

# MODEL PREDICTIVE CONTROL

## HYBRID MPC EXAMPLES

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# COURSE STRUCTURE

- ✓ Basic concepts of model predictive control (MPC) and linear MPC
- ✓ Linear time-varying and nonlinear MPC
- ✓ MPC computations: quadratic programming (QP), explicit MPC
  - Hybrid MPC
  - Stochastic MPC
  - Data-driven MPC

## Course page:

[http://cse.lab.imtlucca.it/~bemporad/mpc\\_course.html](http://cse.lab.imtlucca.it/~bemporad/mpc_course.html)

# HYBRID MPC EXAMPLES

# 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 MPC FOR CRUISE CONTROL - 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

power balance:

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

$i$  = gear

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

# HYBRID MPC FOR CRUISE CONTROL - MODEL

- Gear selection: for each gear # $i$ , define a binary input  $g_i \in \{0, 1\}$   
 $i = R, 1, 2, 3, 4, 5$

- Gear selection (traction force):

$$F_e = \frac{R_g(i)}{k_s} C \quad \text{depends on gear } \#i$$

define auxiliary continuous variables:



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

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

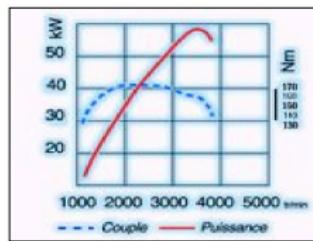
- Gear selection (engine/vehicle speed):

$$\omega = \frac{R_g(i)}{k_s} \dot{x} \quad \text{similarly, also requires 6 auxiliary continuous variables}$$

# HYBRID MPC FOR CRUISE CONTROL - MODEL

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

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



Piecewise-linearization  
(PWL Toolbox, Julián, 2000)



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 !

# HYBRID MPC FOR CRUISE CONTROL - HYSDEL

```
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 Fe, 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;  
        dPWL2 -> dPWL1; dPWL3 -> dPWL2;  
        dPWL3 -> dPWL1; dPWL2 -> dPWL1;  
    }  
}
```

go to demo /demos/cruise/init.m

# HYBRID MPC FOR CRUISE CONTROL - MLD MODEL

- MLD model

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

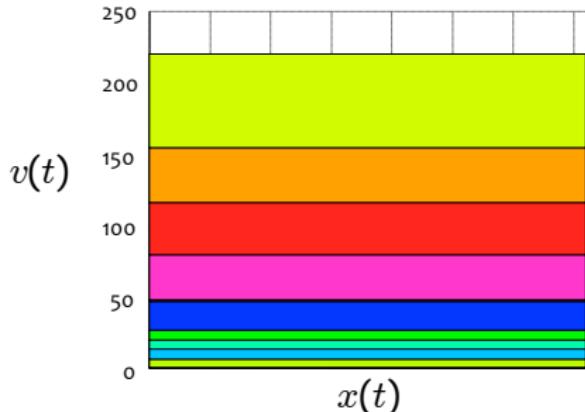
- 2 continuous states:  $x, v$  (vehicle position and speed)
- 2 continuous inputs:  $C, F_b$  (engine torque, brake force)
- 6 binary inputs:  $g_R, g_1, g_2, g_3, g_4, g_5$  (gears)
- 1 continuous output:  $v$  (vehicle speed)
- 18 auxiliary continuous vars: (6+1 traction force, 6+1 engine speed, 4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

# HYBRID MPC FOR CRUISE CONTROL - CONTROLLER

- Maximum-speed controller

$$\begin{aligned} \max_{u(t)} \quad & v(t+1|t) \\ \text{s.t.} \quad & \text{MLD model} \\ & v(t|t) = v(t) \\ & x(t|t) = x(t) \end{aligned}$$

Objective: maximize speed  
(to reproduce max acceleration plots)



$(x(t)$  is irrelevant)

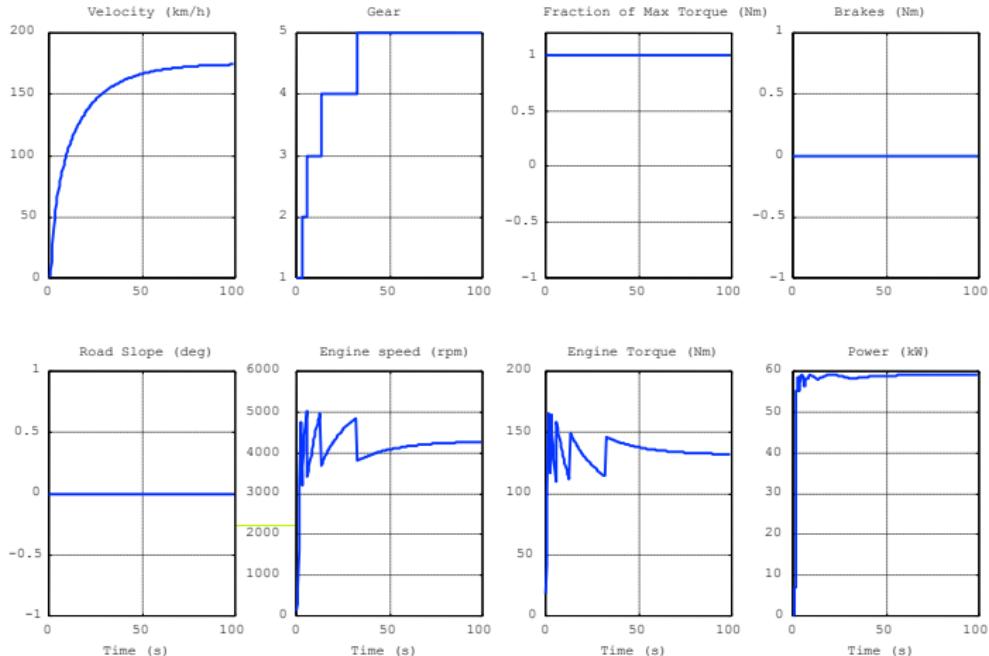
## MILP optimization problem

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>

(parameters: Renault Clio 1.9 DTI RXE)

# HYBRID MPC FOR CRUISE CONTROL - RESULTS

- Maximum-speed controller



# HYBRID MPC FOR CRUISE CONTROL - CONTROLLER

- Cruise controller

$$\min_{u(t)} |v(t+1|t) - v_d(t)| + \rho |\omega|$$

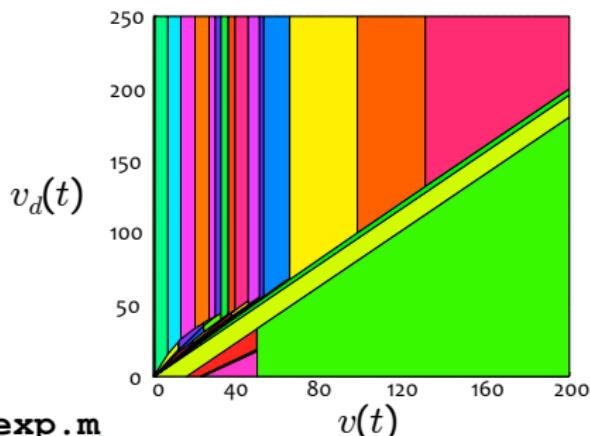
s.t. MLD model

$$v(t|t) = v(t)$$

$$x(t|t) = x(t)$$

## 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>

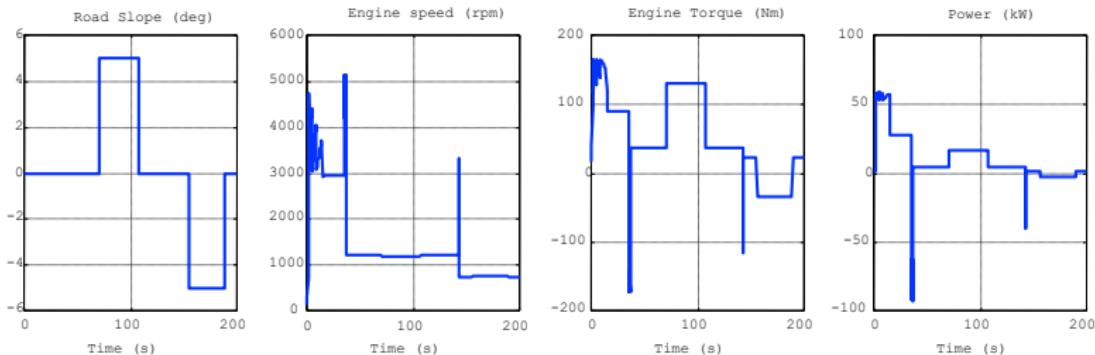
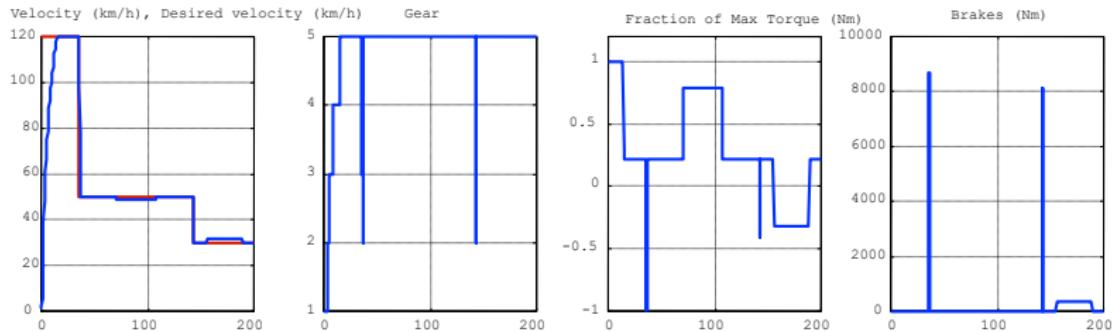


go to demo /demos/cruise/init\_exp.m

# HYBRID MPC FOR CRUISE CONTROL - RESULTS

- **Cruise controller**

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



# HYBRID MPC FOR CRUISE CONTROL - CONTROLLER

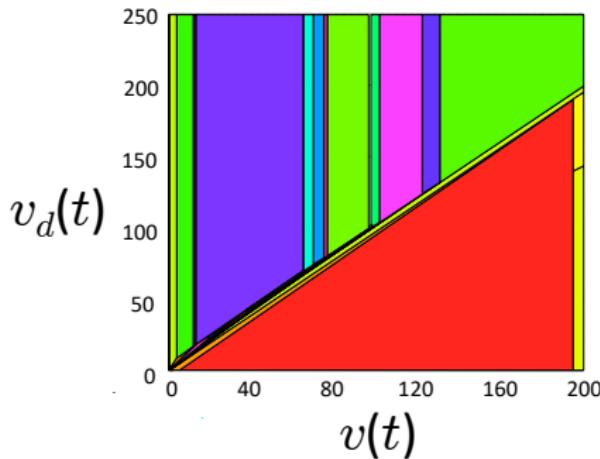
- Smoother cruise controller

$$\begin{aligned} \min_{u(t)} \quad & |v(t+1|t) - v_d(t)| + \rho|\omega| \\ \text{s.t.} \quad & \text{MLD model} \\ & |v(t+1|t) - v(t)| \leq a_{\max} T_s \\ & v(t|t) = v(t) \\ & x(t|t) = x(t) \end{aligned}$$



