# Multi-Parametric Toolbox (MPT)

Michal Kvasnica, Pascal Grieder, Mato Baotic, Miroslav Baric, Frank J. Christophersen, Manfred Morari



Automatic Control Laboratory, ETH Zürich





## Agenda

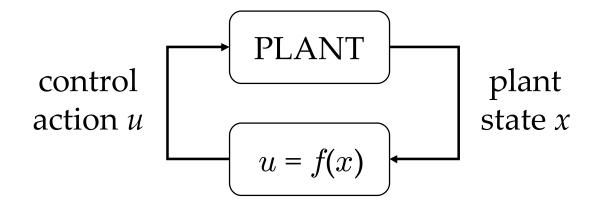
MPT from users' perspective

Overview of basic concepts and functionality

New features in MPT 2.6



## **Typical Problems in Control Theory**



Modeling

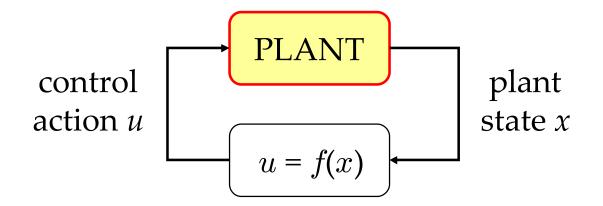
**Control** 

**Analysis** 

**Deployment** 



## **Typical Problems in Control Theory**



Modeling

**Control** 

Analysis

Deployment



#### mpt\_sys and the System Structure

```
sysStruct = mpt_sys(source)
```

#### Possible sources:

- Control toolbox objects
- System identification toolbox objects
- MPC toolbox objects
- HYSDEL source file



#### **Example – Conversion from HYSDEL**

```
>> sysStruct = mpt sys('two tanks.hys')
Conversion from HYSDEL to PWA form finished (0.45 sec)
sysStruct =
        A: \{[2x2 \text{ double}] [2x2 \text{ double}] [2x2 \text{ double}] \}
         B: {[2x2 double] [2x2 double] [2x2 double] [2x2 double]}
         C: {[0 1] [0 1] [0 1] [0 1]}
         D: {[0 0] [0 0] [0 0]}
      umax: [2x1 double]
      umin: [2x1 double]
      xmin: [2x1 double]
      xmax: [2x1 double]
```

- Automatically creates a PWA model (can be disabled)
- Extracts constraints from the model



## Example – Conversion from the Control Toolbox

```
>> M = ss(A, B, C, D)
>> sysStruct = mpt_sys(M, Ts)

sysStruct =

A: [2x2 double]

B: [2x1 double]

C: [2x2 double]

D: [2x1 double]

Ts: 1
```

- Performs discretization if needed
- Constraints have to be provided by the user



#### **Possible Constraints**

Constraints on manipulated variables:

```
sysStruct.umin = umin
sysStruct.umax = umax
```

Constraints on slew rate of manipulated variables:

```
sysStruct.dumin = dumin
sysStruct.dumax = dumax
```

Constraints on system states:

```
sysStruct.xmin = xmin
sysStruct.xmax = xmax
```

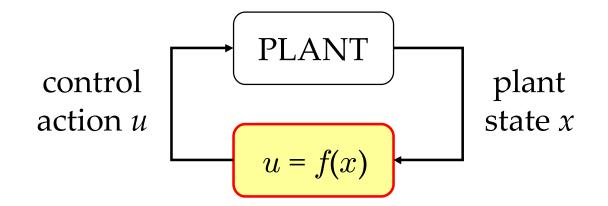
Constraints on system inputs:

```
sysStruct.ymin = ymin
sysStruct.ymax = ymax
```

