

# **EXAMPLES OF HYBRID MODEL PREDICTIVE CONTROL**

**Alberto Bemporad - “Model Predictive Control” course - Academic year 2017/18**

# HYBRID MPC FOR CRUISE CONTROL



Disclaimer: This is  
an academic example

## GOAL:

Command **gear ratio**, **gas pedal**, and **brakes** to **track** a  
desired **speed** and minimize fuel **consumption**

# HYBRID MODEL

- Vehicle dynamics

$$m\ddot{x} = F_e - F_b - \beta\dot{x}$$

$\dot{x}$  = vehicle speed

$F_e$  = traction force

$F_b$  = brake force

→ discretized with sampling time  $T_s = 0.5$  s

- Transmission kinematics

$$\omega = \frac{R_g(i)}{k_s} \dot{x}$$

$\omega$  = engine speed

$C$  = engine torque

$i$  = gear

power balance:

$$F_e \dot{x} = C\omega$$



(drawing by Dong Du Bosque)

# HYBRID MODEL

- Gear selection: for each gear # $i$ , define a binary input

$$i = R, 1, 2, 3, 4, 5$$

$$g_i \in \{0, 1\}$$

- Gear selection (traction force):

$$F_e = \frac{R_g(i)}{k_s} C$$

depends on gear # $i$

define auxiliary continuous variables:

$$\text{IF } g_i = 1 \text{ THEN } F_{ei} = \frac{R_g(i)}{k_s} C \text{ ELSE } 0$$

$$\rightarrow F_e = F_{eR} + F_{e1} + F_{e2} + F_{e3} + F_{e4} + F_{e5}$$



- Gear selection (engine/vehicle speed):

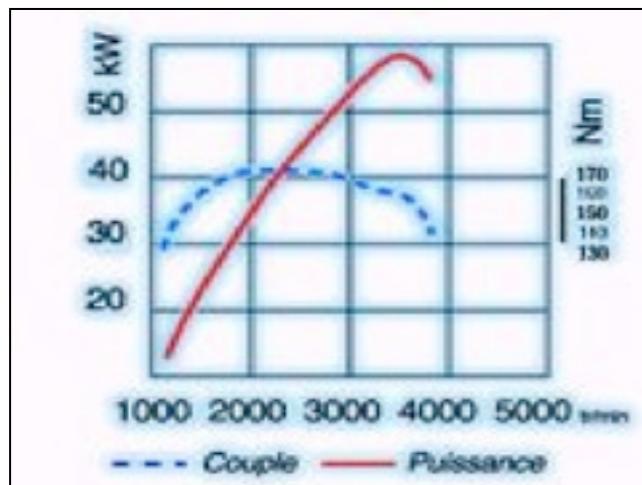
$$\omega = \frac{R_g(i)}{k_s} \dot{x}$$

similarly, also requires 6 auxiliary continuous variables

# HYBRID MODEL

- engine torque  $-C_e^-(\omega) \leq C \leq C_e^+(\omega)$

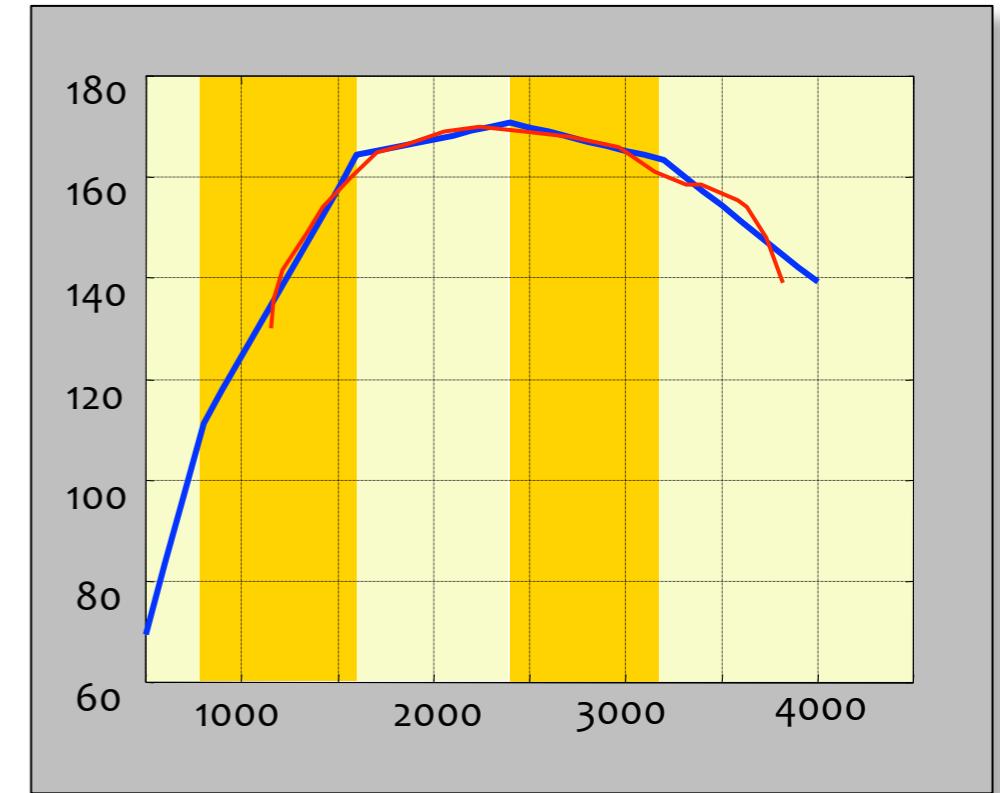
- max engine torque  $C_e^+(\omega)$



requires: 4 binary aux variables  
4 continuous aux variables

- Min engine torque  $C_e^-(\omega) = \alpha_1\omega + \beta_1$

**Note:** in this case the PWL constraint  $C \leq C_e^+(\omega)$  is convex, it could be handled by linear constraints without introducing any binary variable !



# HYSDEL MODEL

```

SYSTEM cruisecontrolmodel {

INTERFACE {
    PARAMETER {
        REAL mass = 1020; /* kg */
        REAL beta_friction = 25; /* W/m*s */
    }

    [snip]
}

STATE { REAL position [0,10000];
        REAL speed [vmin,vmax]; }

INPUT { REAL torque [Cmin,Cmax];
        REAL F_brake [0,max_brake_force];
        BOOL gear1, gear2, gear3, gear4, gear5, gearR;
    }

IMPLEMENTATION {
    AUX {REAL F, Fe1, Fe2, Fe3, Fe4, Fe5, FeR;
        REAL w, w1, w2, w3, w4, w5, wR;
        BOOL dPWL1,dPWL2,dPWL3,dPWL4;
        REAL DCe1,DCe2,DCe3,DCe4; }

    LINEAR {F = Fe1+Fe2+Fe3+Fe4+Fe5+FeR;
            w = w1+w2+w3+w4+w5+wR; }

    AD { dPWL1 = wPWL1-w<=0;
        dPWL2 = wPWL2-w<=0;
        dPWL3 = wPWL3-w<=0;
        dPWL4 = wPWL4-w<=0; }

    DA { Fe1 = {IF gear1 THEN torque/speed_factor*Rgear1};
        Fe2 = {IF gear2 THEN torque/speed_factor*Rgear2};
        Fe3 = {IF gear3 THEN torque/speed_factor*Rgear3};
        Fe4 = {IF gear4 THEN torque/speed_factor*Rgear4};
        Fe5 = {IF gear5 THEN torque/speed_factor*Rgear5};
        FeR = {IF gearR THEN torque/speed_factor*RgearR}; }
}

```

```

w1 = {IF gear1 THEN speed/speed_factor*Rgear1};
w2 = {IF gear2 THEN speed/speed_factor*Rgear2};
w3 = {IF gear3 THEN speed/speed_factor*Rgear3};
w4 = {IF gear4 THEN speed/speed_factor*Rgear4};
w5 = {IF gear5 THEN speed/speed_factor*Rgear5};
wR = {IF gearR THEN speed/speed_factor*RgearR};

DCe1 = {IF dPWL1 THEN (aPWL2-aPWL1)+(bPWL2-bPWL1)*w};
DCe2 = {IF dPWL2 THEN (aPWL3-aPWL2)+(bPWL3-bPWL2)*w};
DCe3 = {IF dPWL3 THEN (aPWL4-aPWL3)+(bPWL4-bPWL3)*w};
DCe4 = {IF dPWL4 THEN (aPWL5-aPWL4)+(bPWL5-bPWL4)*w};

}

CONTINUOUS { position = position+Ts*speed;
             speed = speed+Ts/mass*(F-F_brake-beta_friction*speed);

MUST { /* max engine speed */
       /* wemin <= w1+w2+w3+w4+w5+wR <= wemax */

-w1 <= -wemin; w1 <= wemax;
-w2 <= -wemin; w2 <= wemax;
-w3 <= -wemin; w3 <= wemax;
-w4 <= -wemin; w4 <= wemax;
-w5 <= -wemin; w5 <= wemax;
-wR <= -wemin; wR <= wemax;

-F_brake <=0;
F_brake <= max_brake_force;

-torque-(alpha1+beta1*w) <=0;
torque-(aPWL1+bPWL1*w+DCe1+DCe2+DCe3+DCe4)-1<=0;

-((REAL gear1)+(REAL gear2)+(REAL gear3)+(REAL gear4)+  

      (REAL gear5)+(REAL gearR))<=-0.9999;  

(REAL gear1)+(REAL gear2)+(REAL gear3)+(REAL gear4)+  

      (REAL gear5)+(REAL gearR)<=1.0001;

dPWL4 -> dPWL3; dPWL4 -> dPWL2;  

dPWL4 -> dPWL1; dPWL3 -> dPWL2;  

dPWL3 -> dPWL1; dPWL2 -> dPWL1;  

}
}

```

go to demo /demos/cruise/init.m

# HYBRID MODEL

- MLD model

$$\begin{aligned}x(t+1) &= Ax(t) + B_1 u(t) + B_2 \delta(t) + B_3 z(t) + B_5 \\y(t) &= Cx(t) + D_1 u(t) + D_2 \delta(t) + D_3 z(t) + D_5 \\E_2 \delta(t) + E_3 z(t) &\leq E_4 x(t) + E_1 u(t) + E_5\end{aligned}$$

- 2 continuous states:  $x, v$  (vehicle position and speed)
- 2 continuous inputs:  $C, F_b$  (engine torque, brake force)
- 6 binary inputs:  $g_R, g_1, g_2, g_3, g_4, g_5$  (gears)
- 1 continuous output:  $v$  (vehicle speed)
- 18 auxiliary continuous vars: (6+1 traction force, 6+1 engine speed, 4 PWL max engine torque)
- 4 auxiliary binary vars: (PWL max engine torque breakpoints)
- 100 mixed-integer inequalities

# HYBRID CONTROLLER

- Max-speed controller

$$\max_{u_t} \quad J(u_t, x(t)) \triangleq v(t+1|t)$$

s.t.

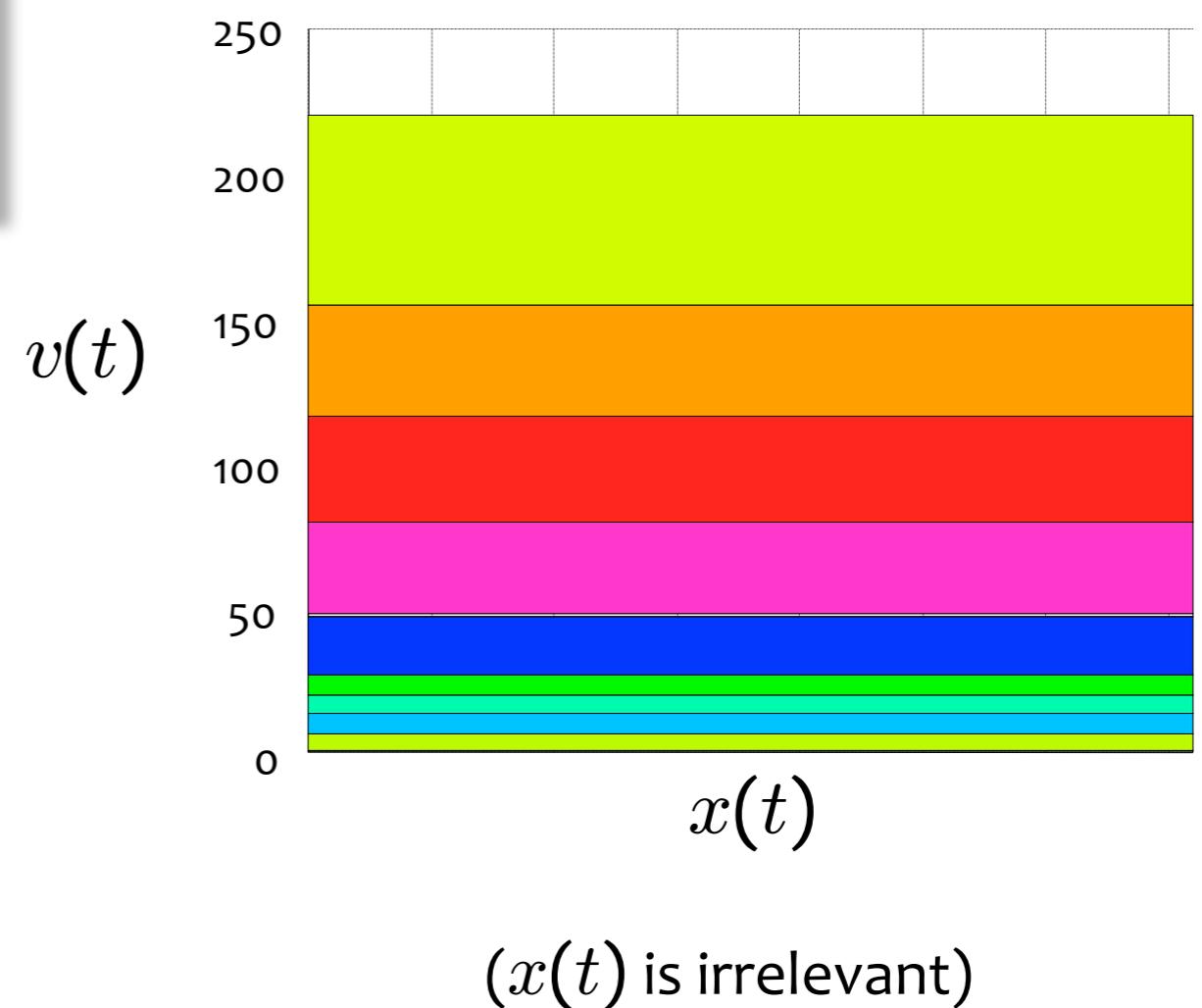
$$\left\{ \begin{array}{l} \text{MLD model} \\ x(t|t) = x(t) \end{array} \right.$$

MILP optimization problem

Linear constraints	96
Continuous variables	18
Binary variables	10
Parameters	1
Time to solve mp-MILP (Sun Ultra 10)	45 s
<b>Number of regions</b>	<b>11</b>

Objective: maximize speed

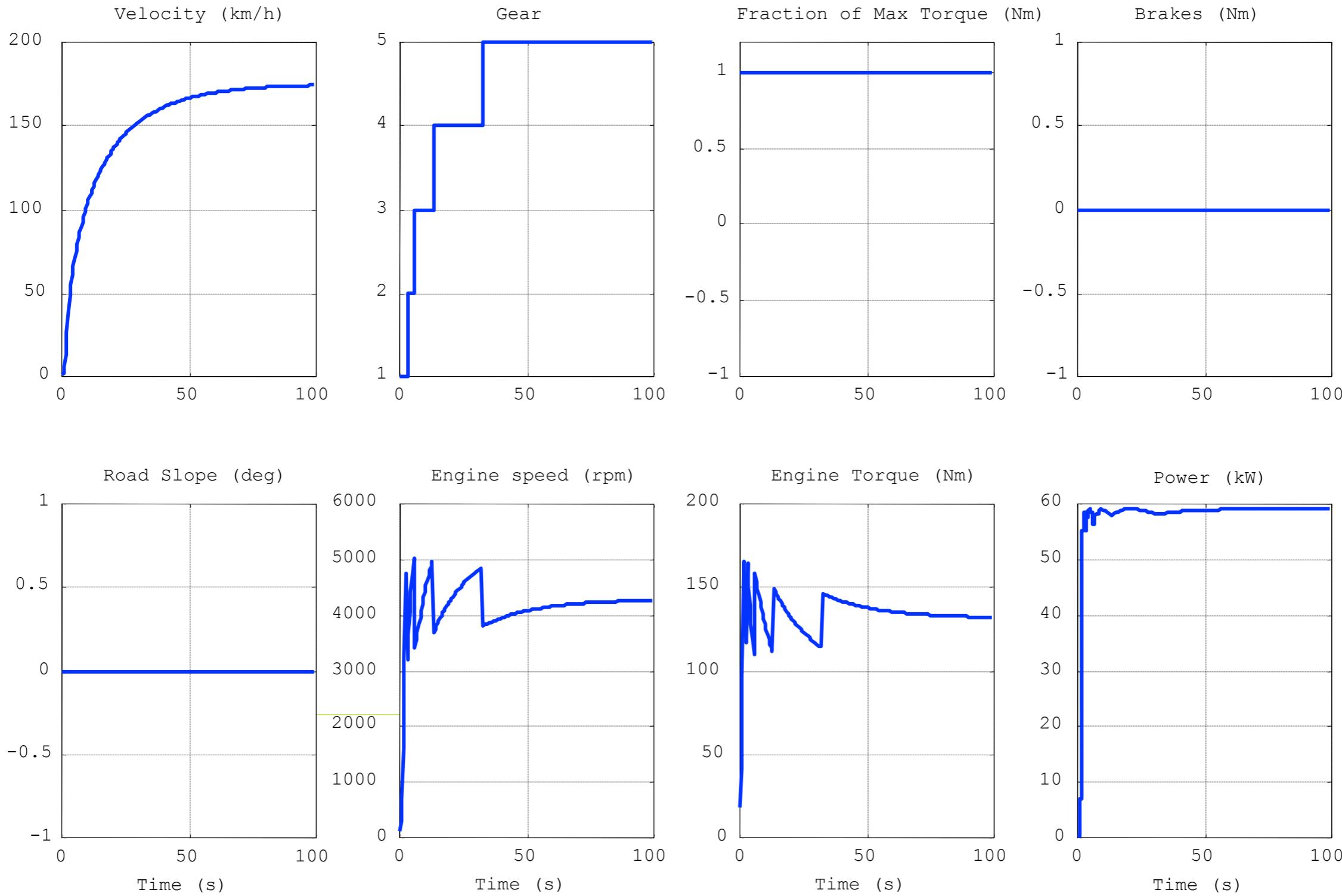
(to reproduce max acceleration plots)



(parameters: Renault Clio 1.9 DTI RXE)

# HYBRID CONTROLLER

- Max-speed controller



# HYBRID CONTROLLER

- Tracking controller

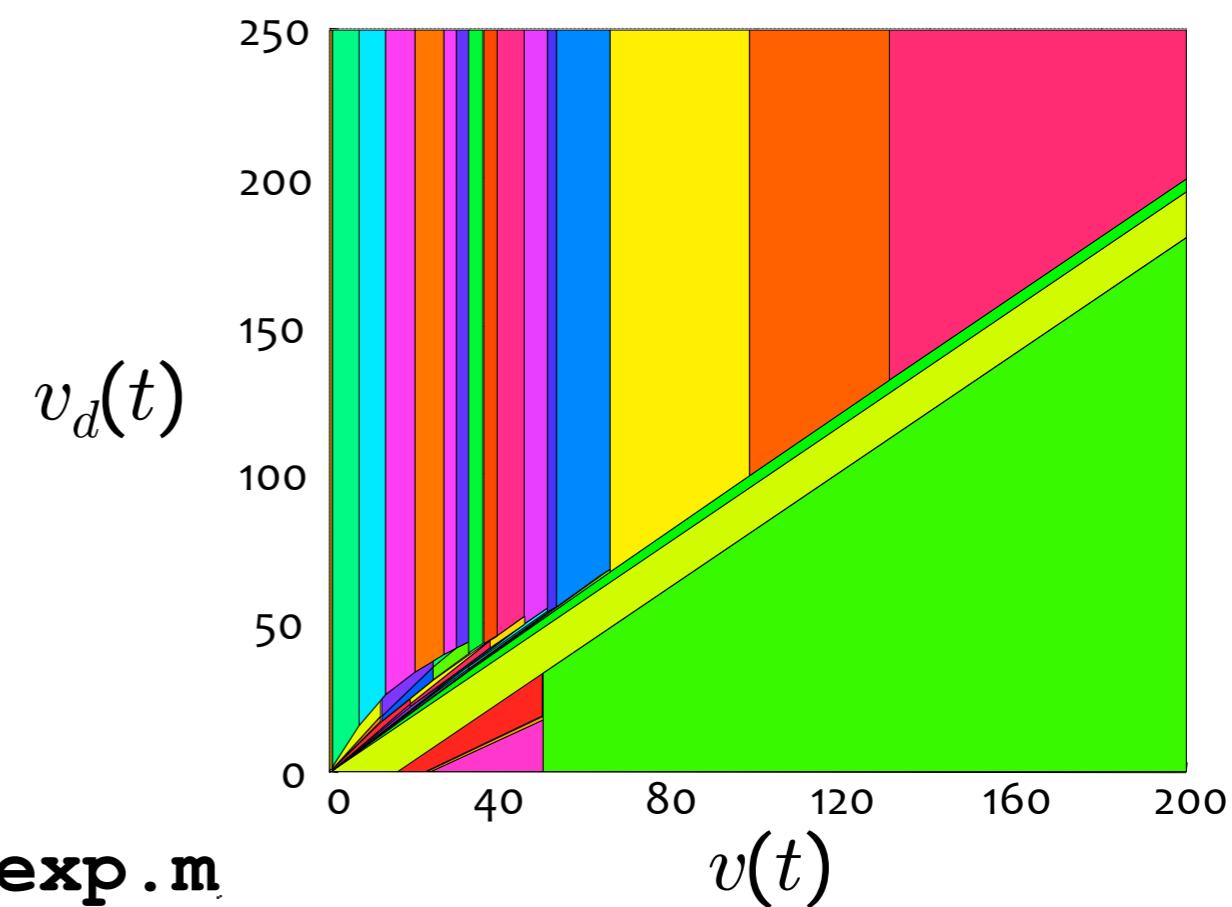
$$\min_{u_t} J(u_t, x(t)) \triangleq |v(t+1|t) - v_d(t)| + \rho|\omega|$$

s.t.

$$\begin{cases} \text{MLD model} \\ x(t|t) = x(t) \end{cases}$$

MILP optimization problem

Linear constraints	98
Continuous variables	19
Binary variables	10
Parameters	2
Time to solve mp-MILP (PC 850Mhz)	43 s
<b>Number of regions</b>	<b>49</b>

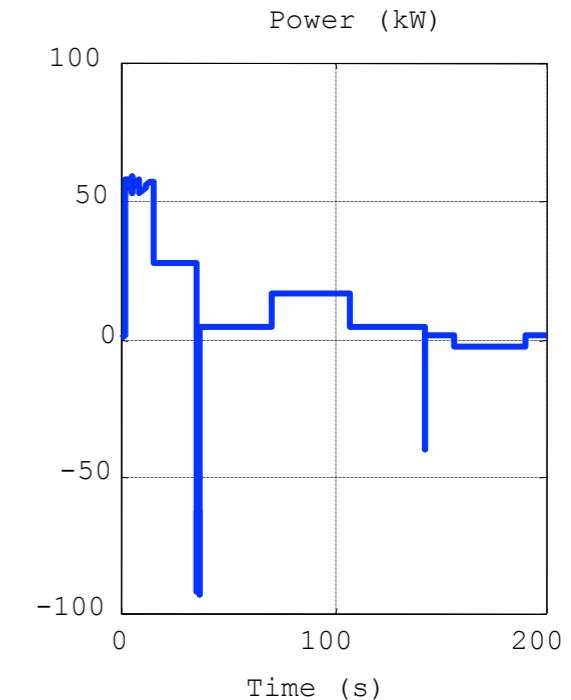
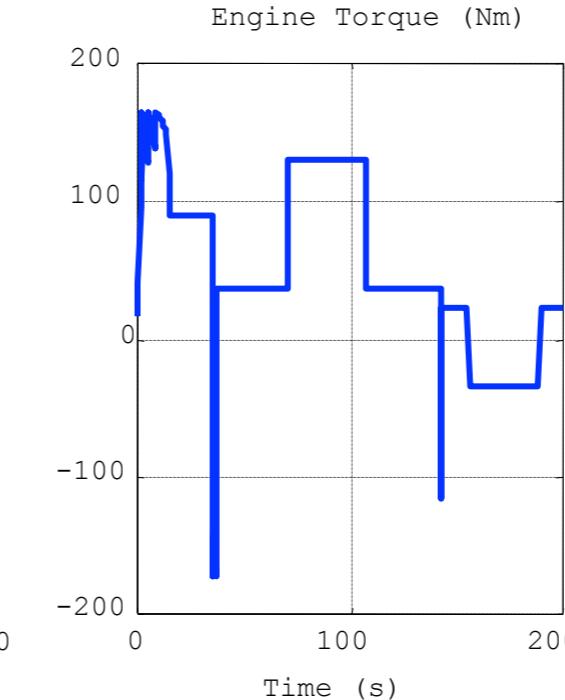
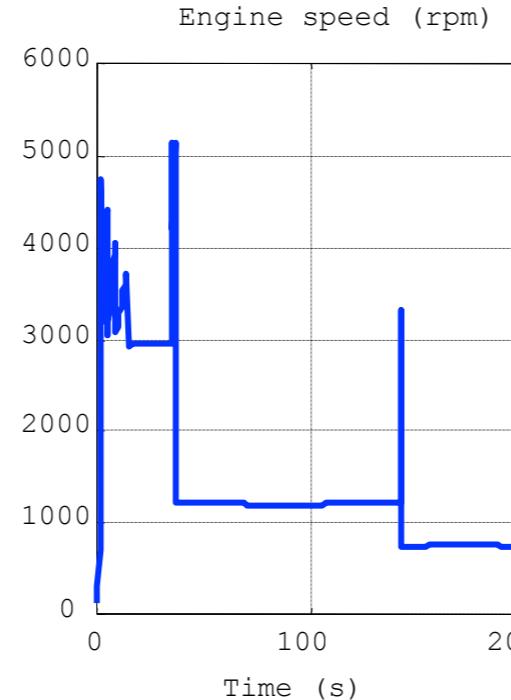
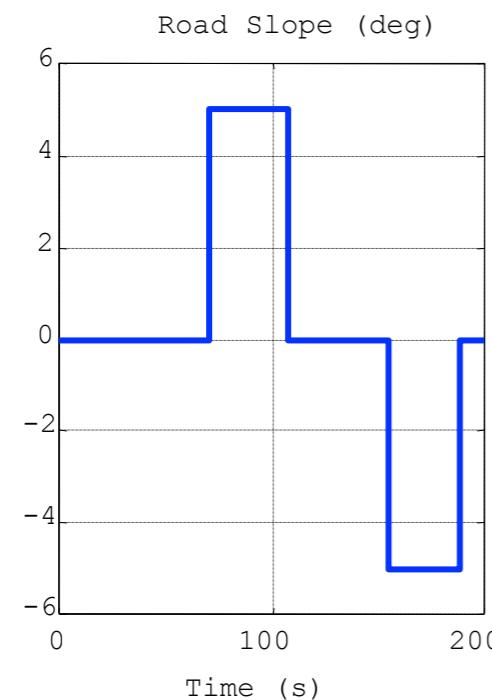
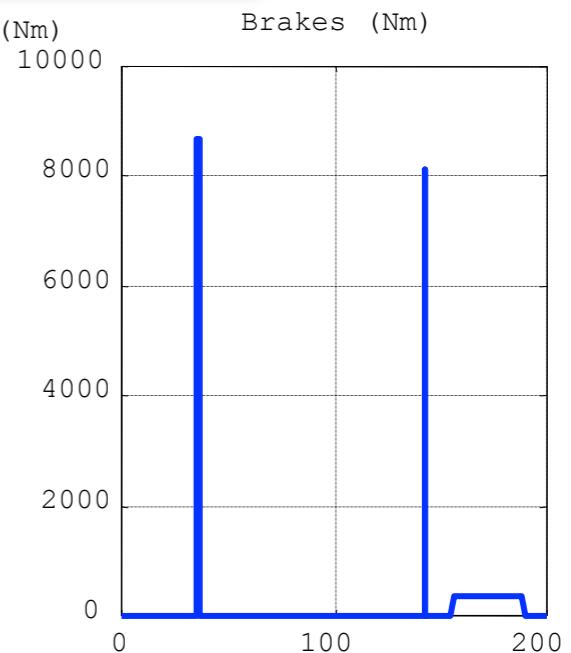
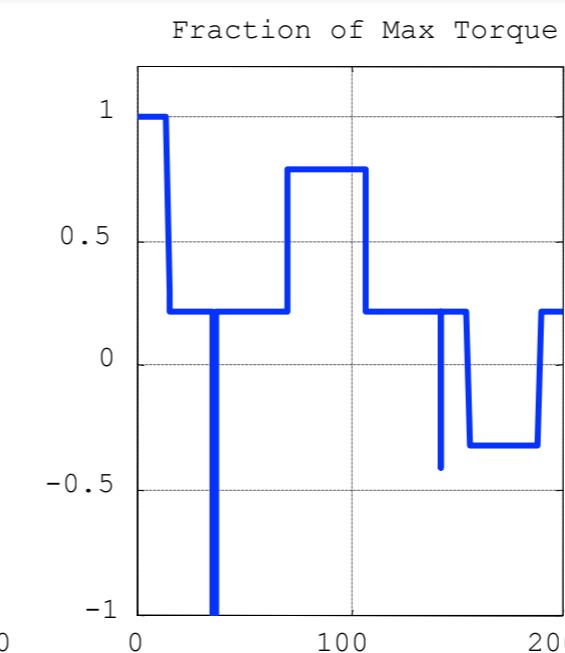
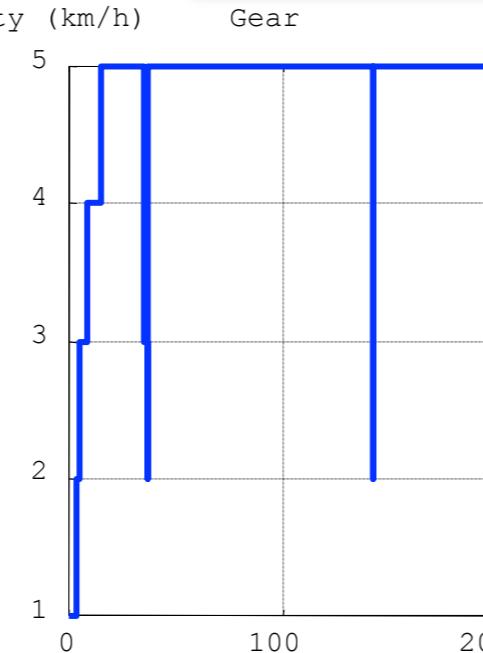
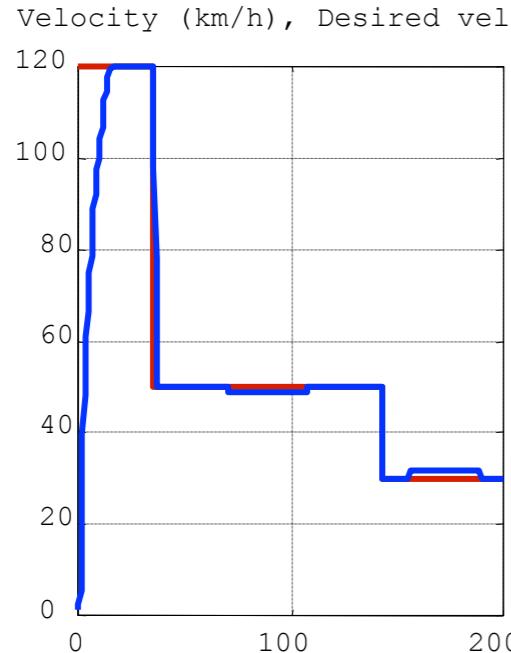


go to demo /demos/cruise/init\_exp.m

# HYBRID CONTROLLER

- Tracking controller

$$\min_{u_t} |v(t+1|t) - v_d(t)| + \rho |\omega| \quad \rho = 0.001$$



# HYBRID CONTROLLER

- Smoother tracking controller

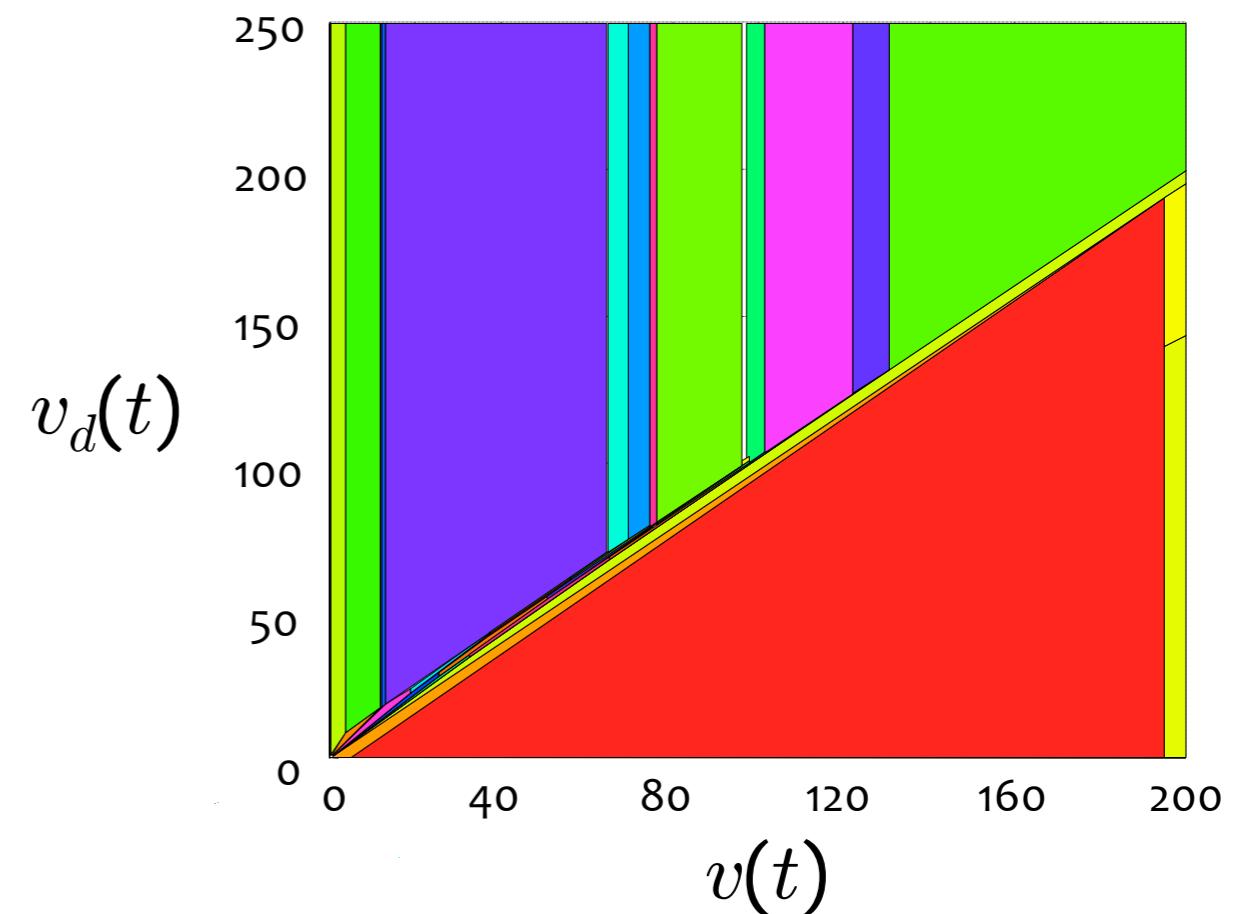
$$\min_{u_t} \quad J(u_t, x(t)) \triangleq |v(t+1|t) - v_d(t)| + \rho|\omega|$$

s.t.