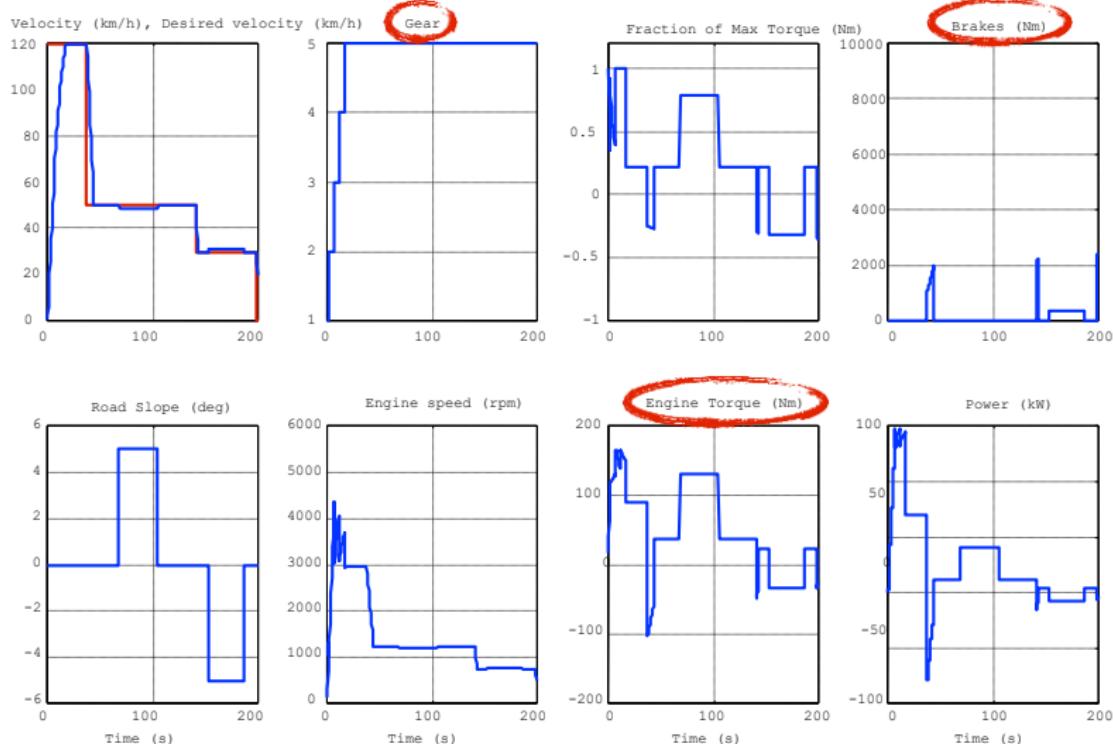
## MILP optimization problem

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 MPC FOR CRUISE CONTROL - RESULTS

- Smoother cruise controller



# HYBRID MPC FOR TRACTION CONTROL

# VEHICLE TRACTION CONTROL PROBLEM

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

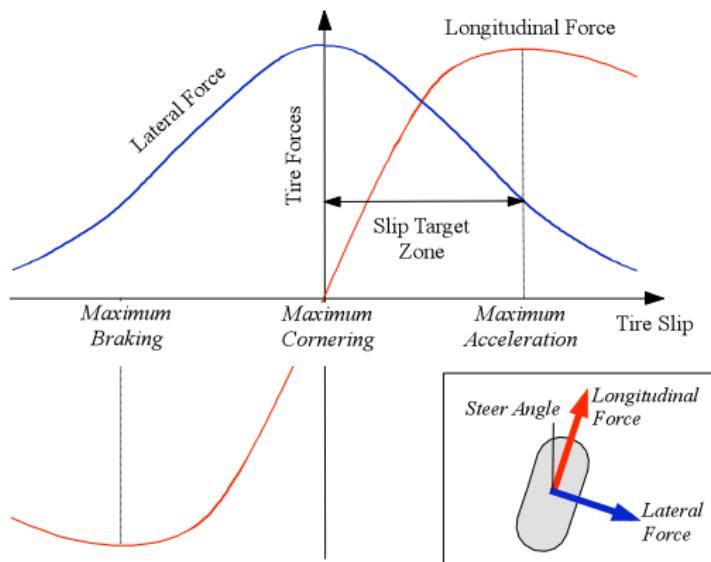


**Model:** nonlinear, uncertain, constraints

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

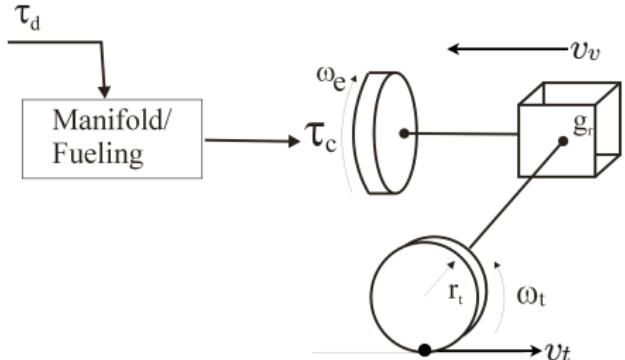
**Solution:** MLD hybrid framework + explicit hybrid MPC strategy

# 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

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

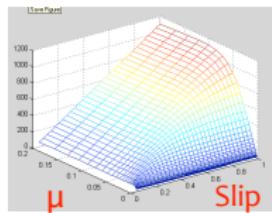
- Manifold/fueling dynamics

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

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

# HYBRIDIZATION OF TIRE CHARACTERISTICS

Torque



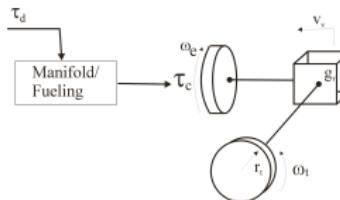
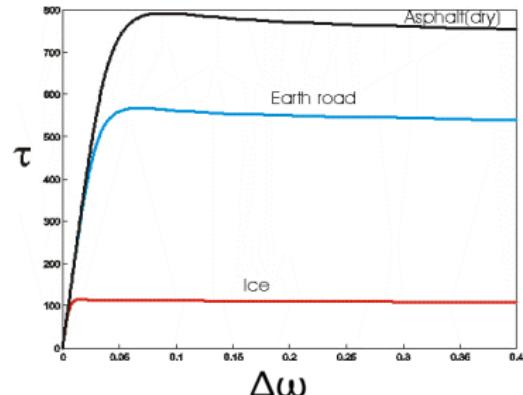
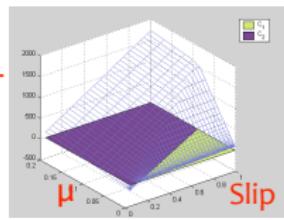
Nonlinear tire torque

$$\tau_t = f(\Delta\omega, \mu)$$



PWA approximation

Torque



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

# MLD TRACTION MODEL

$$\begin{aligned}x(t+1) &= Ax(t) + B_1u(t) + B_2\delta(t) + B_3z(t) + B_5 \\y(t) &= Cx(t) + D_1u(t) + D_2\delta(t) + D_3z(t) + D_5 \\E_2\delta(t) + E_3z(t) &\leq E_4x(t) + E_1u(t) + E_5\end{aligned}$$

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:

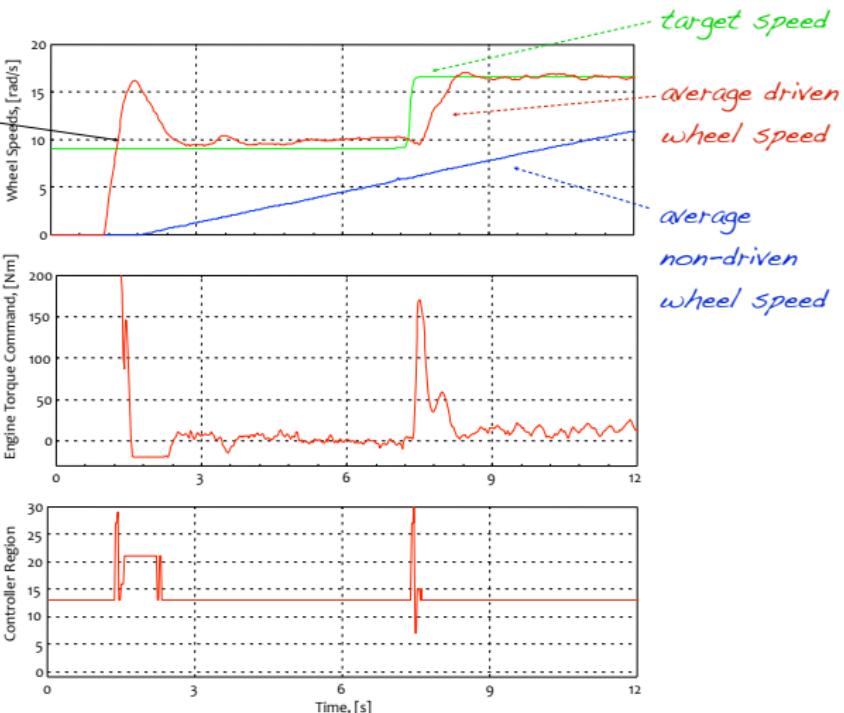
- Limits on the engine torque:

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

# EXPERIMENTAL RESULTS

controller is triggered ON

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



Ford Motor Company

# EXPERIMENTS

(Borrelli, Bemporad, Fodor, Hrovat, 2006)

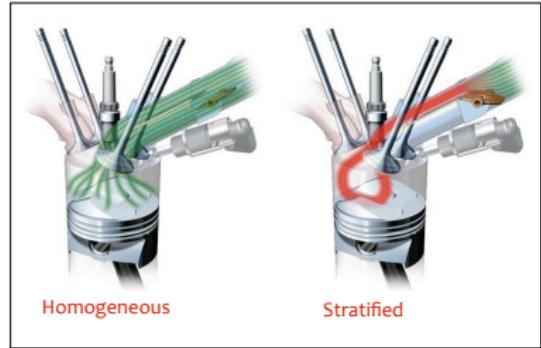


# HYBRID CONTROL OF A DISC ENGINE

# DISC ENGINE CONTROL PROBLEM

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

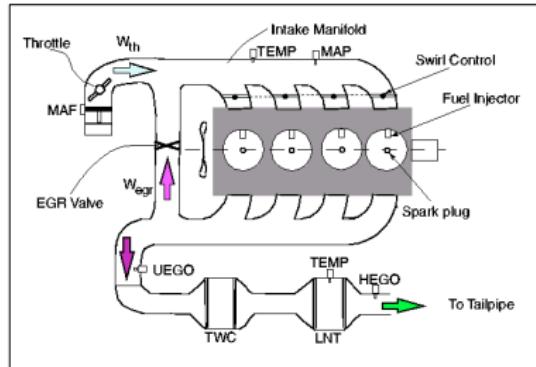
- automatically chooses operating **mode** (homogeneous/stratified)
- can cope with **nonlinear** dynamics
- handles **constraints** on A/F ratio, air-flow, spark
- achieves **optimal** performance (track desired torque and A/F ratio)



# DISC ENGINE

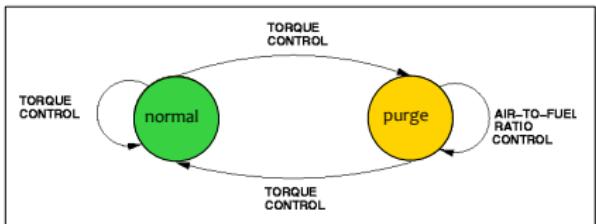
Two distinct regimes:

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

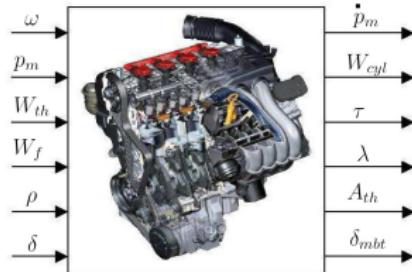
**Periodical cleaning** of the aftertreatment system needed ( $\lambda=14.00$ , homogeneous regime)



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)



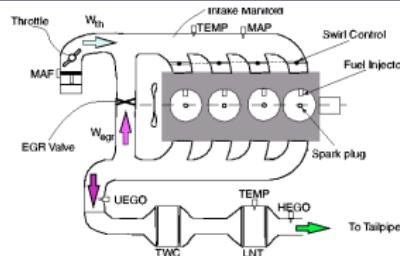
Dynamic equations are **nonlinear**, dynamics and constraints depend on regime  $\rho$

# DISC ENGINE 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:

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

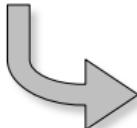
- In-cylinder Air-to-Fuel ratio:

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

- Engine torque:

$$\tau = \tau_{mfr} + \tau_{pump} + \tau_{ind} \quad \text{with } \tau_{mfr}, \tau_{pump} \text{ functions of } p_m$$

