

HYBRID PREDICTIVE OUTPUT FEEDBACK STABILIZATION OF CONSTRAINED LINEAR SYSTEMS

Prashant Mhaskar, Nael H. El-Farra
& Panagiotis D. Christofides

Department of Chemical Engineering
University of California, Los Angeles



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INTRODUCTION

- **Input constraints:**
 - ◇ Finite capacity of control actuators
 - ◇ Impose fundamental limitations on initial conditions
- **Main issues for 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

LINEAR SYSTEMS WITH INPUT CONSTRAINTS

- State space description:

$$\dot{x}(t) = Ax(t) + Bu(t)$$

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

$$u(t) \in \mathcal{U}$$

- ◇ $x(t) \in \mathbb{R}^n$: state vector
- ◇ $u(t) \in \mathbb{R}^m$: control input
- ◇ $y(t) \in \mathbb{R}^k$: output vector
- ◇ (A, B) : controllable pair
- ◇ (C, A) : observable pair
- ◇ $0 \in \text{int. } \mathcal{U}$: compact & convex

- Finite parameterization of controls:

- ◇ U_Δ : piecewise constant functions (with period Δ & values in \mathcal{U})
- ◇ $u(\cdot) \in U_\Delta$ characterized by sequence:

$$\{u[k]\} : u(t) = u(k\Delta), \forall t \in [k\Delta, (k+1)\Delta]$$

- Stabilization of origin under constraints

MODEL PREDICTIVE CONTROL

- Control problem formulation

- ★ Finite-horizon optimal control:

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

- ★ 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

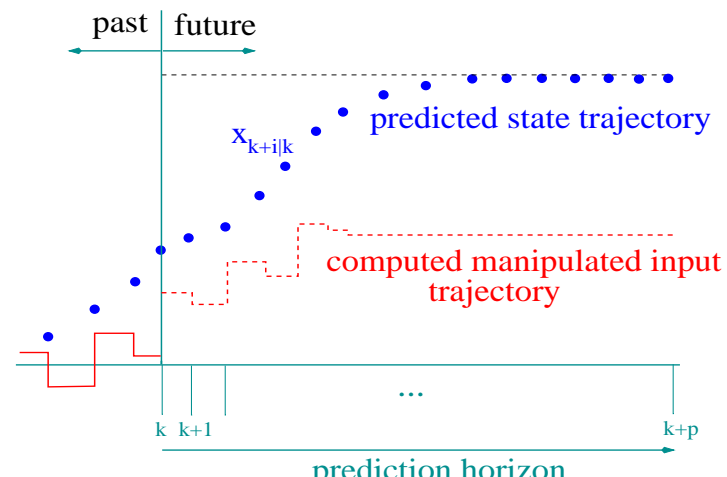
- ◇ $Q, R > 0$: penalty weights

- ◇ $F(\cdot)$: 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: $x(t + T) = 0$

- ★ Terminal inequality constraints: $x(t + T) \in W$

- ★ Control Lyapunov functions: $V(x(t + T)) < V(x(t))$

- **Issues of practical implementation:**

- ◇ Lack of (a priori) explicit characterization of stability region

- ★ Extensive closed-loop simulations

- ★ Restriction to small neighborhoods around origin

BOUNDED LYAPUNOV-BASED CONTROL

- **Explicit bounded nonlinear control law:**

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

◇ An example gain: (Lin & Sontag, 1991)

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

$$\begin{aligned} V &= x^T P x, \quad A^T P + P A - P B B^T P < 0 \\ L_f V &= x^T (A^T P + P A) x, \quad L_G V = 2x^T P B \end{aligned}$$

◇ Nonlinear gain-shaping procedure:

★ **Accounts explicitly for constraints & closed-loop stability**

- **Constrained closed-loop properties:**

◇ **Asymptotic stability**

◇ **Inverse optimality**

$$J = \int_0^\infty (l(x) + u^T R(x) u) dt, \quad l(x) > 0, \quad R(x) > 0, \quad J_{min} = V(x(0))$$

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: $x(0) \in \Omega \implies x(t) \in \Omega \subset D \quad \forall t \geq 0$