help mpt\_sysStruct



## **Typical Problems in Control Theory**



Modeling

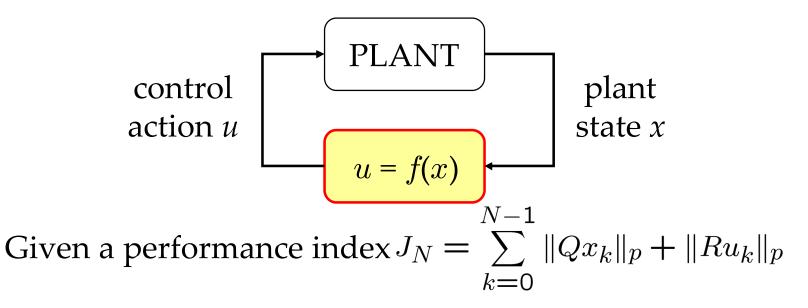
**Control** 

Analysis

Deployment



### **Synthesis: Optimal Control**



Obtain optimal feedback law  $u^* = f(x)$ 

$$J_N^* = \min_{u_0,...,u_{N-1}} J_N,$$
 subj. to System model Constraints



#### Problem Structure probStruct

$$J_N = \sum_{k=0}^{N-1} \|Q(x_k - x_{ref})\|_p + \|R(u_k - u_{ref})\|_p$$

Prediction horizon

Type of objective function

Reference signals

Penalties

help mpt\_probStruct



### mpt\_control

```
ctrl = mpt_control(sysStruct, probStruct, flag)
```

Explicit controllers:

```
ctrl=mpt_control(sysStruct,probStruct)
```

On-line controllers:

```
ctrl=mpt_control(sysStruct,probStruct, 'online')
```



#### **Controllers are Functions**

To obtain the optimizer, simply evaluate the

controller as a function:

$$u = ctrl(x0)$$

#### Example:

$$u = ctrl(-1)$$

u =

0.6180



#### Simulations in Matlab

[X,U,Y] = sim(ctrl, x0, N) simplot(ctrl, x0, N)

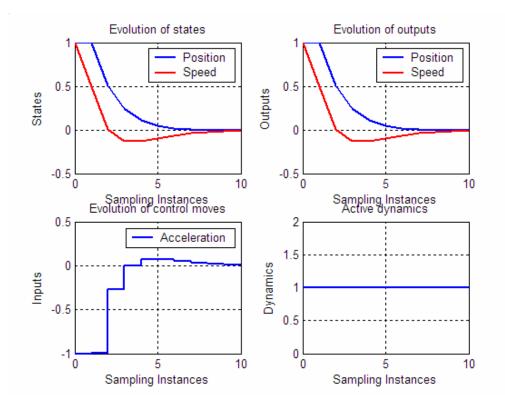
>> X=sim(ctrl, [1;1], 5)

