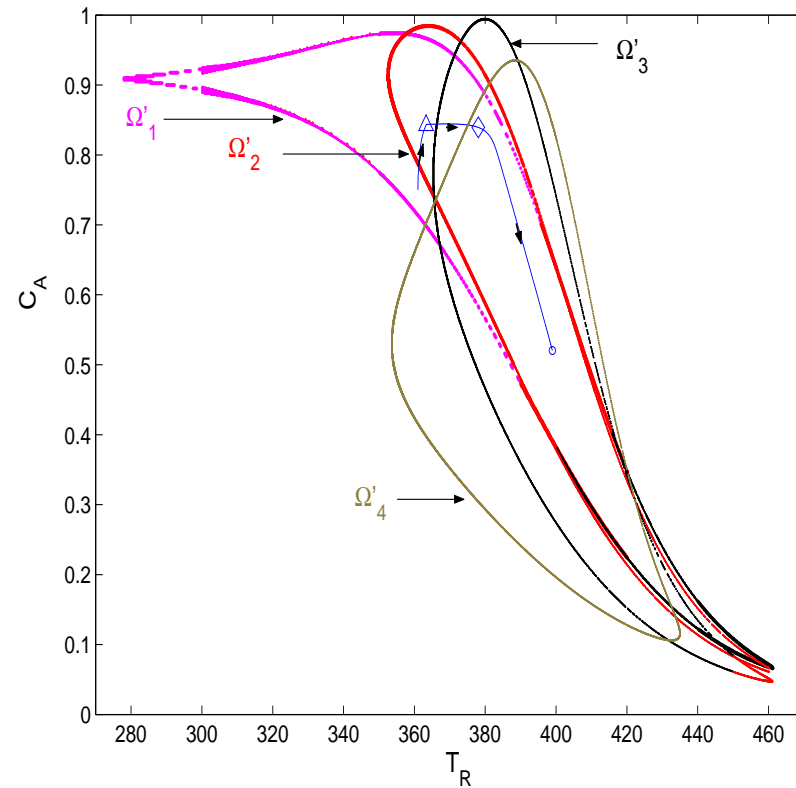
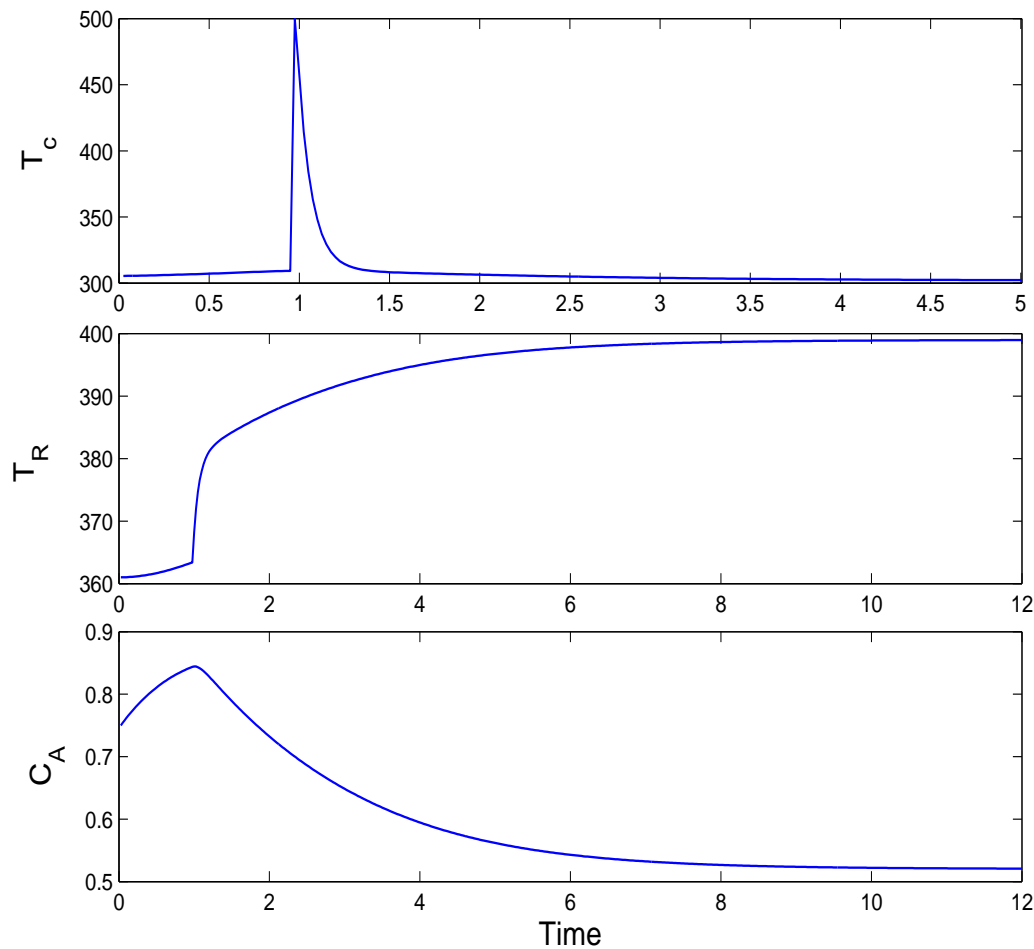
◇ MPC with  $T = 0.25$ ; MPC/BC switching ( $t = 0.125$ ); MPC with  $T = 0.5$