- ◇ Provides larger estimate than saturated linear/nonlinear controllers

HYBRID CONTROL: UNITING BOUNDED CONTROL & MPC

(El-Farra, Mhaskar & Christofides, Automatica, 2004)

- Objectives:

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

- ▷ Reconcile tradeoffs in stability and optimality properties

- ★ Explicit characterization of constrained stability region

- ★ A safety net for the implementation of MPC

- Central idea:

Decoupling “optimality” & “constrained stabilizability”

- ◇ Stability region provided by bounded controller

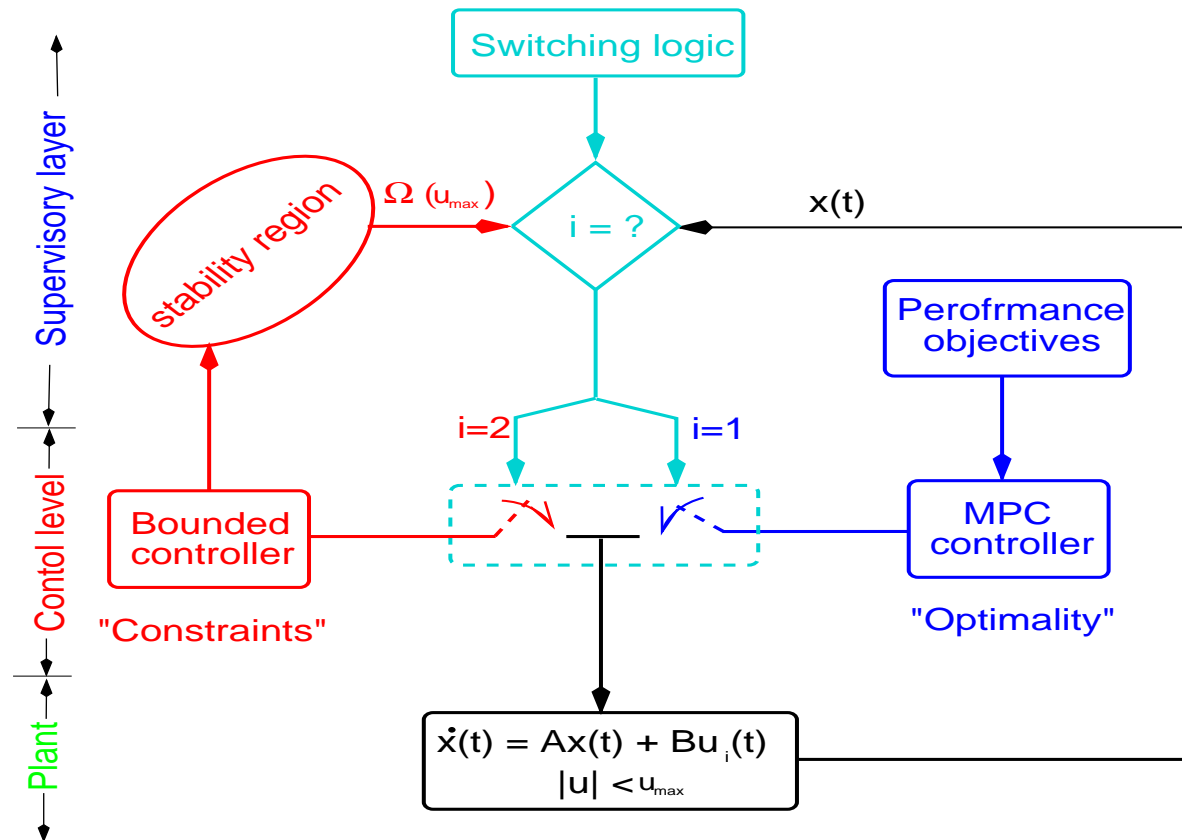
- ◇ Optimal performance supplied by MPC controller

- Approach:

- ◇ Switching between MPC & bounded controller

OVERVIEW OF HYBRID CONTROL STRATEGY

“STATE FEEDBACK”



- Hierarchical control structure
 - ★ Plant level
 - ★ Control level
 - ★ Supervisory level
- Any MPC formulation can be used
 - ★ Switching rules may vary
- Several switching schemes possible