X =

1.00001.00001.00000.50000.51030.00510.2454-0.1299

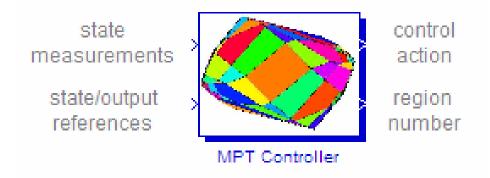
0.1100 -0.1326

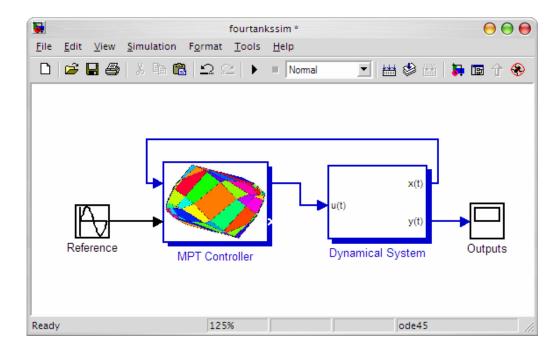
0.0449 -0.0989





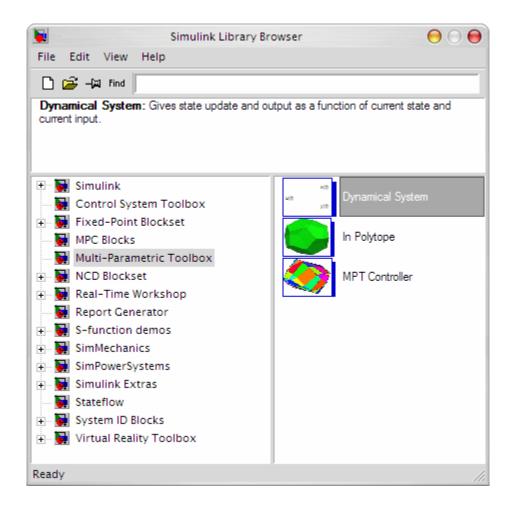
#### Simulations in Simulink







#### The Simulink Library





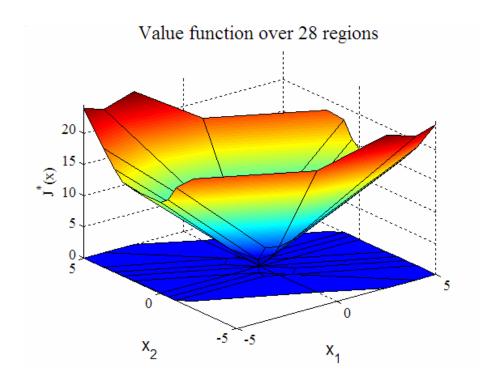
#### Inspection

Direct access to internal fields, e.g.

```
Pn = ctrl.Pn
```

Visual inspection:

```
plot(ctrl)
plotu(ctrl)
plotj(ctrl)
```

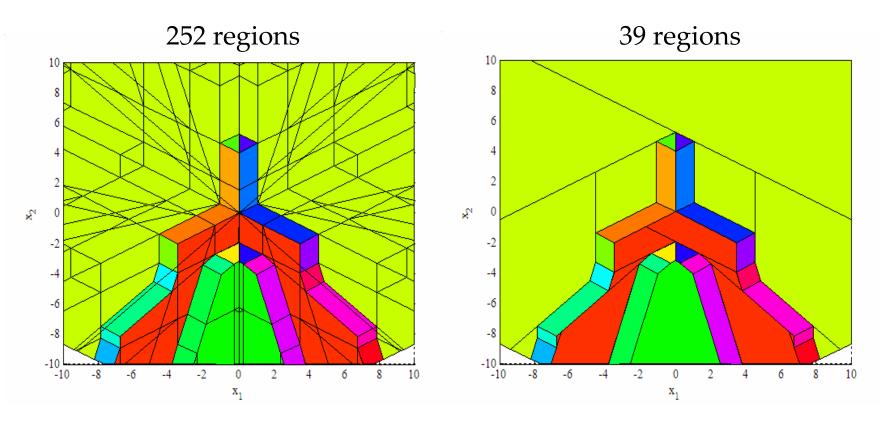




## Post-processing

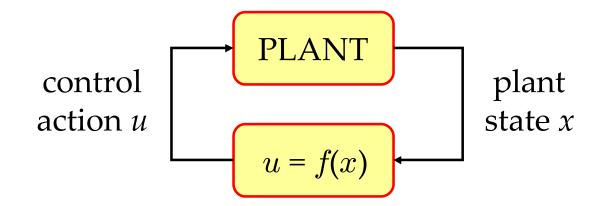
#### Reduce number of regions

ctrl = mpt\_simplify(ctrl)





## **Typical Problems in Control Theory**



Modeling

Control

**Analysis** 

Deployment

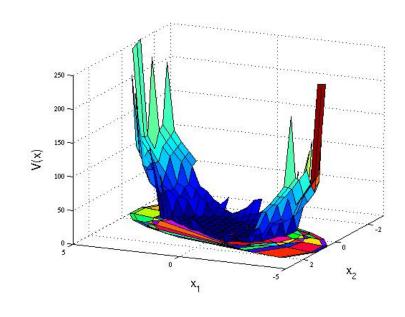


### Lyapunov Analysis

ctrl = mpt\_lyapunov(ctrl, type)

#### Type of Lyapunov functions:

- Quadratic
- Sum of Squares
- Piecewise Affine
- Piecewise Quadratic
- Piecewise Polynomial