$$\left\{ \begin{array}{l} \text{MLD model} \\ |v(t+1|t) - v(t)| \leq a_{\max} T_s \\ x(t|t) = x(t) \end{array} \right.$$

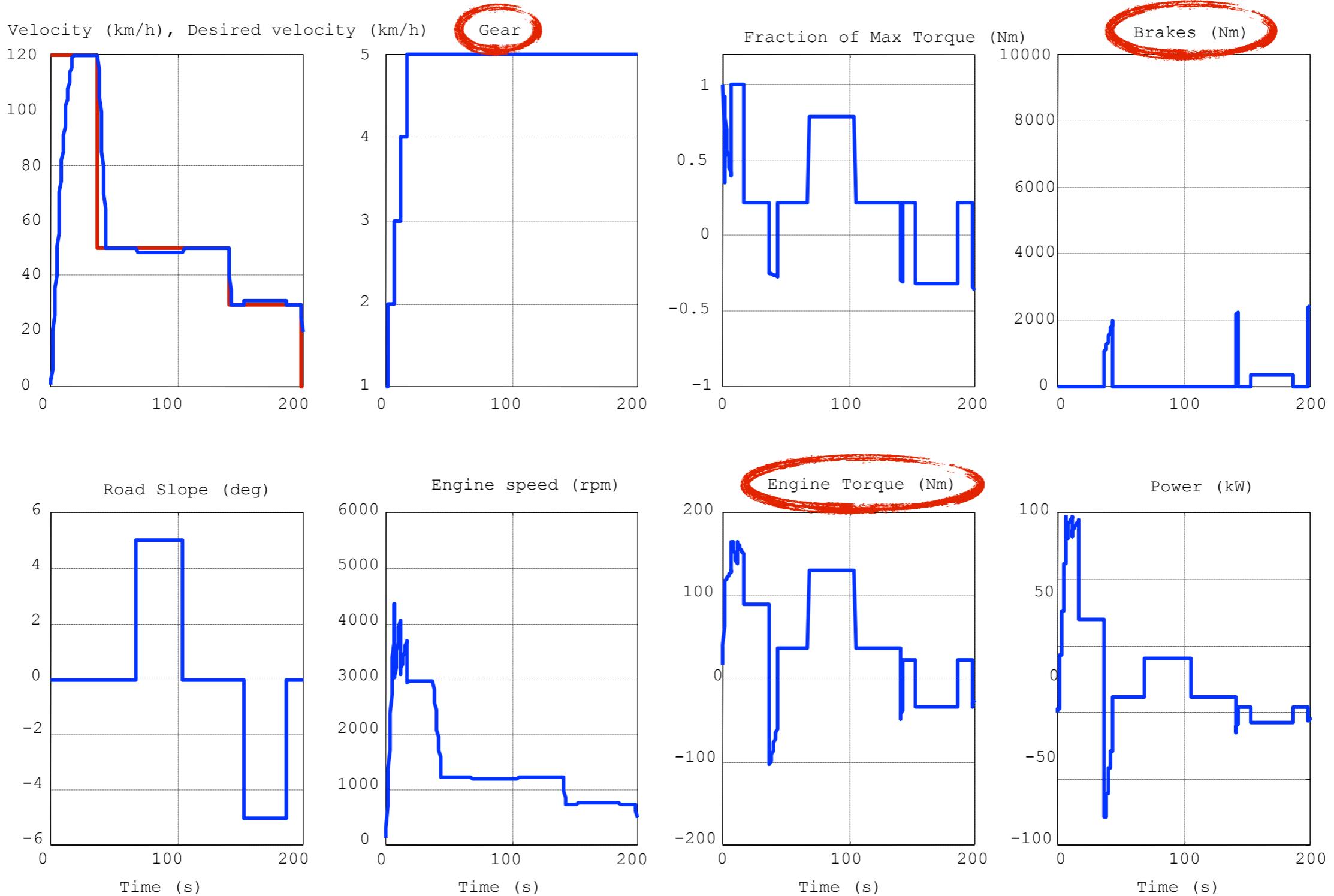
## MILP optimization problem

Linear constraints	100
Continuous variables	19
Binary variables	10
Parameters	2
Time to solve mp-MILP (PC 850Mhz)	47 s
<b>Number of regions</b>	<b>54</b>



# HYBRID CONTROLLER

- Smoother tracking controller



# HYBRID MPC FOR TRACTION CONTROL

F. Borrelli, A. Bemporad, M. Fodor, and D. Hrovat, “[An MPC/Hybrid System Approach to Traction Control](#),” *IEEE Control Syst. Tech.*, vol. 14, n. 3, pp. 541-552, 2006.

# VEHICLE TRACTION CONTROL

Improve driver's ability to control a vehicle under adverse external conditions (wet or icy roads)

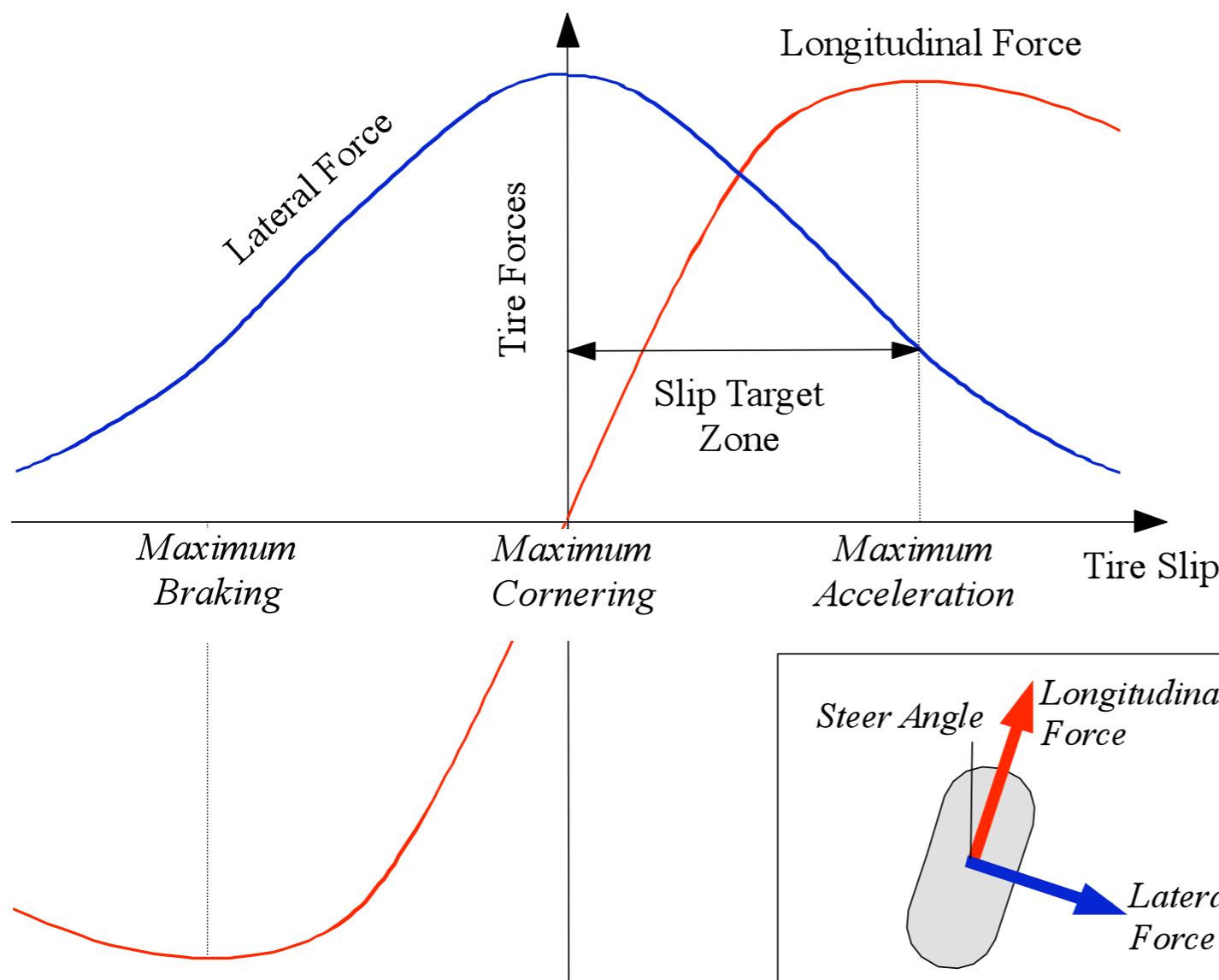


**Model:** nonlinear, uncertain, constraints

**Controller:** suitable for real-time implementation

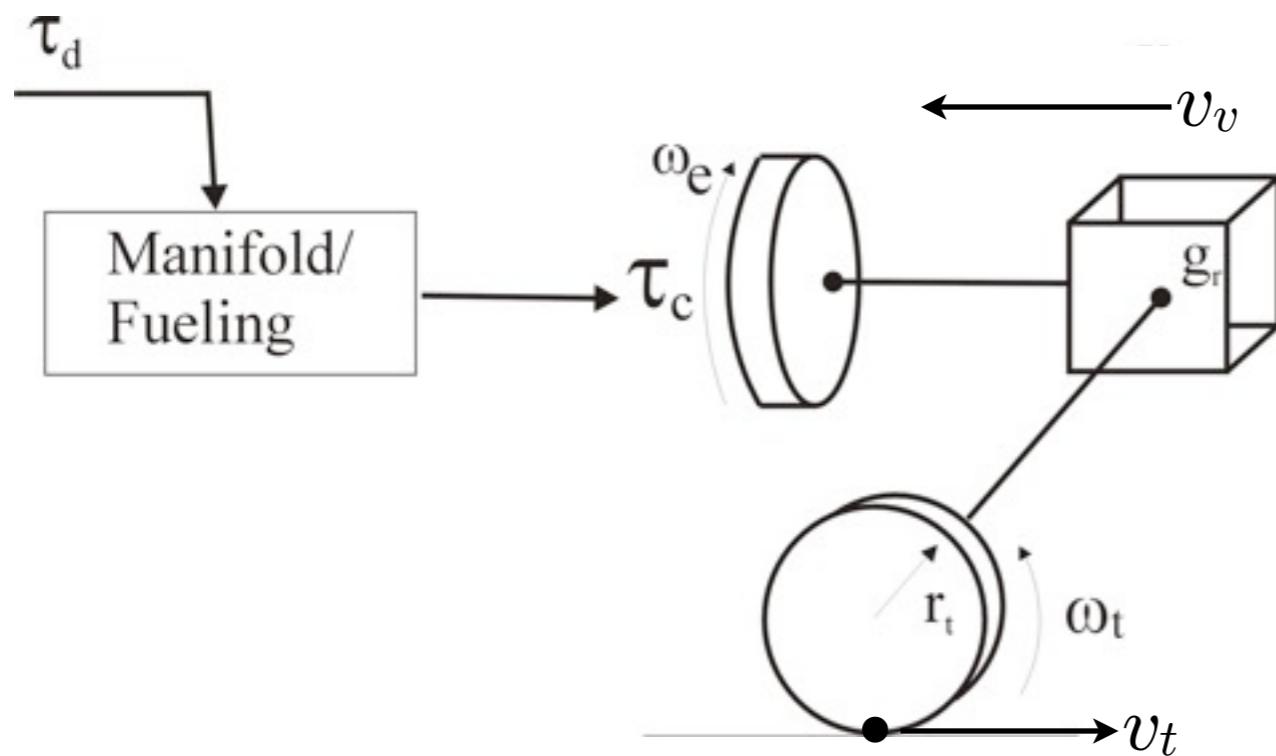
**Solution:** MLD hybrid framework + explicit hybrid MPC strategy

# TIRE FORCE CHARACTERISTICS



# SIMPLE TRACTION MODEL

(Borrelli, Bemporad, Fodor, Hrovat, 2006)



$$v_t = \omega_t r_t = \frac{\omega_e}{g_r} r_t$$

$$\Delta\omega = \frac{1}{r_t}(v_t - v_v) = \frac{\omega_e}{g_r} - \frac{v_v}{r_t} \quad \text{wheel slip}$$

- Mechanical system

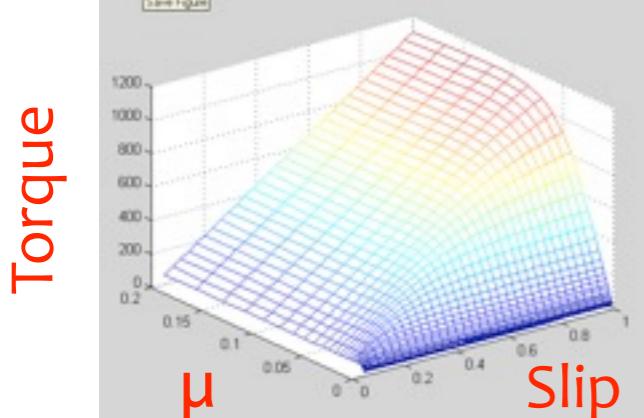
$$\begin{aligned}\dot{\omega}_e &= \frac{1}{J_e} \left( \tau_c - b_e \omega_e - \frac{\tau_t}{g_r} \right) \\ \dot{v}_v &= \frac{\tau_t}{m_v r_t}\end{aligned}$$

- Manifold/fueling dynamics

$$\tau_c = b_i \tau_d(t - \tau_f)$$

- Tire torque  $\tau_t$  is a function of slip  $\Delta\omega$  and road surface adhesion coefficient  $\mu$

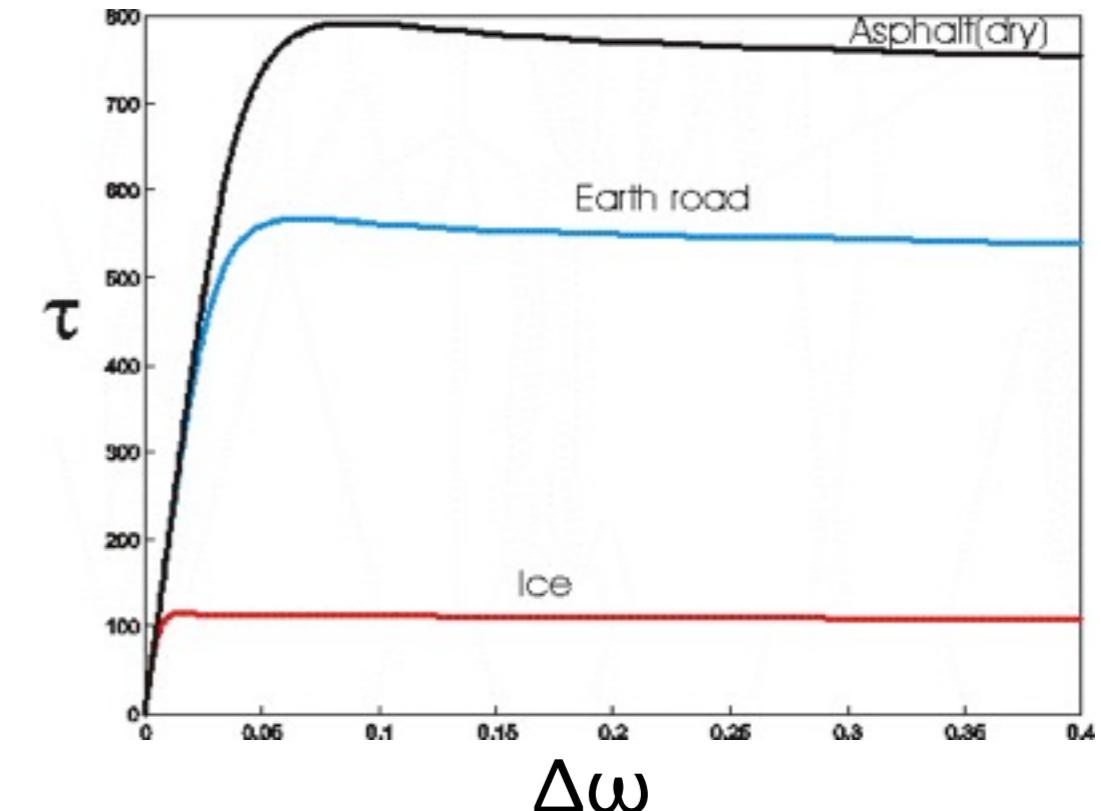
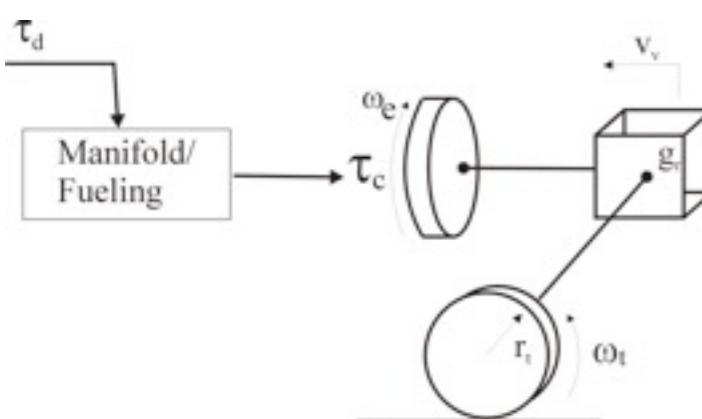
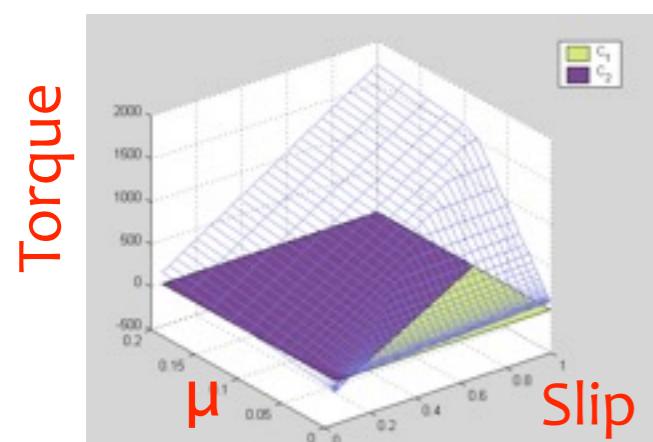
# HYBRID MODEL



Nonlinear tire torque  
 $\tau_t = f(\Delta\omega, \mu)$



PWA approximation



Mixed-Logical  
Dynamical (MLD)  
Hybrid Model  
(discrete time)

# MLD MODEL

$$\begin{aligned}x(t+1) &= Ax(t) + B_1 u(t) + B_2 \delta(t) + B_3 z(t) + B_5 \\y(t) &= Cx(t) + D_1 u(t) + D_2 \delta(t) + D_3 z(t) + D_5 \\E_2 \delta(t) + E_3 z(t) &\leq E_4 x(t) + E_1 u(t) + E_5\end{aligned}$$

state	$x(t) \in \mathbb{R}^4$
input	$u(t) \in \mathbb{R}$
aux. binary	$\delta(t) \in \{0, 1\}$
aux. continuous	$z(t) \in \mathbb{R}^3$

number of mixed-integer inequalities = 14



The MLD matrices are automatically generated in MATLAB format by HYSDEL

# PERFORMANCE AND CONSTRAINTS

- Control objective:

$$\begin{aligned} \min & \quad \sum_{k=0}^N |\Delta\omega(t+k|t) - \Delta\omega_{\text{des}}| \\ \text{s.t.} & \quad \text{MLD dynamics} \end{aligned}$$

- Constraints:

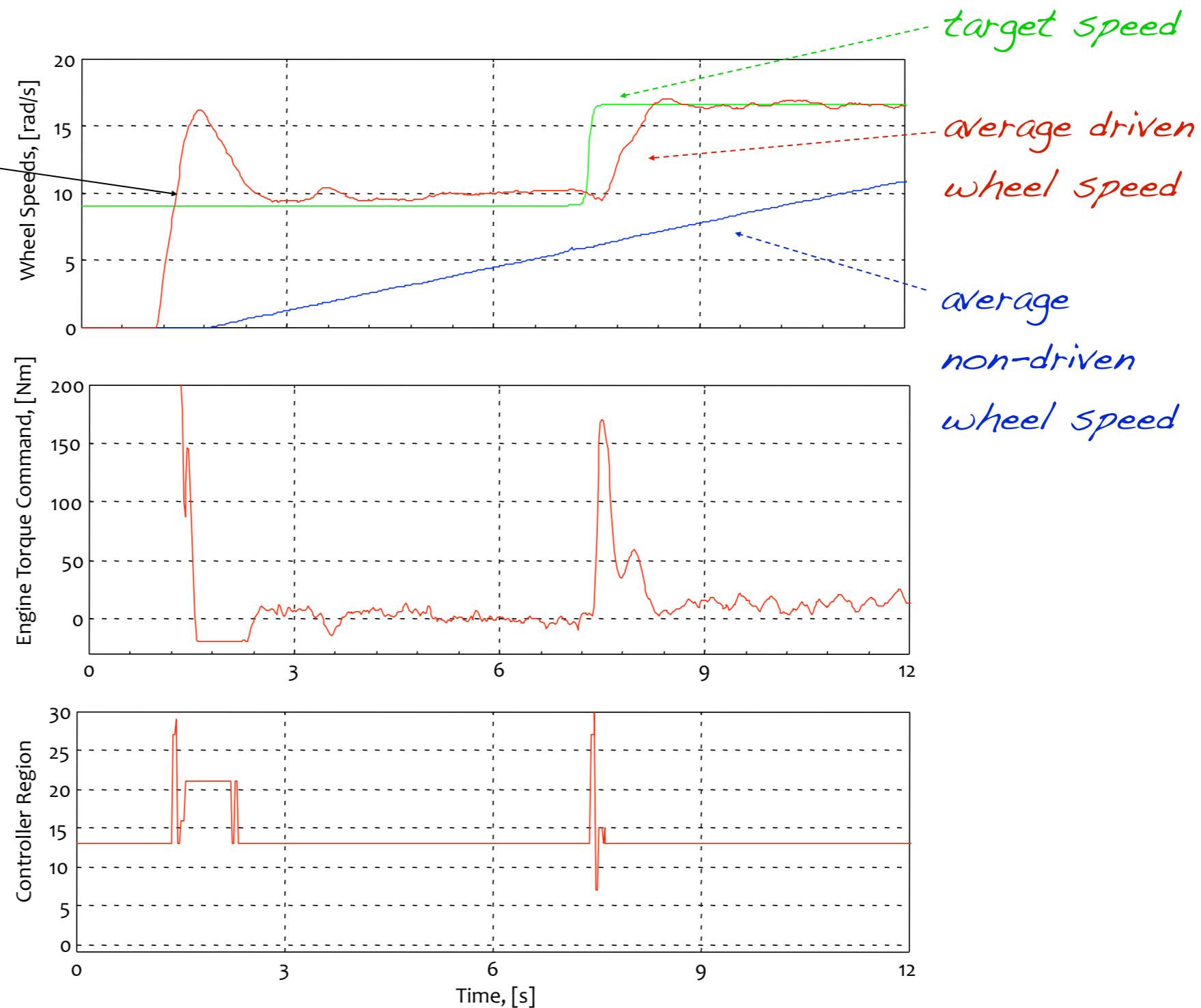
- Limits on the engine torque:

$$-20 \text{ Nm} \leq \tau_d \leq 176 \text{ Nm}$$

# EXPERIMENTAL RESULTS

controller is triggered ON

(250 ms delay from commanded to actual engine torque → initial overspin)



Ford Motor Company

# EXPERIMENTS



indoor ice arena  
 $(\mu \approx 0.2)$

2000 Ford Focus  
2.0l 4-cyl engine  
5-speed manual  
transmission



- 504 regions
- 20ms sampling time
- Pentium 266Mhz +  
Labview

*Ford Motor Company*

# HYBRID CONTROL OF A DISC ENGINE



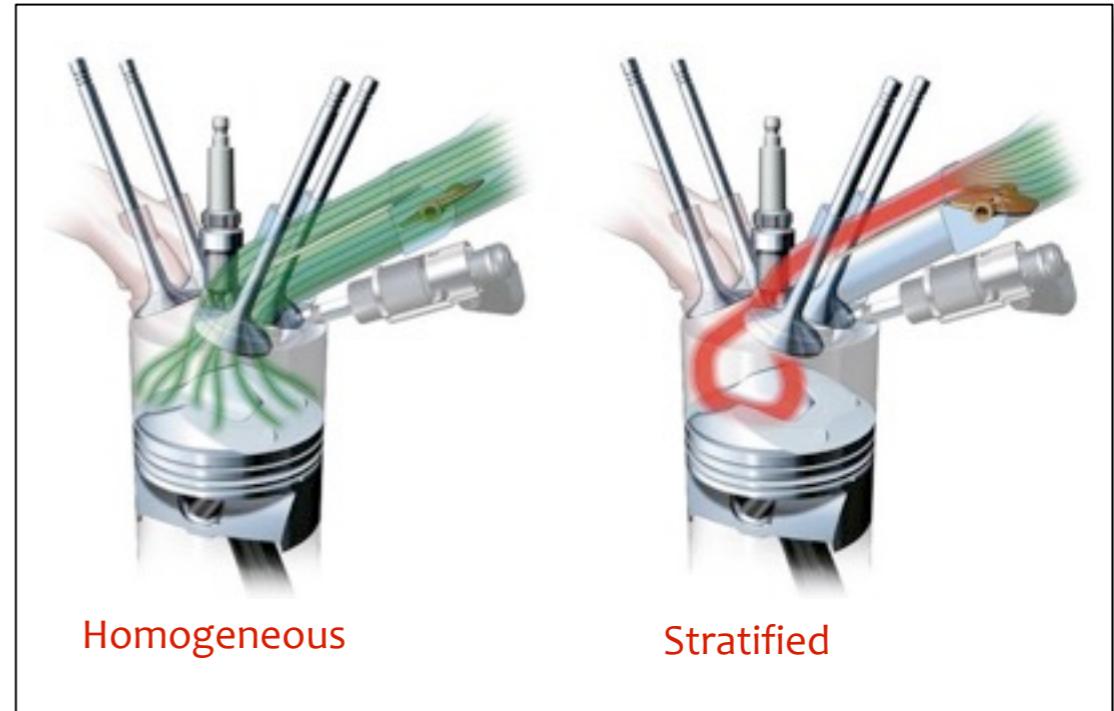
(Photo: Courtesy Mitsubishi)

(N. Giorgetti, G. Ripaccioli, Bemporad, I. Kolmanovsky and D. Hrovat)

# DISC ENGINE CONTROL PROBLEM

**Objective:** develop a controller for a **Direct-Injection Stratified Charge (DISC)** engine that:

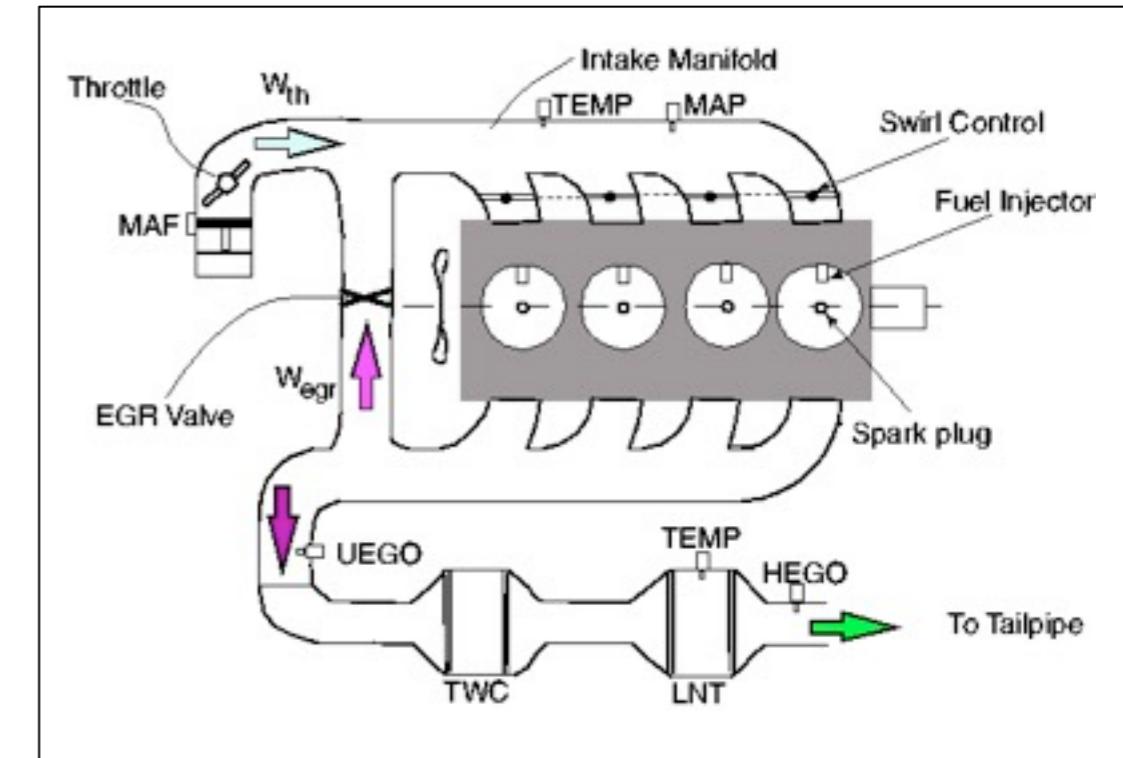
- automatically chooses operating **mode** (homogeneous/stratified)
- can cope with **nonlinear** dynamics
- handles **constraints** on A/F ratio, air-flow, spark
- achieves **optimal** performance (track desired torque and A/F ratio)



# DISC ENGINE

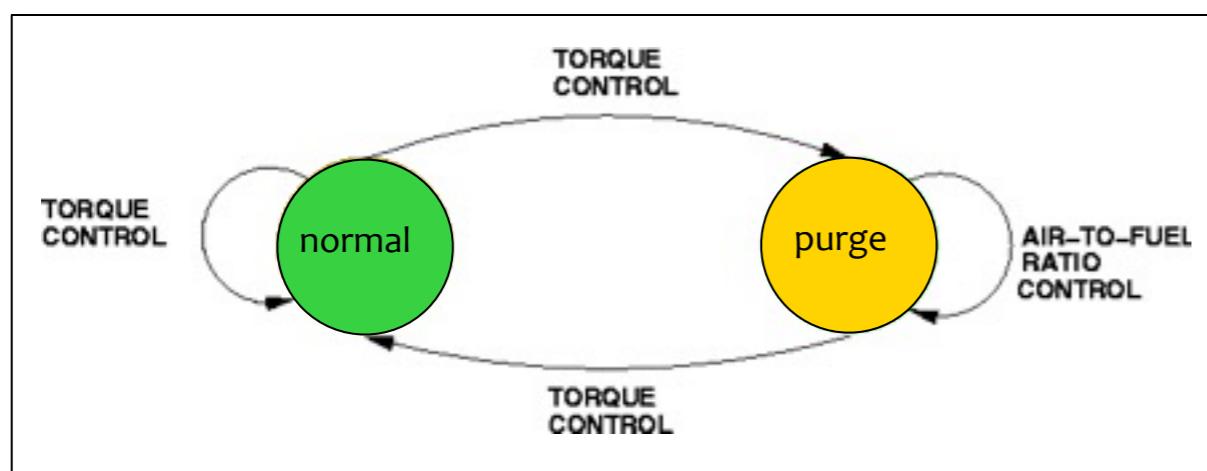
Two distinct regimes:

regime	fuel injection	air-to-fuel ratio
Homogeneous combustion	intake stroke	$\lambda=14.64$
Stratified combustion	compression stroke	$\lambda>14.64$



- Mode is **switched** by changing **fuel injection timing** (late / early)
- Better **fuel economy** during stratified mode

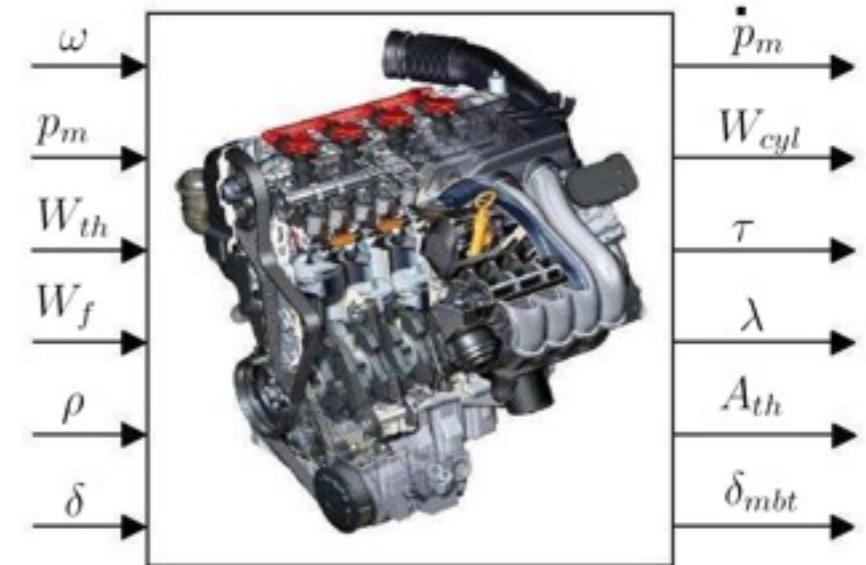
**Periodical cleaning** of the aftertreatment system needed ( $\lambda=14.00$ , homogeneous regime)



the stratified operation  
can only be sustained in a restricted  
part of the engine  
operating range

# DISC ENGINE

- **States:** intake manifold pressure ( $p_m$ )
- **Outputs:** Air-to-fuel ratio ( $\lambda$ ), torque ( $\tau$ ), max-brake-torque spark timing ( $\delta_{mbt}$ )
- **Continuous inputs:** spark advance ( $\delta$ ), air flow ( $W_{th}$ ), fuel flow ( $W_f$ )
- **Binary input:** spark **combustion regime** ( $\rho$ )
- **Disturbance:** engine speed ( $\omega$ ) [measured]
- **Constraints on:**
  - Air-to-fuel ratio (due to engine roughness, misfiring, smoke emiss.)
  - Spark timing (to avoid excessive engine roughness)
  - Mass flow rate on intake manifold (constraints on throttle)

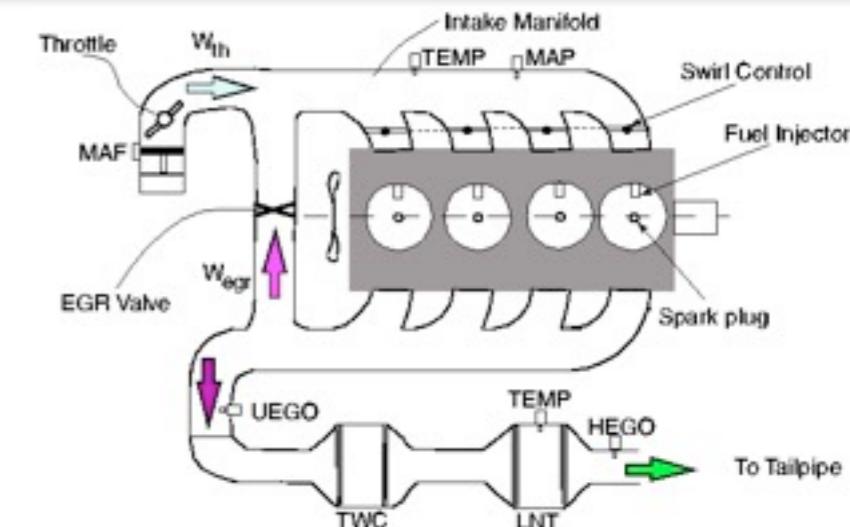


Dynamic equations are **nonlinear**, dynamics and constraints **depend on regime  $\rho$**

# DISC DYNAMICS

Nonlinear model of the engine developed  
and validated at Ford

(Kolmanovsky, Sun, ...)