CONTROLLER SWITCHING SCHEMES

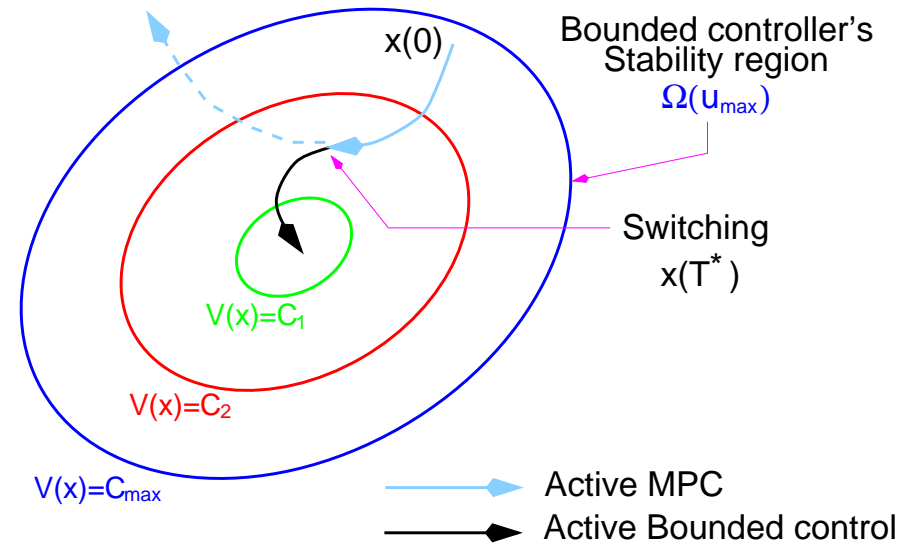
“STATE FEEDBACK”

• Stability-based switching:

(e.g., Classical MPC formulations)

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

$$T^* = \inf\{T^* \geq 0 : V(x^M(T^*)) \geq 0\}$$

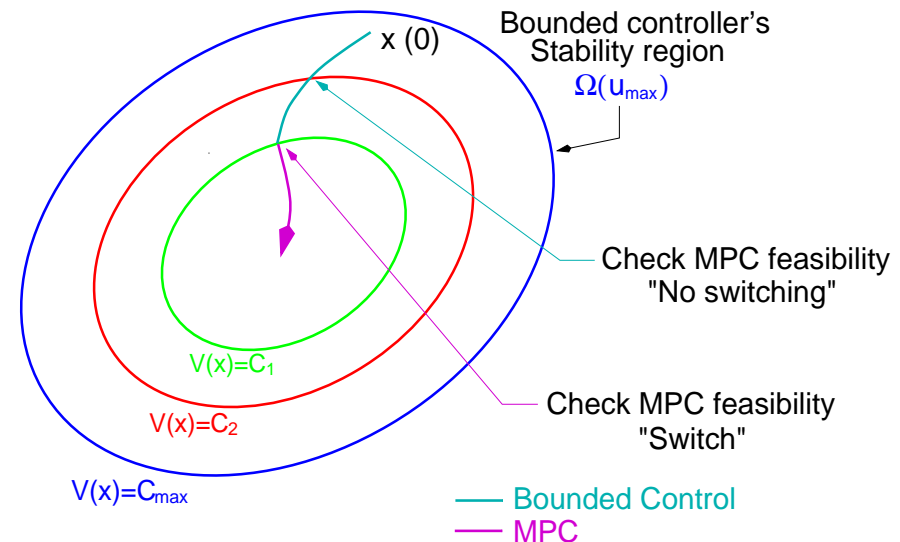


• Feasibility-based switching:

(Stabilizing MPC formulations)

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

★ T^* : earliest time for which MPC yields **feasible solution**



IMPLICATIONS OF SWITCHING SCHEME

- **A safety net for implementing MPC:**
 - ◇ Bounded controller provides a fall-back mechanism
 - ◇ Switched closed-loop system inherits the stability region
 - ★ A priori guarantees for all $x(0) \in \Omega(u_{max})$
 - ★ Stability independent of MPC properties (e.g., horizon length)
 - ★ Can reduce computational load
- **Conceptual differences from other schemes:**
 - ◇ Switching does not occur locally
 - ◇ Provides stability region explicitly
 - ◇ No switching occurs if switching rules are satisfied
 - ★ Only MPC is implemented \implies optimal performance recovered