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



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

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

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

$N$  = control horizon

$x(t)$  = current state

# DISC ENGINE - HYSDEL

```
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, pml, 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
                  );
        }
    }
}

STATE{
    taul={(IF rho THEN taull*pm+...
            tau12*Wth+tau13*Wf+tau14*delta+tau1c
            ELSE tau01*pm+tau02*Wth...
            +tau03*Wf+tau04*delta+tau0c );
}

OUTPUT{
    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);
}

INPUT{
    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  $u(t)$

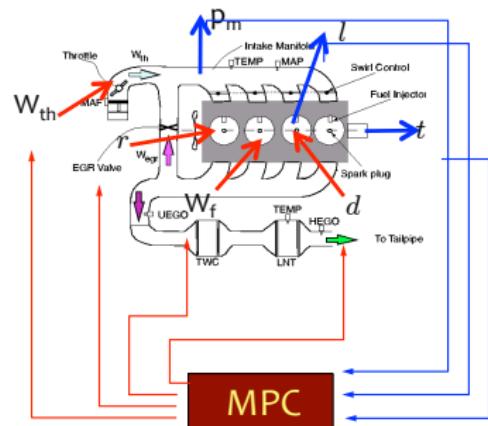
$$u(t) = [W_{th}(t), W_f(t), \delta(t), \rho(t)]$$

Weights:

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

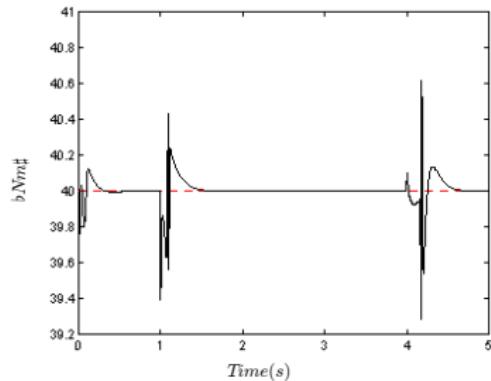
$$Q = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0.001 & 0 \\ 0 & 0 & 0.01 \end{pmatrix} \quad S = \begin{pmatrix} 0.04 & 0 & 0 \\ 0 & 1500 & 0 \\ 0 & 0 & 0.01 \end{pmatrix} \quad s_{\epsilon_T} \quad s_{\epsilon_\lambda}$$

main emphasis on torque



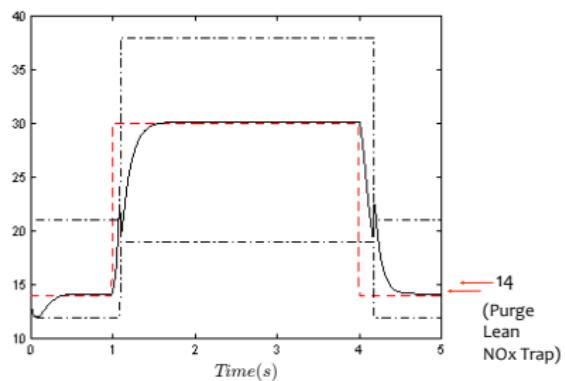
# SIMULATION RESULTS (NOMINAL ENGINE SPEED)

Engine brake torque

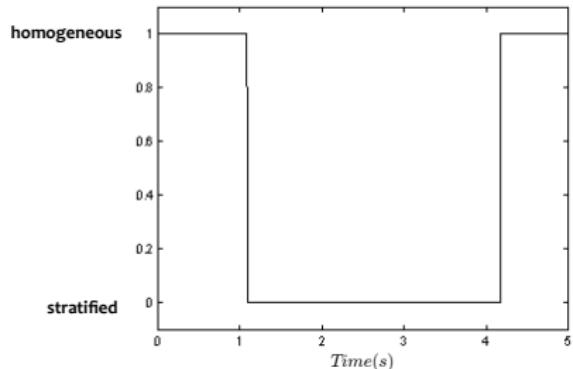


$\omega/2000 \text{ rpm}$

Air-to-fuel ratio



Combustion mode



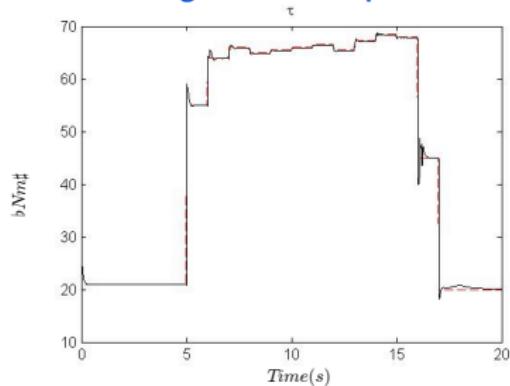
- Control horizon  $N=1$ ;
- Sampling time  $T_s=10 \text{ ms}$ ;
- PC Xeon 2.8 GHz + Cplex 9.1



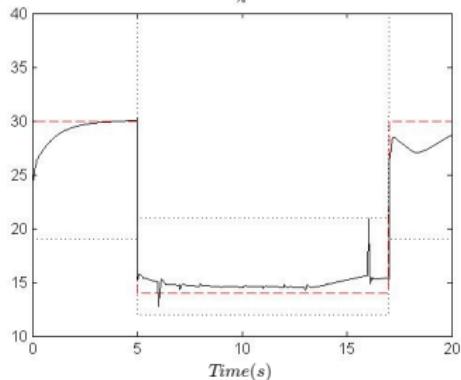
$\approx 3 \text{ ms per time step}$

# SIMULATION RESULTS (VARYING ENGINE SPEED)

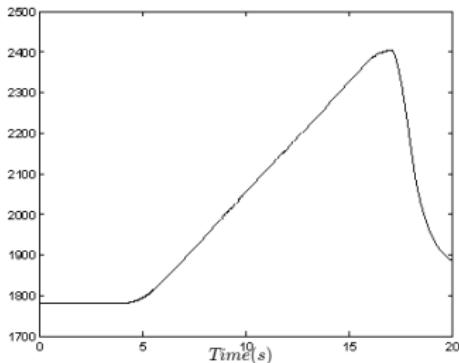
Engine Brake Torque



Air-to-Fuel Ratio



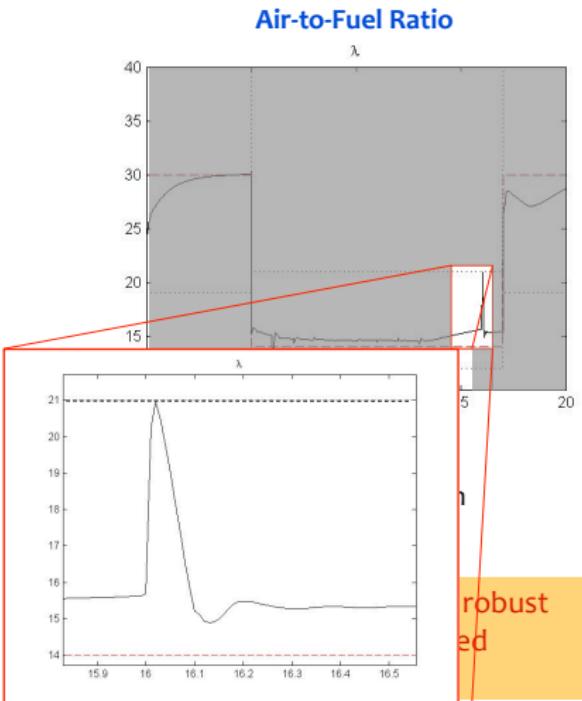
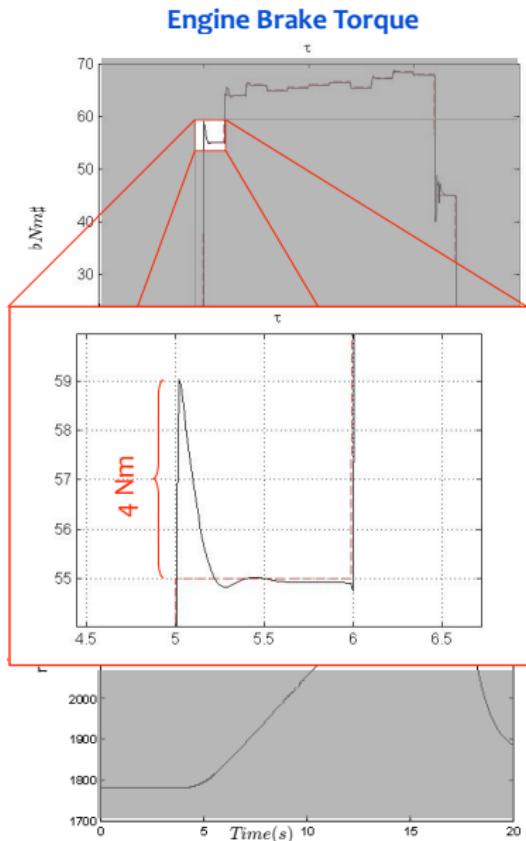
Engine speed



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

Explicit control law:

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

N=1 (control horizon)

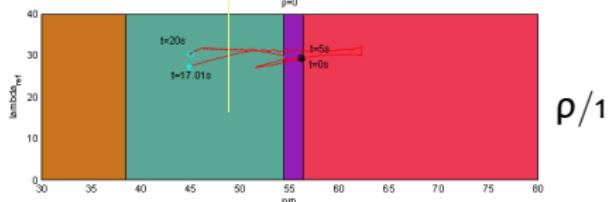
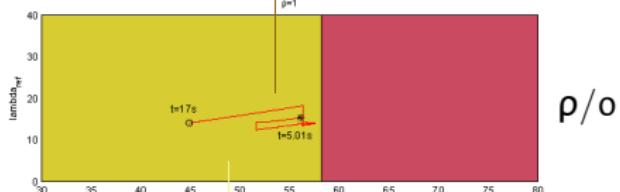
where:  $u = [W_{th} \ W_f \ \delta \ \rho]'$

$$\theta = [p_m \ \epsilon_\tau \ \epsilon_\lambda \ \tau_{ref} \ \lambda_{ref}]$$

$$p_{m,ref} \ W_{th,ref} \ W_{f,ref} \ \delta_{ref}]'$$

42 partitions

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



- Time to compute explicit MPC:

$\approx 3s$ ;

- Sampling time  $T_s = 10$  ms;

- PC Xeon 2.8 GHz + Cplex 9.1

$\rightarrow 8 \mu\text{s}$  per time step

$\approx 3ms$  on

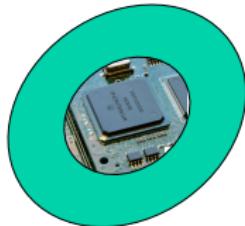
$\mu$ -controller

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

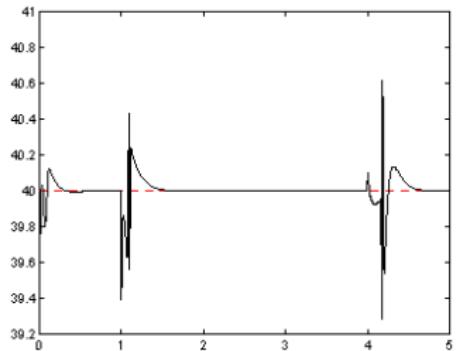


747 partitions

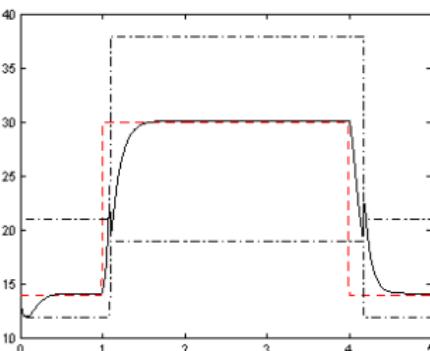
$$\theta = [p_m \ \epsilon_\tau \ \epsilon_\lambda \ \tau_{ref} \ \lambda_{ref}$$

$$p_{m,ref} \ W_{th,ref} \ W_{f,ref} \ \delta_{ref}]'$$

Engine Brake Torque



Air-to-Fuel Ratio



Closed-loop N=2

Closed-loop N=1

adequate !

# EXPLICIT HYBRID MPC OF SEMIACTIVE SUSPENSIONS

# ACTIVE SUSPENSIONS

(Giorgetti, Bemporad, Tseng, Hrovat, 2006)

## Active Suspension System Ford Mercur XR 40i



active  
suspensions

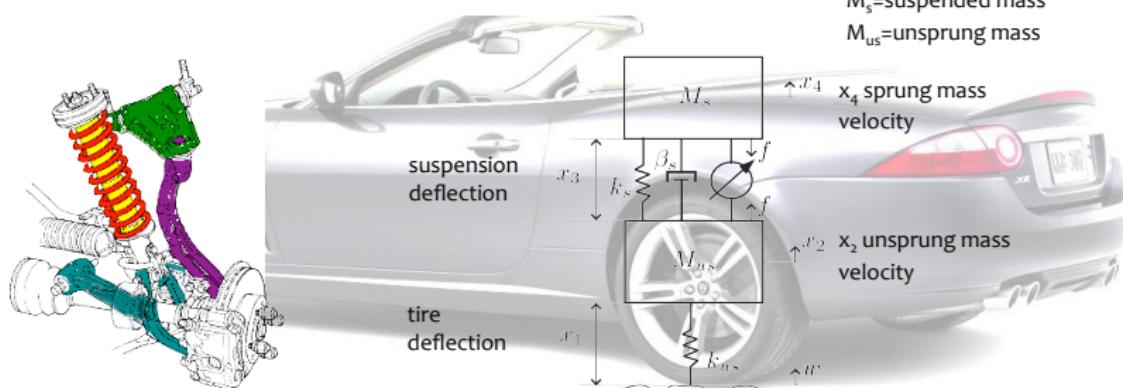


passive  
suspensions

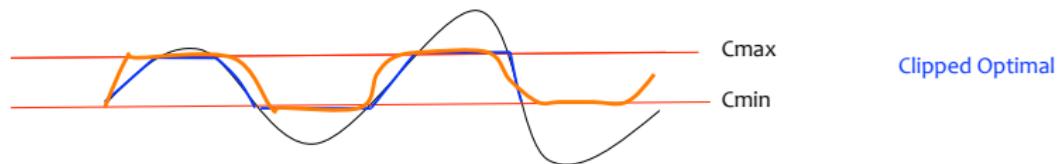


*Ford Motor Company*

# QUEST OF OPTIMAL SEMI-ACTIVE SUSPENSIONS



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

— ? — = —

# SUSPENSION 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

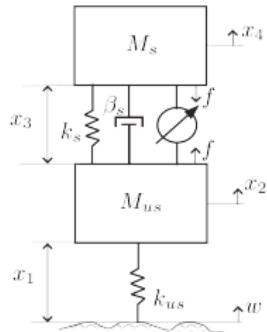
$$\rho = \frac{M_s}{M_{us}}, \omega_{us} = \sqrt{\frac{k_{us}}{M_{us}}}, \omega_s = \sqrt{\frac{k_s}{M_s}}, \zeta = \frac{\beta_s}{2\sqrt{M_s k_s}}, \bar{f} = \frac{f}{M_s}$$

- Output:

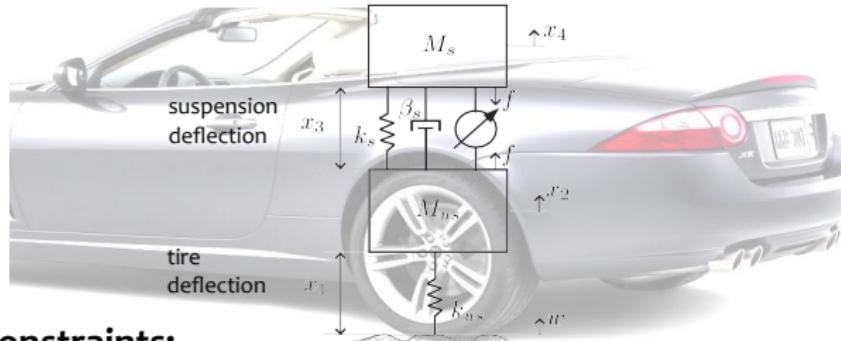
$$y = \frac{dx_4}{dt} = [0 \ 2\zeta\omega_s \ -\omega_s^2 \ -2\zeta\omega_s] x - \bar{f}$$

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

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



# CONSTRAINTS ON SUSPENSION MODEL



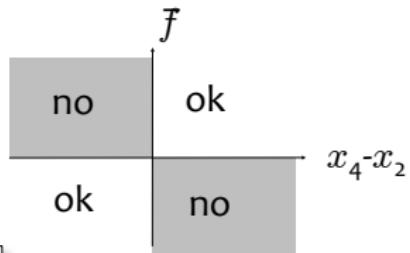
Quarter-car model

→ 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:

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

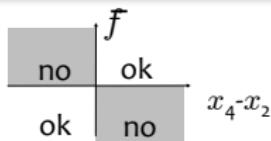
(1), (2) are nonlinear & nonconvex physical constraints

hybrid model

# HYBRID MODEL OF SEMIACTIVE CONSTRAINTS

1) Passivity condition:

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



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

2) Max dissipation power:

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



$$F \geq 0$$

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:

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



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

# SEMIACTIVE SUSPENSION MODEL - HYSDEL