#### Reachability Analysis and Verification

Computation of reachable sets

mpt\_reachSets

Safety and liveness analysis

mpt\_verify

Computation of invariant sets

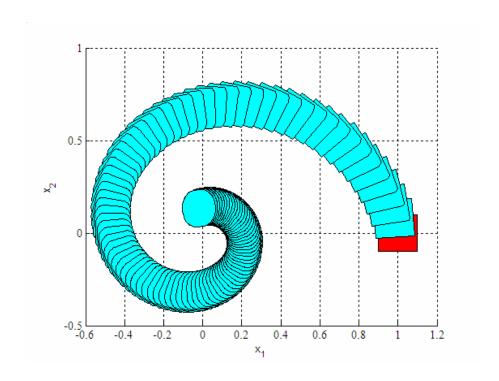
mpt\_invariantSet

Can analyze control laws or dynamical systems



## Computation of Reachable Sets – System Inputs from a Bounded Set

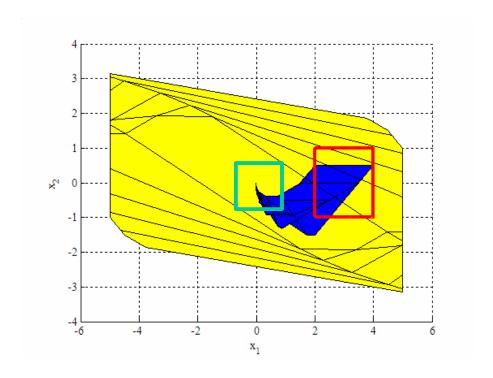
R = mpt\_reachSets(sysStruct, X0, U0, N)





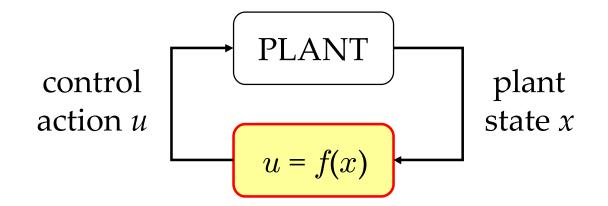
# Verification – System Inputs driven by a Controller

[canreach, Nf] = mpt\_verify(ctrl, X0, Xf, N)





## **Typical Problems in Control Theory**



Modeling

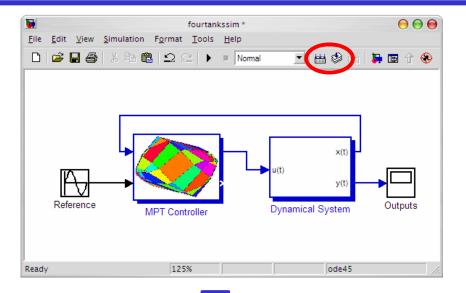
**Control** 

Analysis

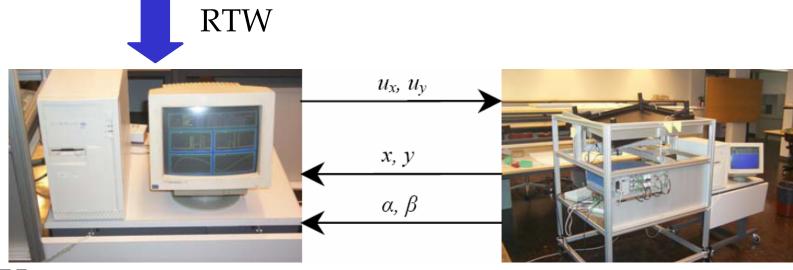
**Deployment** 



#### Real Time Workshop



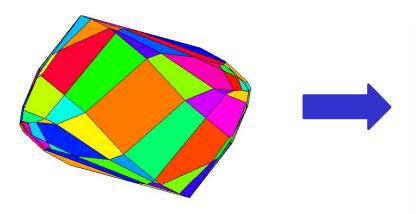
Just press a button...





#### **Export to C code**

#### mpt\_exportc(ctrl, fname)



```
for (iN=0; iN<STOP_TIME; iN++)
{
    printf("time: %d X = [%f %f]\n", iN, X0[0], X0[1]);
    /* tracking controllers require the state vector to be augmented.
    * if the controller is not for tracking, the function will just
    * copy X0 to X
    */
    mpt_augmentState(X, X0, Uprev, reference);
    /* obtain control action for a given state. */
    region = mpt_getInput(X, U);
    if (region < 1)
    {
        printf("No feasible control law found!\n");
        return 0;
    }
}</pre>
```



$$MPT + YALMIP = MPT 2.6$$



#### **Summary of New Features**

- Extended move blocking capabilities
- Control of time-varying systems
- Soft constraints
- MPC for nonlinear systems
- "Design your own MPC" function



#### **Move Blocking**

- Why: to decrease number of free control moves
- How: define a *control horizon* beyond which all control moves are assumed to be fixed
- Example: *N*=5, *M*=2

$$U = \begin{bmatrix} u_0 & u_1 & u_2 & u_3 & u_4 \end{bmatrix}$$
Free Blocked moves moves
$$u_1 = (u_2 = u_3 = u_4)$$