STATE ESTIMATION & OUTPUT FEEDBACK CONTROL

- **Lack of full state measurements:**
 - ◇ Inaccessibility of some process variables for measurement.
 - ◇ Estimation of states from measured outputs necessary.
- **Main objectives for output feedback controller design:**
 - ◇ To establish guaranteed stability from an explicitly characterized set of initial conditions:
 - ▷ Design technique for the state estimator.
 - ▷ Devise switching rules, based on available state measurements.
 - ◇ Controlled rate of convergence of state estimation error.
- **Approach for output feedback controller design:**
 - ◇ Relies on separation principle: combination of
 - ▷ State feedback controllers.
 - ▷ State observers.

DESIGN OF STATE OBSERVER

- State space description of estimator:

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

- ◇ \hat{x} : state estimate.

- ◇ L : Observer gain matrix.

- Structural features:

- ◇ Observer pole placement:

- ▷ Enforce desired decay of estimation error

- ◇ Effect of observer peaking eliminated through:

- ▷ Input saturation (indirect), OR

- ▷ Estimate-saturation (direct)(e.g., El-Farra & Christofides, IJC, 2001).

- Closed-loop analysis:

- ◇ Fall back (bounded controller) robust to a certain allowable error.

- ◇ For a given choice of initial conditions:

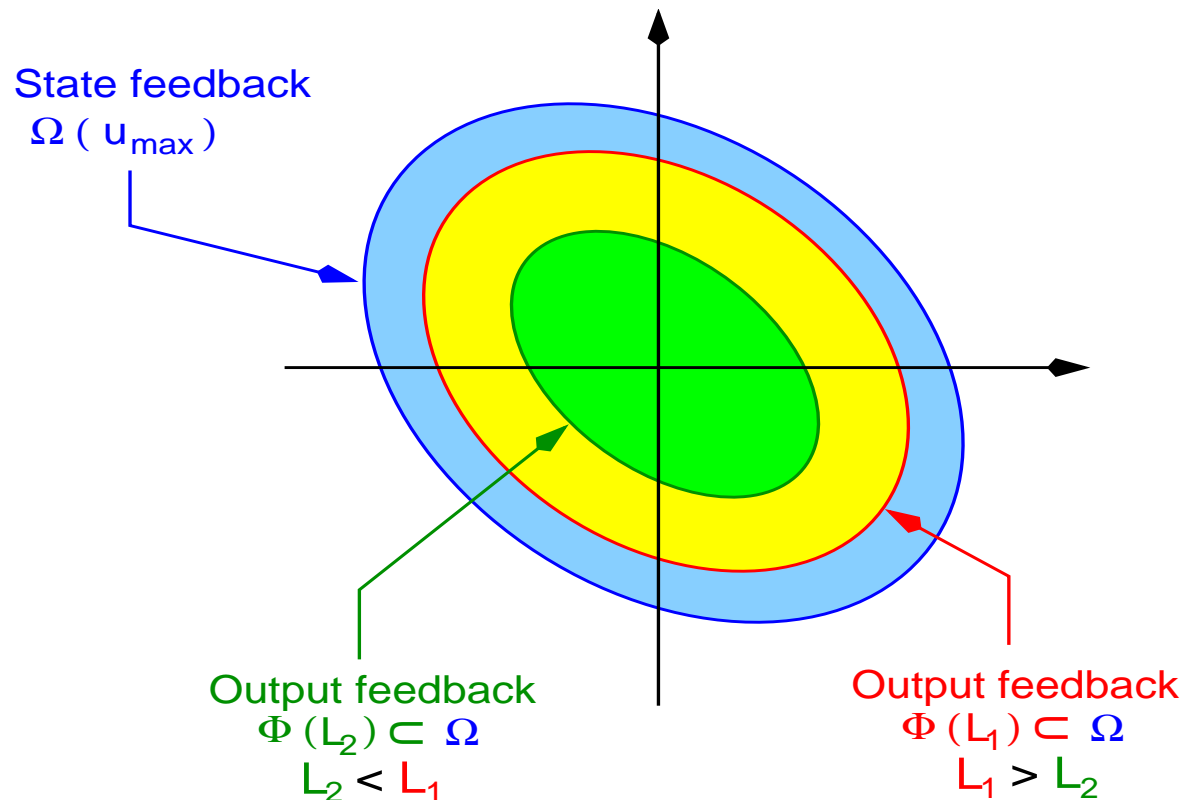
- ▷ State estimator designed to force error under the allowable error.