```
/* 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)
```

get the MLD model in MATLAB

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

simulate the MLD model

# SEMIACTIVE SUSPENSION MODEL - PWA FORM

- 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

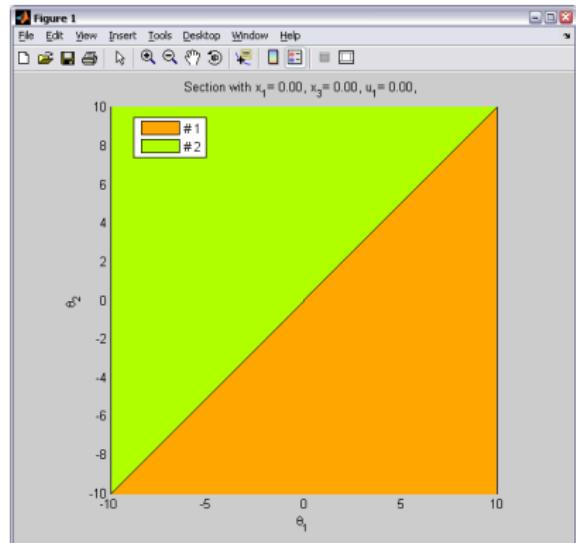
$$(x_1, x_2, x_3, x_4)$$

- 1 continuous input

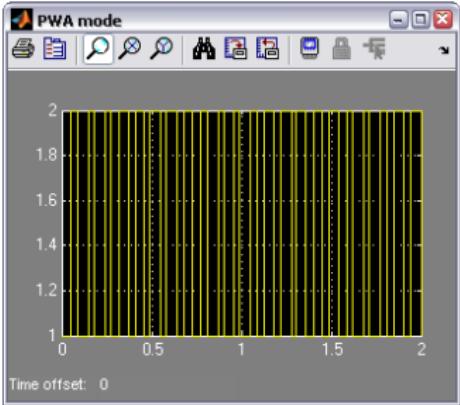
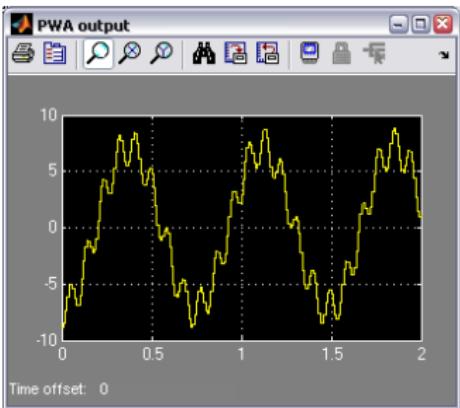
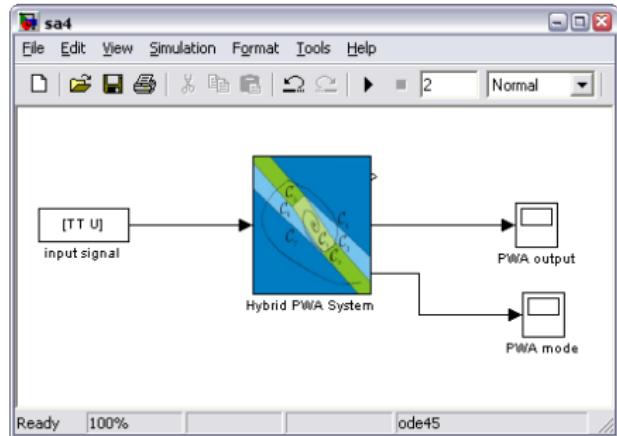
(normalized adjustable damping force  $\bar{f}$ )

- 2 polyhedral regions

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



# OPEN-LOOP SIMULATION OF PWA SUSPENSION MODEL



# MPC PERFORMANCE SPECIFICATIONS

tire deflection

suspension  
deflection

vertical  
acceleration

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

terminal weight  
(Riccati matrix)

# CLOSED-LOOP MPC RESULTS

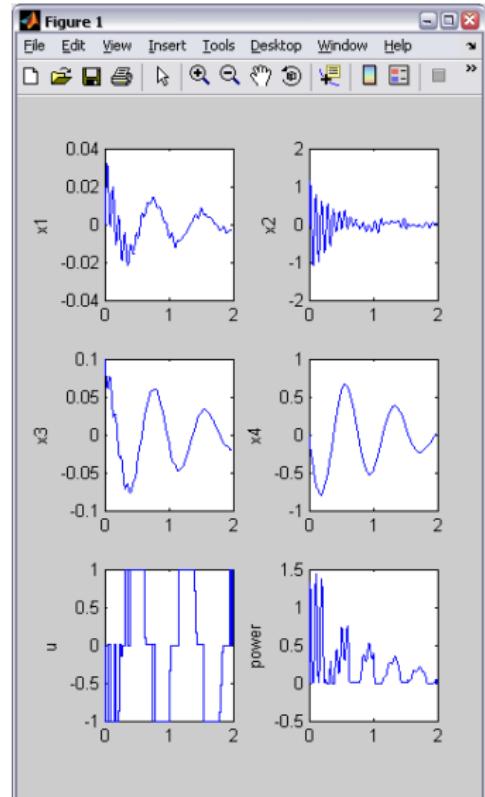
$$J = \int (q_1 x_1^2 + q_3 x_3^2 + \dot{x}_4^2)$$

```
>>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;
```

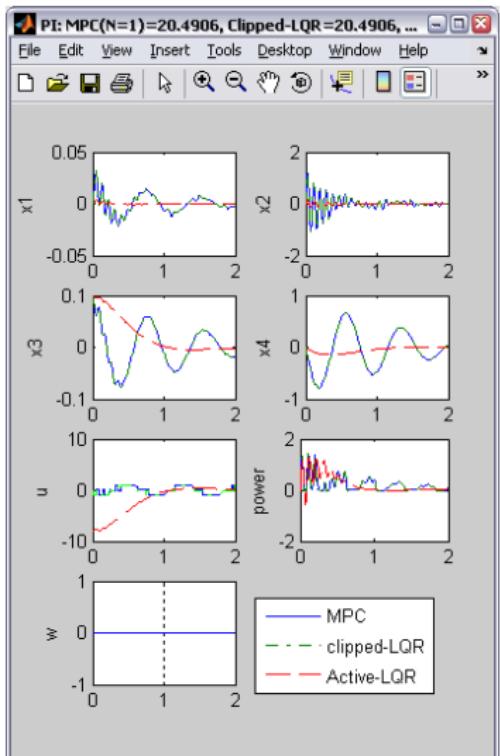
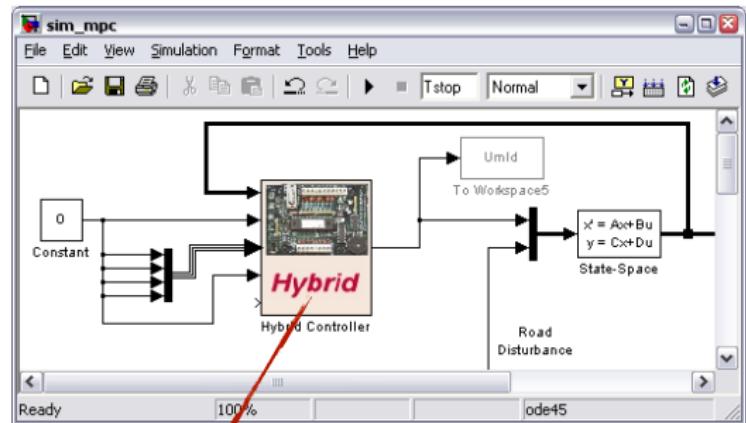
```
>>C=hybcon(S,Q,N,limits,refs);
```

```
>> 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

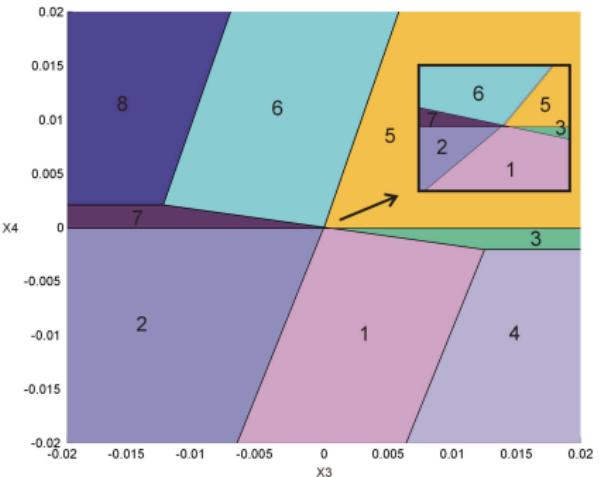
Explicit controller (based on hybrid controller C)
  4 parameter(s)
  1 input(s)
  8 partition(s)
sampling time = 0.01

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

This is a state-feedback controller.