## Assumptions:

- no EGR (exhaust gas recirculation) rate,
- engine speed=2000 rpm.

- Intake manifold pressure:

$$\dot{p}_m = c_m (W_{th} - W_{cyl}) = c_m (W_{th} - k_{cyl} p_m)$$

- In-cylinder Air-to-Fuel ratio:

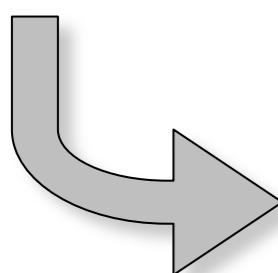
$$\lambda = \frac{W_{cyl}}{W_f} = \frac{k_{cyl} p_m}{W_f}$$

- Engine torque:

$$\tau = \tau_{mfr} + \tau_{pump} + \tau_{ind}$$

with  $\tau_{mfr}, \tau_{pump}$  functions of  $p_m$

$$\tau_{ind} = (\theta_a + \theta_b(\delta - \delta_{mbt})^2) W_f \quad \text{where } \theta_a, \theta_b, \delta_{mbt} \text{ are functions of } \lambda, \delta \text{ and } p_m$$



✓ Good for simulation

✗ Not suitable for optimization-based controller synthesis

# HYBRIDIZATION OF DISC MODEL

DYNAMICS (intake pressure, air-to-fuel ratio, torque):

- Definition of two operating points;
- Linearization of nonlinear dynamics;
- Time discretization of the linear models.



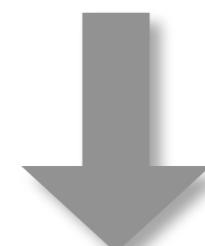
$\rho$ -dependent dynamic equations

CONSTRAINTS on:

- Air-to-Fuel Ratio:  $\lambda_{min}(\rho) \leq \lambda(t) \leq \lambda_{max}(\rho)$
- Mass of air through the throttle:  $0 \leq W_{th} \leq K$
- Spark timing:  $0 \leq \delta(t) \leq \delta_{mbt}(\lambda, \rho)$



$\rho$ -dependent constraints



Hybrid system with 2 modes (switching affine system)

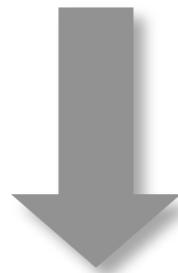
# INTEGRAL ACTION

Integrators on torque error and air-to-fuel ratio error are added to obtain zero offsets in steady-state:

$$\begin{aligned}\epsilon_{\tau,k+1} &= \epsilon_{\tau,k} + T_s(\tau_{\text{ref}}(t) - \tau_k) \\ \epsilon_{\lambda,k+1} &= \epsilon_{\lambda,k} + T_s(\lambda_{\text{ref}}(t) - \lambda_k)\end{aligned}$$

$T_s$  = sampling time

$\tau_{\text{ref}}, \lambda_{\text{ref}}$  = references on brake torque and air-to-fuel ratio



Simulation based on nonlinear model confirms zero offsets in steady-state  
(despite the model mismatch)

# MPC OF DISC ENGINE

$$\min_{\xi} J(\xi, x(t)) = \sum_{k=0}^{N-1} u_k' R u_k + y_k' Q y_k + x_{k+1}' S x_{k+1}$$

subj. to  $\begin{cases} x_0 = x(t), \\ \text{hybrid model} \end{cases}$

$N$ = control horizon

$x(t)$  = current state

$$\xi = [u_0', \gamma_0', z_0', \dots, u_{N-1}', \gamma_{N-1}', z_{N-1}']'$$

where:

$u_k$	$=$	$[W_{th,k} - W_{th,ref}, \ W_{f,k} - W_{f,ref}, \ \delta_k - \delta_{ref}, \ \rho_k - \rho_{ref}]'$
$y_k$	$=$	$[\tau_k - \tau_{ref}, \ \lambda_k - \lambda_{ref}, \ \delta_{mbt} - \delta_k - \Delta\delta_{ref}]'$
$x_k$	$=$	$[p_{m,k} - p_{m,ref}, \ \epsilon_{\tau,k}, \ \epsilon_{\lambda,k}]'$

and:

$$R = \begin{pmatrix} r_{W_{th}} & 0 & 0 & 0 \\ 0 & r_{W_f} & 0 & 0 \\ 0 & 0 & r_{\delta} & 0 \\ 0 & 0 & 0 & r_{\rho} \end{pmatrix} \quad Q = \begin{pmatrix} q_{\tau} & 0 & 0 \\ 0 & q_{\lambda} & 0 \\ 0 & 0 & q_{\Delta\delta} \end{pmatrix} \quad S = \begin{pmatrix} s_{p_m} & 0 & 0 \\ 0 & s_{\epsilon_{\tau}} & 0 \\ 0 & 0 & s_{\epsilon_{\lambda}} \end{pmatrix}$$

Reference values are automatically generated from  $\tau_{ref}$  and  $\lambda_{ref}$  by numerical computations based on the nonlinear model

# DISC ENGINE - HYSDEL LIST

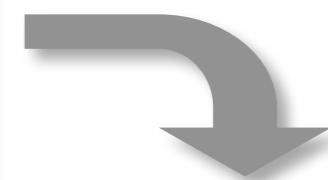
```
SYSTEM hysdisc{
  INTERFACE{
    STATE{
      REAL pm      [1, 101.325];
      REAL xtau    [-1e3, 1e3];
      REAL xlam    [-1e3, 1e3];
      REAL taud    [0, 100];
      REAL lamd    [10, 60];
    }
    OUTPUT{
      REAL lambda, tau, ddelta;
    }
    INPUT{
      REAL Wth      [0, 38.5218];
      REAL Wf       [0, 2];
      REAL delta    [0, 40];
      BOOL rho;
    }
    PARAMETER{
      REAL Ts, pm1, pm2;
      ...
    }
  }

  IMPLEMENTATION{
    AUX{
      REAL lam,taul,dmbtl,lmin,lmax;
    }
    DA{
      lam={IF rho THEN l11*pm+l12*Wth...
            +l13*Wf+l14*delta+l1c
        ELSE    101*pm+102*Wth+103*Wf...
            +104*delta+10c    };
      taul={IF rho THEN tau11*pm+...
            tau12*Wth+tau13*Wf+tau14*delta+tau1c
        ELSE    tau01*pm+tau02*Wth...
            +tau03*Wf+tau04*delta+tau0c } ;
      dmbtl ={IF rho THEN dmbt11*pm+dmbt12*Wth...
            +dmbt13*Wf+dmbt14*delta+dmbt1c+7
        ELSE    dmbt01*pm+dmbt02*Wth...
            +dmbt03*Wf+dmbt04*delta+dmbt0c-1} ;
      lmin ={IF rho THEN 13 ELSE 19};
      lmax ={IF rho THEN 21 ELSE 38};
    }
    CONTINUOUS{
      pm=pm1*pm+pm2*Wth;
      xtau=xtau+Ts*(taud-taul);
      xlam=xlam+Ts*(lamd-lam);
      taud=taud; lamd=lamd;
    }
    OUTPUT{
      lambda=lam-lamd;
      tau=taul-taud;
      ddelta=dmbtl-delta;
    }
    MUST{
      lmin-lam    <=0;
      lam-lmax    <=0;
      delta-dmbtl <=0;
    }
  }
}
```

# MPC – TORQUE CONTROL MODE

$$\min \sum_{k=0}^{N-1} (u_k - u_r)' R (u_k - u_r) + (y_k - y_r)' Q (y_k - y_r) \\ + (x_{k+1} - x_r)' S (x_{k+1} - x_r)$$

subj. to  $\begin{cases} x_0 = x(t), \\ \text{hybrid model} \end{cases}$



Solve  
**MIQP problem**  
to compute  $u(t)$

$$u(t) = [W_{th}(t), W_f(t), \delta(t), \rho(t)]$$

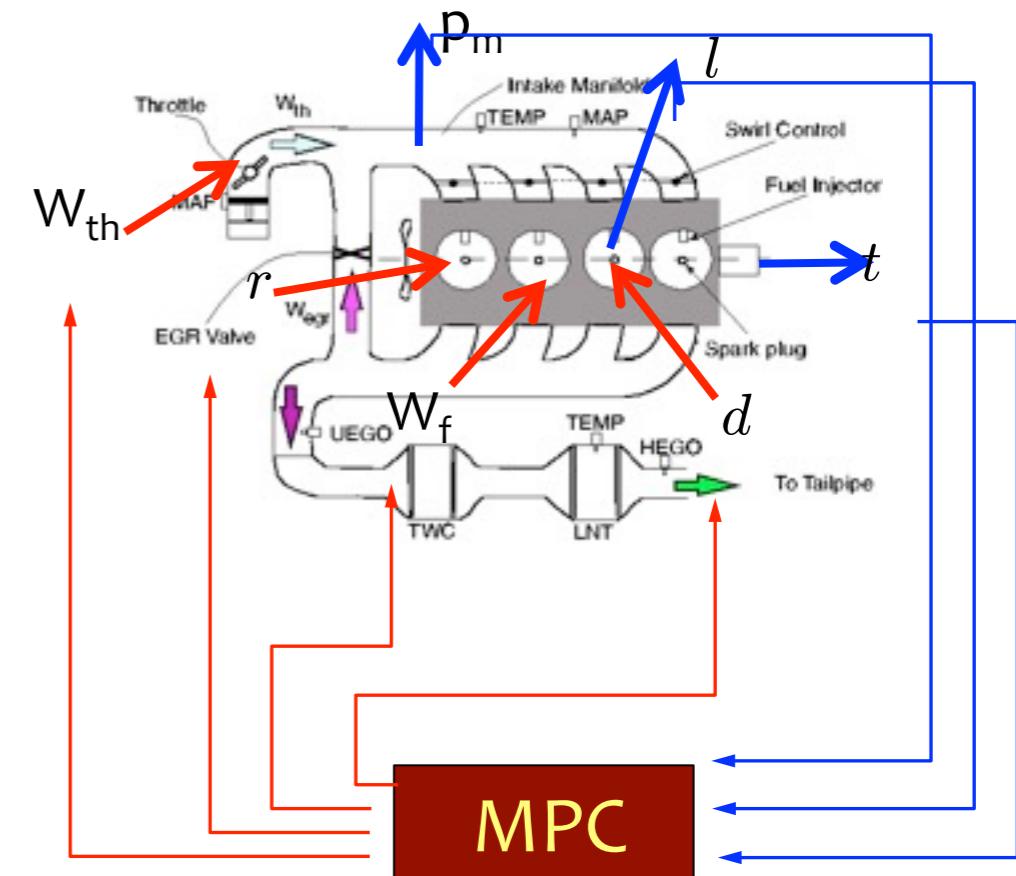
Weights:

$$R = \begin{pmatrix} 0.01 & 0 & 0 & 0 \\ 0 & 0.001 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{pmatrix} \quad r_\rho \quad (\text{prevents unneeded chattering})$$

$$Q = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0.001 & 0 \\ 0 & 0 & 0.01 \end{pmatrix} \quad S = \begin{pmatrix} 0.04 & 0 & 0 \\ 0 & 1500 & 0 \\ 0 & 0 & 0.01 \end{pmatrix}$$

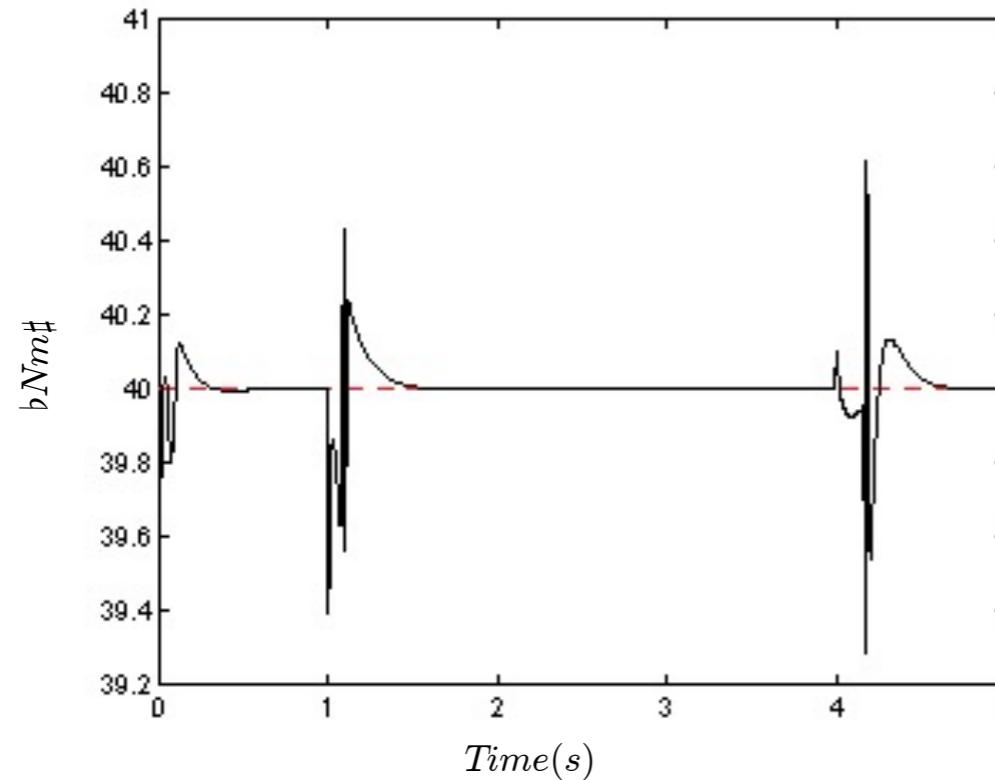
*main emphasis on torque*

$q_T$        $q_\lambda$        $s_{\varepsilon_T}$        $s_{\varepsilon_\lambda}$



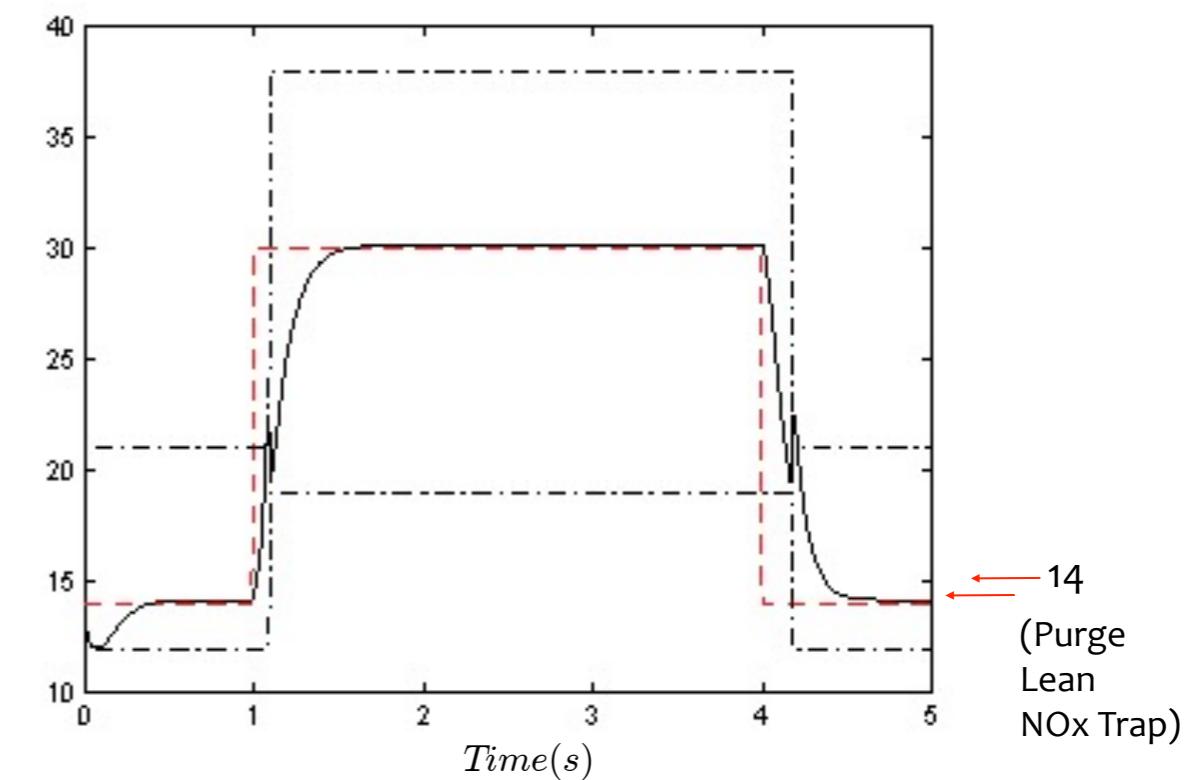
# SIMULATION RESULTS (NOMINAL ENGINE SPEED)

Engine brake torque

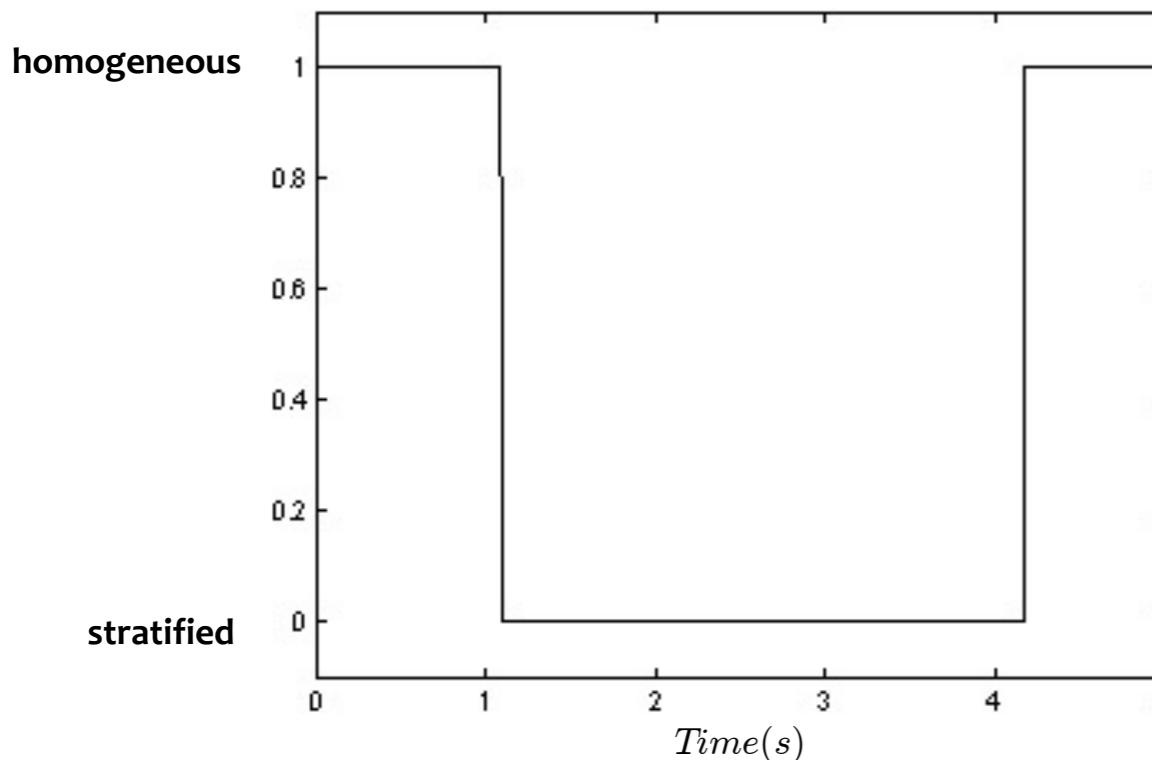


$\omega/2000 \text{ rpm}$

Air-to-fuel ratio



Combustion mode



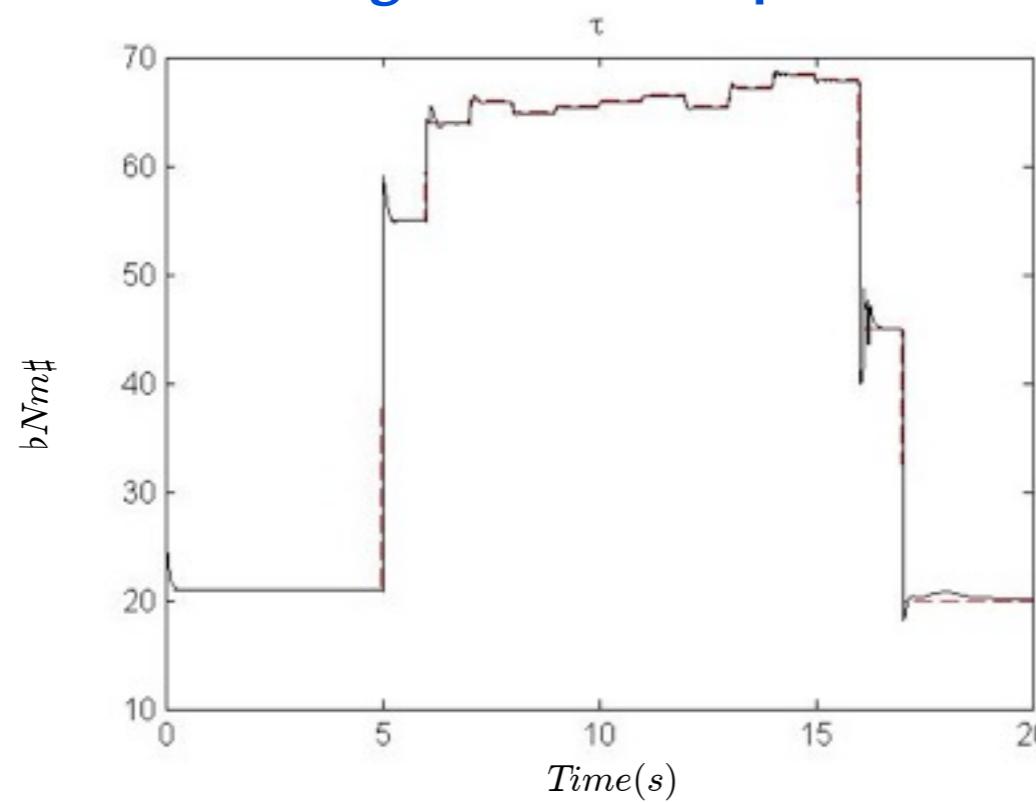
- Control horizon  $N=1$ ;
- Sampling time  $T_s=10 \text{ ms}$ ;
- PC Xeon 2.8 GHz + Cplex 9.1