- ▷ Switching laws, based on the state estimates, for “safe” MPC implementation.

STATE ESTIMATION & OUTPUT FEEDBACK CONTROL

- Practical implications:

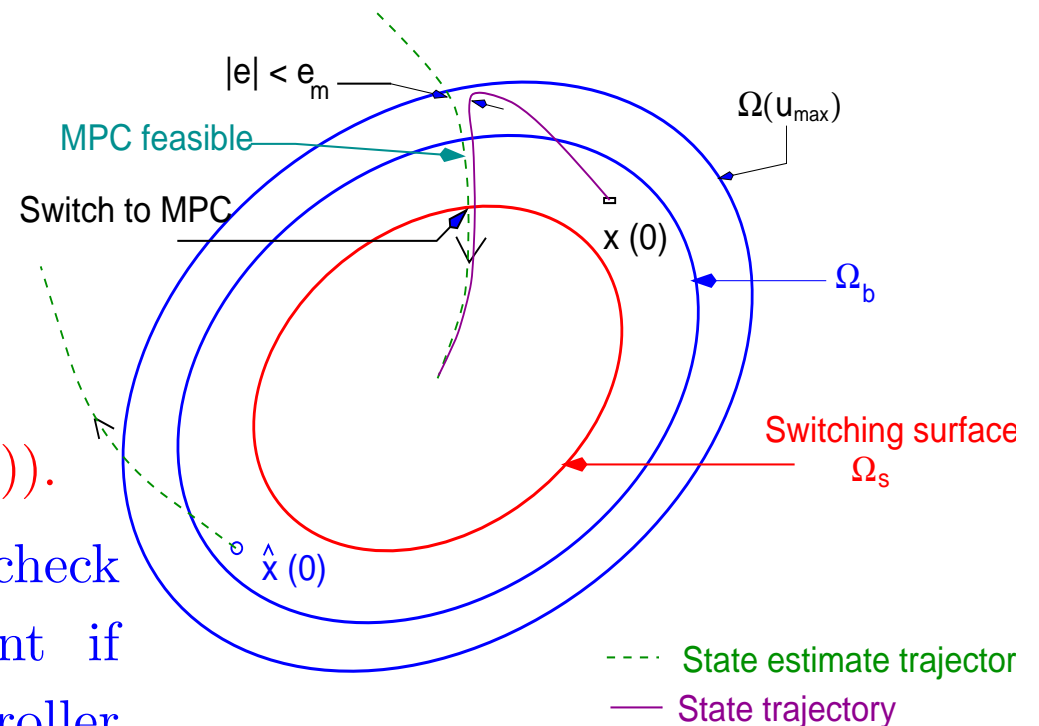
- ◇ Any other estimation scheme, such as Moving Horizon Estimation (MHE), can be used.
- ◇ Requires a transparent relationship between error decay and estimator parameters.
- ◇ MPC implemented in a region where the fall back controller can step in any time to rescue stability.



OUTPUT FEEDBACK IMPLEMENTATION OF SWITCHING

(Mhaskar, El-Farra, & Christofides, AIChE J., to appear)

- Bounded controller design
($u = -k(x)L_g V, \Omega(u_{max})$).
- State observer design
(given $\Omega_b \subset \Omega$, compute L).
- Estimate 'safe' region
(given $\hat{x} \in \Omega_s \Rightarrow x \in \Omega$).
- Initialize: $\hat{x}(0) \in \Omega_b, u(0) = b(\hat{x}(0))$.
- After \hat{x} enters 'safe' region, Ω_s , check feasibility of MPC & implement if $\dot{V}(\hat{x}) < 0$ else keep bounded controller active.