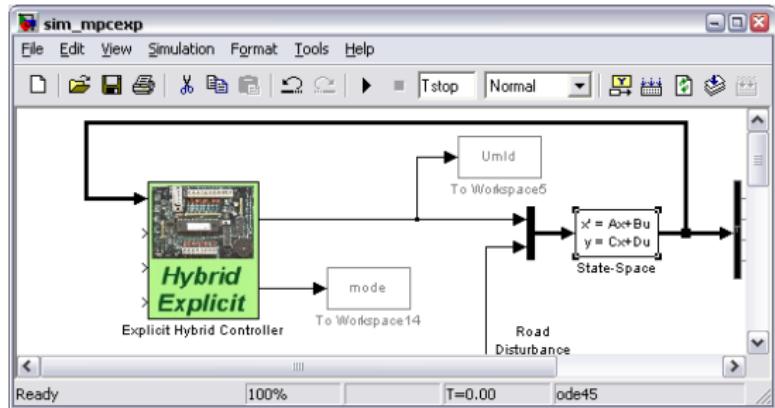
Type "struct(E)" for more details.
>>
```

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



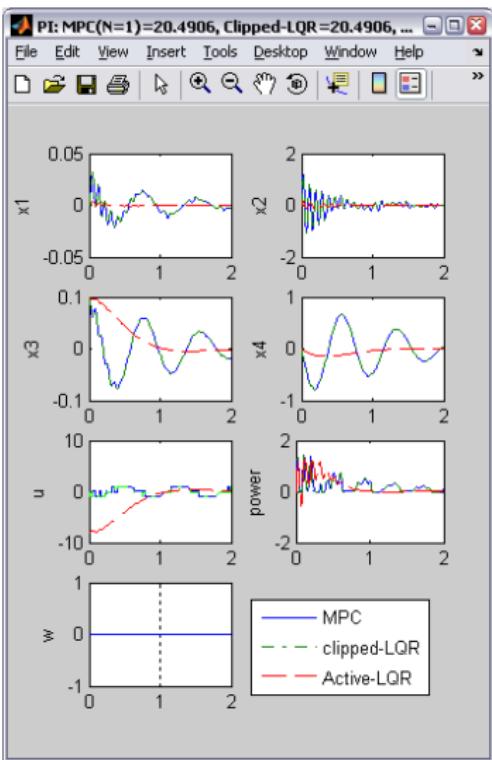
$$u(x) = \begin{cases} 10.4748x_1 + 0.2446x_2 & +79.1519x_3 - 3.9235x_4 \\ (= K_{LQ}) & \text{Regions } \#1, \#6 \\ 0 & \text{Regions } \#2, \#5 \\ (2 \cdot 25.5 \cdot \omega_s)(x_4 - x_2) & \text{Regions } \#3, \#7 \\ -1 & \text{Region } \#4 \\ 1 & \text{Region } \#8 \end{cases}$$

# EXPLICIT HYBRID MPC

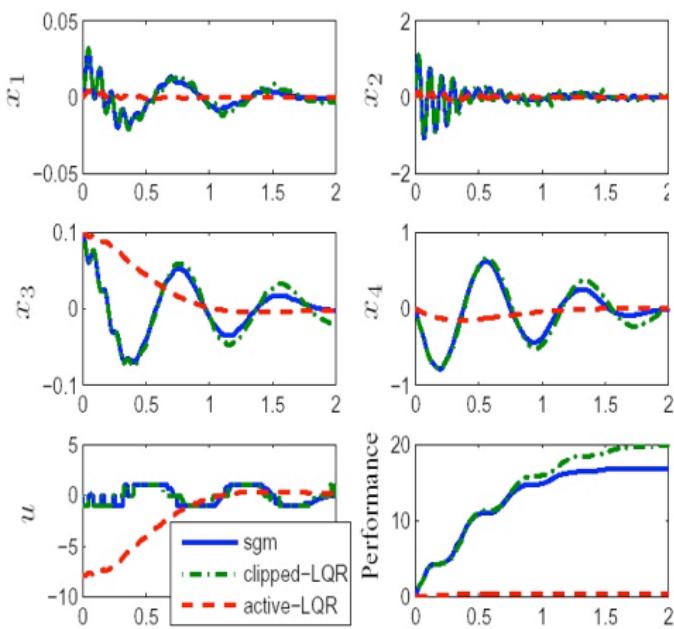


generated  
C-code

```
#define EXPCON_NU 1
#define EXPCON_NZ 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_NB 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	20.4282		
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			

# EXPLICIT HYBRID MPC - 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  $x(0)=[0 \ 0 \ 0]'$
- Simulation time  $T=20$  s, sampling time  $T_s=10$  ms

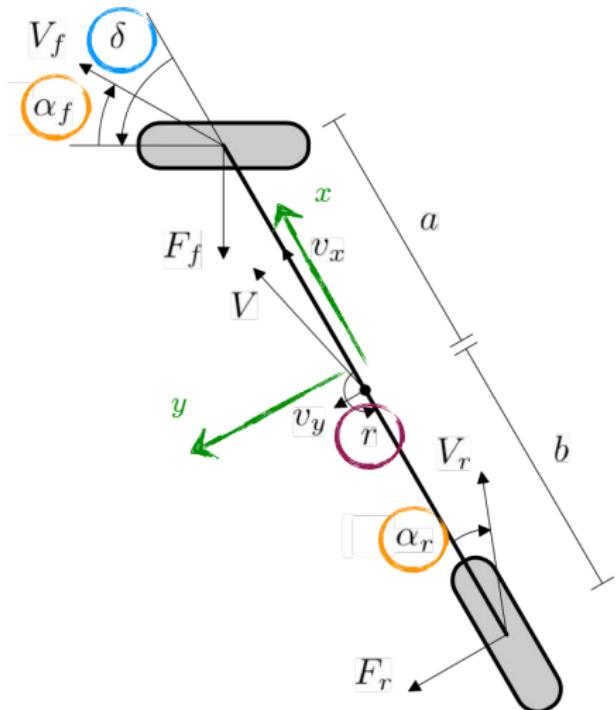
# VEHICLE YAW STABILITY CONTROL

# YAW STABILITY CONTROL PROBLEM

- **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:** Make the **yaw rate** of the vehicle track a time-varying reference computed from the driver's steering angle and current velocity
- **Approach:** Consider the steering command as a reference generator and actuate **steering** and **differential braking** (=coordinated AFS and ESC action)  
(Bernardini, Di Cairano, Bemporad, Tseng, 2009) (Di Cairano, Tseng, Bernardini, Bemporad, 2012)

# VEHICLE MODEL

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



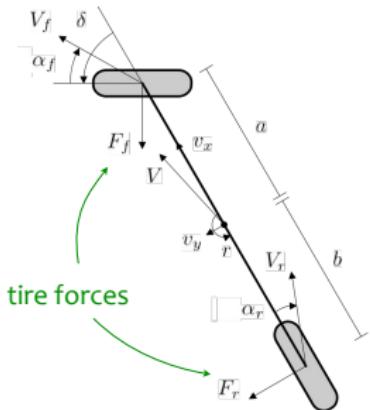
- **reference frame** ( $x,y,z$ ) moving with the vehicle
- **front steering angle**  $\delta$  [rad]
- **tire slip angles**  $\alpha_f, \alpha_r$  [rad]
- **yaw rate**  $r$  [rad/s]

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

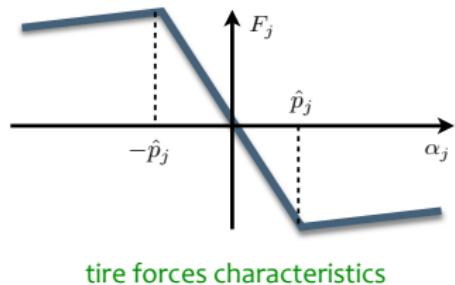
# TIRE FORCE MODEL

- **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**

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

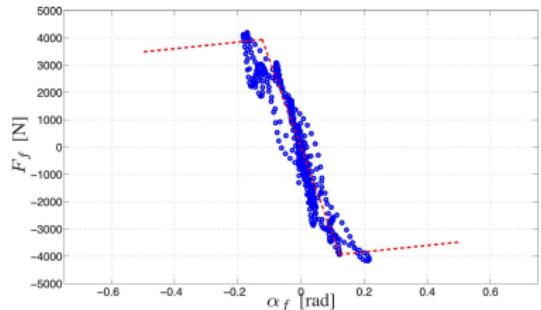


- **Critical slip angles**  $\hat{p}_f, \hat{p}_r$  are threshold values where dynamics switch
- For symmetry, we can restrict to analyze **clockwise turns** (counter-clockwise turns can be handled by opportunely inverting signs)