### Move Blocking in MPT

probStruct.Nc = Nc

```
>> Double_Integrator; probStruct.N=5; probStruct.Nc=2;
>> ctrl = mpt_control(sysStruct, probStruct, 'online');
>> u = ctrl([1; 1], struct('openloop', 1))
```

- -1.0000
- -0.4412
- -0.4412
- -0.4412
- -0.4412

Works for LTI, PWA, MLD

and nonlinear systems



## Control of Time-Varying Systems

- Why: to allow more precise predictions
- How: use one model for each prediction
- Example:

$$x(1) = A_1x(0) + Bu(0)$$
  
 $x(2) = A_2x(1) + Bu(1)$   
 $x(3) = A_3x(2) + Bu(2)$ 

Where  $A_1$ ,  $A_2$ ,  $A_3$  are time-varying matrices



#### **Control of Time-Varying Systems in MPT**

Define one sysStruct for each step of the prediction:

```
S1 = sysStruct; S1.A = A1;
S2 = sysStruct; S2.A = A2;
S3 = sysStruct; S3.A = A3;
model = \{ S1, S2, S3 \};
probStruct.N = 3;
ctrl = mpt_control(model, probStruct)
```



#### Control of Time-Varying Systems in MPT

#### Anything can be time-varying, also constraints:

```
S1 = sysStruct; S1.ymax = ymax1; S1.ymin = S1.ymin1;
S2 = sysStruct; S2.ymax = ymax2; S2.ymin = S1.ymin2;
S3 = sysStruct; S3.umax = umax3; S3.umin = S3.umin3;
model = { S1, S2, S3 };

probStruct.N = 3;
ctrl = mpt_control(model, probStruct)
```



#### Control of Time-Varying Systems in MPT

One can also freely combine LTI/PWA/MLD models:

However, dimensions must stay identical!



#### **Soft Constraints**

- Why: hard constraints can lead to infeasibility
- How: introduce a *slack* variable which quantifies by how much was a given constraint violated. Penalize the slack to heavily to stay within bounds if possible.
- Example:

$$x \le x_{max} + s_x$$
$$s_x > 0$$



#### **Soft Constraints in MPT**

- State, input and output constraints can be softened
- To enable soft constraints, specify the corresponding penalty matrix:

```
- probStruct.Sx
```

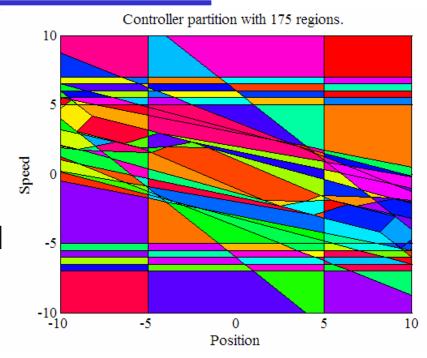
- probStruct.Sy
- probStruct.Su

probStruct.{Sx | Sy | Su} can be combined



### **Soft Constraints in MPT**

sysStruct.ymax = [5; 5]
sysStruct.ymin = [-5; -5]
probStruct.Sy = 1000



sysStruct.Pbnd = unitbox(2, 10)

ctrl = mpt\_control(sysStruct, probStruct)
plot(ctrl)

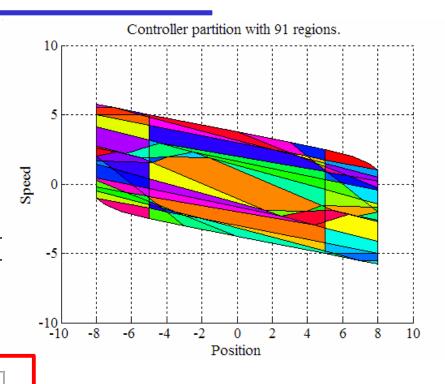


### **Soft Constraints in MPT**

sysStruct.ymax = [5; 5]sysStruct.ymin = [-5; -5]

probStruct.Sy = 1000

probStruct.symax = [3; 3]



ctrl = mpt\_control(sysStruct, probStruct)
plot(ctrl)



### **Soft Constraints in MPT**

probStruct. {sxmax | symax | sumax} can also be used to denote only a subset of constraints to be soft, e.g.

```
probStruct.sxmax = [6.5; 0]
```

Here, constraint on the 2<sup>nd</sup> state is still hard.