SIMULATION EXAMPLE

- State-space description:

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t)\end{aligned}$$

$$A = \begin{bmatrix} 0.55 & 0.15 & 0.05 \\ 0.15 & 0.40 & 0.20 \\ 0.10 & 0.15 & 0.45 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- ★ Origin unstable (three real positive eigenvalues)
- ★ Input constraints: $u_i \in [-1, 1]$, $i = 1, 2$
- ★ Only the first and the third states are measured

CONTROLLER DESIGN

- **Bounded controller:**

- ◇ Lyapunov function:

$$V = x^T P x, \quad P = \begin{bmatrix} 6.5843 & 4.2389 & -3.8307 \\ 4.2389 & 3.6091 & -2.6672 \\ -3.8307 & -2.6672 & 2.8033 \end{bmatrix}$$

- ◇ Stability region: $x^T (A^T P + P A) x < \|B^T P x\|$

- **Model predictive controller:**

- ◇ Performance index:

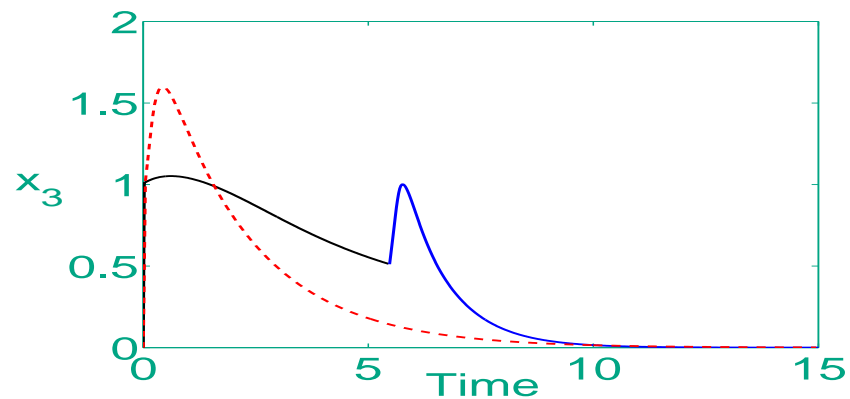
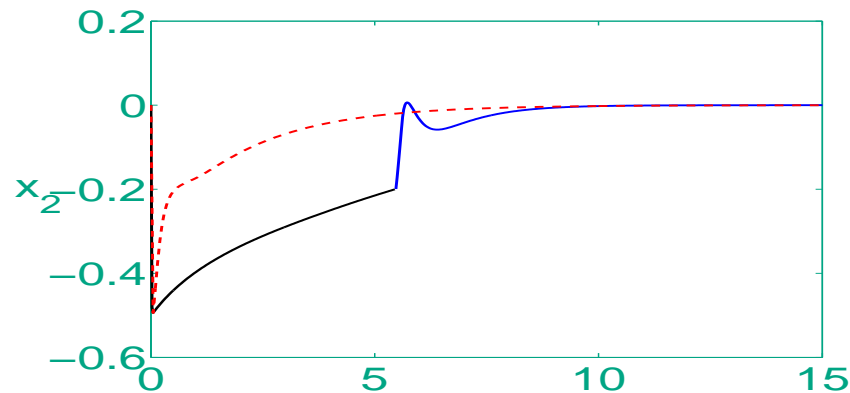
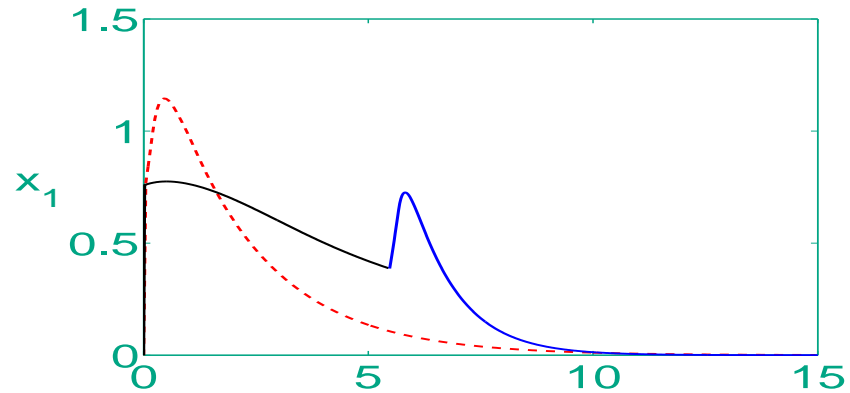
$$J = \int_t^{t+T} [\|x(\tau)\|_Q^2 + \|u(\tau)\|_R^2 + \|\dot{u}(\tau)\|_S^2] d\tau$$

