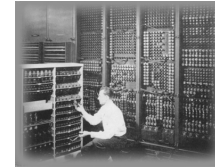


Hybrid Systems: Modeling

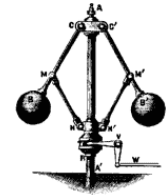
Hybrid systems



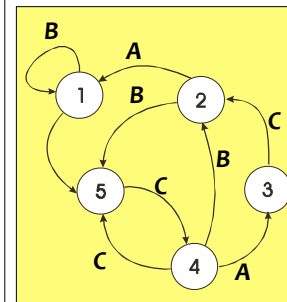
$x \in \{1, 2, 3, 4, 5\}$
 $u \in \{A, B, C\}$

Computer Science
 Finite state machines

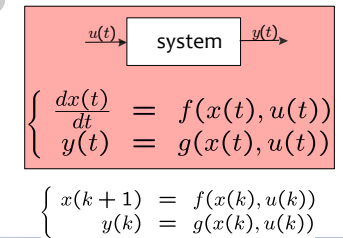
Control Theory
 Continuous dynamical systems



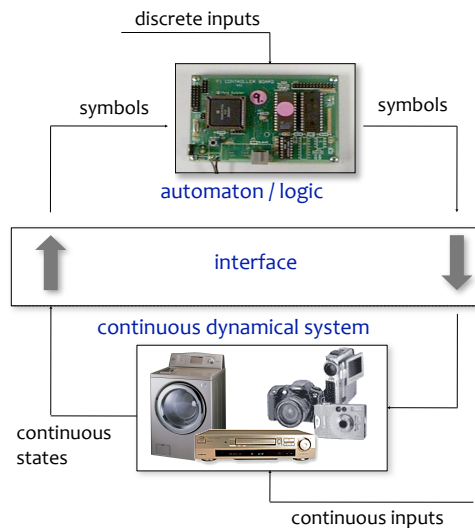
$x \in \mathbb{R}^n$
 $u \in \mathbb{R}^m$
 $y \in \mathbb{R}^p$



Hybrid systems

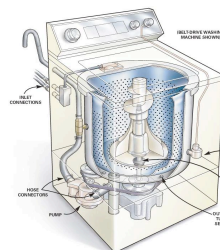


Embedded systems

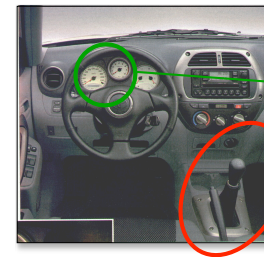


- Automobiles
- Industrial processes
- Consumer electronics
- Home appliances
- ...

Example:

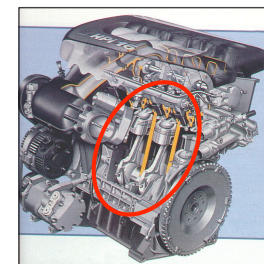


"Intrinsically hybrid" systems



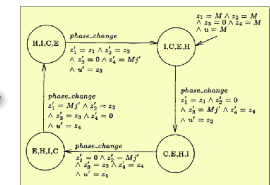
- Transmission

discrete command $(R, N, 1, 2, 3, 4, 5)$ + continuous dynamical variables (velocities, torques)



- Four-stroke engines

Automaton, dependent on power train motion



Example of hybrid control problem

Cruise control problem



GOAL:

command gear ratio, gas pedal, and brakes to track a desired speed and minimize consumptions

CHALLENGES:

- continuous and discrete inputs
- dynamics depends on gear
- nonlinear torque/speed maps

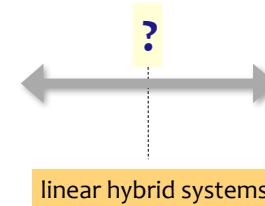


Key requirements for hybrid models

- **Descriptive** enough to capture the behavior of the system
 - continuous dynamics (physical laws)
 - logic components (switches, automata, software code)
 - interconnection between logic and dynamics
- **Simple** enough for solving analysis and synthesis problems

$$\begin{cases} x' = Ax + Bu \\ y = Cx + Du \end{cases}$$

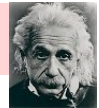
linear systems



$$\begin{cases} x' = f(x, u, t) \\ y = g(x, u, t) \end{cases}$$

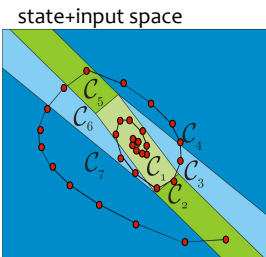
nonlinear systems

“Make everything as simple as possible, but not simpler.”
— Albert Einstein



Piecewise affine systems

$$\begin{aligned} x(k+1) &= A_{i(k)}x(k) + B_{i(k)}u(k) + f_{i(k)} \\ y(k) &= C_{i(k)}x(k) + D_{i(k)}u(k) + g_{i(k)} \\ i(k) &\text{ s.t. } H_{i(k)}x(k) + J_{i(k)}u(k) \leq K_{i(k)} \end{aligned}$$