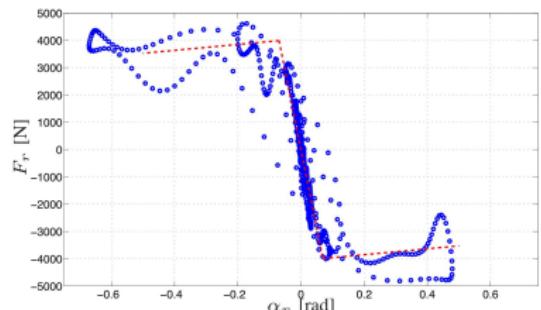


# TIRE FORCE MODEL

## 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

# VEHICLE DYNAMICAL MODEL

slip angles

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

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

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

yaw rate derivative

# VEHICLE DYNAMICAL MODEL - PWA FORM

- 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 ( $\hat{Y}(k) = 0$ )



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

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

driver's steering angle

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

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

- 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**

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

- 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

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

slip angles tracking & int. action

yaw rate tracking

penalty on actuators action

hybrid dynamics

state and input constraints



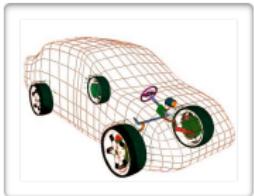
MIQP

- State and input constraints:

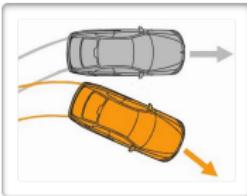
$$\begin{array}{rclcrcl} [z]_1(k) & \geq & -\hat{p}_f & \text{front slip angle} & |[u]_1(k)| & \leq & 1000 \text{ [Nm]} & \text{yaw moment} \\ [z]_2(k) & \geq & -\hat{p}_r & \text{rear slip angle} & |[u]_2(k)| & \leq & 0.35 \text{ [rad]} & \text{front steering} \end{array}$$

# SIMULATION SETUP

- 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:

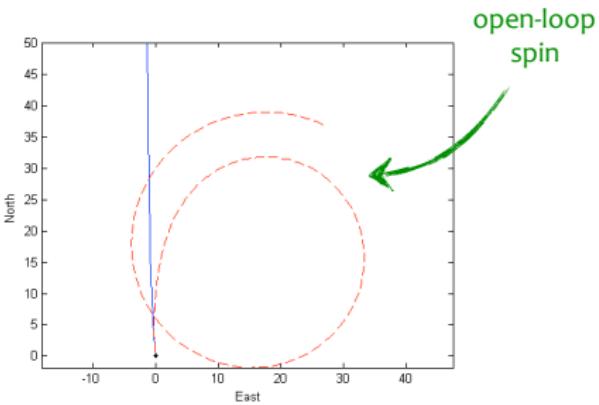
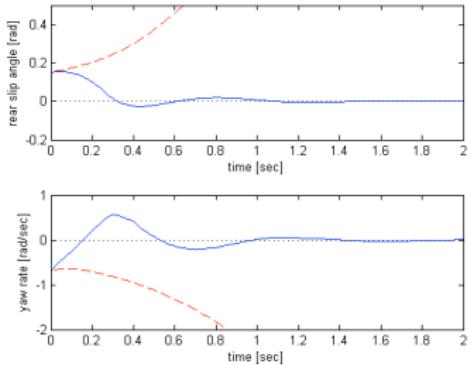
- ▶ Prediction horizon  $N = 3$

- ▶ Weight matrices  $Q_z = \begin{bmatrix} .1 & 0 & 0 & 0 \\ 0 & .1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ ,  $Q_u = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $Q_y = 1$

# CLOSED-LOOP SIMULATIONS RESULTS

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

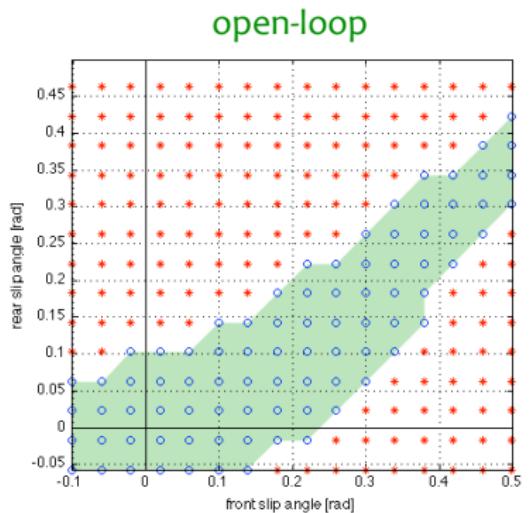
open-loop vs closed-loop



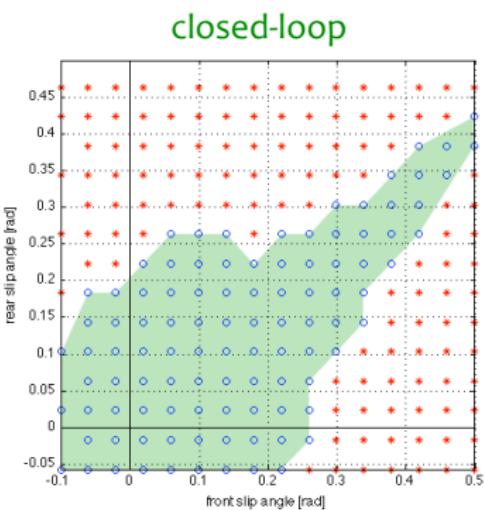
- Controller has to cope with **linearization errors**

# CLOSED-LOOP SIMULATIONS RESULTS

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



control action

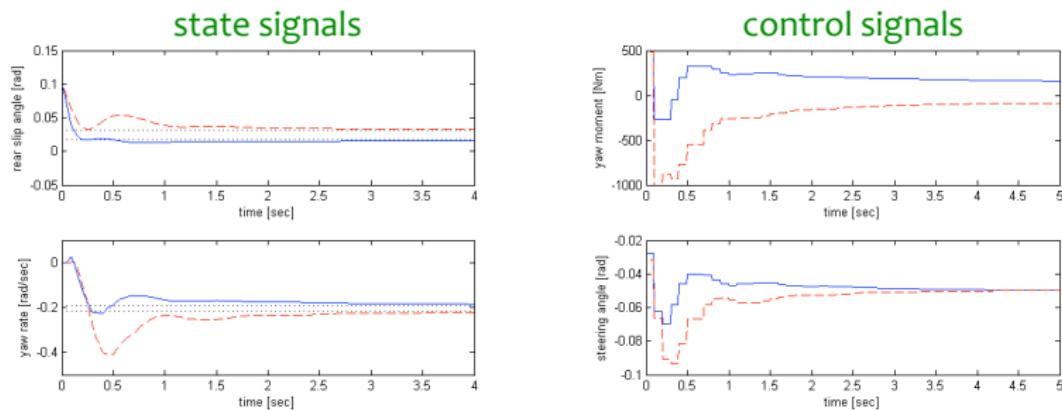


stable initial conditions  
unstable initial conditions

closed-loop provides a  
larger stability region

# CLOSED-LOOP SIMULATIONS RESULTS

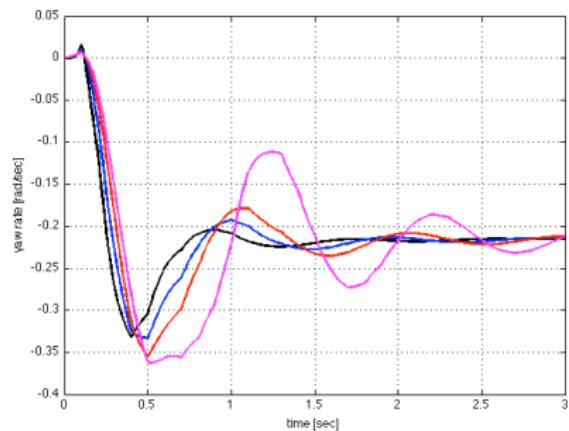
- Robustness analysis w.r.t. model mismatches (with  $\hat{\delta} = -0.05$ ):
  - ▶ nominal longitudinal velocity  $\hat{v}_x = 20 \text{ m/s}$
  - ▶ real longitudinal velocity  $v_x = 15 \text{ m/s}$  and  $v_x = 25 \text{ m/s}$



- Stability and fast tracking response are provided

# CLOSED-LOOP 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

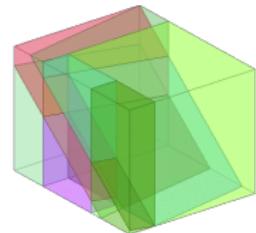
# COMPUTATIONAL ISSUES

- **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

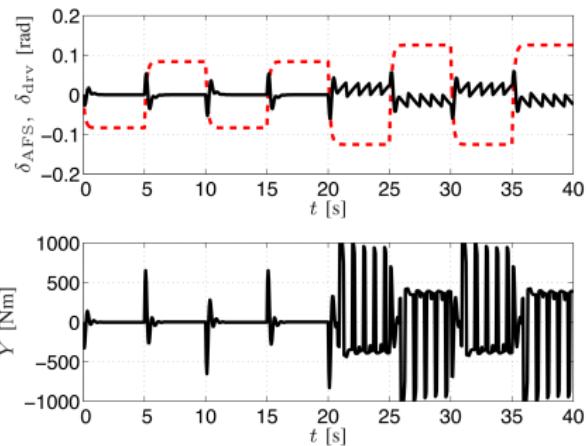
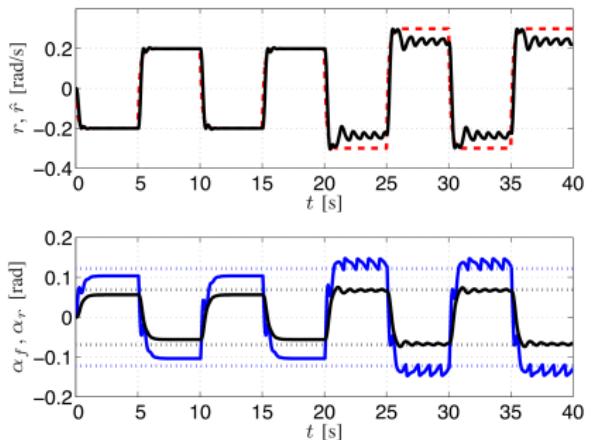
- **Explicit solution** of hybrid MPC control problem:

- 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



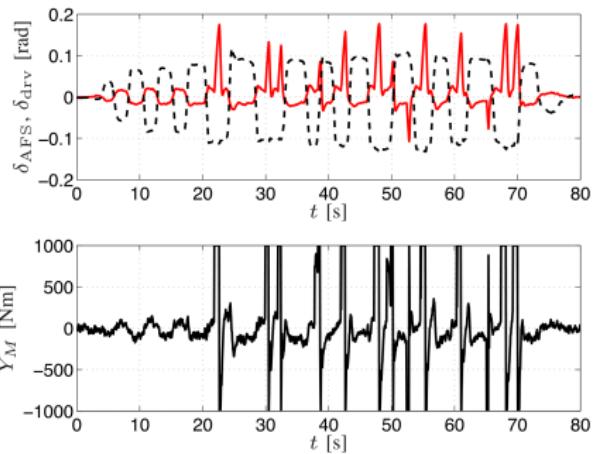
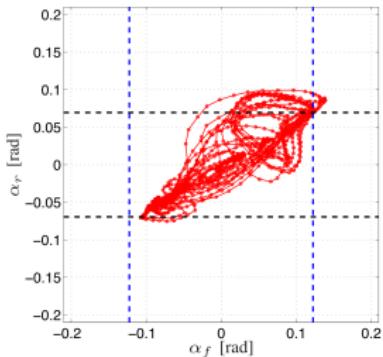
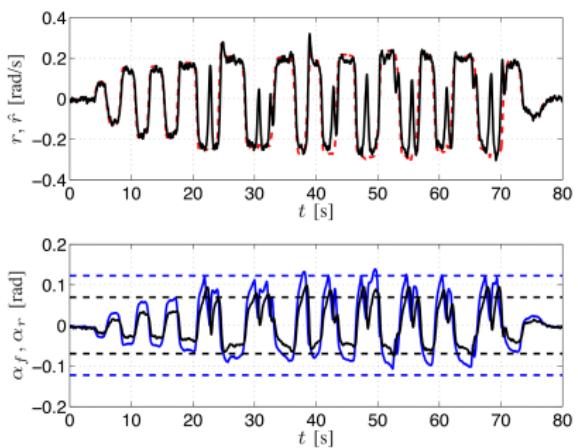
# SWITCHED MPC SOLUTION

- 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:



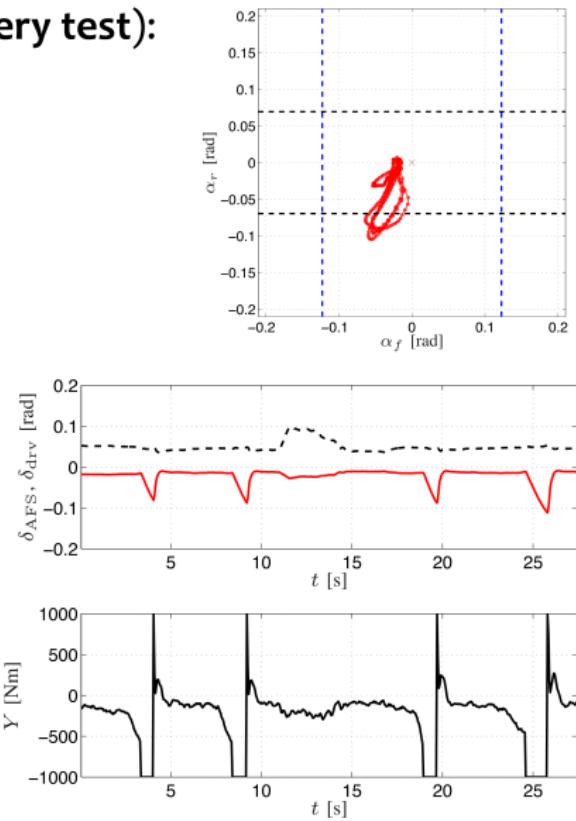
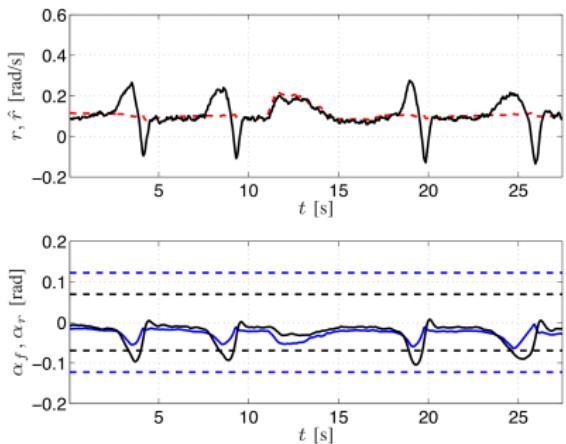
# EXPERIMENTAL RESULTS

- Experimental results (slalom test):



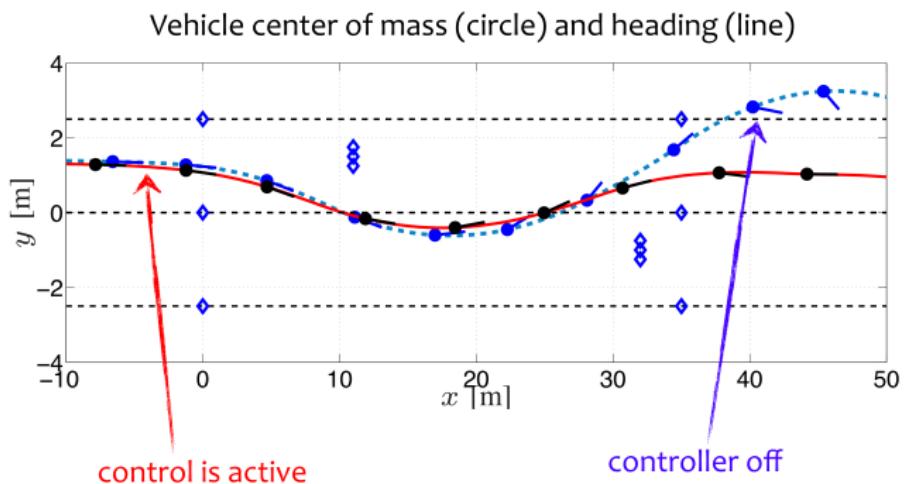
# EXPERIMENTAL RESULTS

- Experimental results (stability recovery test):



# EXPERIMENTAL RESULTS

- Experimental results (**double-lane change**):



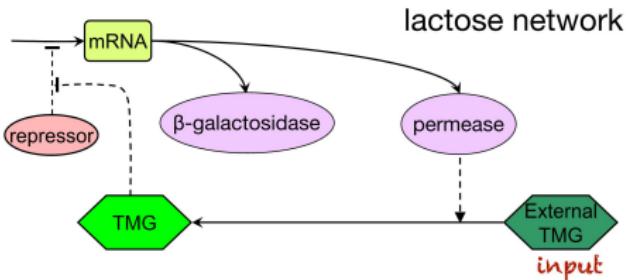
- Alternative: use switched linear MPC and mpQP (Di Cairano, Tseng, 2010)

# HYBRID MPC OF INDUCTION OF ESCHERICHIA COLI

# HYBRID MPC OF INDUCTION OF ESCHERICHIA COLI

(Julius, Sakar, Bemporad, Pappas, 2007)

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

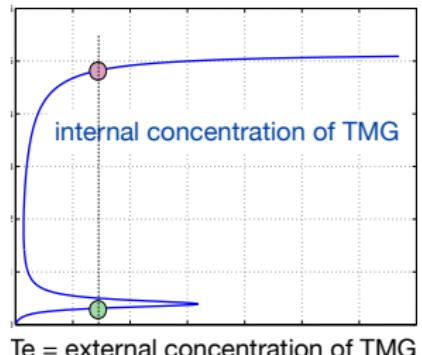
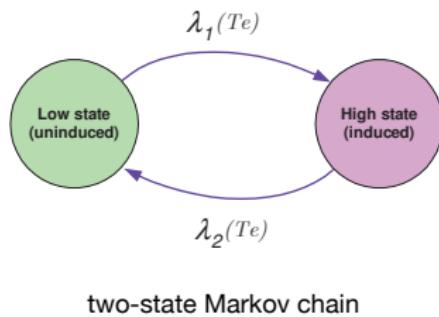


TMG = *thio-methyl galactosidase* concentration

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

# HYBRID MPC OF INDUCTION OF ESCHERICHIA COLI

- Bistable lactose regulation system of E. coli



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

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

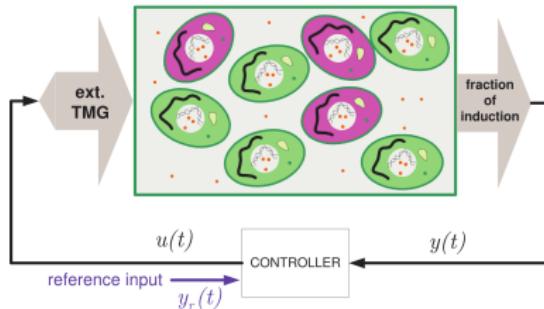
- Transition rates  $\lambda_1$ ,  $\lambda_2$  modeled as **piecewise constant** functions of  $T_e$

$T_e[10^{-3}\text{mM}]$	$\lambda_1(T_e)[\text{min}^{-1}]$	$\lambda_2(T_e)[\text{min}^{-1}]$
[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**