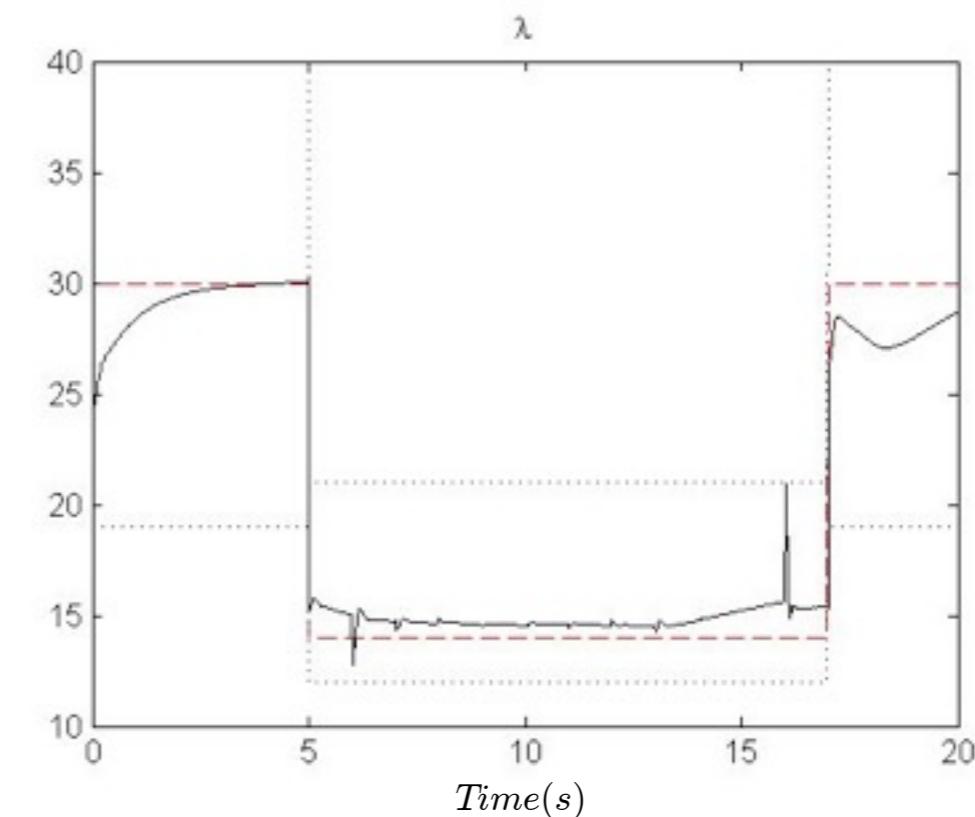
$\approx 3 \text{ ms per time step}$

# SIMULATION RESULTS (VARYING ENGINE SPEED)

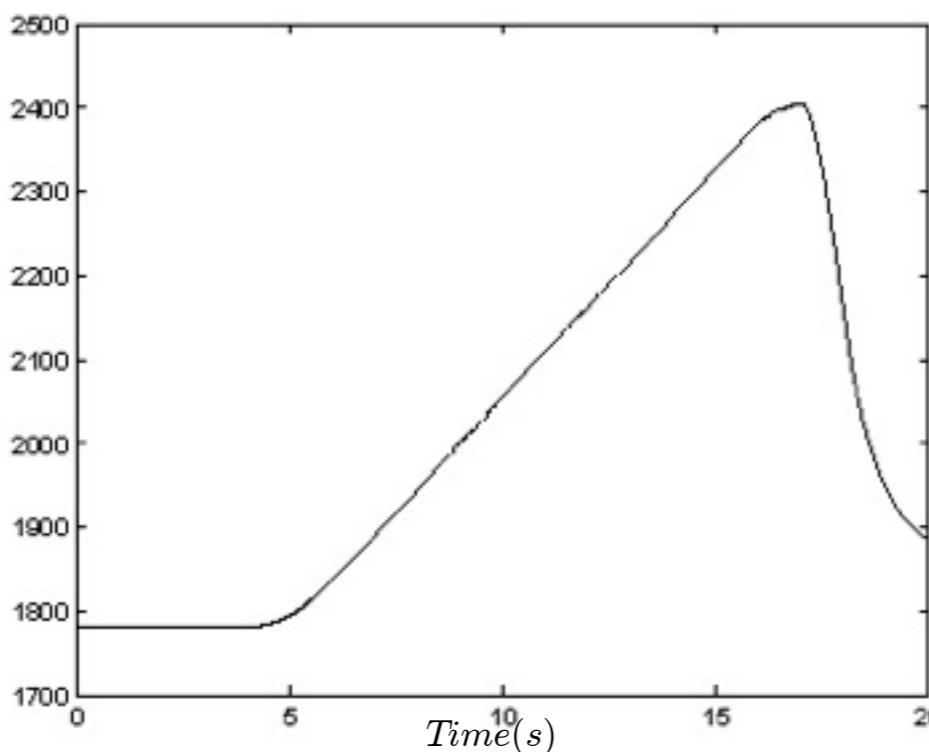
Engine Brake Torque



Air-to-Fuel Ratio



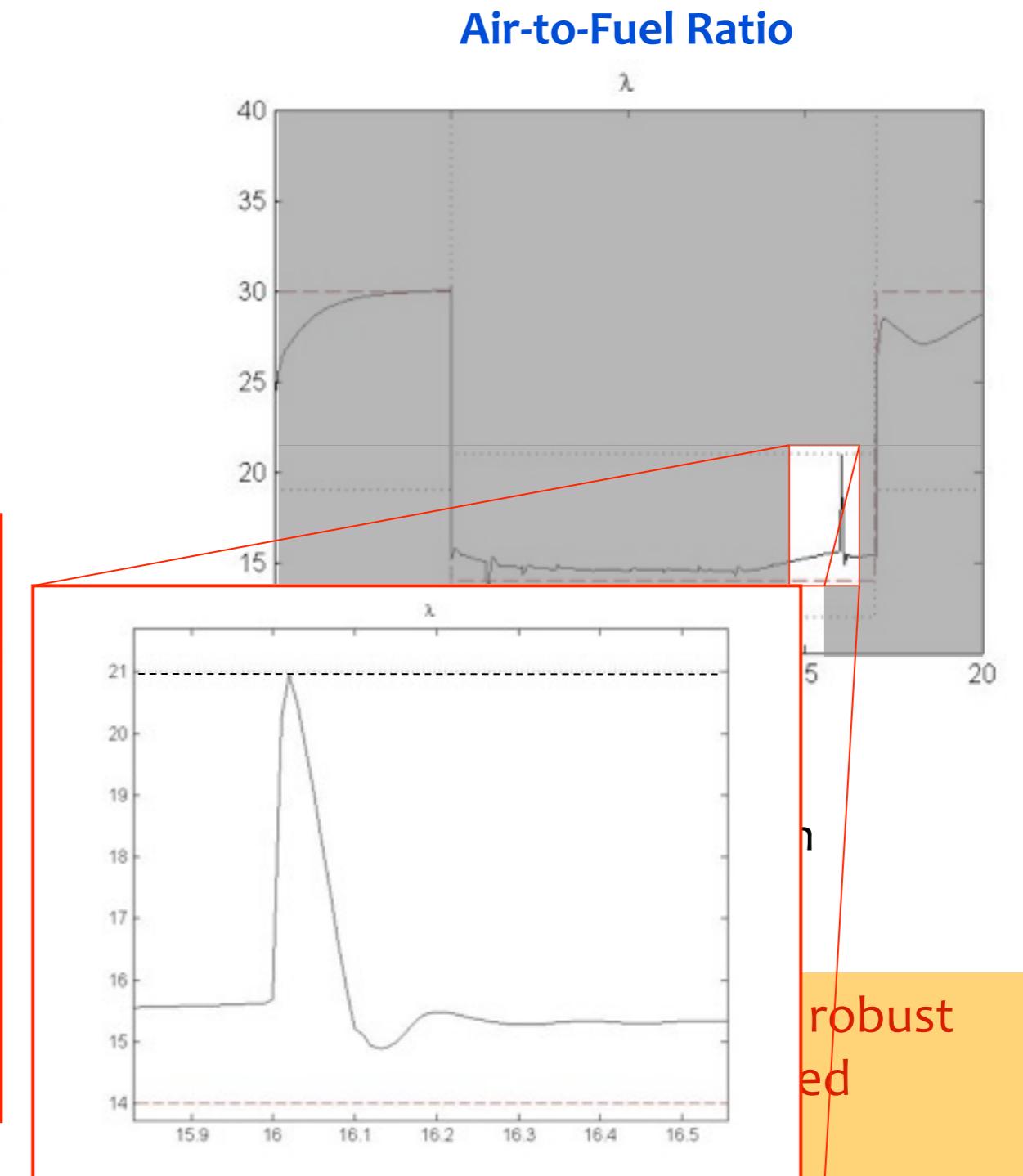
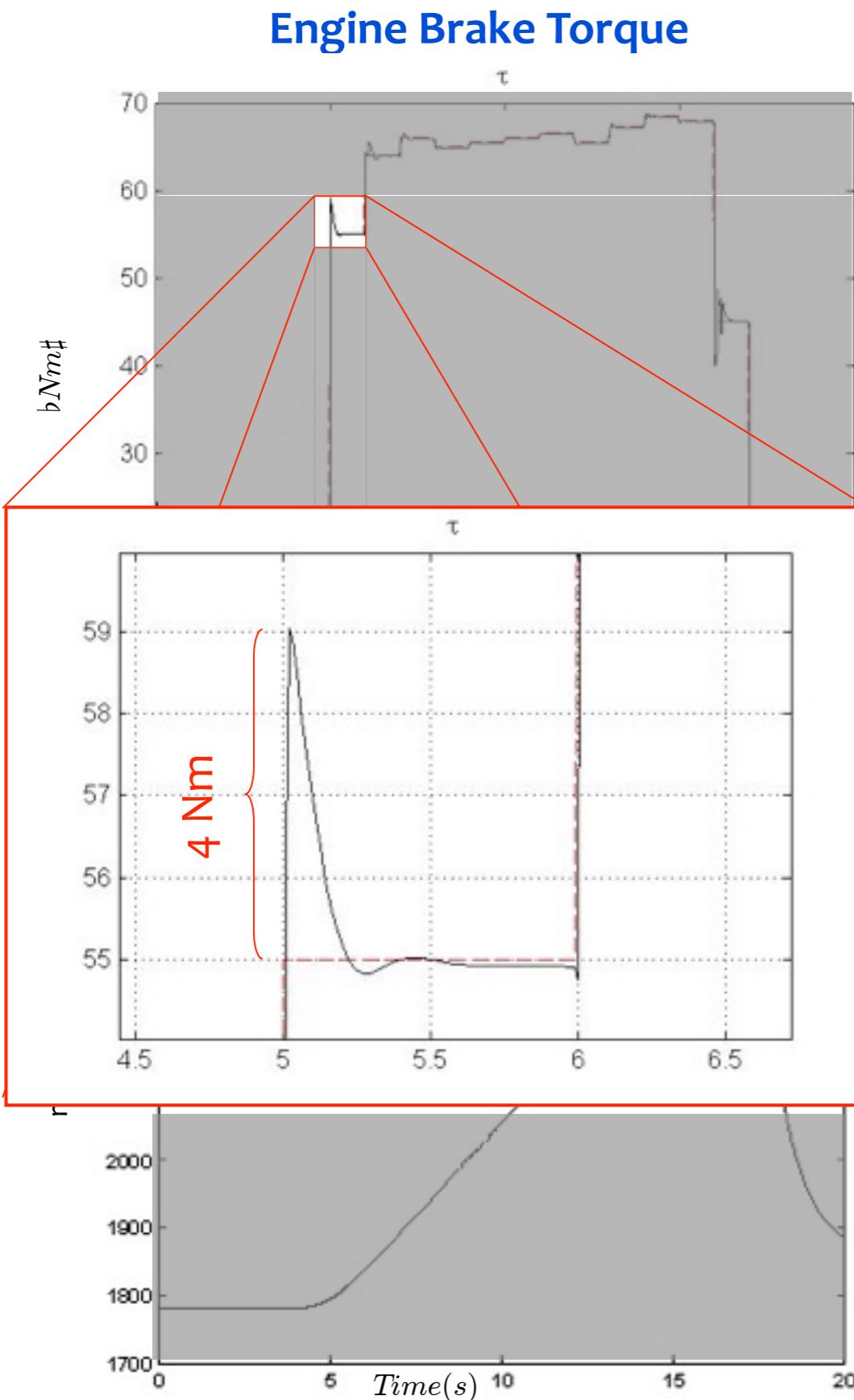
Engine speed



20 s segment of the European drive cycle (NEDC)

Hybrid MPC design is quite robust with respect to engine speed variations

# SIMULATION RESULTS (VARYING ENGINE SPEED)



Control code too complex  
(MIQP) → not implementable !

# EXPLICIT MPC CONTROLLER

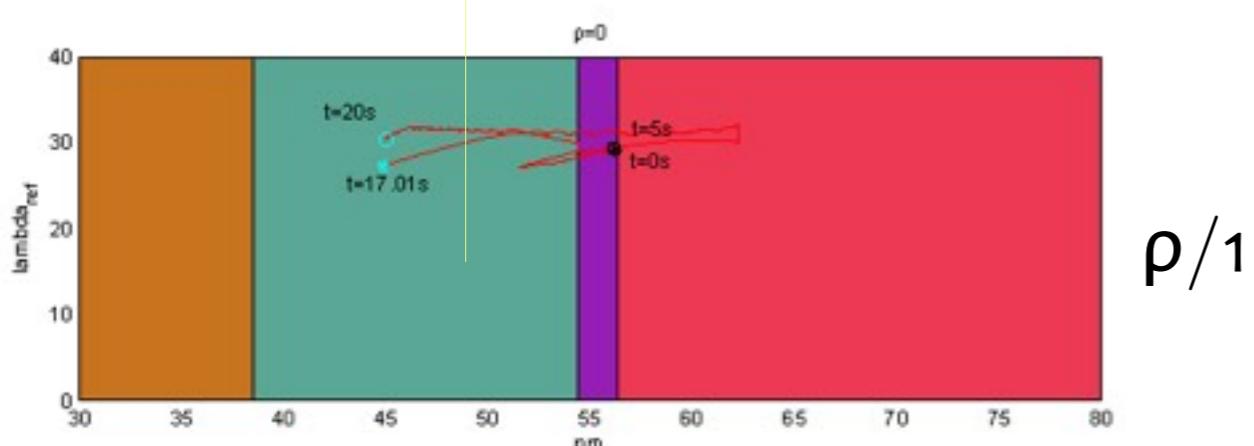
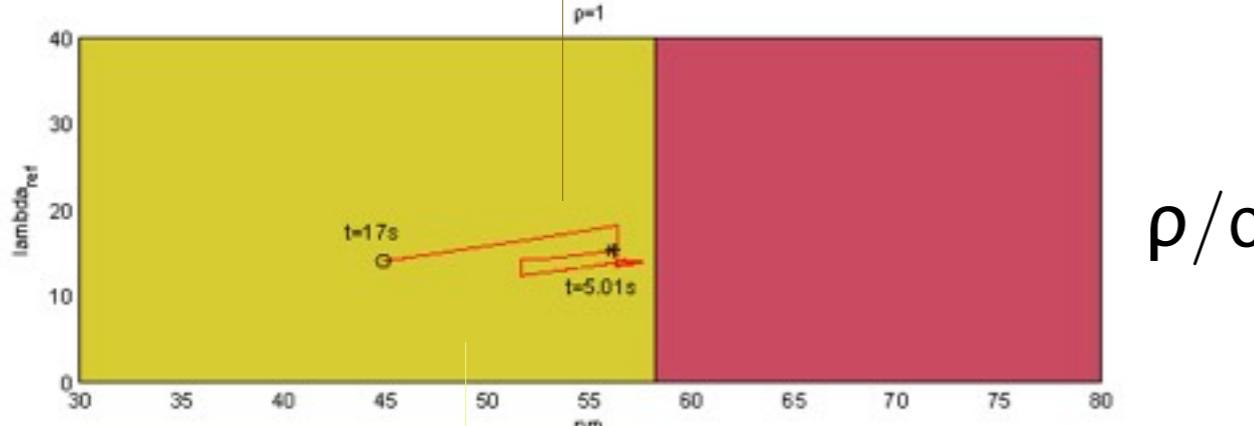
Explicit control law:

$$u(t) = f(\theta(t))$$

where:  $u = [W_{th} \ W_f \ \delta \ \rho]'$

$$\theta = [p_m \ \epsilon_\tau \ \epsilon_\lambda \ \tau_{ref} \ \lambda_{ref} \\ p_{m,ref} \ W_{th,ref} \ W_{f,ref} \ \delta_{ref}]'$$

Cross-section by the  $\tau_{ref}$ - $\lambda_{ref}$  plane



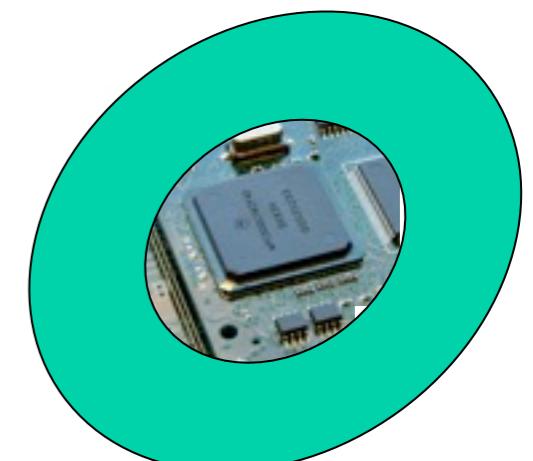
$N=1$  (control horizon)

42 partitions

- Time to compute explicit MPC:  
 $\approx 3s$ ;
  - Sampling time  $T_s=10$  ms;
  - PC Xeon 2.8 GHz + Cplex 9.1
- 8  $\mu$ s per time step

≈3ms on

$\mu$ -controller  
Motorola  
MPC 555  
43kb RAM  
(custom made for Ford)



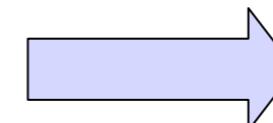
# EXPLICIT MPC CONTROLLER (N=2)

Explicit control law:

$$u(t) = f(\theta(t))$$

N=2 (control horizon)

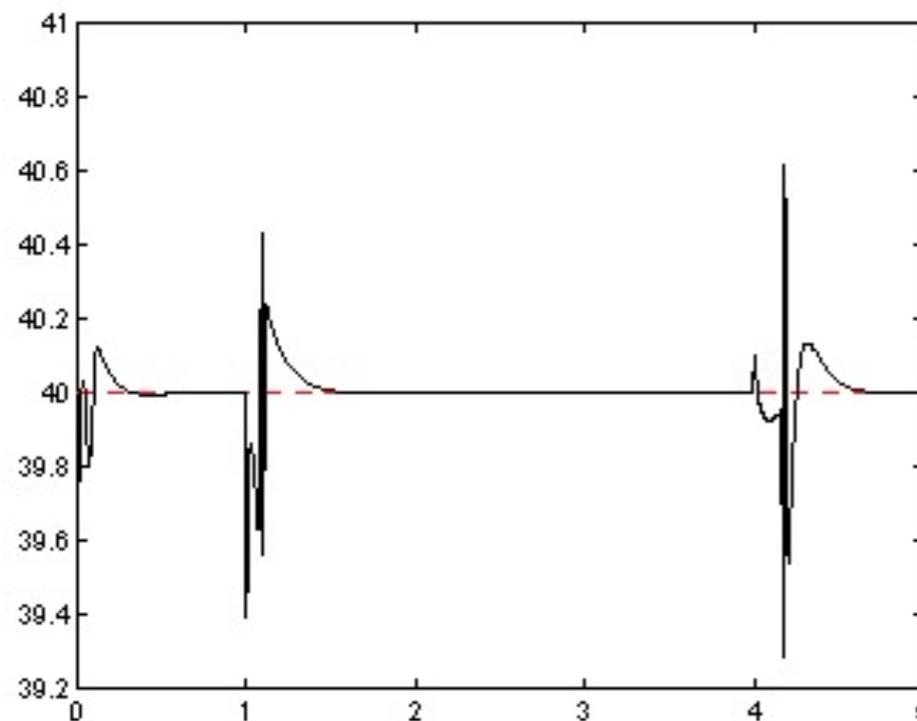
where:  $u = [W_{th} \ W_f \ \delta \ \rho]'$



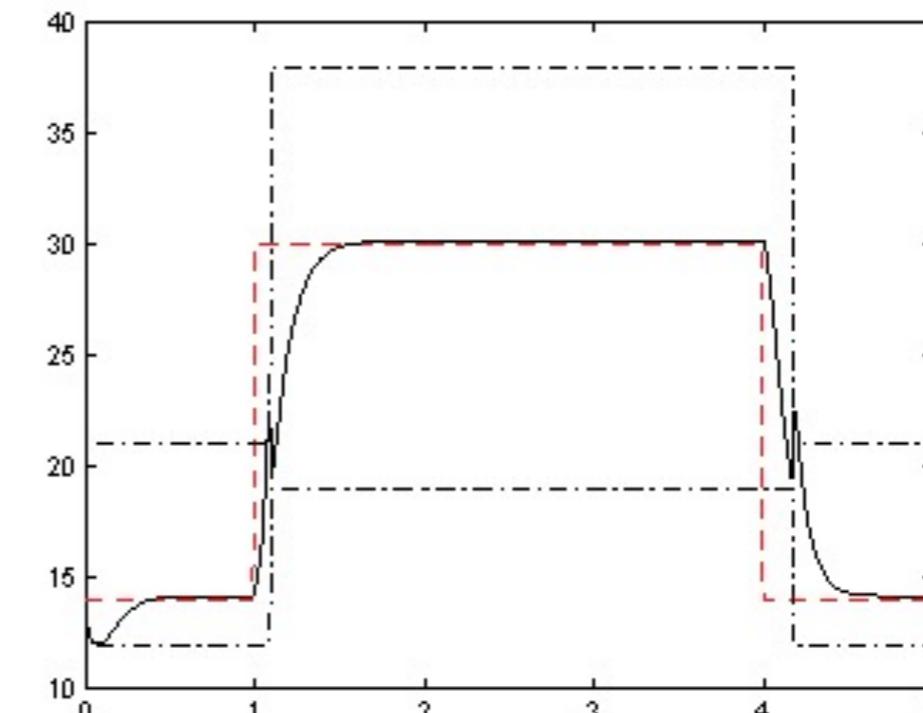
747 partitions

$$\begin{aligned}\theta = & [p_m \ \epsilon_\tau \ \epsilon_\lambda \ \tau_{\text{ref}} \ \lambda_{\text{ref}} \\ & p_{m,\text{ref}} \ W_{th,\text{ref}} \ W_{f,\text{ref}} \ \delta_{\text{ref}}]'\end{aligned}$$

Engine Brake Torque



Air-to-Fuel Ratio



Closed-loop N=2

Closed-loop N=1

adequate !

# EXPLICIT HYBRID MPC OF SEMIACTIVE SUSPENSIONS

N. Giorgetti, A. Bemporad, H. E. Tseng, and D. Hrovat, “*Hybrid model predictive control application towards optimal semi-active suspension*,” International Journal of Control, vol. 79, no. 5, pp. 521–533, 2006.

# ACTIVE SUSPENSIONS

Active Suspension System  
Ford Mercur XR 40i



active  
suspensions

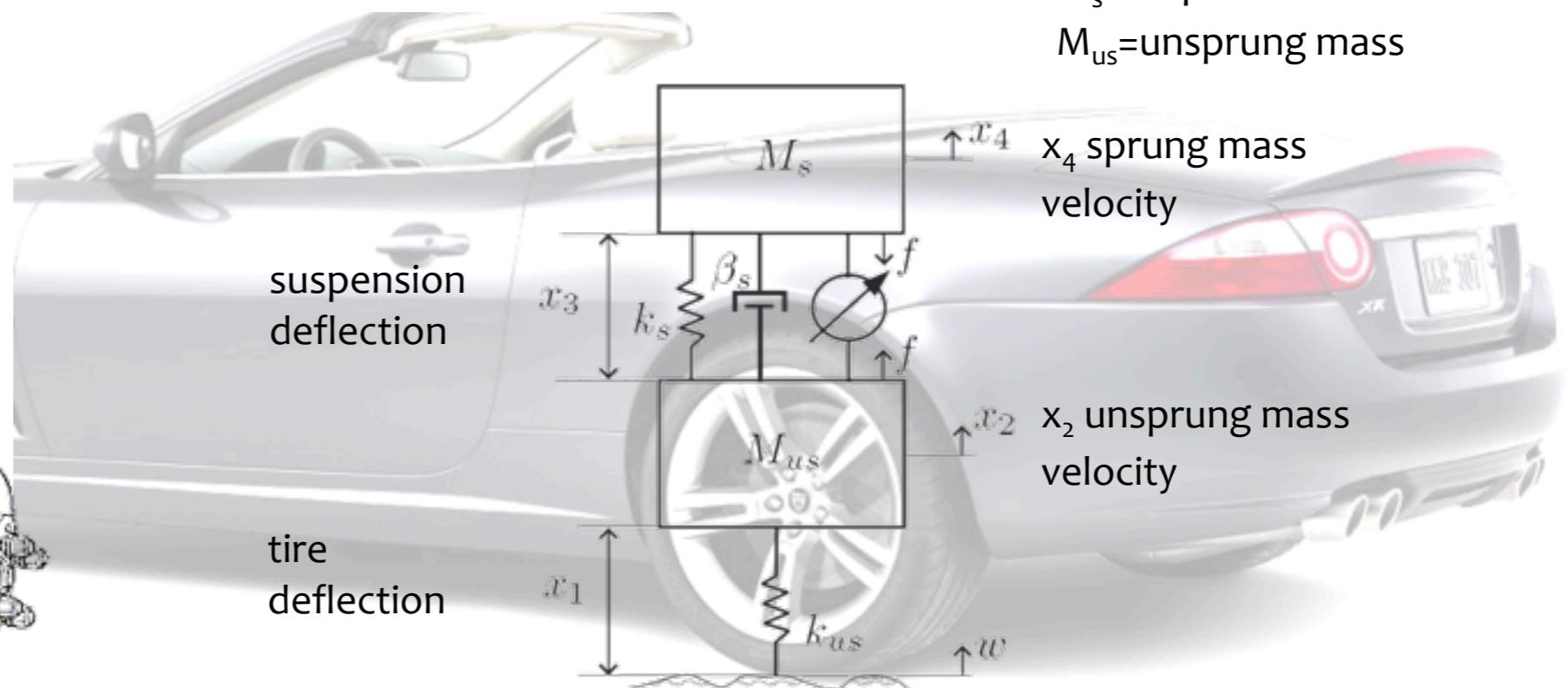
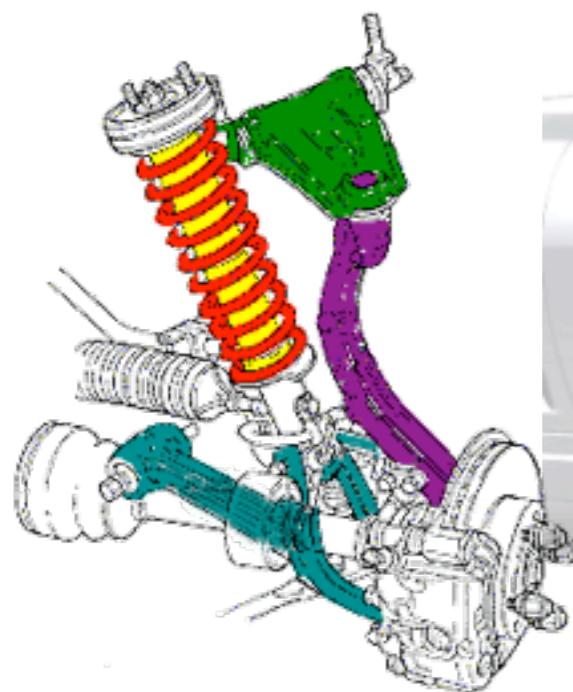


passive  
suspensions



Ford Motor Company

# QUEST OF OPTIMAL SEMI-ACTIVE SUSPENSIONS



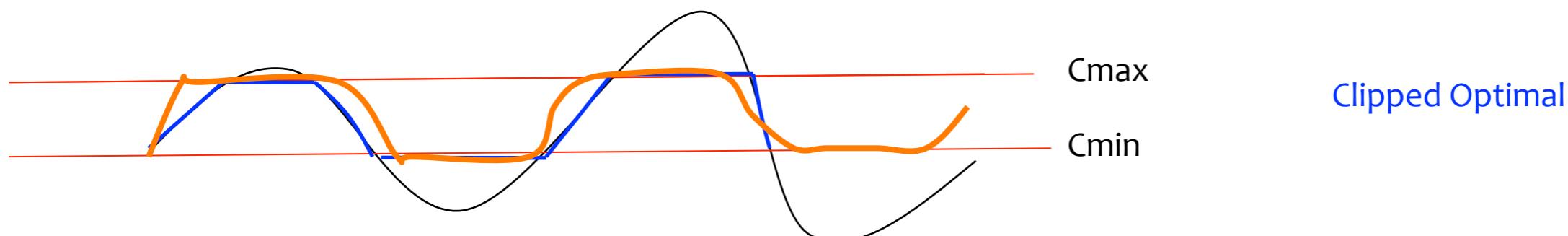
$M_s$ =suspended mass

$M_{us}$ =unsprung mass

$x_4$  sprung mass  
velocity

$x_2$  unsprung mass  
velocity

For Semi-Active with Variable Damping,  $f(x)=C^*(x_4-x_2)$



—  $C=f(x)/(x_4-x_2)$ , where  $f(x)$  is the optimal active suspension force

—  $C=\text{sat}[f(x)/(x_4-x_2)]$

— Optimal

— ? — = —

Clipped Optimal

# MODEL

- State-space model

$$\dot{x} = Ax + B\bar{f} + B_w w$$

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 \\ -\omega_{us}^2 & -2\rho\zeta\omega_s & \rho\omega_s^2 & 2\rho\zeta\omega_s \\ 0 & -1 & 0 & 1 \\ 0 & 2\zeta\omega_s & -\omega_s^2 & -2\zeta\omega_s \end{bmatrix}$$

$$B = \begin{bmatrix} 0 \\ \rho \\ 0 \\ -1 \end{bmatrix} \quad B_w = \begin{bmatrix} -1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$x_1$  = tire deflection from equilibrium  
 $x_2$  = unsprung mass velocity  
 $x_3$  = suspension deflection from equilibrium  
 $x_4$  = sprung mass velocity  
 $\bar{f}$  = normalized adjustable force  
 $w$  = road velocity disturbance

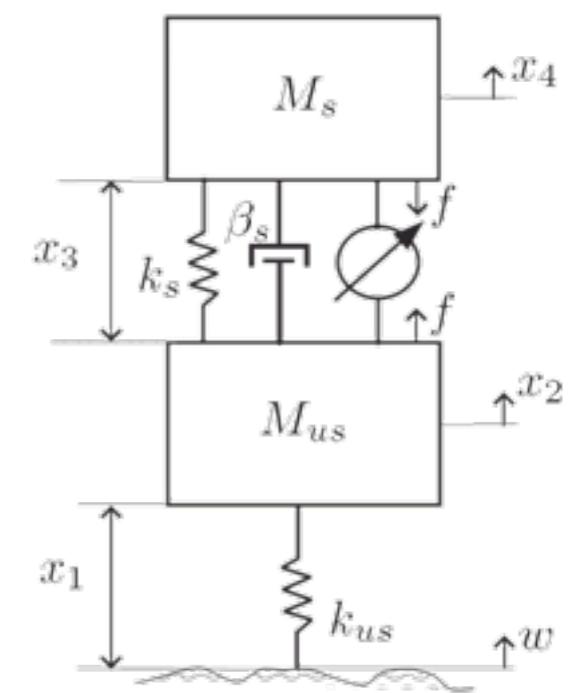
$$\rho = \frac{M_s}{M_{us}}, \quad \omega_{us} = \sqrt{\frac{k_{us}}{M_{us}}}, \quad \omega_s = \sqrt{\frac{k_s}{M_s}}, \quad \zeta = \frac{\beta_s}{2\sqrt{M_s k_s}}, \quad \bar{f} = \frac{f}{M_s}$$

- Output:

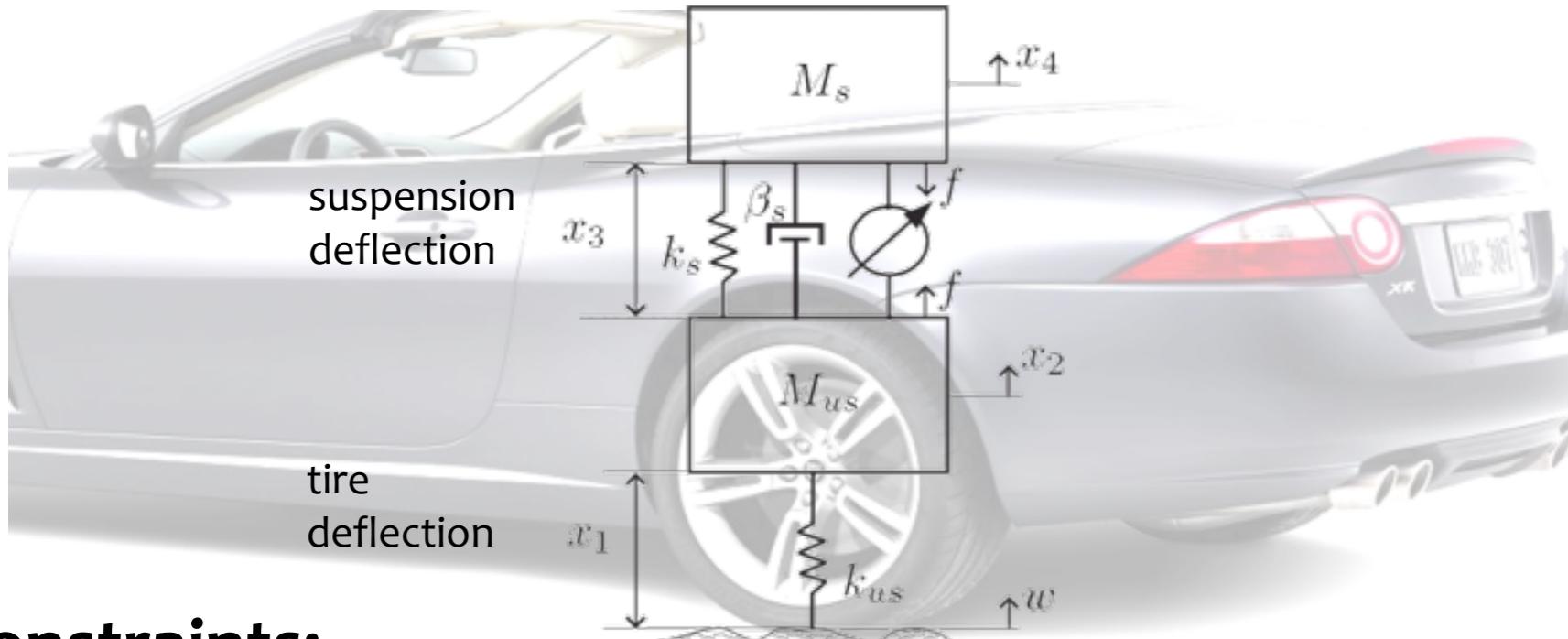
$$y = \frac{dx_4}{dt} = [0 \ 2\zeta\omega_s \ -\omega_s^2 \ -2\zeta\omega_s] x - \bar{f}$$

$$\begin{aligned}
 \bullet \text{ Cost: } J &= \int (q_{x_1} x_1^2 + q_{x_3} x_3^2 + \dot{x}_4^2) dt \\
 &= \int (x' Q x + \dot{x}_4^2) dt
 \end{aligned}$$

$$\bullet \text{ Time-discretization: } T_s = 10 \text{ ms}$$



# CONSTRAINTS

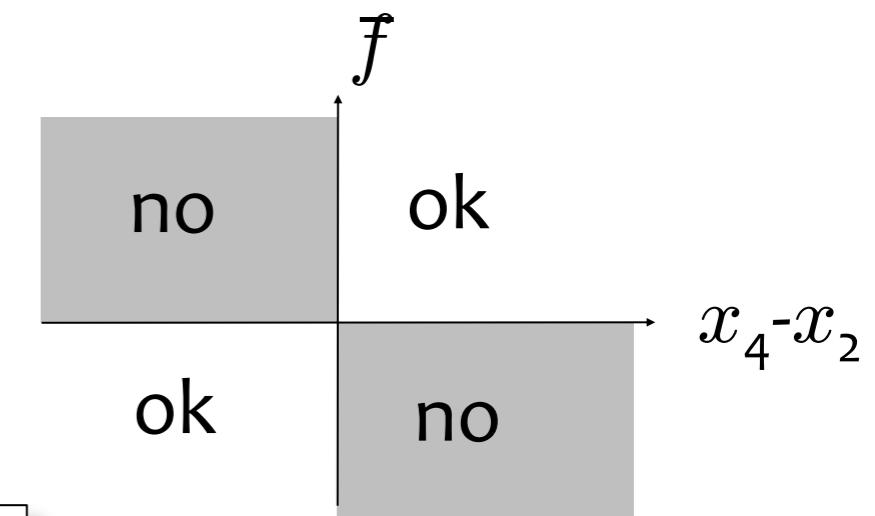


Quarter-car model  
→ linear model

**Constraints:**

1) Passivity condition:

$$\bar{f}(x_4 - x_2) \geq 0$$



2) Max dissipation power:

$$\bar{f}(x_4 - x_2) \leq (2 \cdot 25.5 \cdot \omega_s)(x_4 - x_2)^2$$

3) Saturation:

$$|\bar{f}| \leq \sigma$$

(1), (2) are **nonlinear & nonconvex**  
physical constraints

hybrid model

# CONSTRAINTS

1) Passivity condition:

$$\bar{f}(x_4 - x_2) \geq 0 \quad \longleftrightarrow \quad \begin{array}{c|c} \bar{f} & \\ \hline \text{no} & \text{ok} \\ \text{ok} & \text{no} \end{array} \quad x_4 - x_2$$

$$\begin{aligned} [\delta_v = 1] &\leftrightarrow [x_4 - x_2 \geq 0] \\ [\delta_{\bar{f}} = 1] &\leftrightarrow [\bar{f} \geq 0] \\ [\delta_v = 1] &\rightarrow [\delta_{\bar{f}} = 1] \\ [\delta_v = 0] &\rightarrow [\delta_{\bar{f}} = 0] \end{aligned}$$

2) Max dissipation power:

$$\bar{f}(x_4 - x_2) \leq (2 \cdot 25.5 \cdot \omega_s)(x_4 - x_2)^2 \quad \longleftrightarrow \quad F \geq 0$$

where

$$F = \begin{cases} \bar{f} - (2 \cdot 25.5 \cdot \omega_s)(x_4 - x_2) & \text{if } (x_4 - x_2) \leq 0 \\ -\bar{f} + (2 \cdot 25.5 \cdot \omega_s)(x_4 - x_2) & \text{otherwise} \end{cases}$$

3) Saturation:

$$|\bar{f}| \leq \sigma \quad \longleftrightarrow \quad \begin{array}{c} \bar{f} \leq \sigma \\ \bar{f} \geq -\sigma \end{array}$$