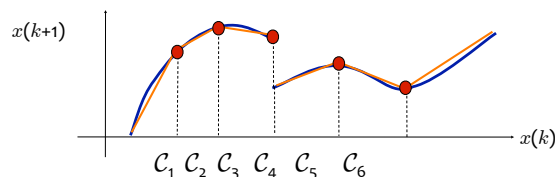


(Sontag 1981)

$$x \in \mathbb{R}^n, u \in \mathbb{R}^m, y \in \mathbb{R}^p$$

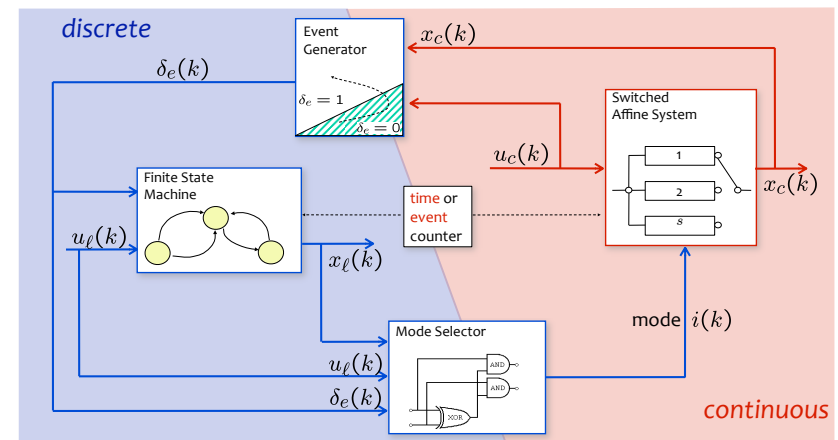
$$i(k) \in \{1, \dots, s\}$$

Can approximate nonlinear and/or discontinuous dynamics arbitrarily well



Discrete Hybrid Automaton (DHA)

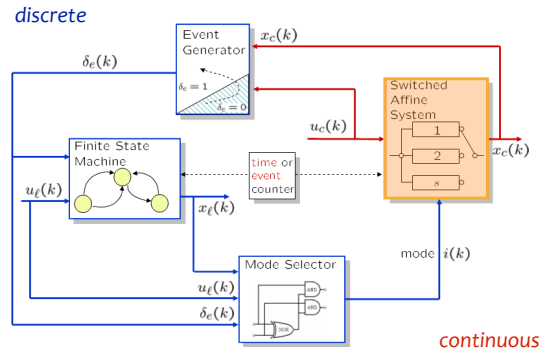
(Torrìsi, Bemporad, 2004)



$x_\ell \in \{0, 1\}^{m_b}$ = binary states
 $u_\ell \in \{0, 1\}^{m_b}$ = binary inputs
 $\delta_e \in \{0, 1\}^{n_e}$ = event variables

$x_c \in \mathbb{R}^{n_c}$ = continuous states
 $u_c \in \mathbb{R}^{m_c}$ = continuous inputs
 $i \in \{1, 2, \dots, s\}$ = current mode

Switched affine system

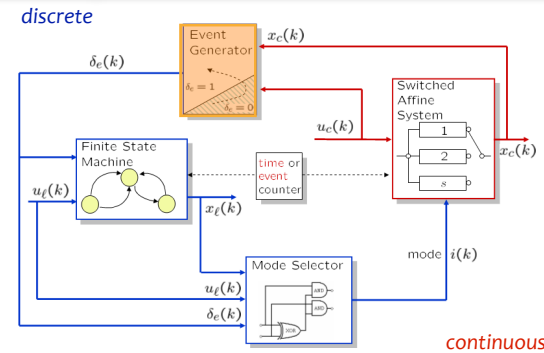


The affine dynamics depend on the current mode $i(k)$:

$$x_c(k+1) = A_{i(k)}x_c(k) + B_{i(k)}u_c(k) + f_{i(k)}$$

$$x_c \in \mathbb{R}^{n_c}, u_c \in \mathbb{R}^{m_c}$$

Event generator



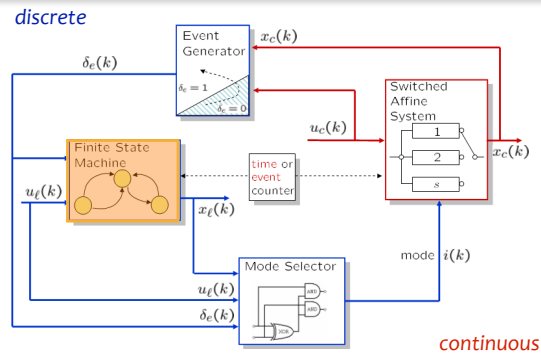
Event variables are generated by linear threshold conditions over continuous states, continuous inputs, and time:

$$[\delta_e^i(k) = 1] \iff [H^i x_c(k) + K^i u_c(k) \leq W^i]$$

$$x_c \in \mathbb{R}^{n_c}, u_c \in \mathbb{R}^{m_c}, \delta_e \in \{0, 1\}^{n_e}$$

$$\text{Example: } [\delta(k) = 1] \iff [x_c(k) \geq 0]$$

Finite state machine

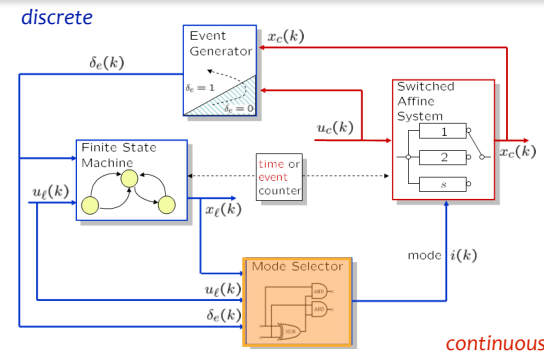


The binary state of the finite state machine evolves according to a Boolean state update function:

$$x_\ell(k+1) = f_B(x_\ell(k), u_\ell(k), \delta_e(k)) \quad x_\ell \in \{0, 1\}^{n_\ell}, u_\ell \in \{0, 1\}^{m_\ell}, \delta_e \in \{0, 1\}^{n_e}$$

$$\text{Example: } x_\ell(k+1) = \neg \delta_e(k) \vee (x_\ell(k) \wedge u_\ell(k))$$

Mode selector



The mode selector can be seen as the output function of the discrete dynamics

The active mode $i(k)$ is selected by a Boolean function of the current binary states, binary inputs, and event variables:

$$i(k) = f_M(x_\ell(k), u_\ell(k), \delta_e(k)) \quad x_\ell \in \{0, 1\}^{n_\ell}, u_\ell \in \{0, 1\}^{m_\ell}, \delta_e \in \{0, 1\}^{n_e}$$

Example:

$$i(k) = \begin{bmatrix} \neg u_\ell(k) \vee x_\ell(k) \\ u_\ell(k) \wedge x_\ell(k) \end{bmatrix} \implies \begin{array}{c|cc} u_\ell/x_\ell & 0 & 1 \\ \hline 0 & i = \begin{bmatrix} 1 \\ 0 \end{bmatrix} & i = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ 1 & i = \begin{bmatrix} 0 \\ 0 \end{bmatrix} & i = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \end{array}$$

the system has 3 modes

Logic → inequalities

$$X_1 \vee X_2 = \text{TRUE} \quad \text{---} \quad \delta_1 + \delta_2 \geq 1, \quad \delta_1, \delta_2 \in \{0, 1\}$$

(Glover 1975, Williams 1977, Hooker 2000)

0. Given a Boolean statement

$$F(X_1, X_2, \dots, X_n) = \text{TRUE}$$

1. Convert to Conjunctive Normal Form (CNF):

$$\bigwedge_{j=1}^m \left(\bigvee_{i \in P_j} X_i \vee \bigvee_{i \in N_j} \bar{X}_i \right) = \text{TRUE}$$