## MPC for Nonlinear Systems

- Why: to make a basis for comparison of PWA approximations
- How: formulate (easy) and solve (extremely difficult) a nonlinear problem
- Only polynomial type of nonlinearities allowed
- Support for piecewise nonlinear systems (even harder to solve)
- Available solvers:
  - fmincon (Matlab)
  - PENBMI (commercial)
  - YALMIP's bmibnb (bilinear branch & bound)



## MPC for Nonlinear Systems in MPT

```
sysStruct = mpt_sys(@duffing_oscillator)
```

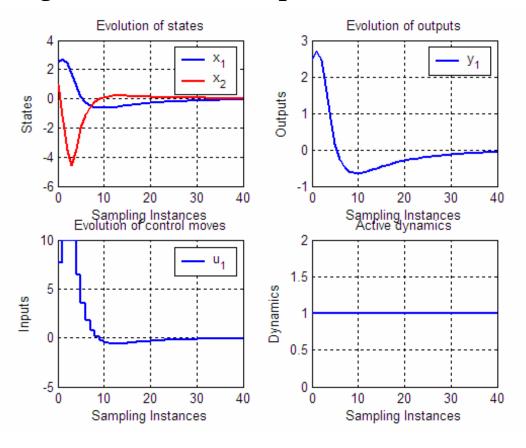
```
probStruct.norm = 2;
probStruct.Q = [1 0; 0 1];
probStruct.R = 0.1;
probStruct.N = 3;

ctrl = mpt_control(sysStruct, probStruct, 'online')
```



## MPC for Nonlinear Systems in MPT

Duffing oscillator example (Fotiou et al., 2006)



Computation time below 2 secs per sample



# Design Your Own MPC Problem

- Why: to allow (almost) arbitrary MPC problem formulations
- How: generate a skeleton of an MPC problem and allow users to add/remove constraints and/or create a new objective function
- Goal: make the whole procedure entirely general, easy to use and fit the results into our framework

$$J_N^* = \min_{u_0, \dots, u_{N-1}} J_N,$$
 probStruct + user subj. to System dynamics Constraints - sysStruct + user



## 3 Phases of mpt\_ownmpc

### 1. Design phase

[C, O, V] = mpt\_ownmpc(sysStruct, probStruct, flag)

### 2. Modification phase

Modify the constraints "C" and/or the objective "O"

### 3. Computation phase

ctrl = mpt\_ownmpc(sysStruct, probStruct, C, O, V)



## Design Phase

- All previously mentioned features can be combined together, e.g. move blocking, timevarying systems, soft constraints, nonlinear models
- Generates a skeleton of the MPC problem based on sysStruct/probStruct
- Returned variables are YALMIP objects



## **Design Phase**

```
[C,O,V] = mpt_ownmpc(sysStruct,probStruct)
>> V
      x: {[1x1 sdpvar] [1x1 sdpvar] [1x1 sdpvar]}
      u: {[1x1 sdpvar] [1x1 sdpvar]}
      y: {[1x1 sdpvar] [1x1 sdpvar]}
>> C
| ID|
      Constraint
                                    Type
                                                         Tag
| #1| Numeric value|
                        Element-wise 2x1 | umin < u 1 < umax |
 #2| Numeric value
                        Element-wise 2x1 | ymin < y 1 < ymax |
| #3| Numeric value| Equality constraint 1x1| x_2 == A*x_1 + B*u_1|
     Numeric value | Equality constraint 1x1 | y_1 == C*x_1 + D*u_1 |
 #5 | Numeric value
                         Element-wise 2x1
                                                  x 2 in Tset
 #6 | Numeric value
                         Element-wise 2x1
                                                  x 0 in Pbnd
 #7 | Numeric value
                         Element-wise 2x1 | umin < u_0 < umax |
| #8| Numeric value|
                         Element-wise 2x1 | ymin < y_0 < ymax |
| #9| Numeric value| Equality constraint 1x1| x_1 == A*x_0 + B*u_0|
|#10| Numeric value | Equality constraint 1x1 | y 0 == C*x 0 + D*u 0 |
```



# Polytopic Constraints on all Predicted States

- Task: add polytopic constraints  $Hx_k \leq K$
- Implementation:

```
[C, O, V] = mpt_ownmpc(sysStruct, probStruct);
x = V.x;
```

```
for k = 1:length(x)

C = C + set(H * x{k} <= K);

end
```

```
ctrl = mpt_ownmpc(sysStruct, probStruct, C, O, V);
```