# HYSDEL MODEL

```
/* Semiactive suspension system

(C) 2003-2005 by A.Bemporad, D.Hrovat,
E.Tseng, N.Giorgetti

*/
SYSTEM suspension {
INTERFACE {
STATE {
    REAL x1 [-0.05,0.05];
    REAL x2 [-5,5];
    REAL x3 [-0.2,0.2];
    REAL x4 [-2,2];
}
INPUT{
    REAL u [-10,10]; /* m/s^2 */
}
OUTPUT {
    REAL y;
}
PARAMETER {
    REAL A1dot,A2dot,A3dot,A4dot,B4dot,ws;
    REAL A11,A12,A13,A14,B1,A21,A22,A23,A24,B2;
    REAL A31,A32,A33,A34,B3,A41,A42,A43,A44,B4;
}
}
}
```

```
IMPLEMENTATION {
AUX {
    BOOL sign;
    BOOL usign;
    REAL F;
}
AD {
    sign = x4-x2<=0;
    usign = u<=0;
}
DA {
    F={ IF sign THEN u-(2*25.5*ws)*(x4-x2)
        ELSE -u+(2*25.5*ws)*(x4-x2) };
}
OUTPUT { y=A1dot*x1+A2dot*x2+A3dot*x3
        +A4dot*x4+B4dot*u;
}
CONTINUOUS {
    x1 = A11*x1+A12*x2+A13*x3+A14*x4+B1*u;
    x2 = A21*x1+A22*x2+A23*x3+A24*x4+B2*u;
    x3 = A31*x1+A32*x2+A33*x3+A34*x4+B3*u;
    x4 = A41*x1+A42*x2+A43*x3+A44*x4+B4*u;
}
MUST {
    sign -> usign;
    ~sign -> ~usign;
    F>=0;
} } }
```

```
>>S=mld ('semiact3',Ts)
```

```
>>[X,T,D,Z,Y]=sim(S,x0,U);
```

get the MLD model in MATLAB

simulate the MLD model

# HYBRID PWA MODEL

- PWA model

$$x(k+1) = A_{i(k)}x(k) + B_{i(k)}u(k) + f_{i(k)}$$

$$y(k) = C_{i(k)}x(k) + D_{i(k)}u(k) + g_{i(k)}$$

$$i(k) \text{ s.t. } H_{i(k)}x(k) + J_{i(k)}u(k) \leq K_{i(k)}$$

- 4 continuous states

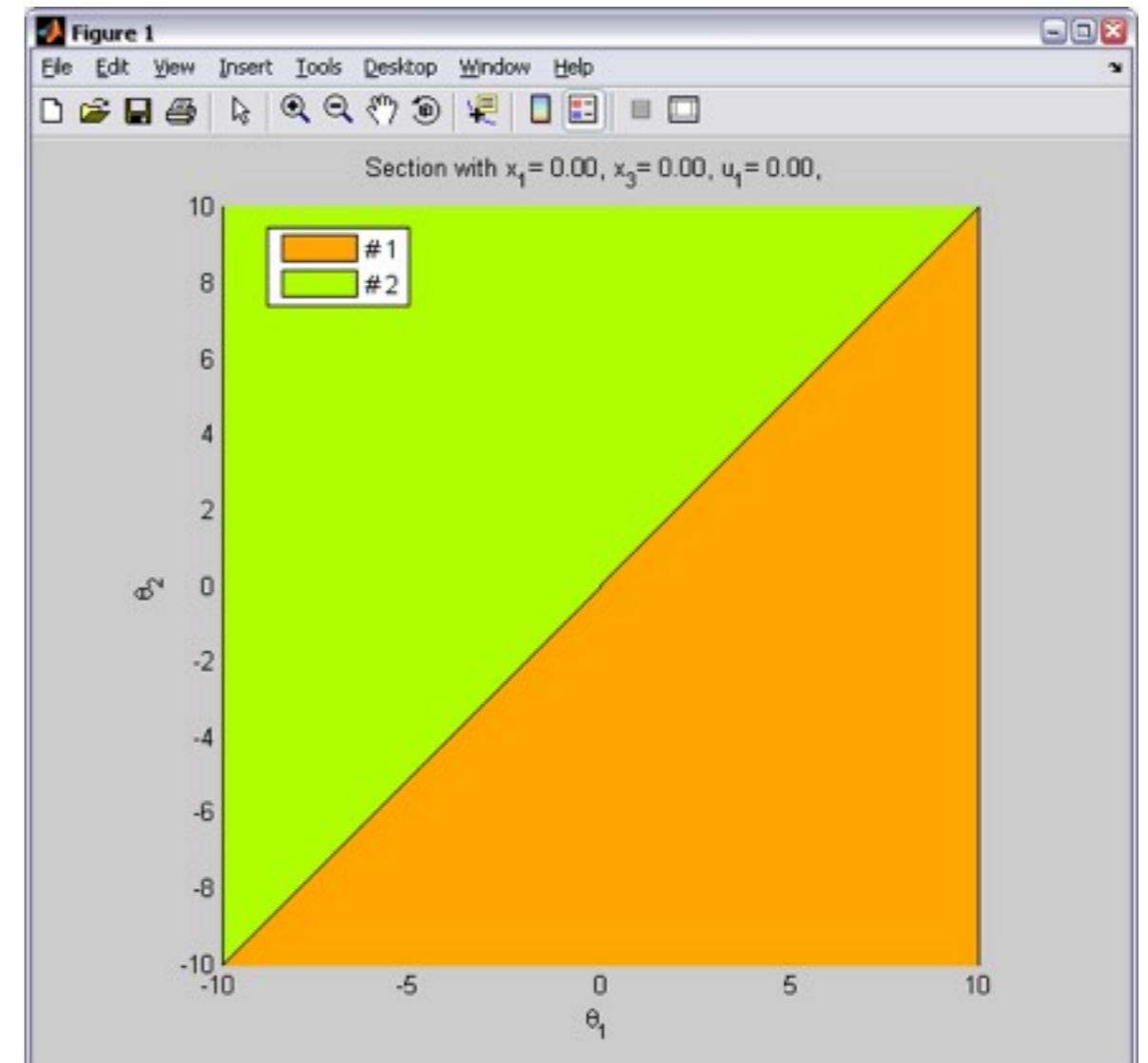
$(x_1, x_2, x_3, x_4)$

- 1 continuous input

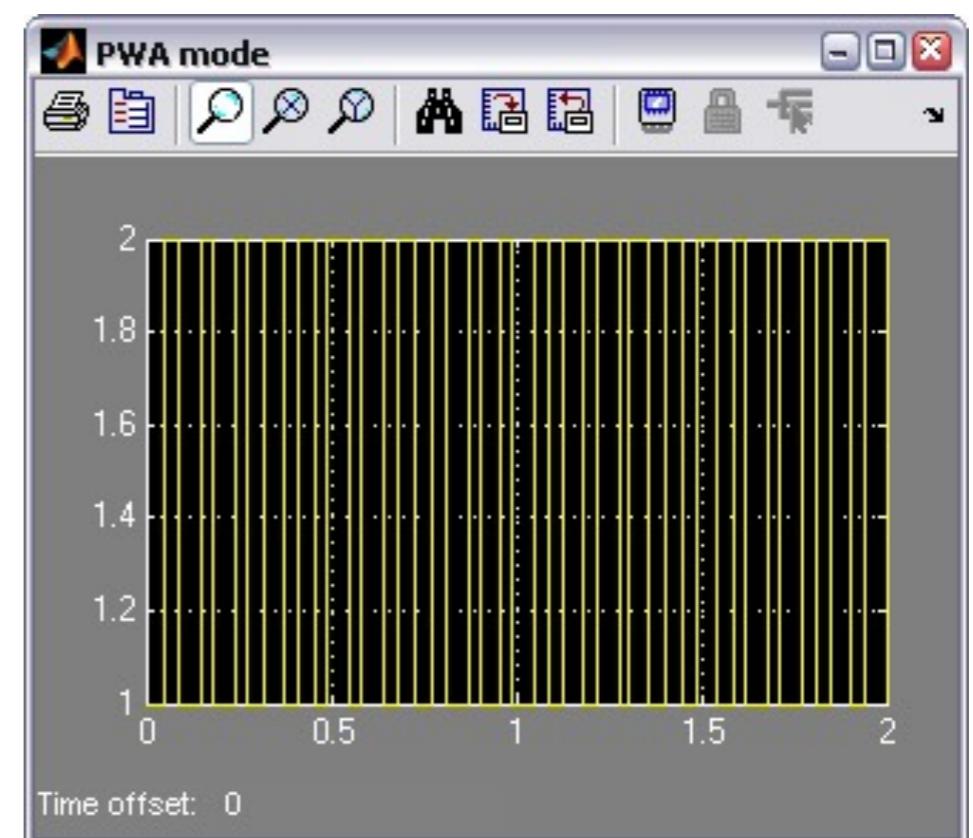
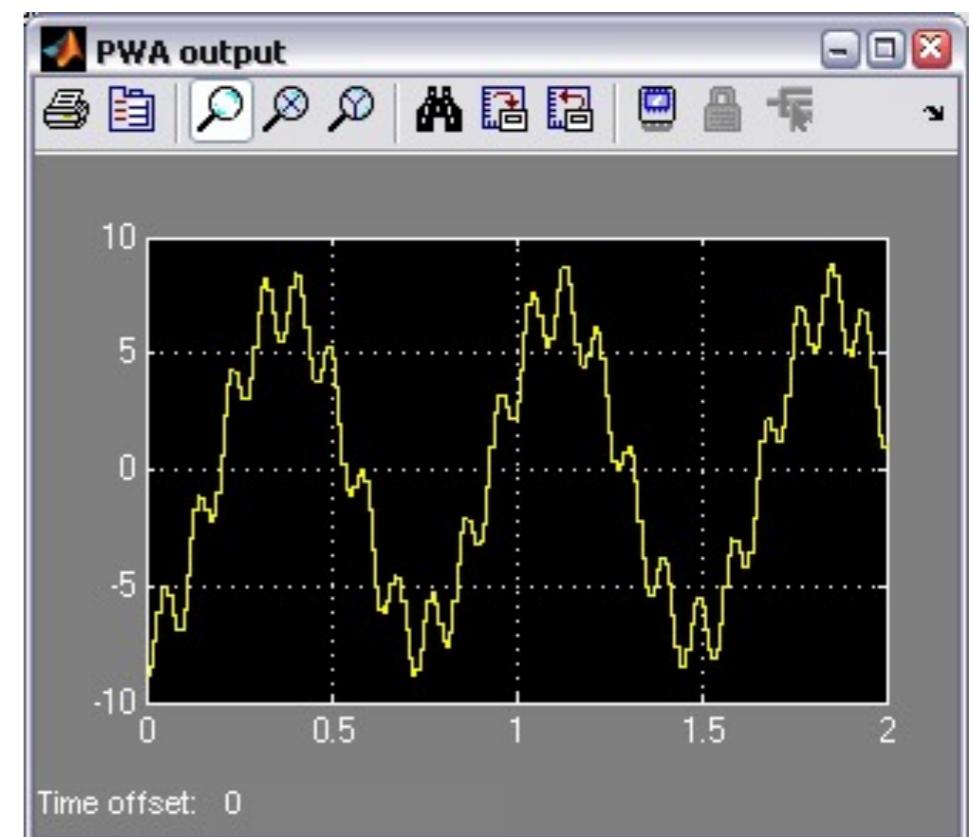
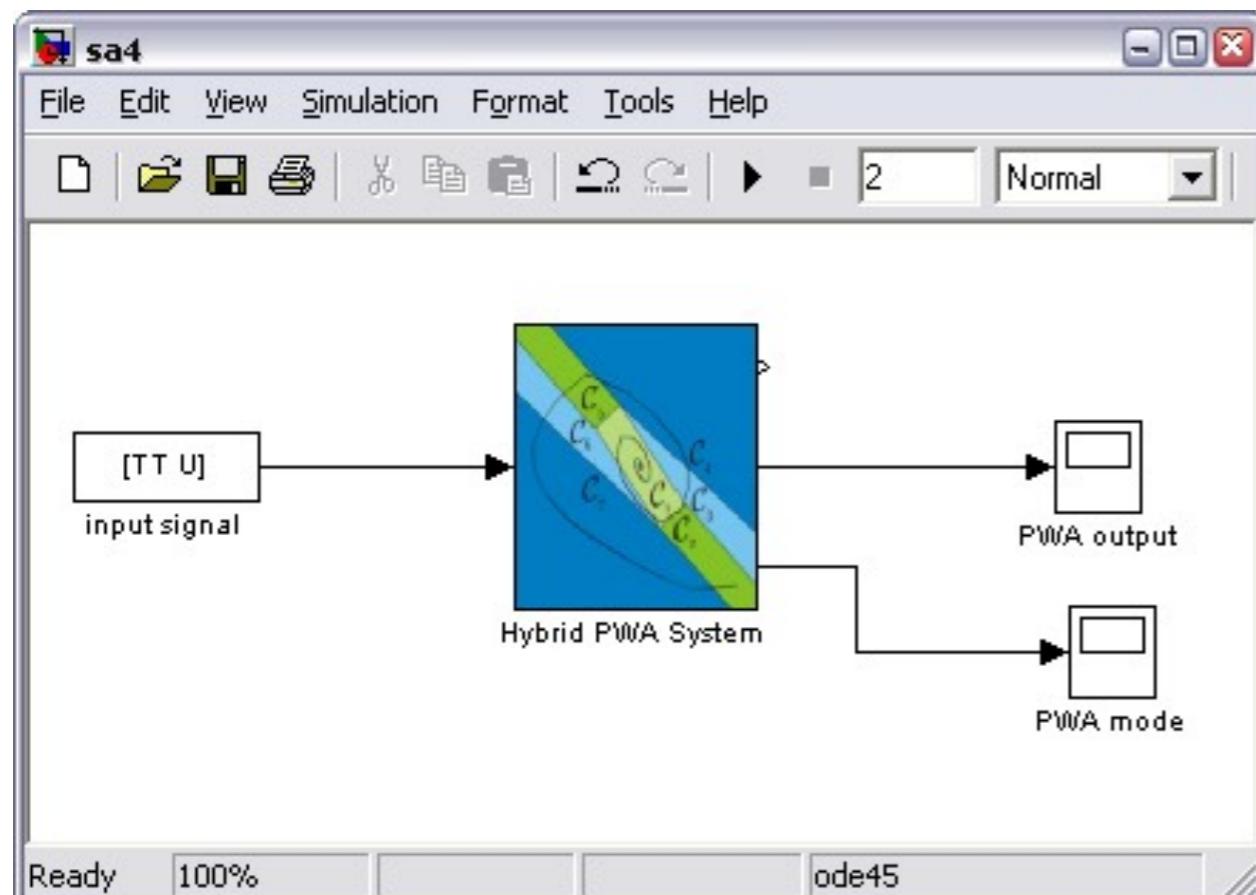
(normalized adjustable damping force  $\bar{f}$ )

- 2 polyhedral regions

```
>>P=pwa (S) ;
```



# SIMULATION IN SIMULINK



# PERFORMANCE SPECS

tire deflection

suspension  
deflection

vertical  
acceleration

$$\min \left( \sum_{k=1}^{N-1} 1100x_{1,k}^2 + 100x_{3,k}^2 + \dot{x}_{4,k}^2 \right) + x_N' P x_N$$

terminal weight  
(Riccati matrix)

# CLOSED-LOOP MPC RESULTS (COMMAND LINE)

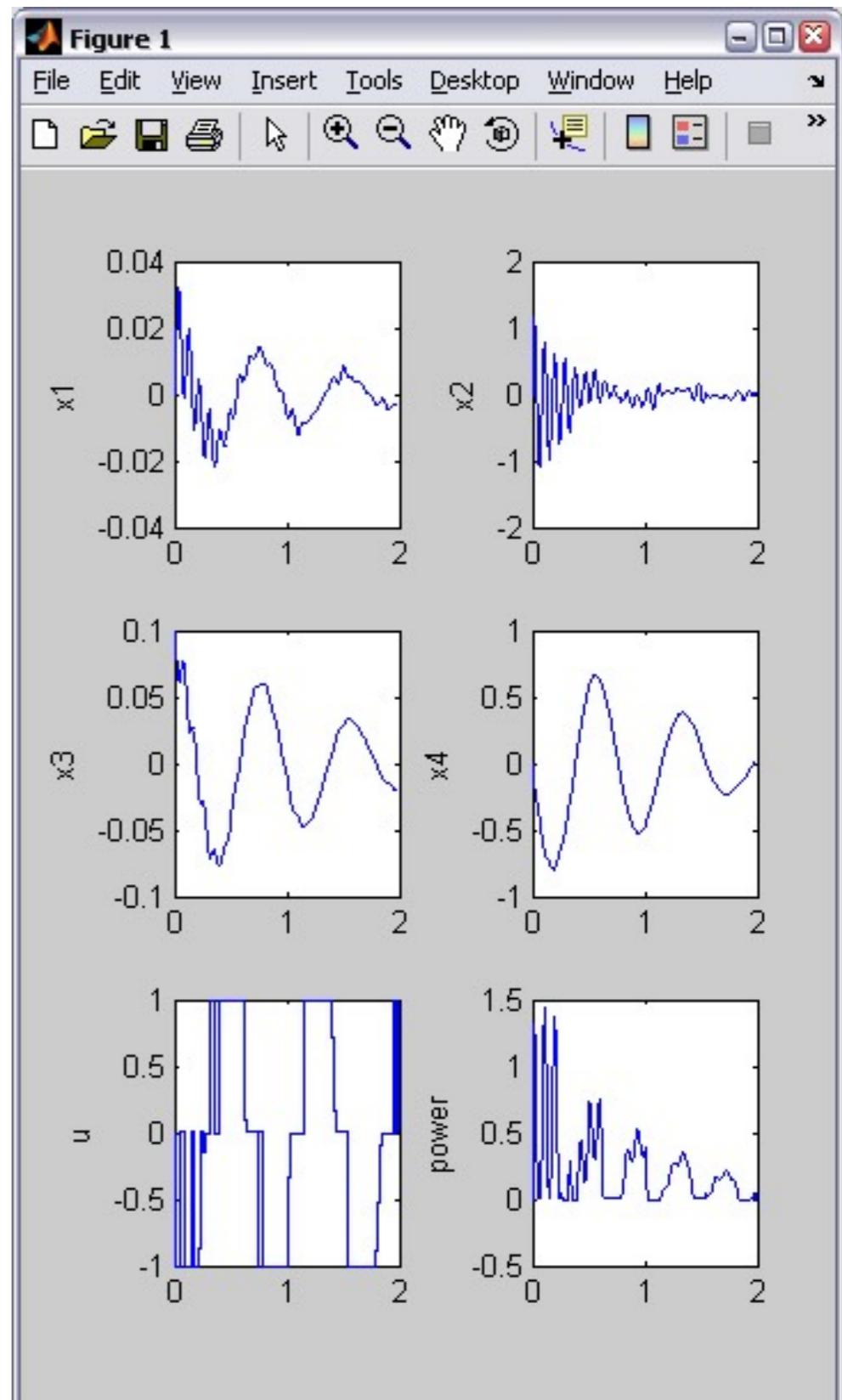
$$J = \int (q_1 x_1^2 + q_3 x_3^2 + \dot{x}_4^2)$$

```
>>refs.y=1; % weights output #1  
>>Q.y=Ts*rx4d;% output weight  
...  
>>Q.norm=2; % quadratic costs  
>>N=1; % optimization horizon  
>>limits.umin=umin;  
>>limits.umax=umax;
```

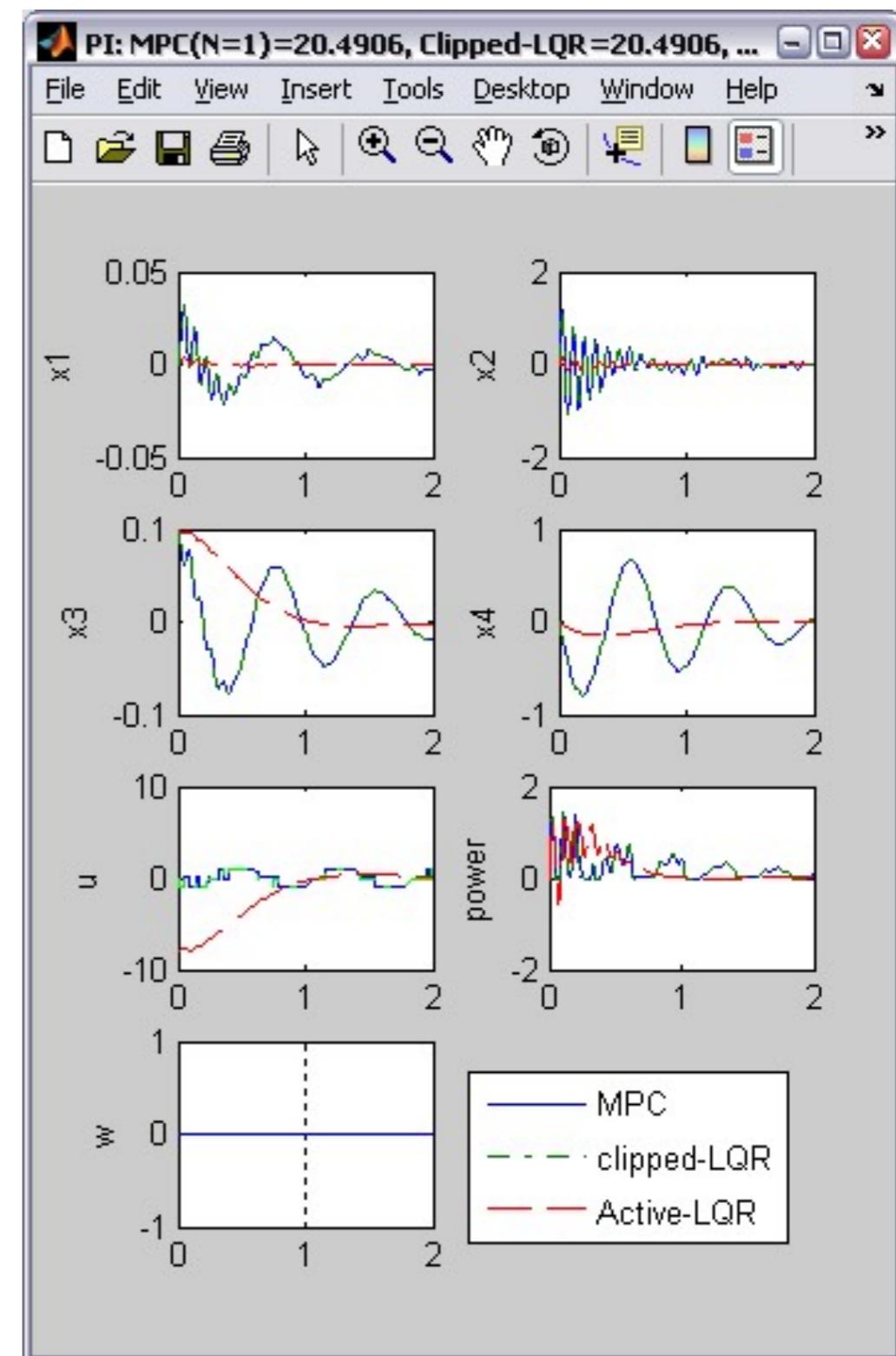
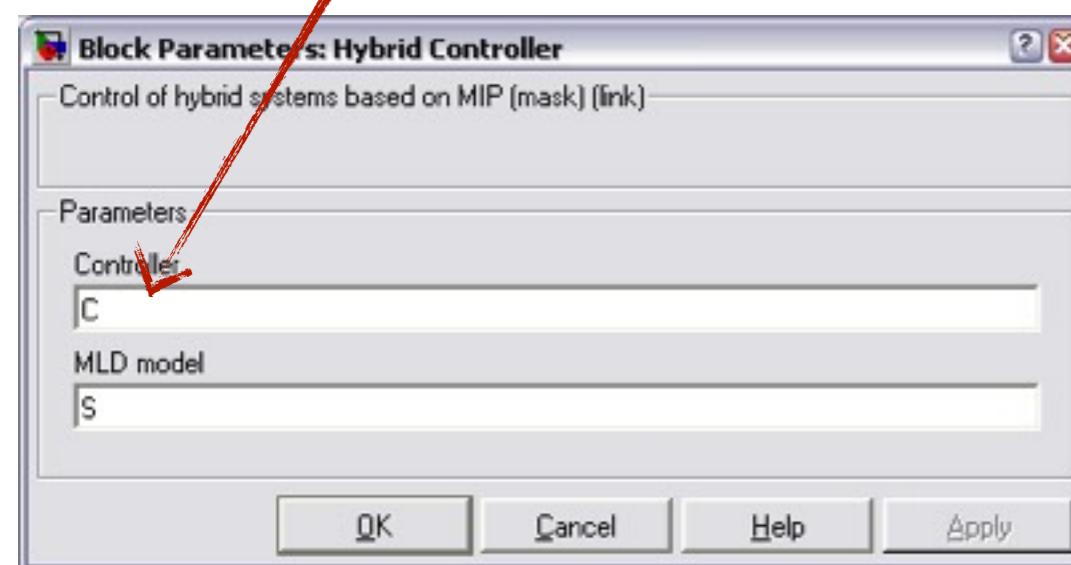
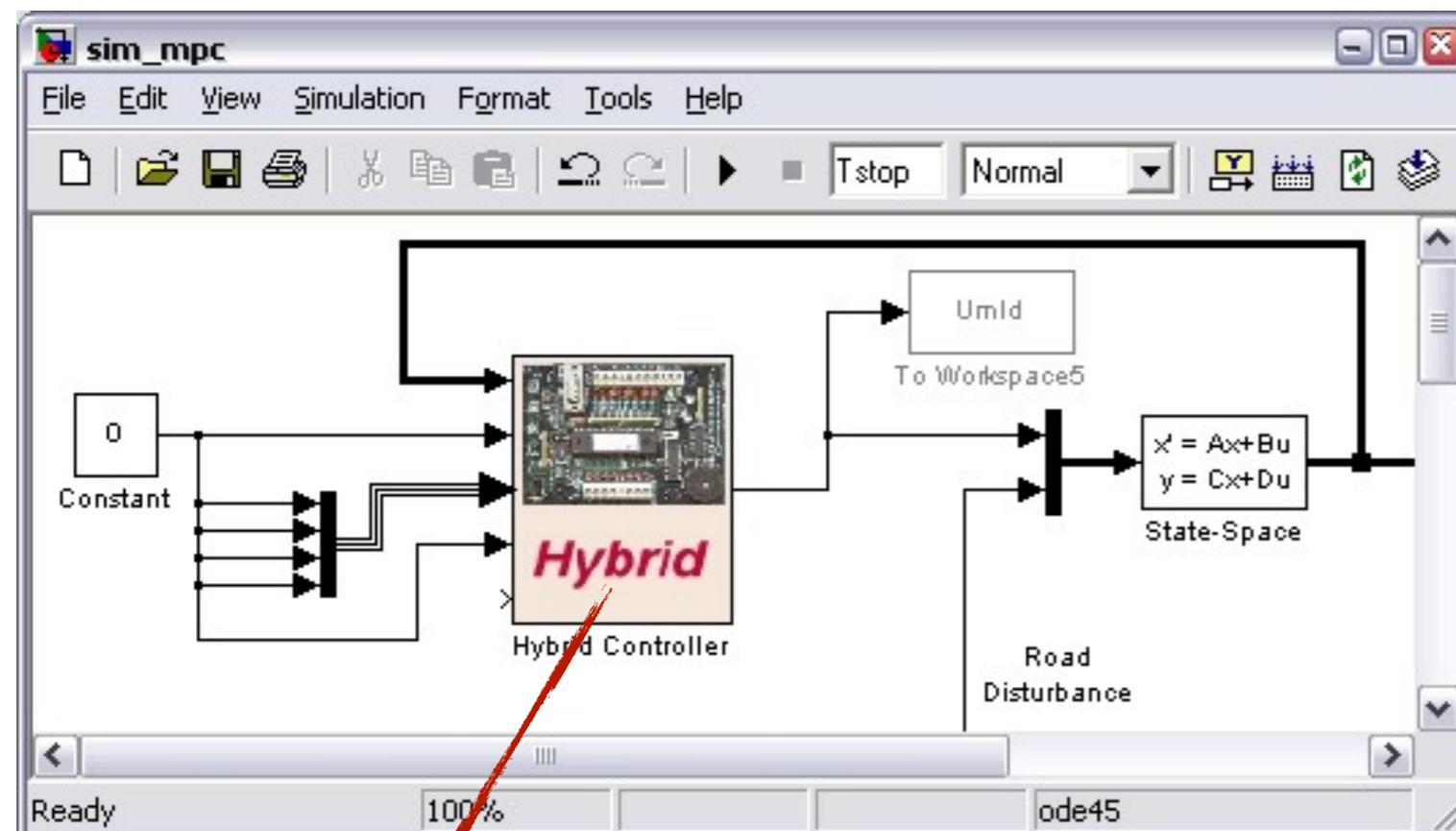
```
>>C=hybcon(S,Q,N,limits,refs);
```

```
>> C  
  
Hybrid controller based on MLD model S <semiact3.hys> [2-norm]  
  
4 state measurement(s)  
1 output reference(s)  
1 input reference(s)  
4 state reference(s)  
0 reference(s) on auxiliary continuous z-variables  
  
4 optimization variable(s) (2 continuous, 2 binary)  
13 mixed-integer linear inequalities  
sampling time = 0.01, MIQP solver = 'cplex'  
  
Type "struct(C)" for more details.  
>>
```

```
>>[XX,UU,DD,ZZ,TT]=sim(C,S,r,x0,Tstop);
```



# CLOSED-LOOP MPC RESULTS (SIMULINK)



# EXPLICIT HYBRID MPC

```
>> E=expcon(C, range, options);
```

```
>> E

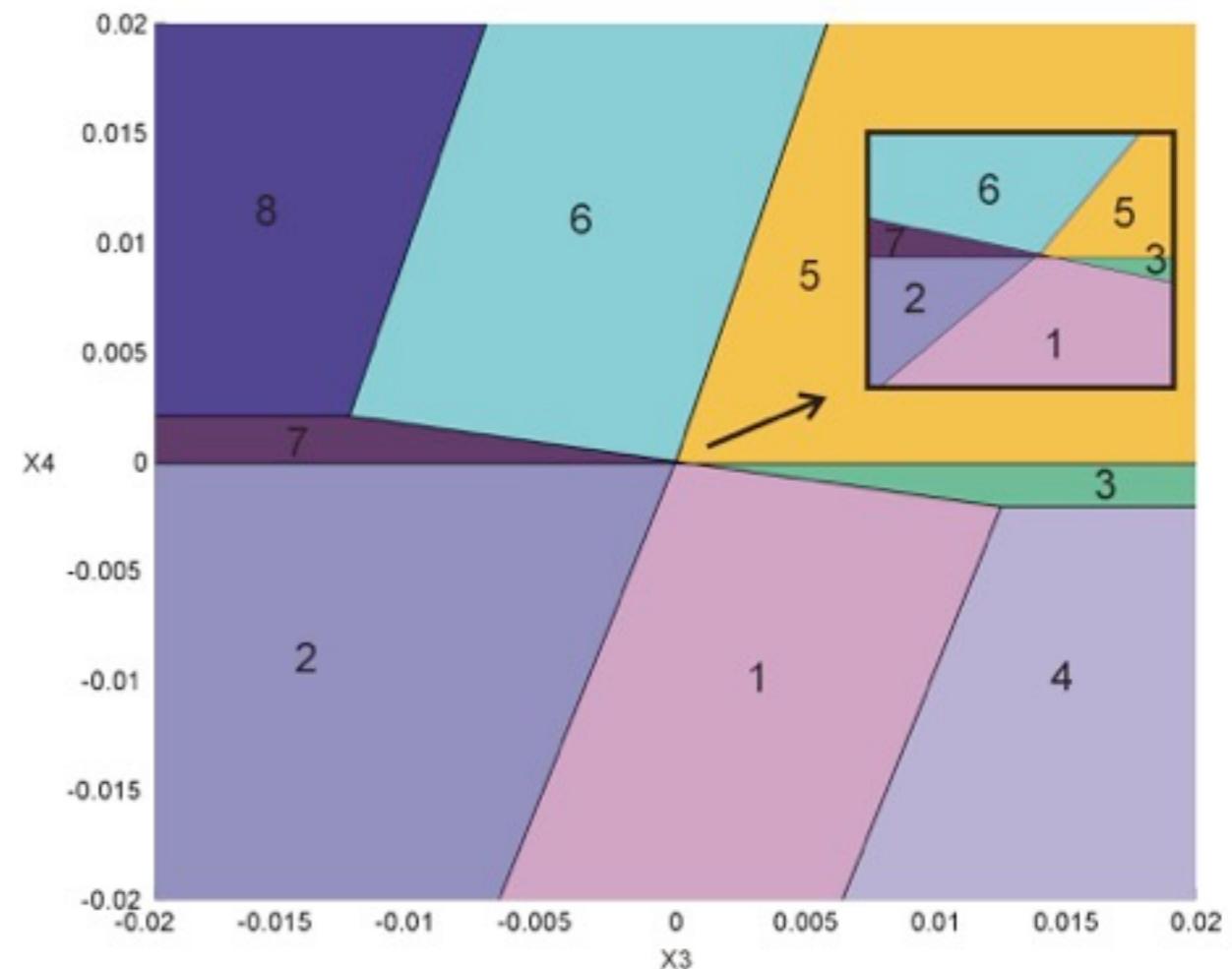
Explicit controller (based on hybrid controller C)
  4 parameter(s)
  1 input(s)
  8 partition(s)
sampling time = 0.01

The controller is for hybrid systems (tracking)
[2-norm]

This is a state-feedback controller.

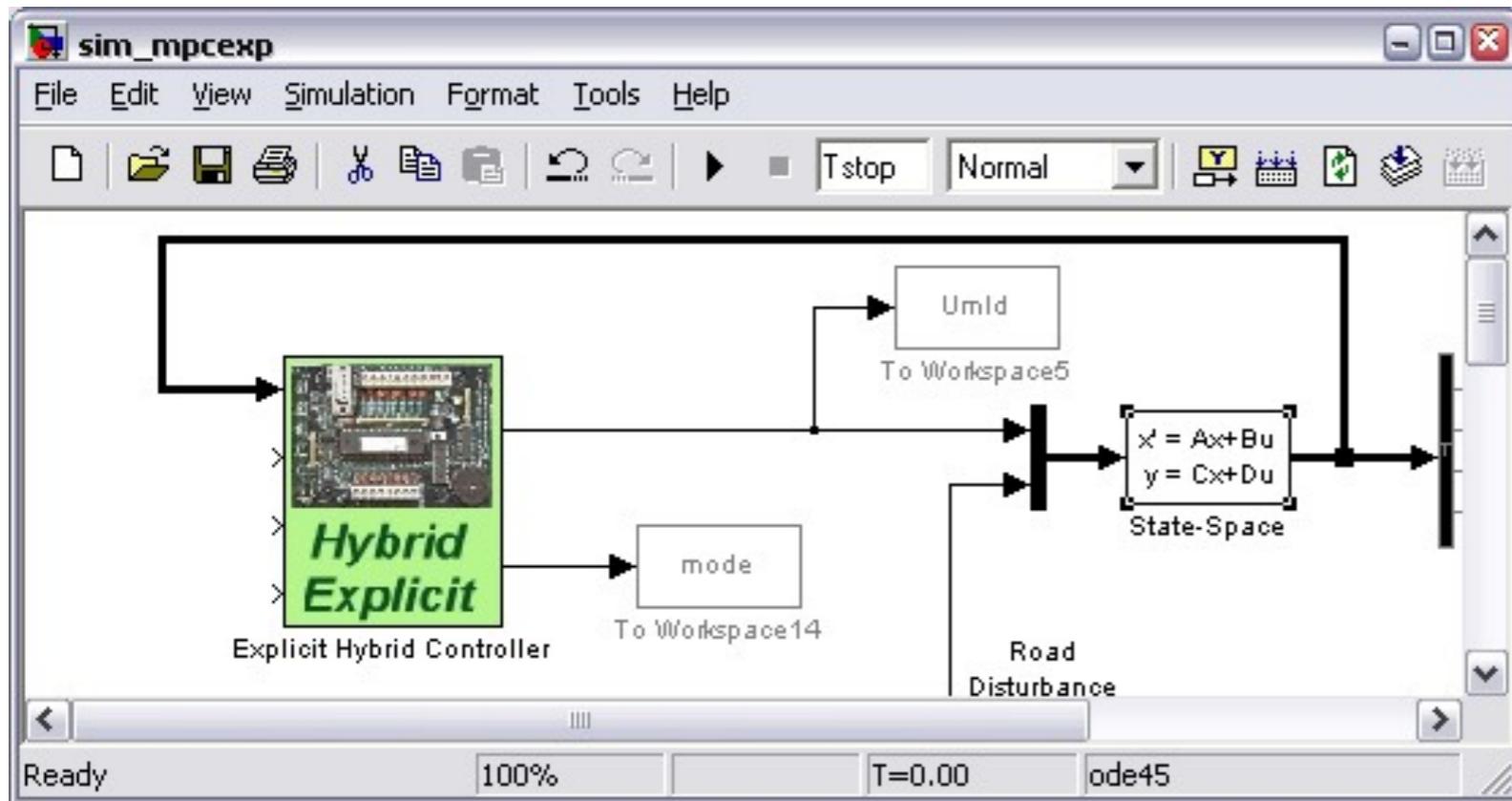
Type "struct(E)" for more details.
>>
```