2. Transform into inequalities:

$$\begin{array}{l} \sum_{i \in P_1} \delta_i + \sum_{i \in N_1} (1 - \delta_i) \geq 1 \\ \vdots \\ \sum_{i \in P_m} \delta_i + \sum_{i \in N_m} (1 - \delta_i) \geq 1 \end{array}$$

$$P_j \cup N_j \subseteq \{1, n\}$$

polyhedron

$$A\delta \leq b, \quad \delta \in \{0, 1\}^n$$

Any logic proposition can be translated into integer linear inequalities

Logic → inequalities: symbolic approach

Example: $F(X_1, X_2, X_3) = [X_3 \leftrightarrow X_1 \wedge X_2]$

1. Convert to Conjunctive Normal Form (CNF):

(see e.g. <http://www.coursland.net/aima/propositionApplet.html> or just google for “CNF + applet” ...)

$$(X_3 \vee \neg X_1 \vee \neg X_2) \wedge (X_1 \vee \neg X_3) \wedge (X_2 \vee \neg X_3)$$

2. Transform into inequalities:

$$\begin{cases} \delta_3 + (1 - \delta_1) + (1 - \delta_2) \geq 1 \\ \delta_1 + (1 - \delta_3) \geq 1 \\ \delta_2 + (1 - \delta_3) \geq 1 \end{cases}$$

Logic → inequalities: geometric approach

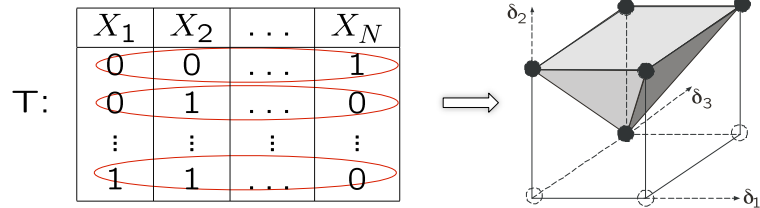
Boolean statement

$$F(X_1, X_2, \dots, X_n) = \text{TRUE}$$

polyhedron

$$A\delta \leq b, \quad \delta \in \{0, 1\}^n$$

The polytope $P = \{\delta : A\delta \leq b\}$ is the convex hull of the rows of the truth table T associated with formula $F(X_1, \dots, X_N)$



Convex hull algorithms: **cdd**, **lrs**, **qhull**, **chD**, **Hull**, **Porto**

CDDMEX package included in the Hybrid Toolbox

Logic → inequalities: geometric approach

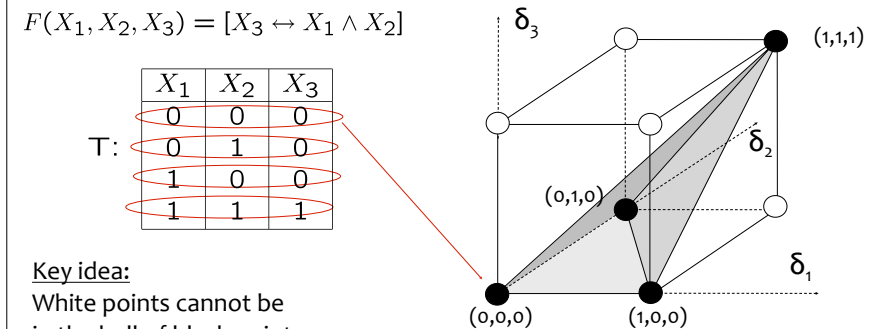
Example: logic “AND”

$$F(X_1, X_2, X_3) = [X_3 \leftrightarrow X_1 \wedge X_2]$$

	X_1	X_2	X_3
T :	0	0	0
	0	1	0
	1	0	0
	1	1	1

Key idea:

White points cannot be in the hull of black points



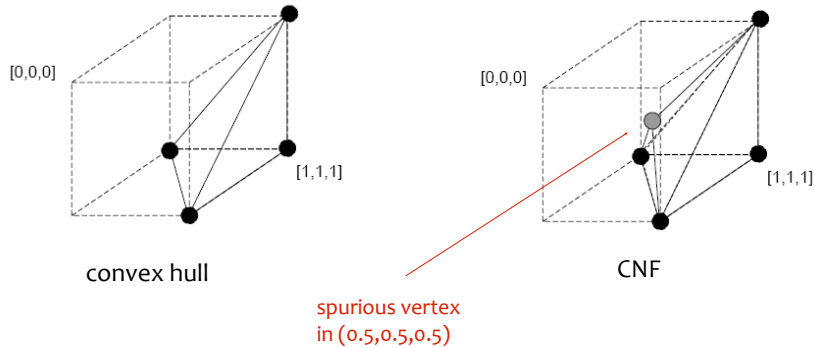
$$\text{conv} \left(\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \right) = \left\{ \delta \in \mathbb{R}^3 : \begin{array}{l} -\delta_1 + \delta_3 \leq 0 \\ -\delta_2 + \delta_3 \leq 0 \\ \delta_1 + \delta_2 - \delta_3 \leq 1 \end{array} \right\}$$

```
>> V=struct('V',[0 0 0;0 1 0;1 0 0;1 1 1]);
>> H=cddmex('hull',V);A=H.A,b=H.B
```