◇ MPC/BC switching ( $t = 0.45$ );



# CLOSED-LOOP SIMULATION RESULTS

“Performance-driven switching”



◇ Closed-loop state trajectory

◇ Input & state profiles

◇  $\dot{V}_1(t = 0.95) > 0$  ( $\dot{V}_2(t = 0.95) < 0$ ),  $\dot{V}_2(t = 1.1) > 0$  ( $\dot{V}_3(t = 1.1) < 0$ )  $\implies$

▷ Scheme 1: switch to bounded controller,  $J = 1.81 \times 10^6$

▷ Scheme 2: no switching - optimal performance,  $J = 1.64 \times 10^5$

# APPLICATION TO A CRYSTALLIZER MOMENTS MODEL

- **State–space description:**

$$\begin{aligned}\dot{x}_0 &= -x_0 + (1 - x_3)Dae \frac{-F}{y^2} \\ \dot{x}_1 &= -x_1 + yx_0 \\ \dot{x}_2 &= -x_2 + yx_1 \\ \dot{x}_3 &= -x_3 + yx_2 \\ \dot{y} &= \frac{1 - y - (\alpha - y)yx_2}{1 - x_3} + \frac{u}{1 - x_3}\end{aligned}$$

- ◇ Unstable equilibrium point surrounded by limit cycle
- ◇ Input constraints:  $u \in [-1, 1]$