Explicit solution ( $N=1, x_1=x_2=0$ ):



$$u(x) = \begin{cases} 10.4748x_1 + 0.2446x_2 & +79.1519x_3 - 3.9235x_4 \\ (= K_{LQ}) & \text{Regions } \#1, \#6 \\ 0 & \text{Regions } \#2, \#5 \\ (2 \cdot 25.5 \cdot \omega_s)(x_4 - x_2) & \text{Regions } \#3, \#7 \\ -1 & \text{Region } \#4 \\ 1 & \text{Region } \#8 \end{cases}$$

# EXPLICIT HYBRID MPC

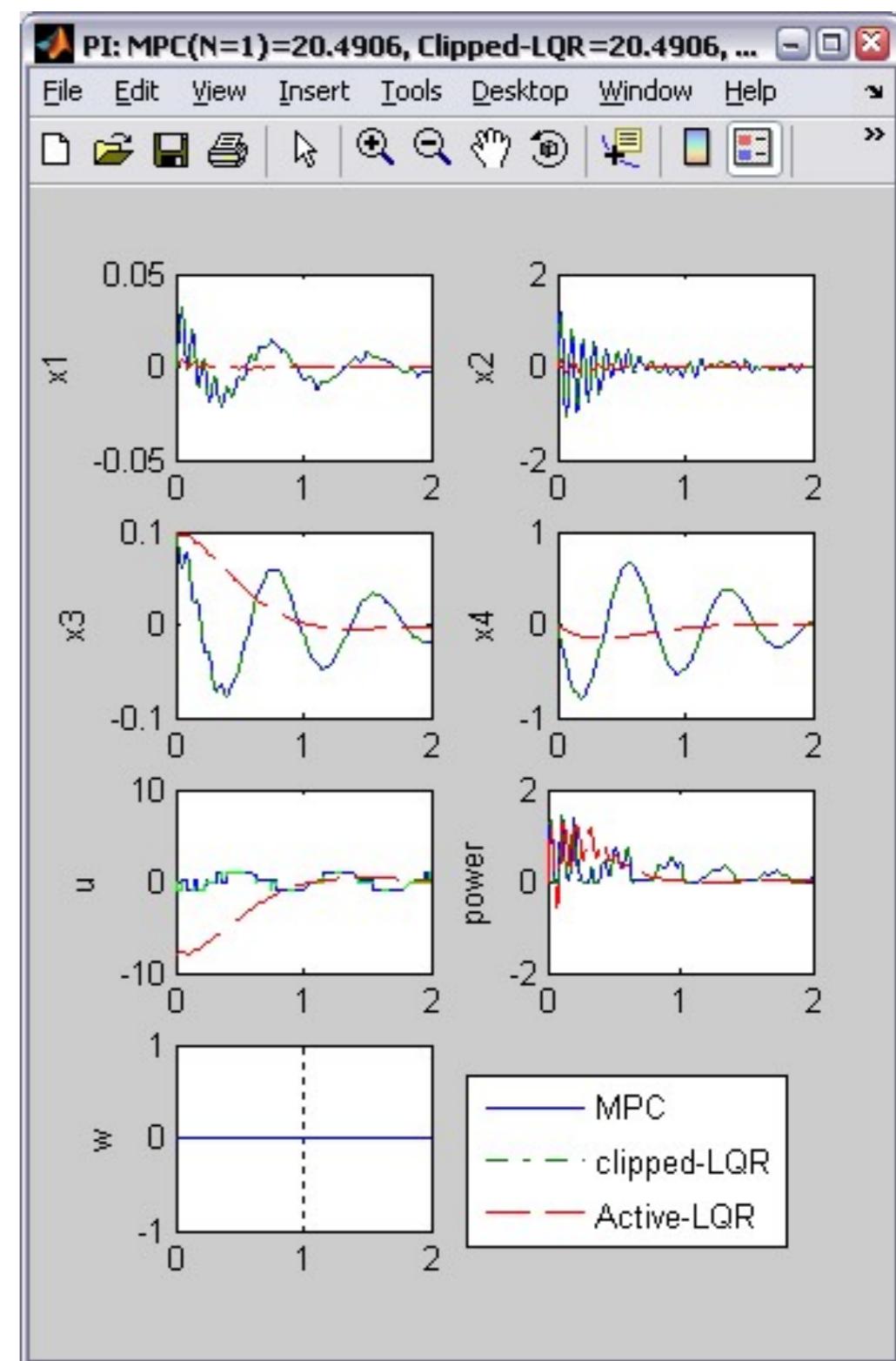


generated  
C-code

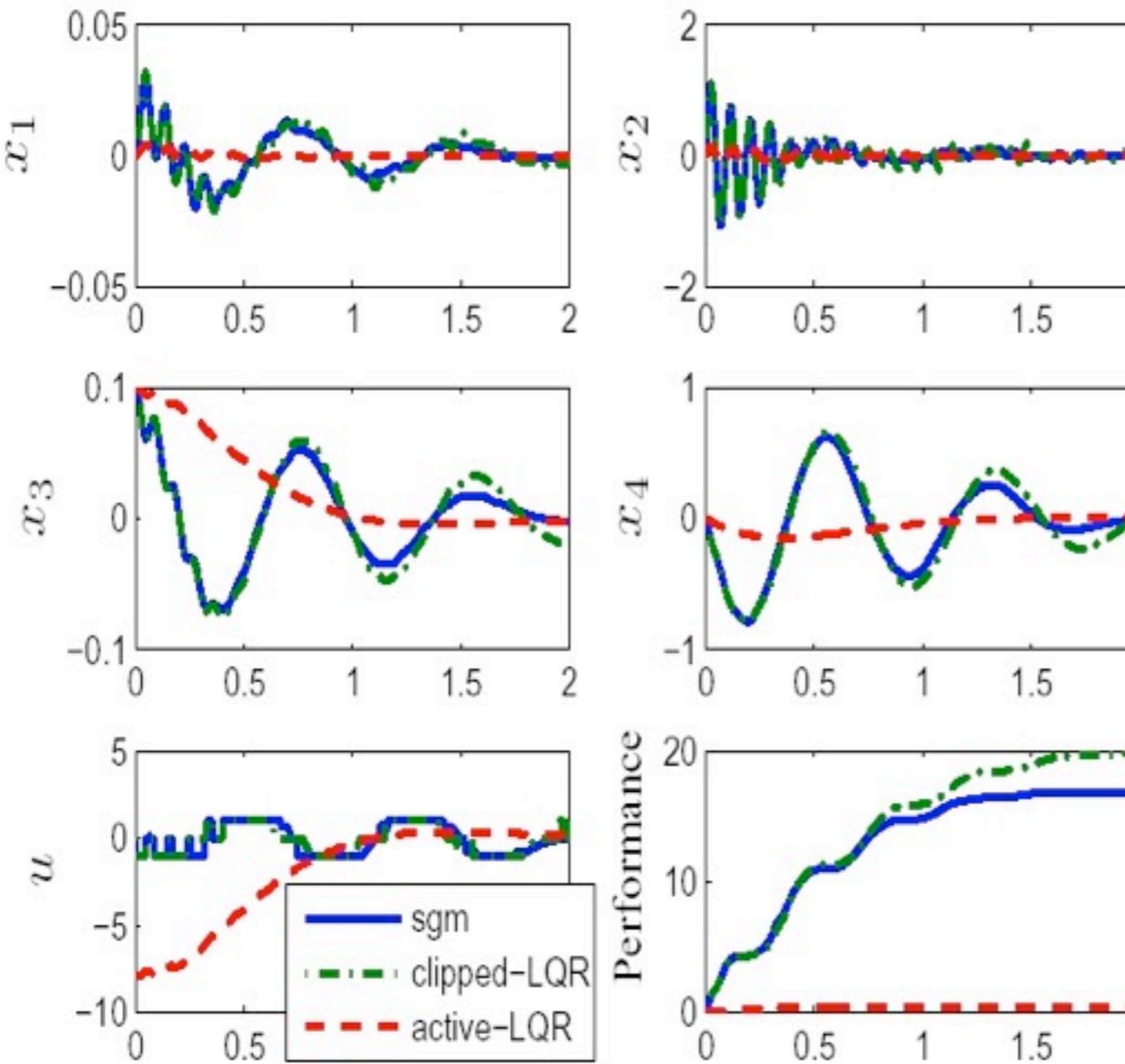


```
#define EXPCON_NU 1
#define EXPCON_NX 4
#define EXPCON_NY 1
#define EXPCON_TS 0.01000000
#define EXPCON_REG 8
#define EXPCON_NTH 4
#define EXPCON_NYM 4
#define EXPCON_NUC 1
#define EXPCON_NUB 0
#define EXPCON_NGAIN 1
#define EXPCON_NH 21
#define EXPCON_NF 8
static double EXPCON_F[]={
    10.4748,0,0,0,10.4748,0,0,0,-0.244594,0,
    480.664,0,
    3.92349,0,
    480.664,0
};

static double EXPCON_G[]={
    0,1e-006,-1e-006,-1,0,0,1e-006,1
```



# QUEST OF OPTIMAL SEMIACTIVE SUSPENSIONS



PARAMETER VALUES USED IN SIMULATION

Parameter	Value	Description
$T_s$	10 ms	Sampling time
$\omega_s$	1.5 Hz	Sprung mass natural frequency
$\omega_{us}$	10 Hz	Wheel-hop natural frequency
$\rho$	10	Sprung-to-unprung mass ratio
$\zeta$	0	Damping ratio
$\sigma$	1	Maximum force capacity
$q_1$	1100	Weight on tire deflection
$q_3$	100	Weight on suspension deflection

TABLE II  
SHOCK TEST: MPC COST VALUE FOR DIFFERENT CONTROL HORIZONS SUBJECTED TO I.C. = [0 0 0.1 0].

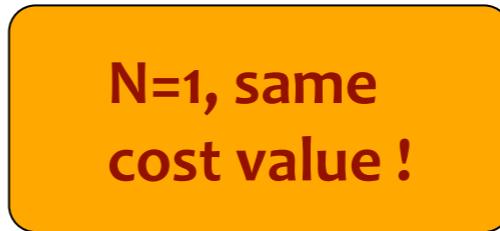
N	MPC	Clipped LQR	SGM	LQR
1	20.4282	20.4282	17.4944	0.4446
2	20.4054			
3	20.3290			
4	20.1100			
5	19.7380			
10	20.9840			
12	19.3084			
14	18.4842			
15	18.5996			
16	19.3212			
20	18.0764			
30	17.1494			
40	17.1304			

# SIMULATION RESULTS

- Horizon N=1: same as Clipped-LQR !
- Better closed-loop performance for increasing N

Performance Index

N	MPC	Clipped-LQR
1	1.5155	1.5155
5	1.4416	
10	1.5238	
15	1.3083	
20	1.2204	
30	1.1456	
40	1.1462	



N=1, same cost value !

- Simulations with road noise.
- Initial condition  $x(0)=[0 \ 0 \ 0 \ 0]'$
- Simulation time  $T=20$  s, sampling time  $T_s=10$  ms

# VEHICLE YAW STABILITY CONTROL BY COORDINATED ACTIVE FRONT STEERING AND DIFFERENTIAL BRAKING

(D. Bernardini, S. Di Cairano, A. Bemporad and H.E. Tseng, CDC 2009)  
(S. Di Cairano, H.E. Tseng, D. Bernardini, A. Bemporad, IEEE TCST, 2012)



# OVERVIEW

- **Problem:**

Control vehicle stability while tracking driver's desired trajectory

- ▶ **Electronic Stability Control (ESC)**

(Koibuchi *et al.*, 1996)

- ▶ **Active Front Steering (AFS)**

(Ackermann, 1997)

- **Main control objective:**

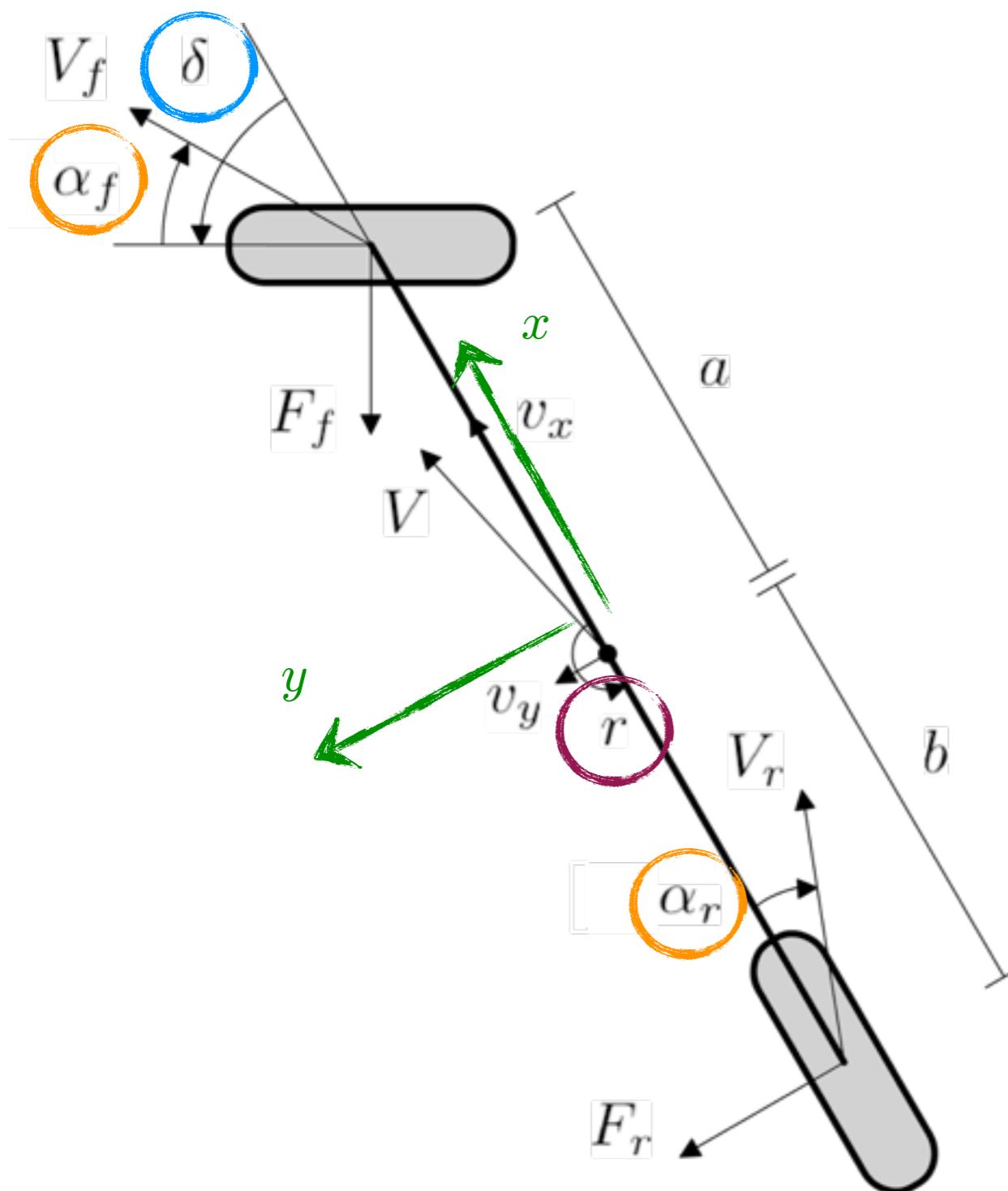
Force the vehicle **yaw rate** to track a time-varying reference computed by the driver's steering angle and the current vehicle velocity

- **Approach:**

Consider the steer as a reference generator and actuate steering and differential braking (**coordinated AFS and ESC action**)

# VEHICLE MODEL

- **Bicycle model** appropriate in high speed turns (Gillespie, 1992)



- **reference frame** ( $x,y,z$ ) moving with the vehicle
- **front steering angle**  $\delta$  [rad]
- **tire slip angles**  $\alpha_f, \alpha_r$  [rad]
- **yaw rate**  $r$  [rad/s]

$$\tan(\alpha_f + \delta) = \frac{v_y + ar}{v_x}$$
$$\tan \alpha_r = \frac{v_y - br}{v_x}$$

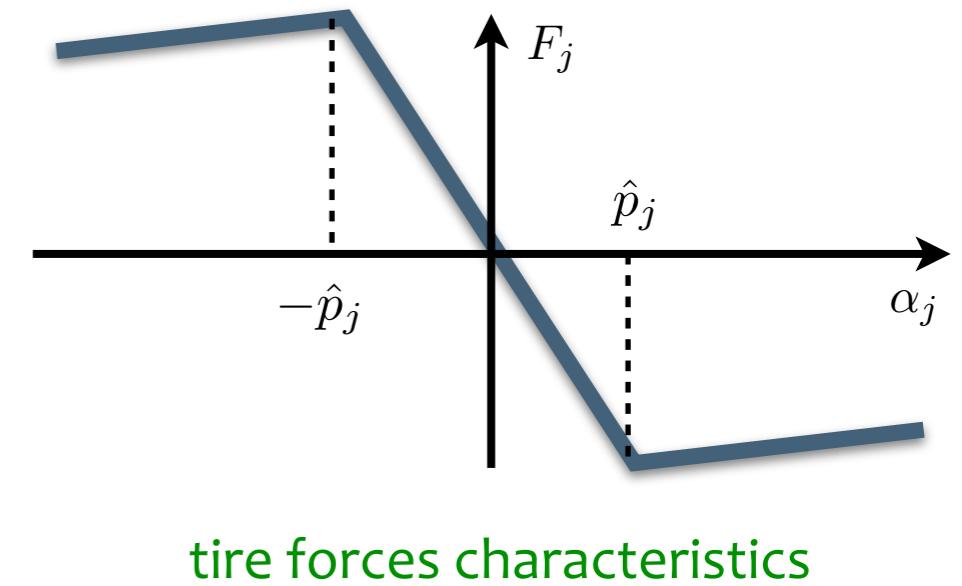
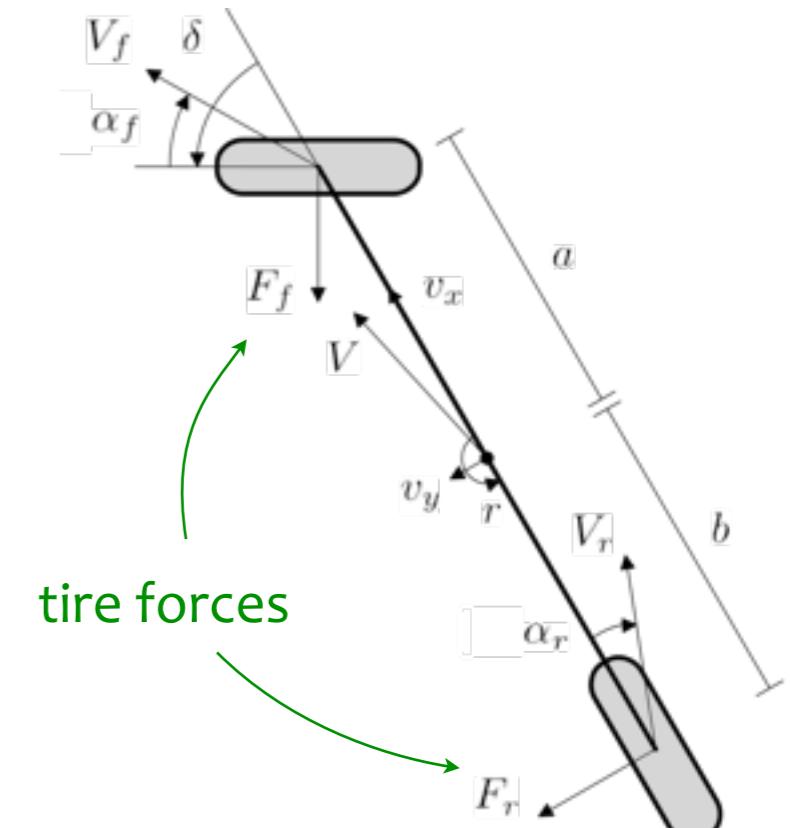
# TIRE FORCE MODEL

- **Tire force characteristics:** are nonlinear functions of the **slip angles** and of the **longitudinal slip**
- For a constant longitudinal slip, we use a **piecewise affine model**

$$F_f(\alpha_f) = \begin{cases} -c_f \alpha_f & \text{if } -\hat{p}_f \leq \alpha_f \leq \hat{p}_f \\ -(d_f \alpha_f + e_f) & \text{if } \alpha_f > \hat{p}_f \end{cases}$$

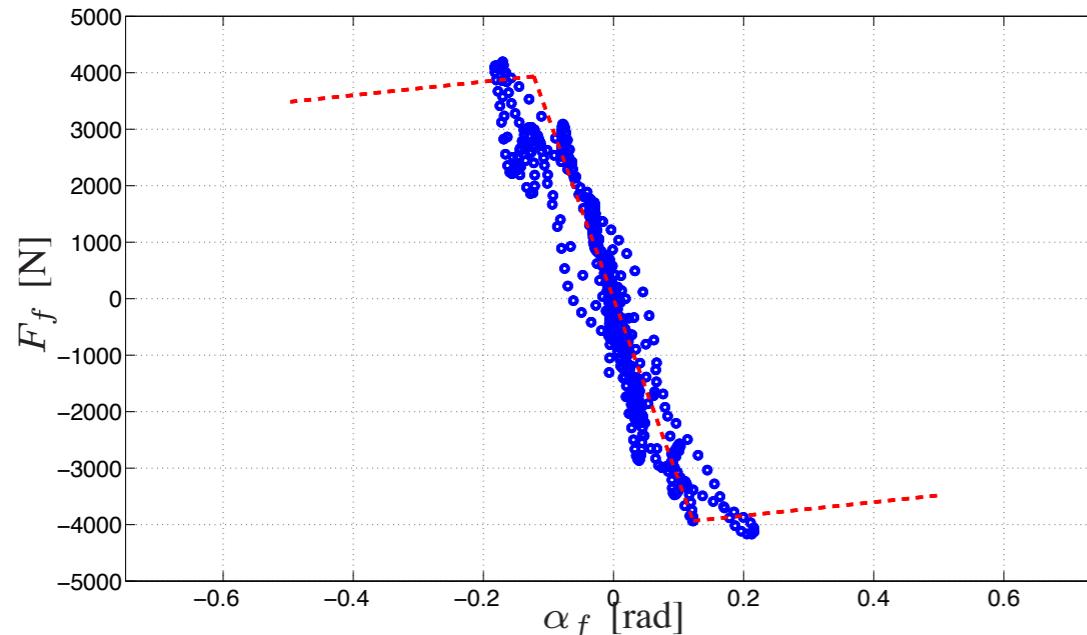
$$F_r(\alpha_r) = \begin{cases} -c_r \alpha_r & \text{if } -\hat{p}_r \leq \alpha_r \leq \hat{p}_r \\ -(d_r \alpha_r + e_r) & \text{if } \alpha_r > \hat{p}_r \end{cases}$$

- **Critical slip angles**  $\hat{p}_f, \hat{p}_r$  are threshold values where dynamics switch
- For symmetry, we can restrict to analyze **clockwise turns** (counter-clockwise turns can be handled by opportunely inverting signs)

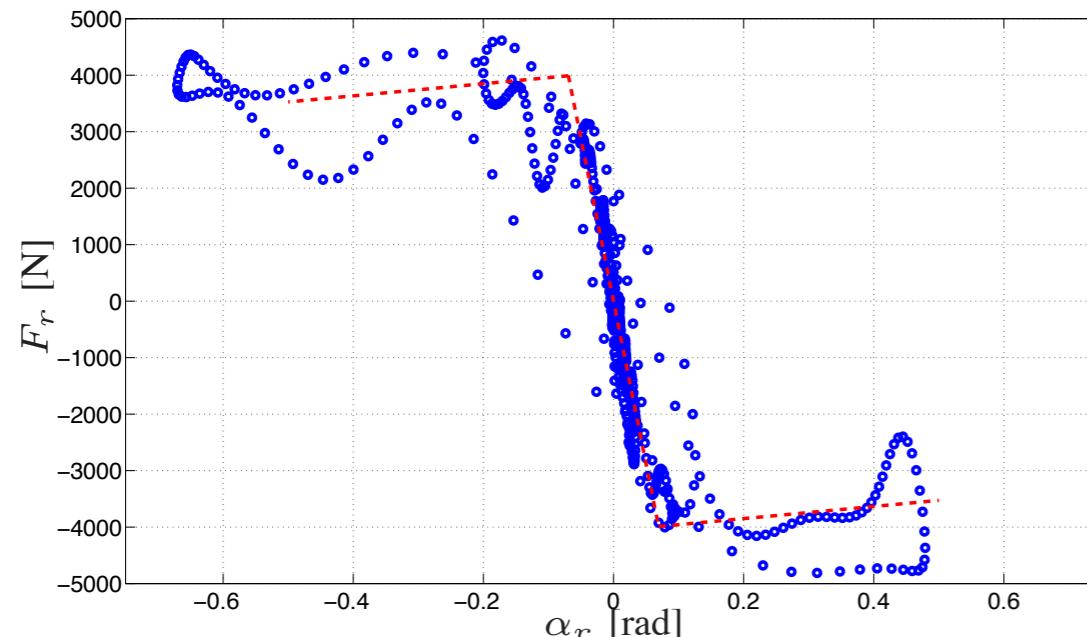


# TIRE FORCE MODEL

## Sideslip angle-force characteristics



(a) Front tires



(b) Rear tires

Experimental tire data and piecewise linear approximation of the tire



Rear-wheel drive test vehicle equipped with active front steering and differential braking used for experimental validation

# DYNAMICAL MODEL

slip angles

$$\begin{aligned}\dot{\alpha}_f &= \frac{\dot{v}_y + a\dot{r}}{v_x} - \dot{\delta} \\ \dot{\alpha}_r &= \frac{\dot{v}_y - b\dot{r}}{v_x}\end{aligned}$$

static yaw rate

$$r = \frac{v_x}{a+b}(\alpha_f - \alpha_r + \delta)$$

overall dynamical model

$$\begin{aligned}\dot{\alpha}_f &= \frac{F_f + F_r}{mv_x} - \frac{v_x}{a+b}(\alpha_f - \alpha_r + \delta) + \frac{a}{v_x I_z}(aF_f - bF_r + Y) \\ \dot{\alpha}_r &= \frac{F_f + F_r}{mv_x} - \frac{v_x}{a+b}(\alpha_f - \alpha_r + \delta) - \frac{b}{v_x I_z}(aF_f - bF_r + Y) \\ r &= \frac{v_x}{a+b}(\alpha_f - \alpha_r + \delta)\end{aligned}$$

$$\dot{v}_y = \frac{F_f \cos \delta + F_r}{m} - rv_x$$

lateral velocity

yaw rate derivative

$$\dot{r} = \frac{aF_f \cos \delta - bF_r + Y}{I_z}$$

# PIECEWISE-AFFINE MODEL

- The overall dynamics model is recast as a PWA system by introducing the Boolean variables

$$\begin{aligned}\gamma_f = 0 &\leftrightarrow \alpha_f \leq \hat{p}_f \\ \gamma_r = 0 &\leftrightarrow \alpha_r \leq \hat{p}_r\end{aligned}$$

- By discretizing with sampling period  $T_s = 0.1$  s we obtain

$$\begin{aligned}x(k+1) &= A_i x(k) + B_i u(k) + f_i \\ y(k) &= C x(k) + D u(k) \\ i \in \{1, \dots, 4\} &: H_i x(k) \leq K_i\end{aligned}$$

where

$$x = [\alpha_f \ \alpha_r]'$$

slip angles

$$u = [Y \ \delta]'$$

yaw moment front steering

$$y = r$$

yaw rate

# REFERENCE GENERATION

- **Control goal:**

stabilize the system at the equilibrium obtained with  $\delta(k) = \hat{\delta}(k)$   
while minimizing the use of the brake actuator ( $\hat{Y}(k) = 0$ )



driver's steering angle

- **Equilibrium condition in the linear region:**

$$\begin{aligned}\dot{\alpha}_f^0 &= \frac{F_f + F_r}{mv_x} - \frac{v_x}{a+b}(\alpha_f - \alpha_r + \delta) + \frac{a}{v_x I_z}(aF_f - bF_r + Y^0) \\ \dot{\alpha}_r^0 &= \frac{F_f + F_r}{mv_x} - \frac{v_x}{a+b}(\alpha_f - \alpha_r + \delta) - \frac{b}{v_x I_z}(aF_f - bF_r + Y^0)\end{aligned}$$

- **Time-varying set-points** are defined using the overall dynamical model

$$\begin{aligned}\hat{\alpha}_f &= \frac{m\tilde{v}_x^2 bc_r \hat{\delta}}{m\tilde{v}_x^2 (ac_f - bc_r) - c_f c_r (a+b)^2} \\ \hat{\alpha}_r &= \hat{\alpha}_f \frac{ac_f}{bc_r} \\ \hat{r} &= \frac{\tilde{v}_x}{a+b} (\hat{\alpha}_f - \hat{\alpha}_r + \hat{\delta})\end{aligned}$$

- Current longitudinal velocity  $v_x(k)$  is used to **update set-points**

# HYBRID PREDICTION MODEL

- **Yaw rate tracking:**

zero tracking error in steady state is provided by **integral action**

$$\begin{aligned} \text{integral of} & \quad I_r(k+1) = I_r(k) + r(k) - r_s(k) \\ \text{tracking error} & \\ \text{yaw rate set-} & \quad r_s(k+1) = r_s(k) \\ \text{point} & \end{aligned}$$

- The **global hybrid dynamical model** of the vehicle is given by

$$\begin{aligned} z(k+1) &= \tilde{A}_i z(k) + \tilde{B}_i u(k) + \tilde{f}_i && \text{augmented state} \\ y(k) &= \tilde{C} z(k) + \tilde{D} u(k) \\ i &\in \{1, \dots, 4\} : \tilde{H}_i z(k) \leq \tilde{K}_i \\ \tilde{A}_i &= \begin{bmatrix} A_i & 0 & 0 \\ \frac{v_x}{a+b} & \frac{-v_x}{a+b} & 1 & -1 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \quad \tilde{B}_i = \begin{bmatrix} B_i \\ 0 & \frac{v_x}{a+b} \\ 0 & 0 \end{bmatrix}, \quad \tilde{f}_i = \begin{bmatrix} f_i \\ 0 \\ 0 \end{bmatrix}, \\ \tilde{C} &= \begin{bmatrix} \frac{v_x}{a+b} & \frac{-v_x}{a+b} & 0 & 0 \end{bmatrix}, \quad \tilde{D} = \begin{bmatrix} 0 & \frac{v_x}{a+b} \end{bmatrix}, \\ \tilde{H}_i &= \begin{bmatrix} H_i & 0 & 0 \end{bmatrix}, \quad \tilde{K}_i = K_i . \end{aligned}$$

$$z = \begin{bmatrix} \alpha_f \\ \alpha_r \\ I_r \\ r_s \end{bmatrix}$$

# CONTROL PROBLEM FORMULATION

- The optimal control problem solved at every time step  $k$  is

$$\begin{aligned} \min_{\mathbf{u}_k} \quad & \sum_{j=0}^{N-1} \left\{ (z_{k+j|k} - \hat{z})' Q_z (z_{k+j|k} - \hat{z}) \right. \\ & + (y_{k+j|k} - \hat{y})' Q_y (y_{k+j|k} - \hat{y}) \left. + (u_{k+j|k} - \hat{u})' Q_u (u_{k+j|k} - \hat{u}) \right\} \\ \text{s.t.} \quad & z_{k|k} = z(k) \\ & \text{hybrid dynamics} \\ & \text{state and input constraints} \end{aligned}$$