$$Q = qI > 0, \quad R = rI > 0, \quad S = sI > 0$$

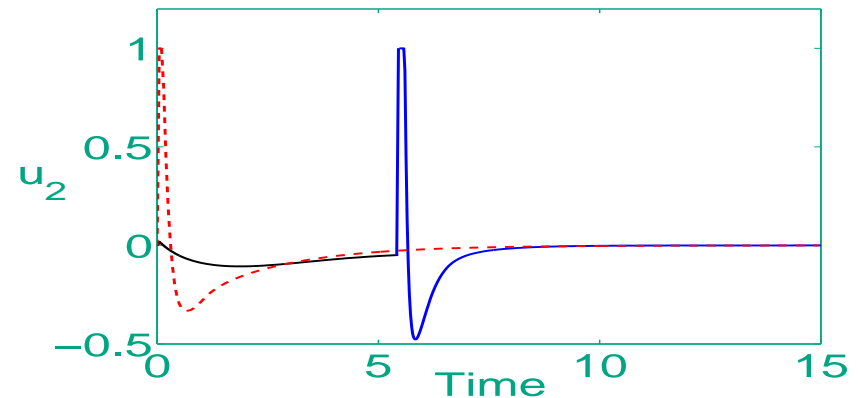
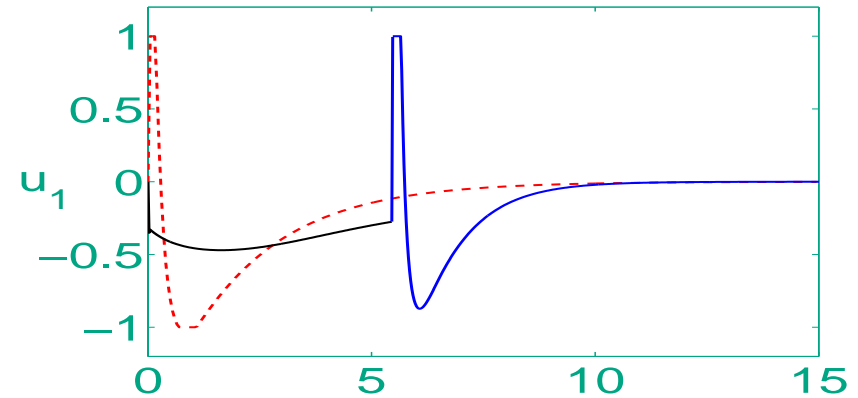
- **State observer:** ($\lambda = 500, a = 2$)