# Polytopic Constraints on a Subset of Predicted States

- Task: add polytopic constraints  $Hx_k \le K$  on  $x_0$ ,  $x_2$  and  $x_3$
- Implementation:

```
for k = [1 3 4],
    % x{1} corresponds to x(0)
    % x{2} corresponds to x(1)
    % x{3} corresponds to x(2)
    % x{4} corresponds to x(3)
C = C + set(H * x{k} <= K);
end
```



## **Complex Move Blocking**

Task: add complex move-blocking type of constraints:

```
1. u_0 = u_1

2. (u_1 - u_2) = (u_2 - u_3)

3. u_3 = K x_2
```

Implementation:

```
% u_0 == u_1
C = C + set( V.u{1} == V.u{2})

% (u_1-u_2) == (u_2-u_3)
C = C + set( (V.u{2} - V.u{3} ) == ( V.u{3} - V.u{4} ) )

% u_3 == K*x_2
C = C + set( V.u{4} == K * V.x{3})
```



# Using sum() and abs() Operators

- Task: bound the overall energy consumption during the whole prediction horizon
- Implementation:

```
C = C + set(sum(abs([V.u{:}])) <= bound)
```



#### **Contraction Constraints**

- Task: force state  $x_{k+1}$  to be closer (in a 1-norm sense) to the origin that  $x_k$  has been
- Implementation:

```
for k = 1:length(V.x)-1

C = C + set(norm(V.x\{k+1\}, 1) \le norm(V.x\{k\}, 1));

end
```



# **Logic Constraints**

- Task:  $x_0 \le 0 \Rightarrow u_0 \ge 1$
- Implementation:

$$C = C + set(implies(V.x{1} <= 0, V.u{1} >= 1))$$

• However: doesn't say anything about  $u_0$  if  $x_0$  is > 0



# **Logic Constraints**

- Task:  $x_0 \le 0 \iff u_0 \ge 1$
- Implementation:

$$C = C + set(iff(V.x{1} <= 0, V.u{1} >= 1))$$



### **Nonlinear Constraints**

- Task: add a ball constraint  $x_N^2 \le 1$
- Implementation:

$$C = C + set(V.x\{end\}'*V.x\{end\} <= 1)$$

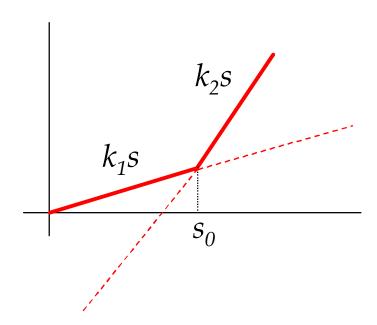
 However: the problem will become nonlinear with all the negative implications



## **Custom Objectives**

Task: use piecewise affine penalties on slacks

```
probStruct.sxmax = Inf
probStruct.Sx = 0
[C, O, V] = mpt_ownmpc(sysStruct, probStruct)
V =
      x: {[1x1 sdpvar] [1x1 sdpvar] [1x1 sdpvar]}
      u: {[1x1 sdpvar] [1x1 sdpvar]}
      y: {[1x1 sdpvar] [1x1 sdpvar]}
     sx: {[1x1 sdpvar] [1x1 sdpvar] [1x1 sdpvar]}
   type: 'explicit'
for k = 1: length(V.sx),
  s = V.sx\{k\};
  0 = 0 + \max([k1*s, k2*s - s0*(k2 - k1)])
```



We can use  $max(k_1s, k_2s)$ 

end

## **Custom Objectives**

- Task: use time varying reference signals
- By default MPT assume the reference is fixed for all predictions steps
- But the user can change it!

```
O = norm(V.x{1} - rt1, 1) + ...
norm(V.x{2} - rt2, 1) + ...
norm(V.x{3} - rt3, 1) + ...
V.u{1}'*V.u{1} + V.u{2}'*V.u{2}
```



## "Unpack and Use" Toolbox

- Works on Windows, Linux, Solaris, Macs
- HYSDEL
- Hybrid Identification Toolbox
- YALMIP
- Ellipsoidal Toolbox
- ESP
- Interfaces to many solvers

















#### http://control.ee.ethz.ch/~mpt/