← slip angles tracking & int. action  
← yaw rate tracking  
← penalty on actuators action

 MIQP

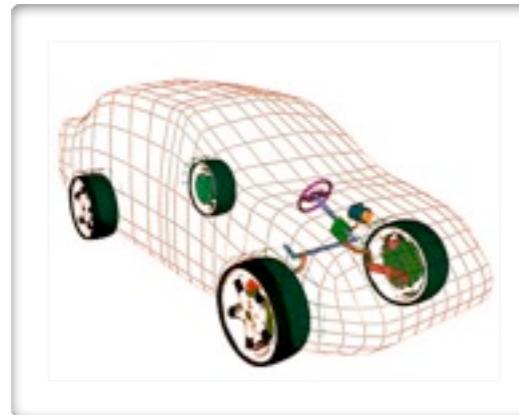
- State and input constraints:

$$\begin{aligned} [z]_1(k) &\geq -\hat{p}_f && \text{front slip angle} \\ [z]_2(k) &\geq -\hat{p}_r && \text{rear slip angle} \end{aligned}$$

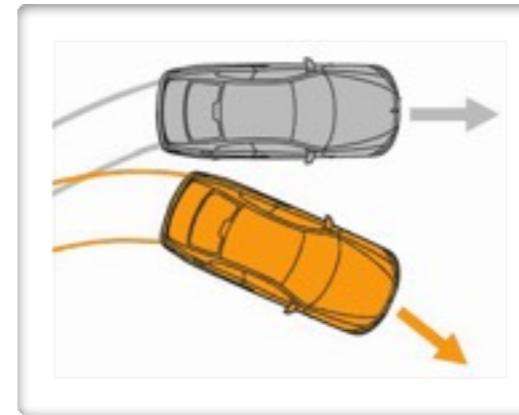
$$\begin{aligned} |[u]_1(k)| &\leq 1000 \text{ [Nm]} && \text{yaw moment} \\ |[u]_2(k)| &\leq 0.35 \text{ [rad]} && \text{front steering} \end{aligned}$$

# SIMULATIONS RESULTS

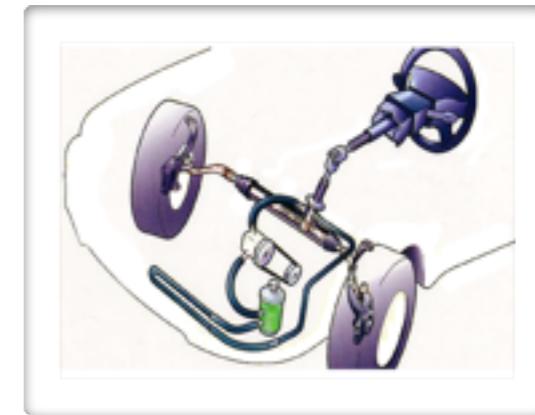
- Simulations run on a **nonlinear vehicle model** including:



longitudinal and lateral  
vehicle dynamics



yaw rate dynamics



steering actuation  
dynamics

- Driver's steering command **constant** over the simulation interval
- **Controller setup** after calibration:

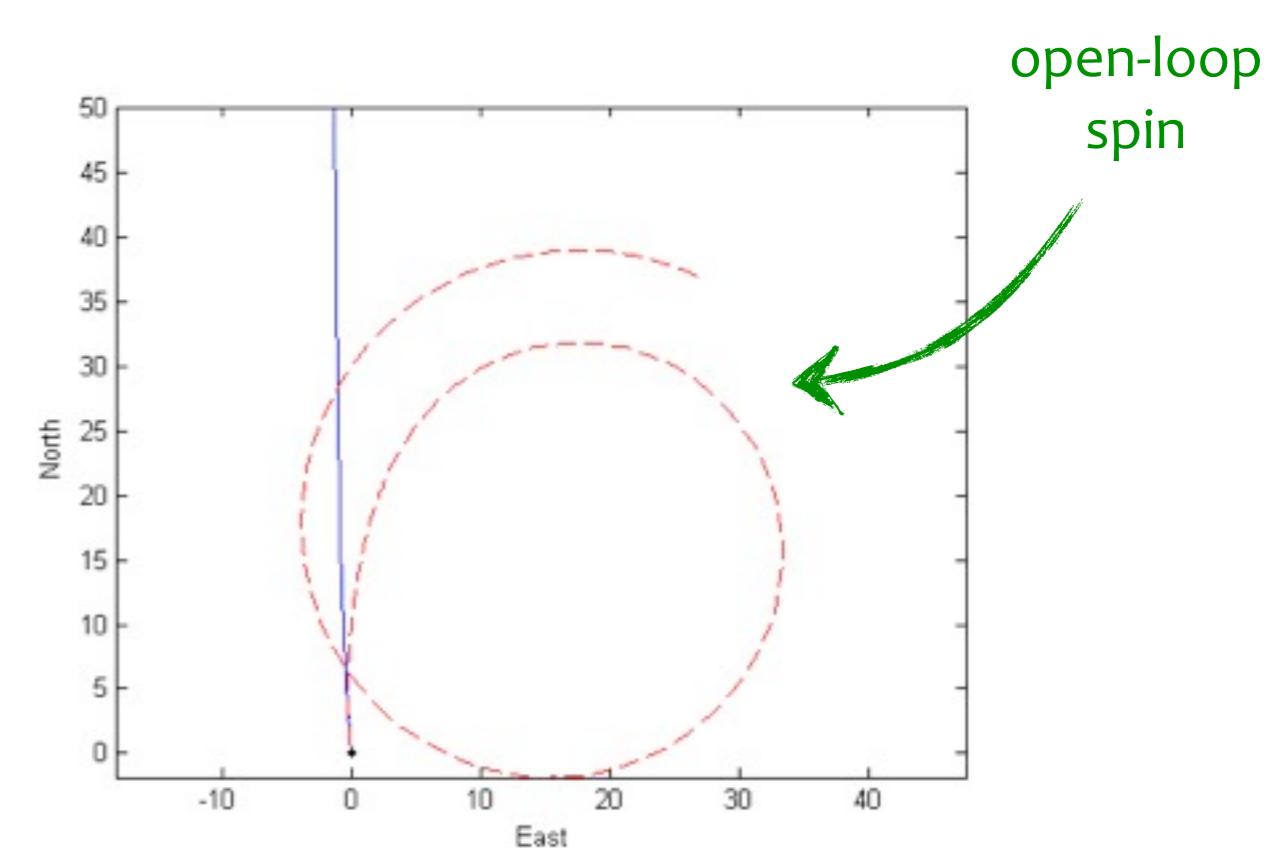
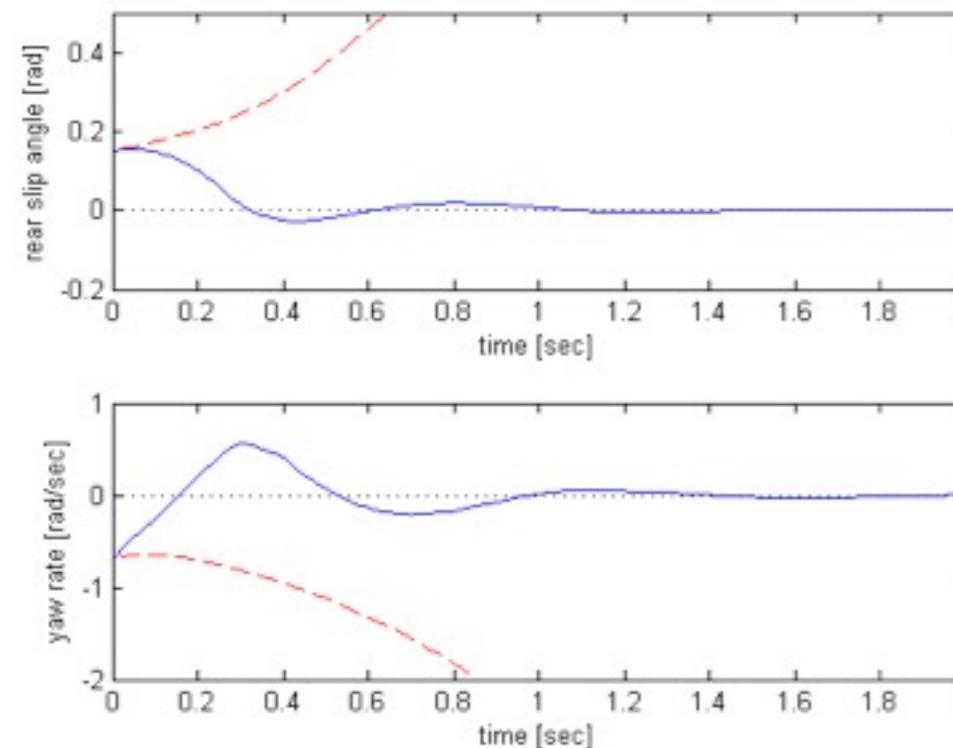
► Prediction horizon  $N = 3$

► Weight matrices  $Q_z = \begin{bmatrix} .1 & 0 & 0 & 0 \\ 0 & .1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ ,  $Q_u = [1 \ 0]$ ,  $Q_y = 1$

# SIMULATIONS RESULTS

- **Stability analysis** under nominal conditions (with  $\hat{\delta} = 0$ ):

open-loop vs closed-loop

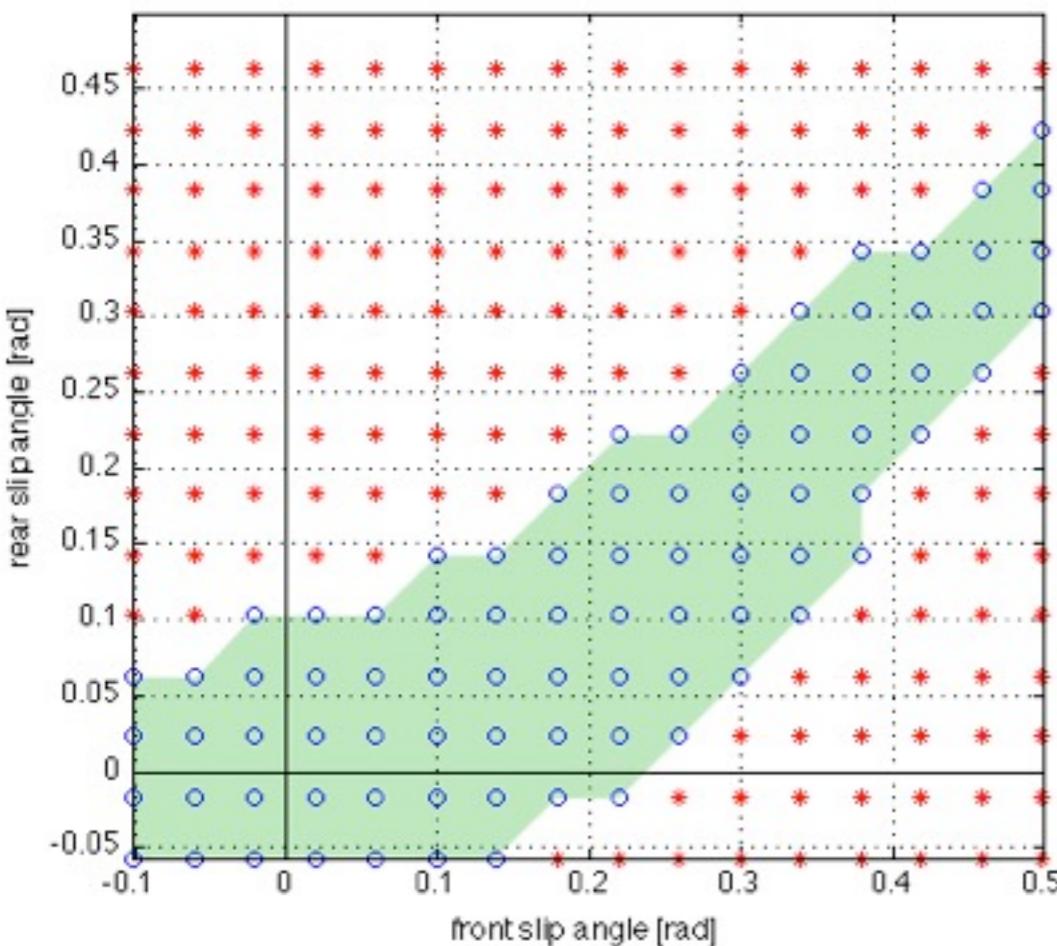


- Controller has to cope with **linearization errors**

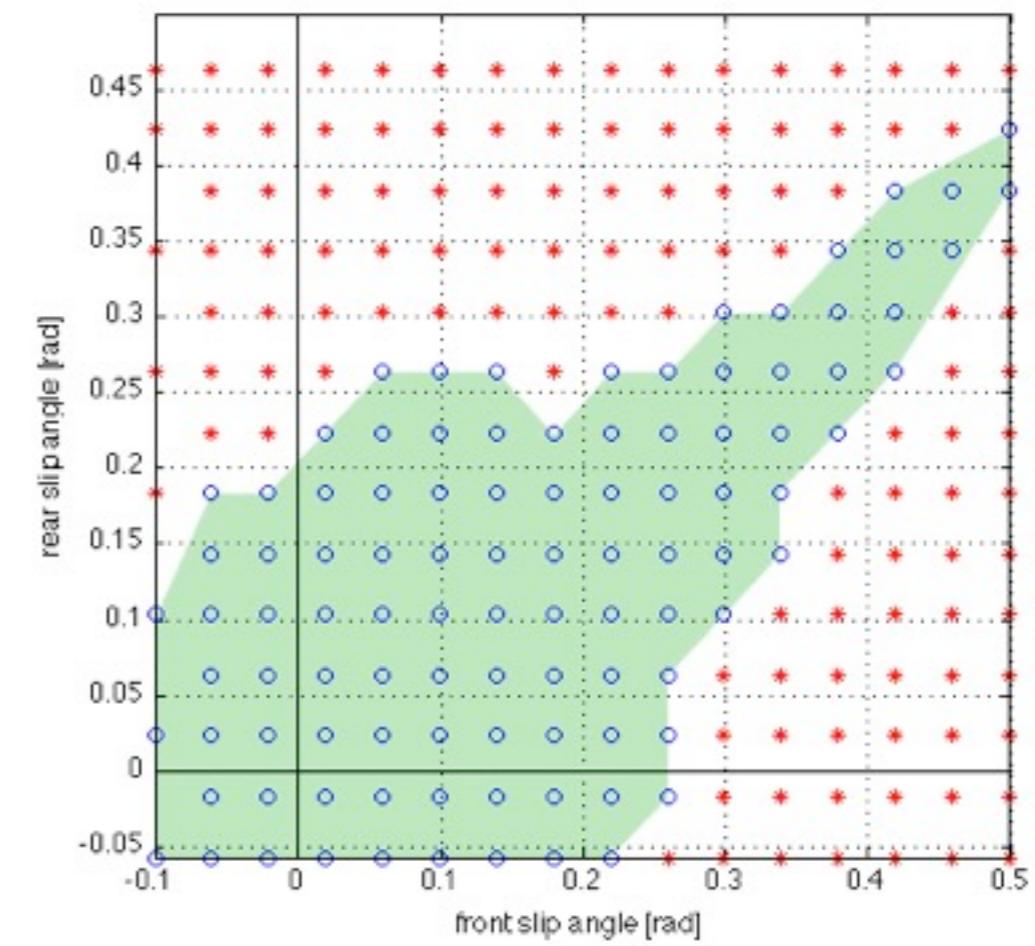
# SIMULATIONS RESULTS

- **Stability analysis** under nominal conditions (with  $\hat{\delta} = 0$ ):

open-loop



closed-loop



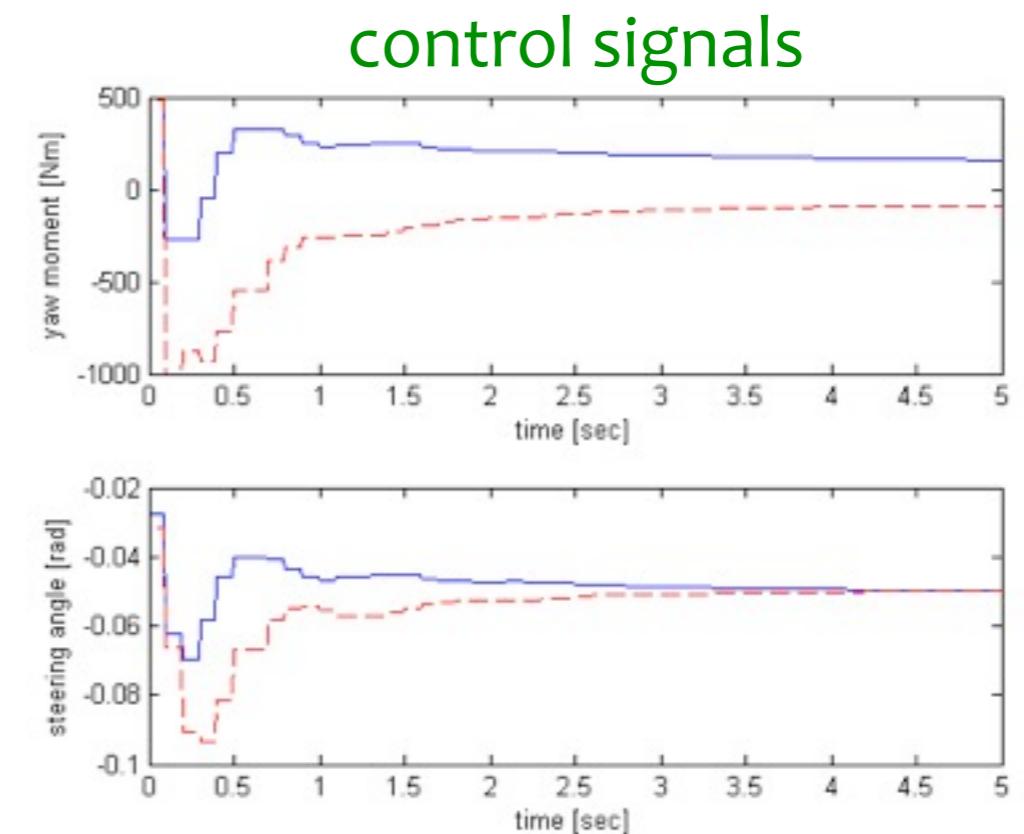
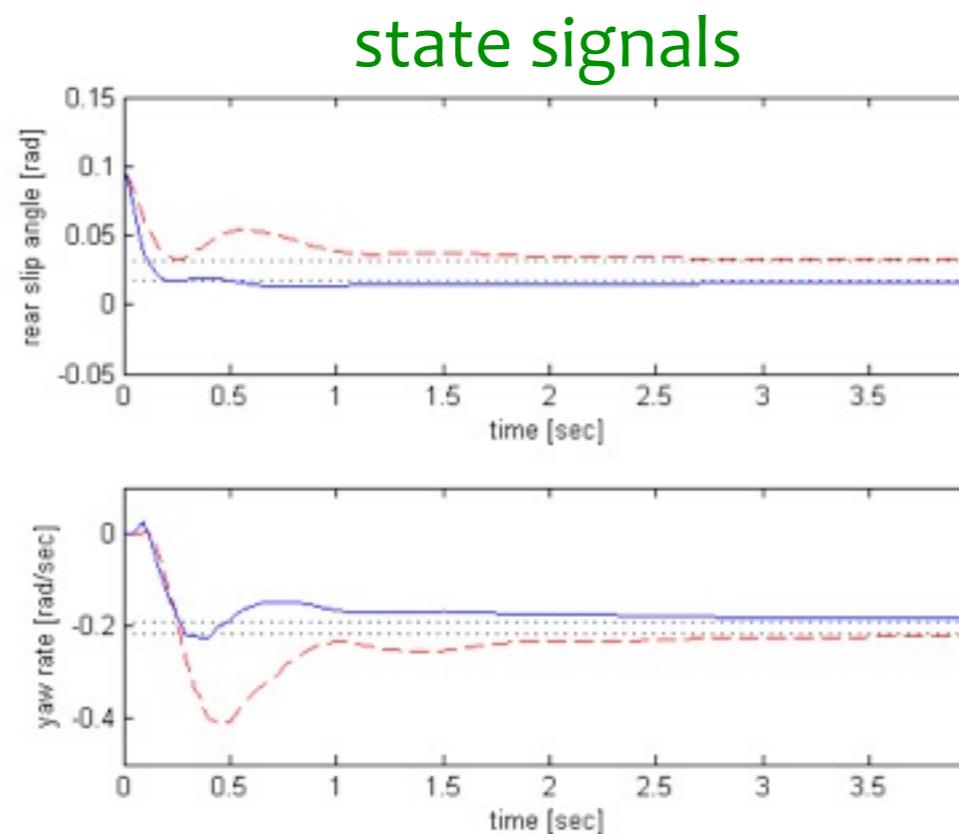
stable initial conditions

unstable initial conditions

closed-loop provides a  
**larger** stability region

# SIMULATIONS RESULTS

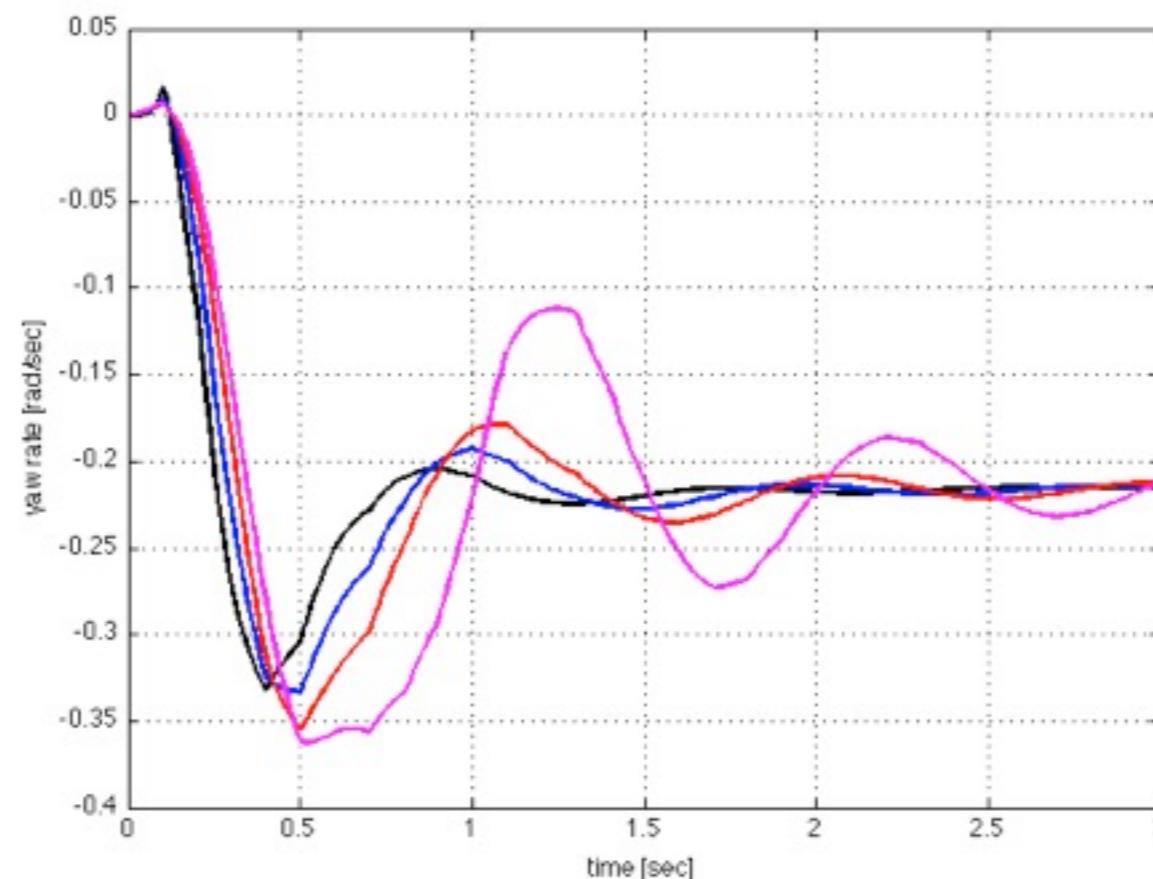
- Robustness analysis w.r.t. model mismatches (with  $\hat{\delta} = -0.05$ ):
  - ▶ nominal longitudinal velocity  $\hat{v}_x = 20 \text{ m/s}$
  - ▶ real longitudinal velocity  $v_x = 15 \text{ m/s}$  and  $v_x = 25 \text{ m/s}$



- Stability and fast tracking response are provided

# SIMULATIONS RESULTS

- Turns on slippery road surface (with  $\hat{\delta} = -0.05$ ):
  - ▶ several values tested:  $\hat{s} = 0$ ,  $s = 0.20$ ,  $s = 0.30$ ,  $s = 0.35$



- Good degree of robustness with respect to slip mismatches

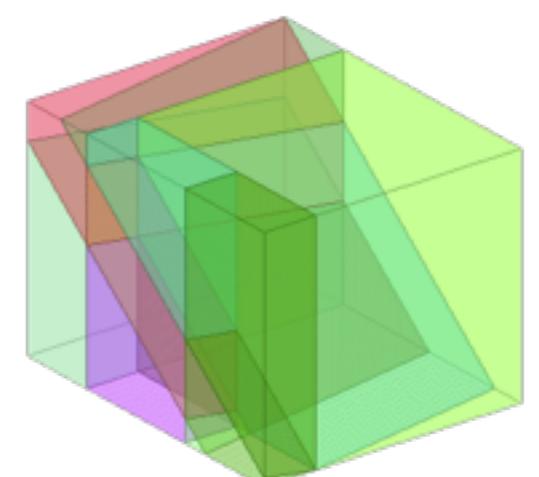
# SIMULATIONS RESULTS

- **Computational issues:**

- ▶ MPC-based approach is **viable for experimental tests** (average CPU time 17ms, worst-case CPU time 63 ms in MATLAB), but requires a MIQP **solver** in the ECU

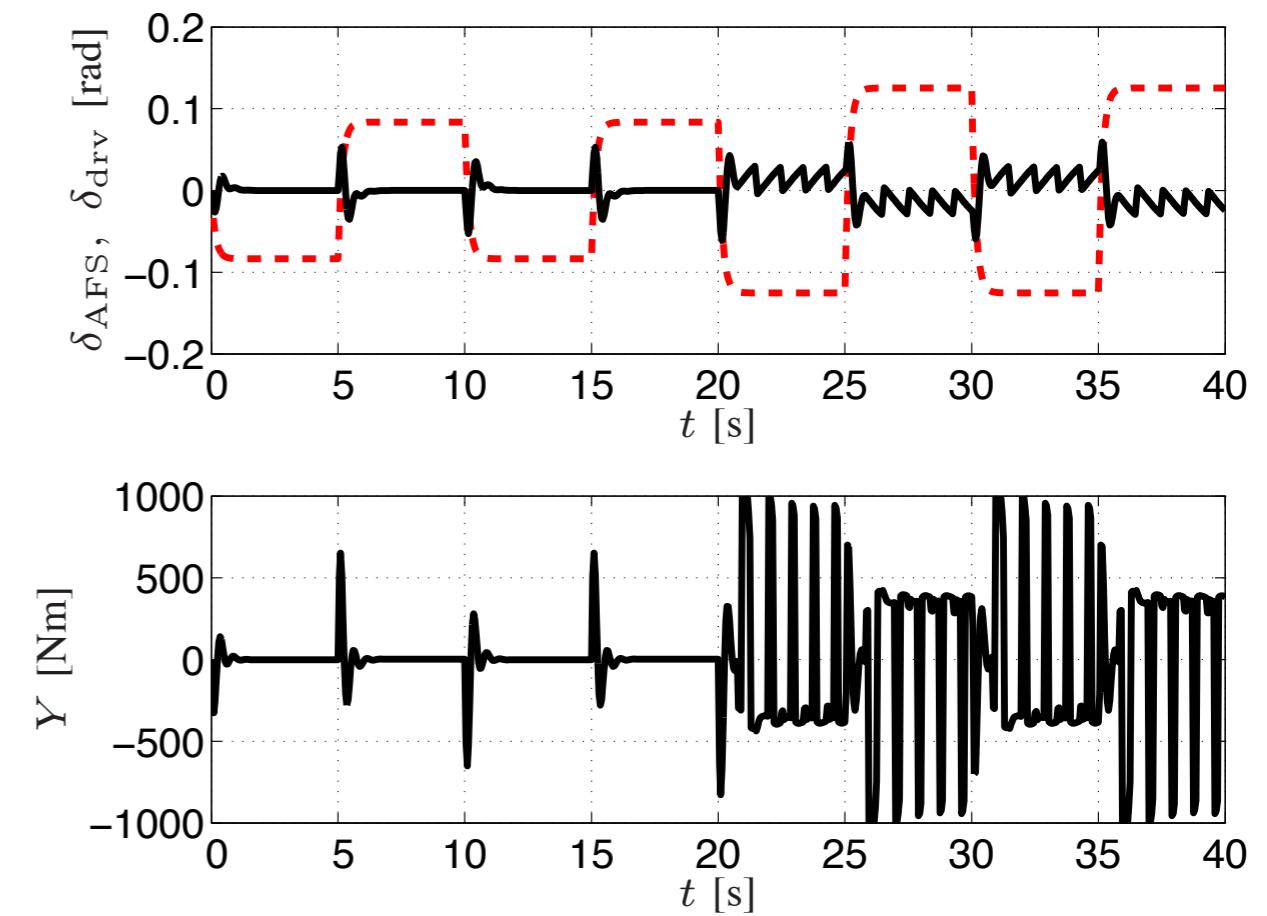
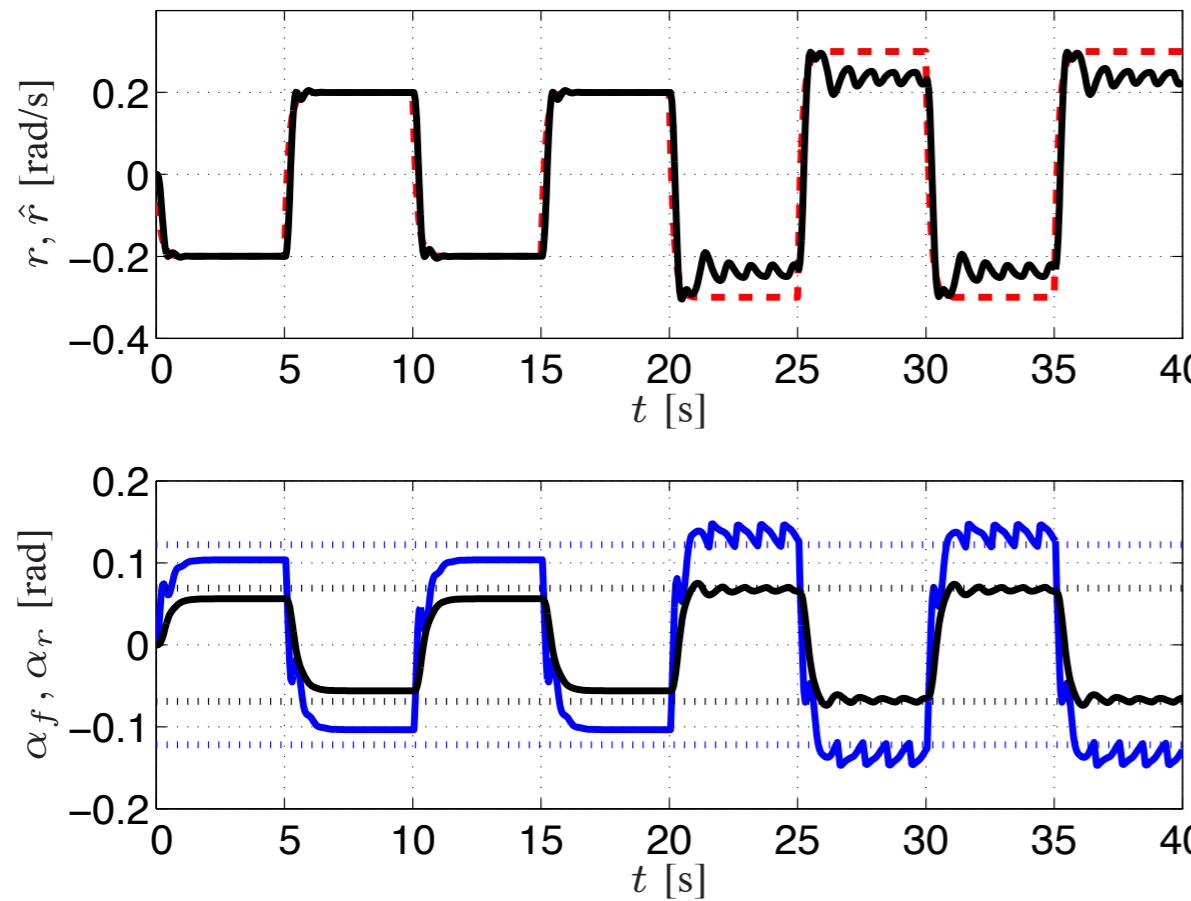
- **Explicit solution** of hybrid MPC control problem:

- ▶ Exploits **multiparametric programming** techniques to provide a description of the control law as an **explicit** function of the state
- ▶ All the computation is executed **off-line**, only simple set-membership tests and function evaluations are performed on-line to compute the control action
- ▶ However, the explicit solution requires **memory** (around **5000 polytopes** to be stored, in this case!)
- ▶ An **approximation** of the solution is needed for real implementation



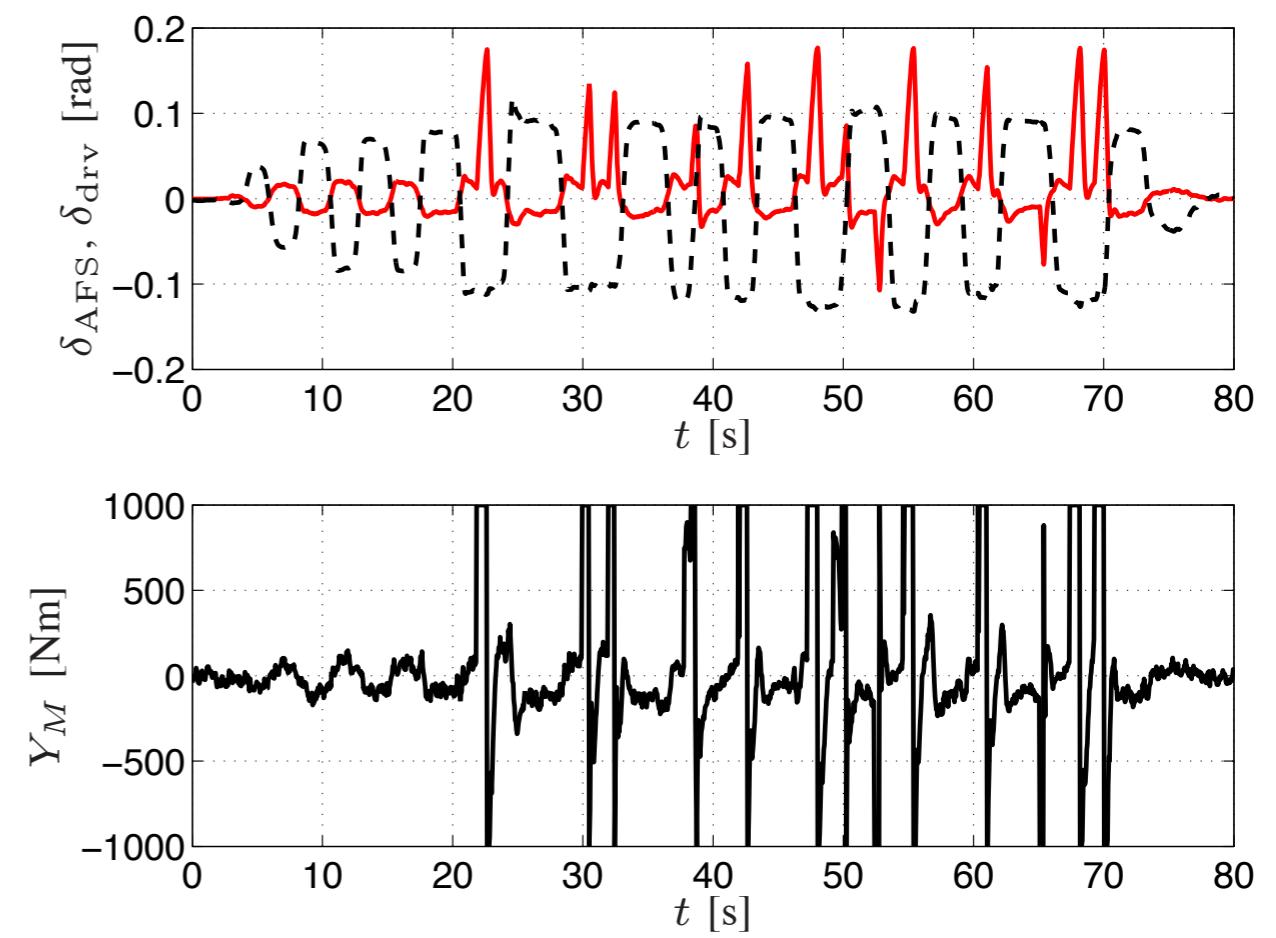
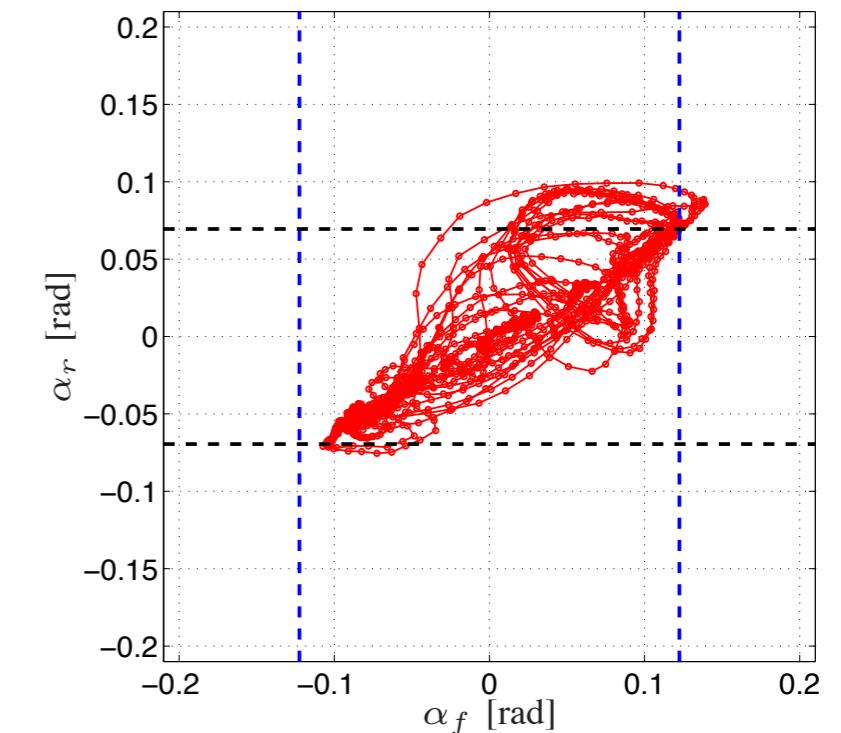
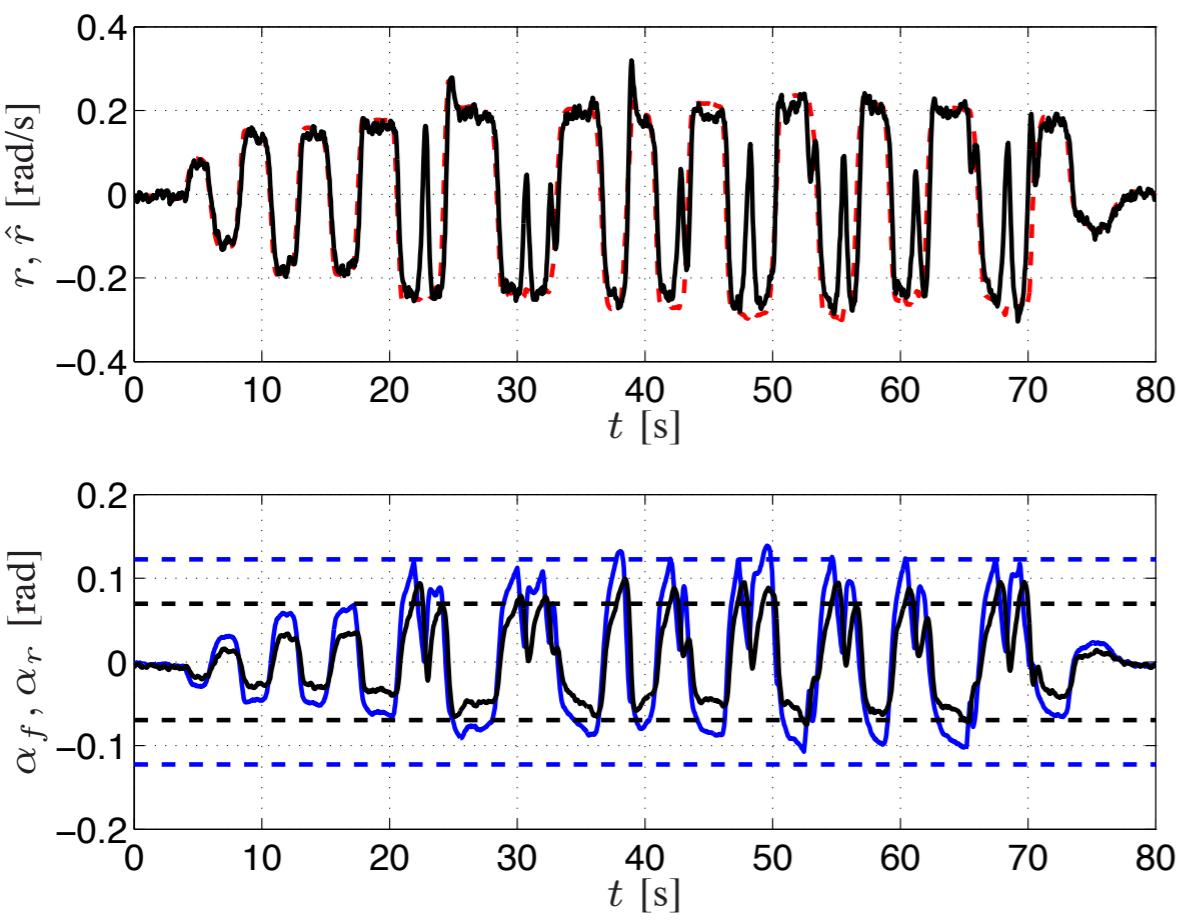
# SWITCHED MPC SOLUTION

- Simpler solution: assume PWA mode remains constant in prediction
- Design a linear MPC controller for each mode, make it explicit
- 4 linear explicit MPC's are enough (linear/saturation  $\times$  front/rear)
- Simulation results:



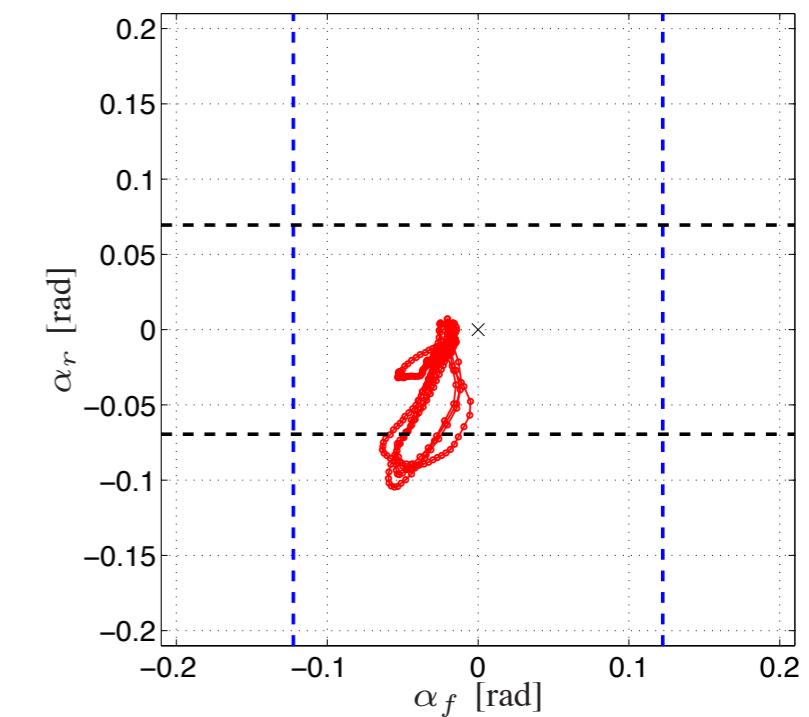
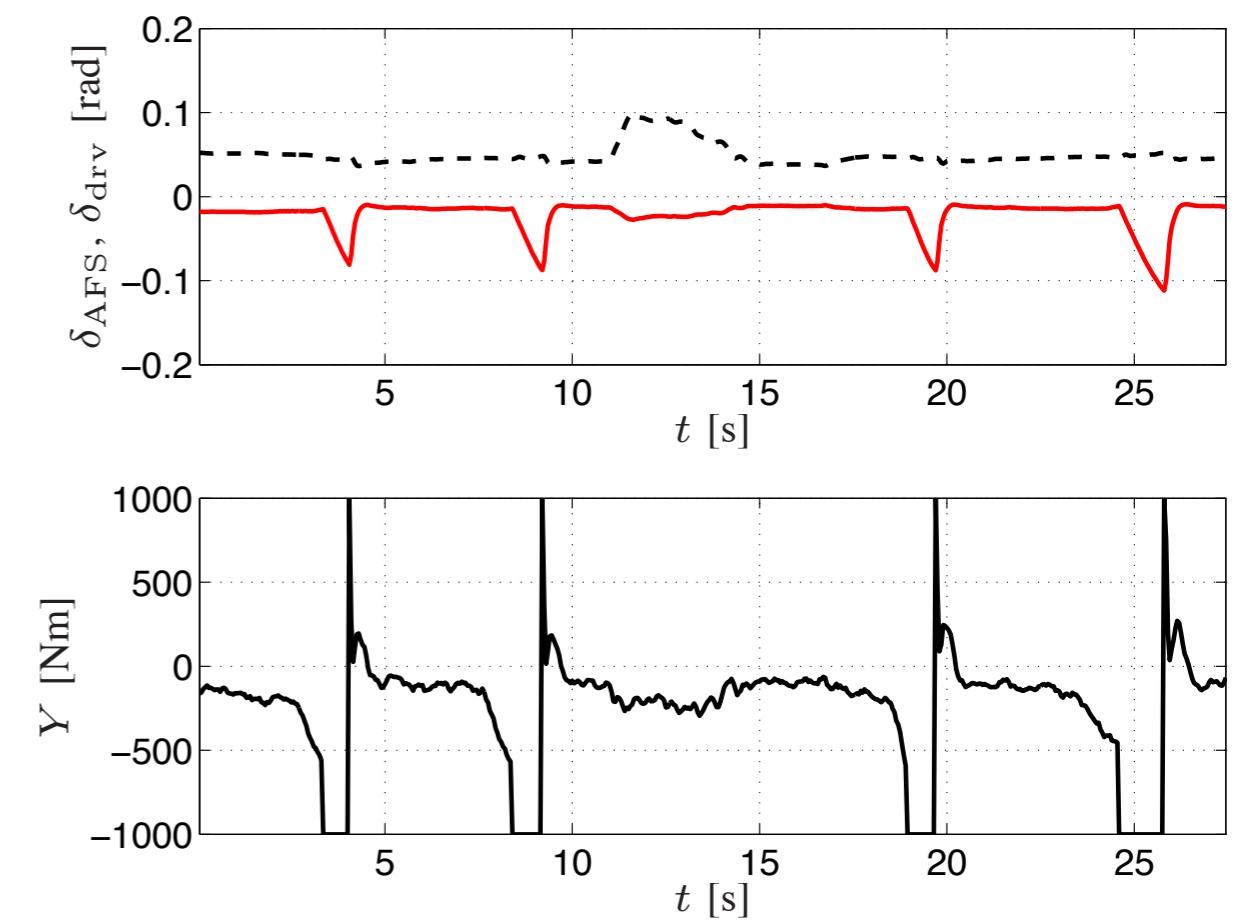
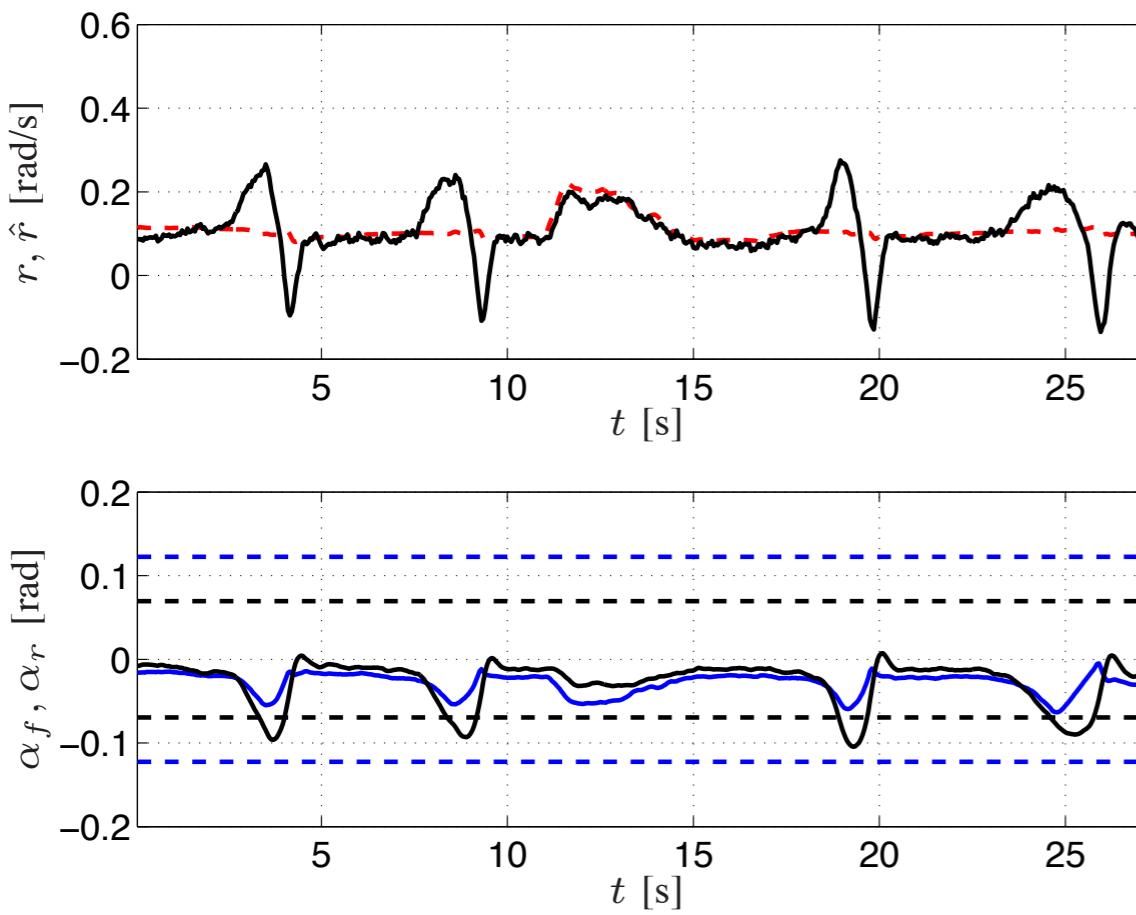
# EXPERIMENTS

- Experimental results (slalom test):



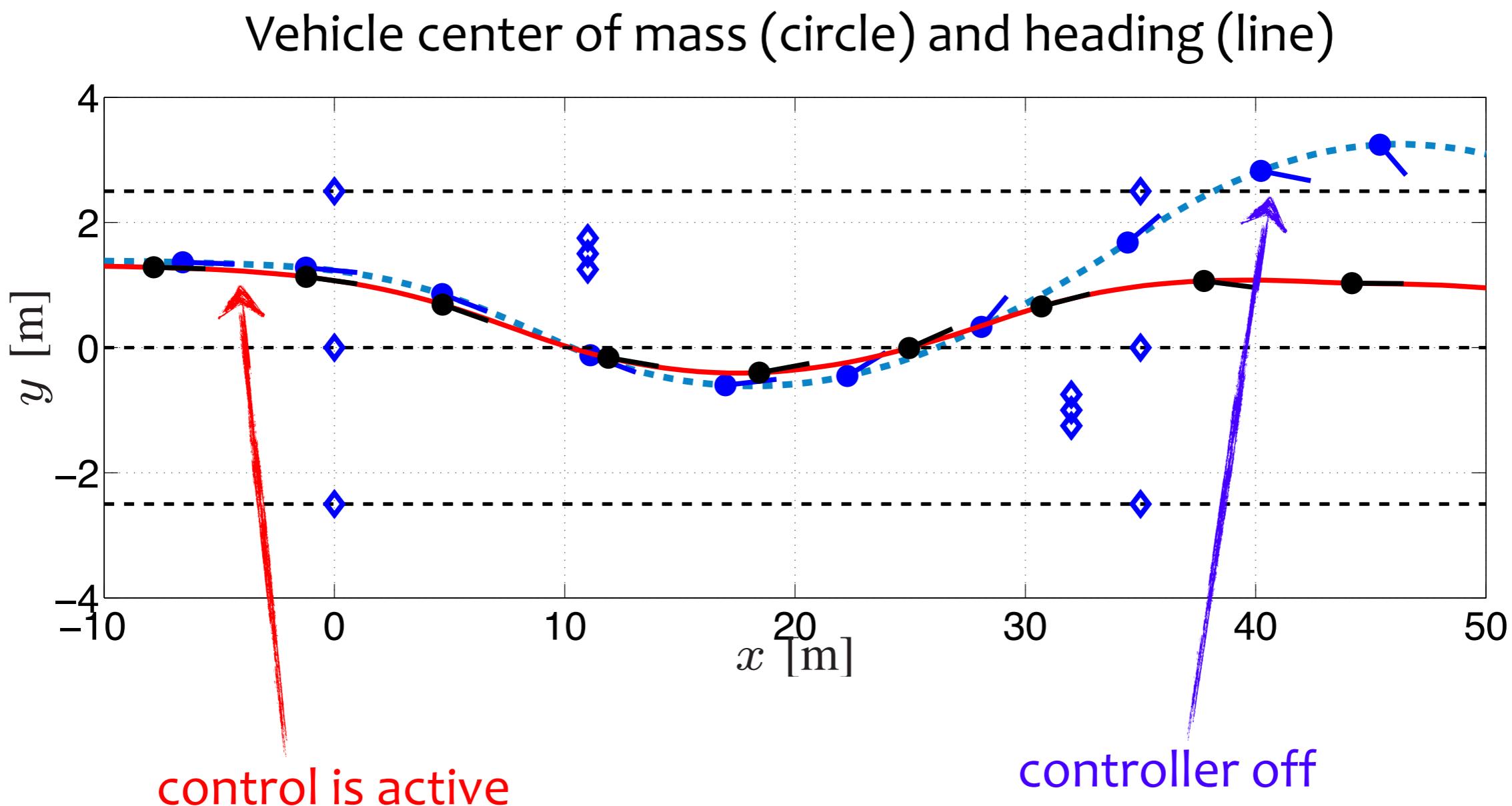
# EXPERIMENTS

- Experimental results (stability recovery test):



# EXPERIMENTS

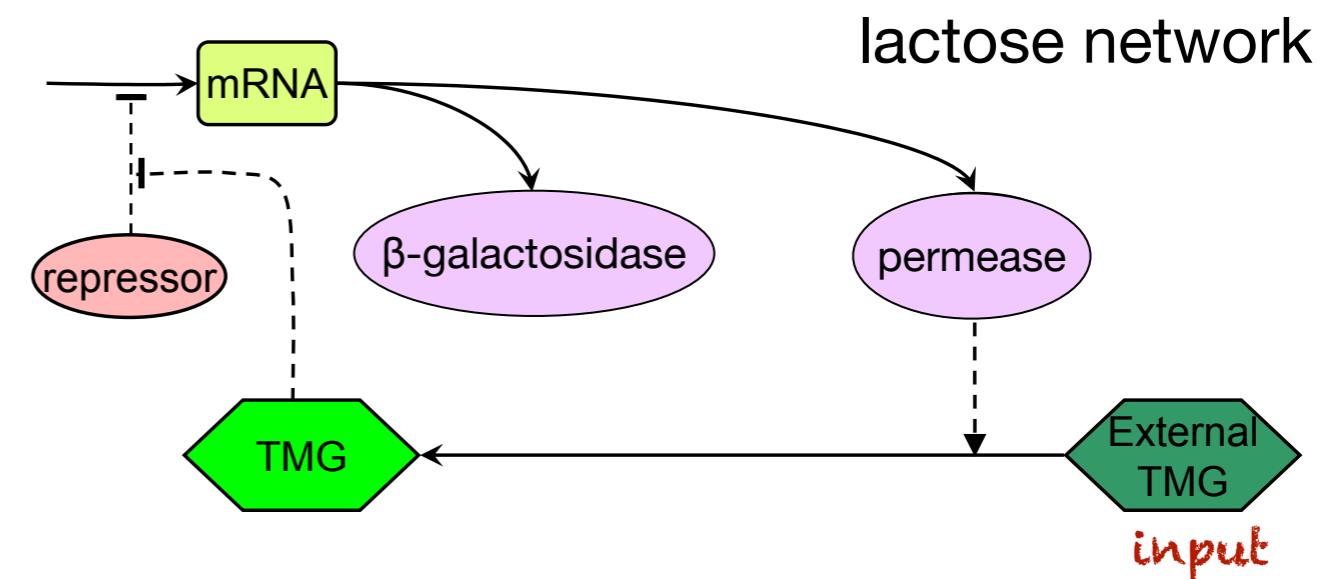
- Experimental results (double lane change):



# HYBRID MPC OF INDUCTION OF ESCHERICHIA COLI

(Julius, Sakar, Bemporad, Pappas, 2007)

- Goal: control the lactose regulation system of a colony of *E. coli*

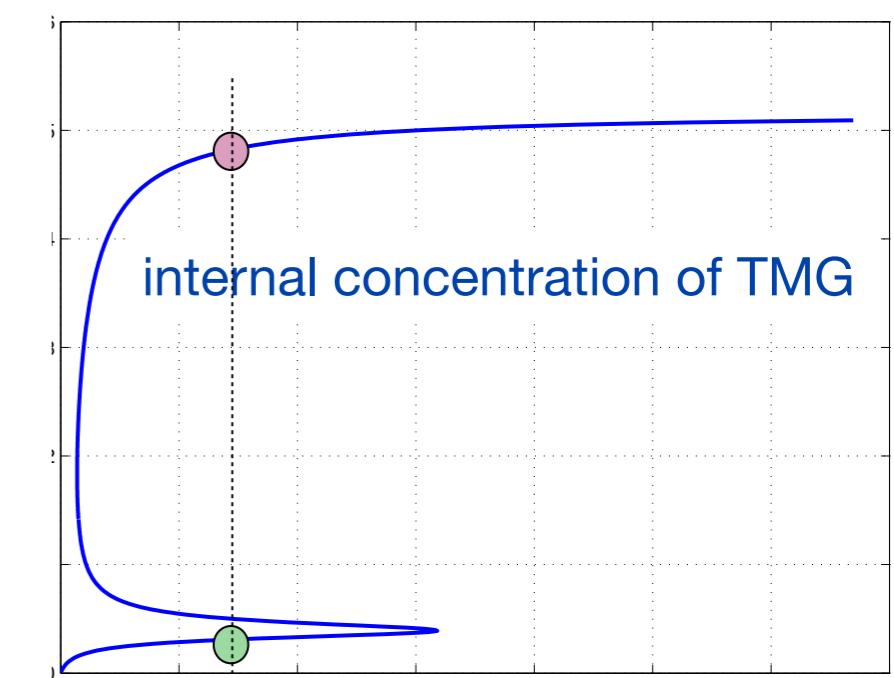
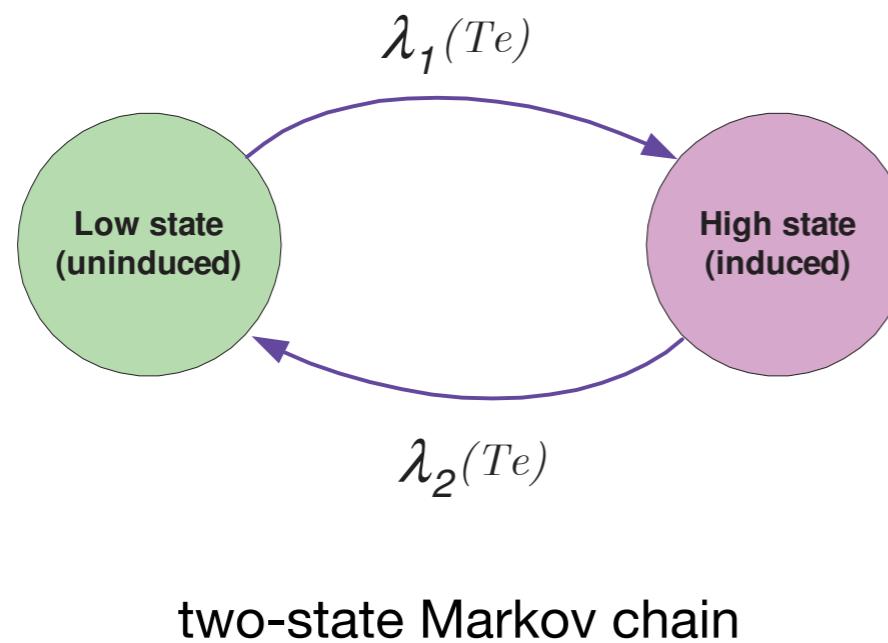


TMG = *thio-methyl galactosidase* concentration

- Model, measurements, and actuation are at the **entire colony** level

# HYBRID MPC OF INDUCTION OF ESCHERICHIA COLI

- **Bistable** lactose regulation system of E. coli



$T_e$  = external concentration of TMG

- The probabilities  $x_{lo}$ ,  $x_{hi}$  to be in low/high state satisfy the dynamics

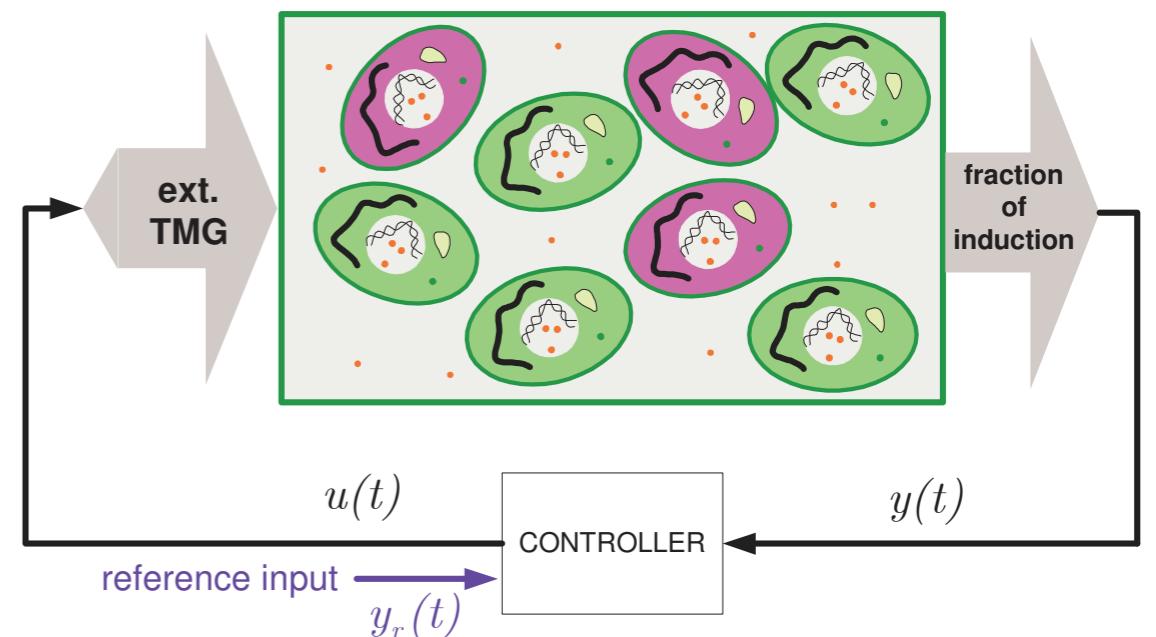
$$\frac{d}{dt} \begin{bmatrix} x_{lo} \\ x_{hi} \end{bmatrix} = \begin{bmatrix} -\lambda_1(T_e) & \lambda_2(T_e) \\ \lambda_1(T_e) & -\lambda_2(T_e) \end{bmatrix} \begin{bmatrix} x_{lo} \\ x_{hi} \end{bmatrix}$$

- Transition rates  $\lambda_1$ ,  $\lambda_2$  modeled as **piecewise constant** functions of  $T_e$

$T_e[10^{-3}\text{mM}]$	$\lambda_1(T_e)[\text{min}^{-1}]$	$\lambda_2(T_e)[\text{min}^{-1}]$
[1.4, 1.5)	$8.68 \cdot 10^{-4}$	$5.91 \cdot 10^{-3}$
[1.5, 1.6)	$9.27 \cdot 10^{-4}$	$3.61 \cdot 10^{-3}$
[1.6, 1.7)	$1.13 \cdot 10^{-3}$	$2.36 \cdot 10^{-3}$
[1.7, 1.8)	$1.39 \cdot 10^{-3}$	$1.54 \cdot 10^{-3}$
[1.8, 1.9)	$1.67 \cdot 10^{-3}$	$9.53 \cdot 10^{-4}$
[1.9, 2.0)	$1.93 \cdot 10^{-3}$	$5.54 \cdot 10^{-4}$

# HYBRID MPC OF INDUCTION OF ESCHERICHIA COLI

- Hybrid MPC problem
  - switched linear system
  - constraints on input  $T_e$  and  $dT_e/dt$
  - penalties on tracking error  $y - y_r$  and input rate  $dT_e/dt$



- Closed-loop results
  - MPC controller developed with **Hybrid Toolbox** in MATLAB
  - Mixed-Integer Linear Program solver GLPK
  - solution time: **32 ms** (worst case=**280 ms**) on 1.2 GHz laptop
  - sampling time = **10 min**