CLOSED-LOOP SIMULATION RESULTS

★ State & estimate profiles



★ Input profiles



★ $x_0 = [0.75 \ -0.5 \ 1.0]^T \in \Omega_s$

★ $\hat{x}_0 = [0 \ 0 \ 0]^T \in \Omega_s$

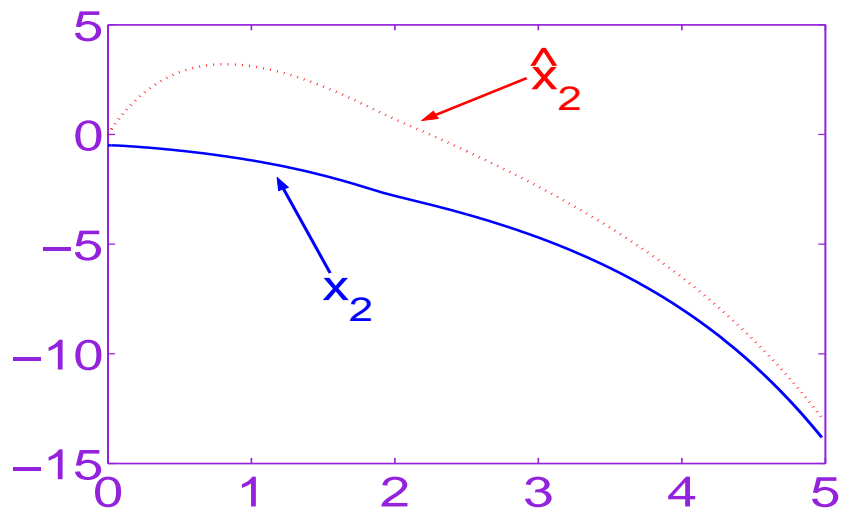
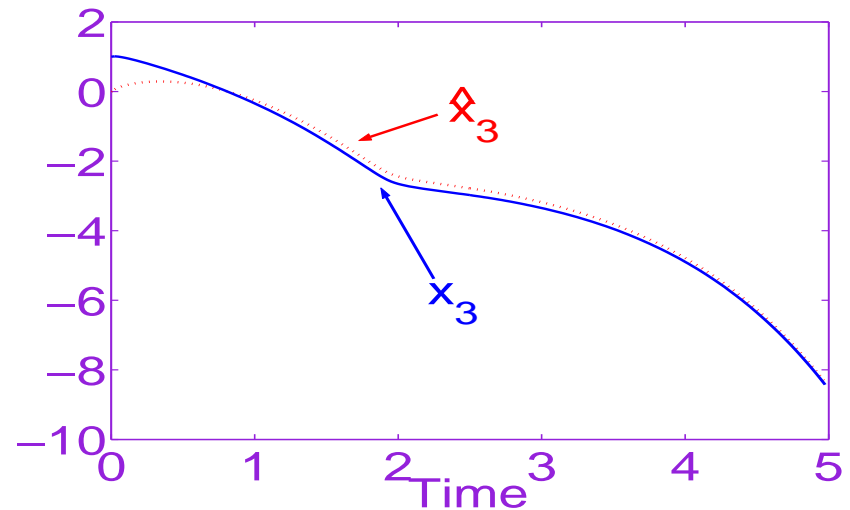
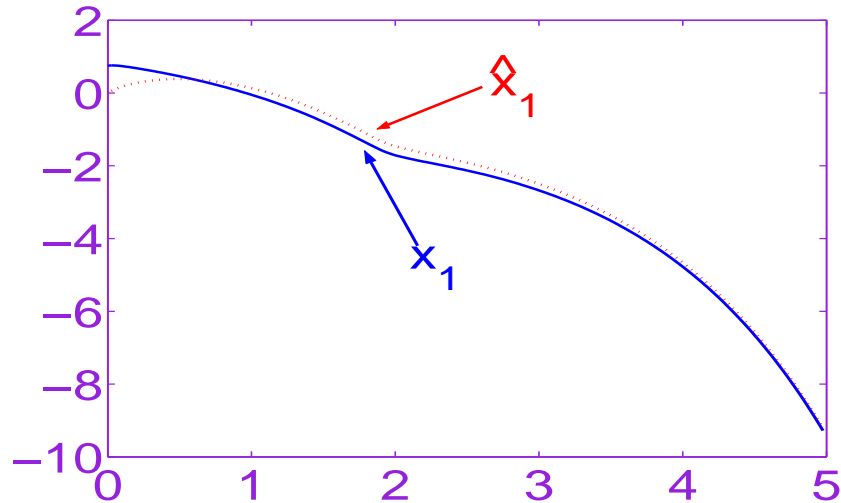
★ BC /MPC switching

($T = 1.5$, $t_{switch} = 5.45$)

★ MPC with $T = 3.5$

CLOSED-LOOP SIMULATION RESULTS

★ State & estimate profiles



★ $x_0 = [0.75 \ -0.5 \ 1.0]^T \in \Omega_s$

★ Observer gain: $\lambda = 0.5$

★ Bounded controller

(MPC infeasible, no switching)

CONCLUSIONS

- Output feedback stabilization of constrained linear systems
- A hybrid control structure uniting MPC & bounded control
 - ◇ Decoupling:
 - ★ Optimality (model predictive control)
 - ★ Constrained stability region (bounded control)
 - ◇ Output feedback implementation of switching strategy:
 - ★ State feedback controller design
 - ★ State observers
 - ◇ A safety net for implementation of MPC

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