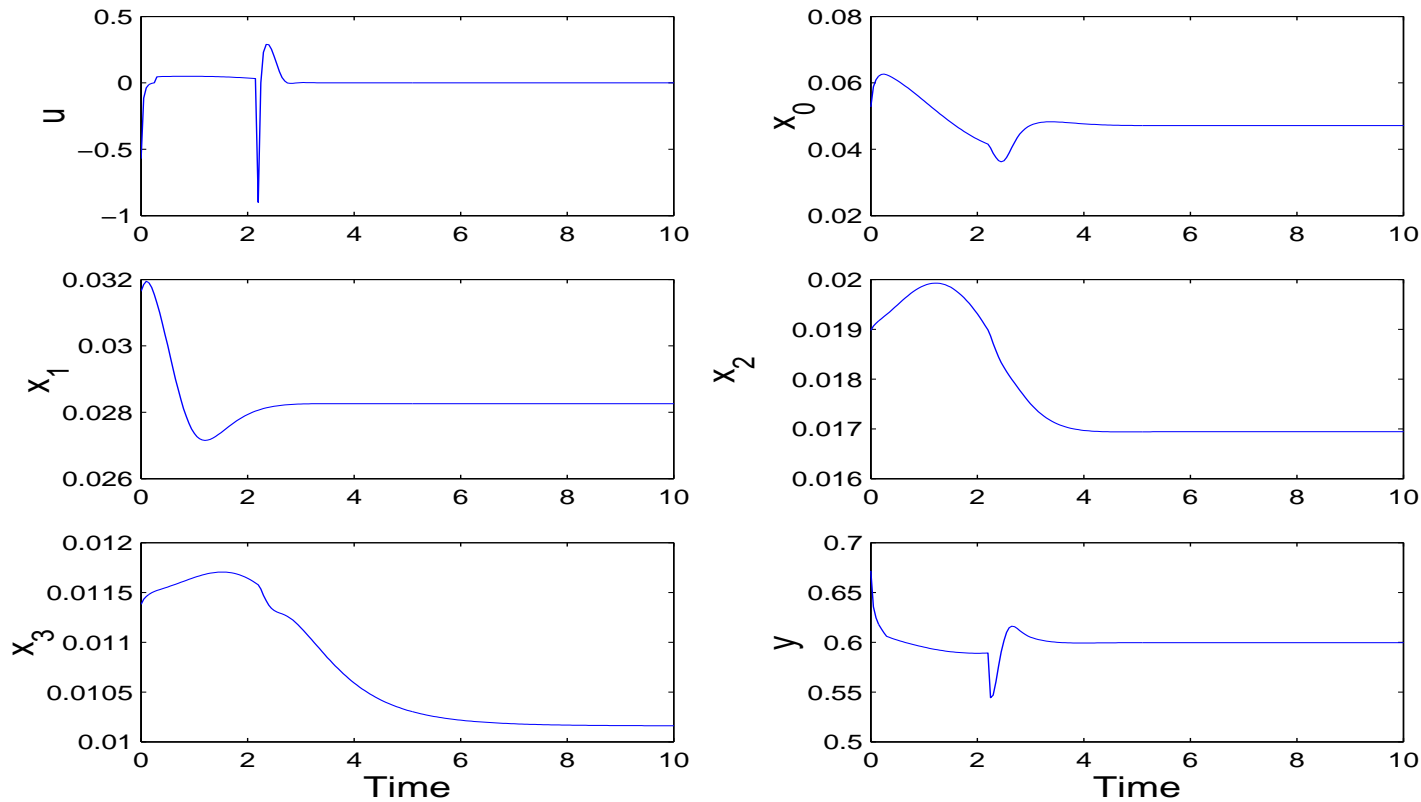
- **Bounded controller:**

- ◇ Normal form representation:

$$\begin{aligned}\dot{\xi} &= A\xi + bl(\xi, \eta) + b\alpha(\xi, \eta)u \\ \dot{\eta} &= \Psi(\xi, \eta)\end{aligned}$$

# CLOSED-LOOP SIMULATION RESULTS

“Stability-based switching”



◇ Input & state profiles

◇  $x(0) = [0.053 \ 0.03 \ 0.02 \ 0.01 \ 0.67]^T \in \Omega_{system}(u_{max})$

◇ MPC with  $T = 0.25$ , switching ( $t = 2.1$ ),  $J = 0.246$

## CONCLUSIONS

- Stabilization of nonlinear systems with input constraints
- A hybrid approach uniting MPC & bounded control
  - ◇ Decoupling:
    - ▷ Optimality (model predictive control)
    - ▷ Constrained stability region (bounded control)
  - ◇ Design & implementation of switching laws:
    - ▷ Stability, performance, etc.
    - ▷ Accommodate any MPC formulation
  - ◇ Use of computationally inexpensive MPC implementations
    - ▷ Off-line explicit characterization of stability region

## ACKNOWLEDGEMENT

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