Geometric vs. symbolic approach

- The polyhedron obtained via convex hull is the smallest one
- The one obtained via CNF may be larger. Example:

$$(X_1 \vee X_2) \wedge (X_1 \vee X_3) \wedge (X_2 \vee X_3) = \text{TRUE}$$



Note: no other example with 3 vars but $(X_1 \vee X_2) \wedge (X_1 \vee X_3) \wedge (X_2 \vee X_3) \wedge (\bar{X}_1 \vee \bar{X}_2 \vee \bar{X}_3)$

Big-M technique (iff)

- Consider the *if-and-only-if* condition

$$[\delta = 1] \leftrightarrow [a'x_c - b \leq 0]$$

$$x_c \in \mathcal{X} \subset \mathbb{R}^{n_c}$$

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

- Assume \mathcal{X} bounded and let M and m such that

$$M > a'x_c - b, \forall x_c \in \mathcal{X}$$

$$m < a'x_c - b, \forall x_c \in \mathcal{X}$$

- The *if-and-only-if* condition is equivalent to

$$\begin{cases} a'x_c - b \leq M(1 - \delta) \\ a'x_c - b > m\delta \end{cases}$$

- To avoid strict inequalities, replace by $a'x_c - b \geq \epsilon + (m - \epsilon)\delta$ where $\epsilon > 0$ is a small number (e.g., machine precision)

Big-M technique (if-then-else)

- Consider the *if-then-else* condition

$$z = \begin{cases} a'_1 x_c + f_1 & \text{if } \delta = 1 \\ a'_2 x_c + f_2 & \text{otherwise} \end{cases}$$

$$x_c \in \mathcal{X} \subset \mathbb{R}^{n_c}$$

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

$$z \in \mathbb{R}$$

- Assume \mathcal{X} bounded and let M_1, M_2 and m_1, m_2 such that

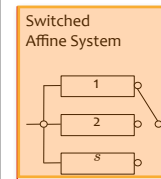
$$M_1 > a'_1 x_c + f_1 > m_1, \forall x_c \in \mathcal{X}$$

$$M_2 > a'_2 x_c + f_2 > m_2, \forall x_c \in \mathcal{X}$$

- The *if-then-else* condition is equivalent to

$$\begin{cases} (m_1 - M_2)(1 - \delta) + z \leq a_1 x_c + f_1 \\ (m_2 - M_1)(1 - \delta) - z \leq -a_1 x_c - f_1 \\ (m_2 - M_1)\delta + z \leq a_2 x_c + f_2 \\ (m_1 - M_2)\delta - z \leq -a_2 x_c - f_2 \end{cases}$$

Switched affine system



The state-update equation can be rewritten as a difference equation + *if-then-else* conditions:

$$z_1(k) = \begin{cases} A_1 x_c(k) + B_1 u_c(k) + f_1, & \text{if } i(k) = 1 \\ 0, & \text{otherwise,} \end{cases}$$

$$\vdots$$

$$z_s(k) = \begin{cases} A_s x_c(k) + B_s u_c(k) + f_s, & \text{if } i(k) = s, \\ 0, & \text{otherwise,} \end{cases}$$

$$x_c(k+1) = \sum_{i=1}^s z_i(k)$$

where $z_i(k) \in \mathbb{R}^{n_c}, i = 1, \dots, s$

Output equations $y_c(k) = C_i x_c(k) + D_i u_c(k) + g_i$ admit a similar transformation

Logic and inequalities

$$X_1 \vee X_2 = \text{TRUE} \longrightarrow \delta_1 + \delta_2 \geq 1, \quad \delta_1, \delta_2 \in \{0, 1\} \quad (\text{Glover 1975, Williams 1977, Hooker 2000})$$

Any logic statement

$$f(X) = \text{TRUE}$$

$$\bigwedge_{j=1}^m (\forall_{i \in P_j} X_i \vee \forall_{i \in N_j} \neg X_i) \quad (\text{CNF})$$

$N_j, P_j \subseteq \{1, \dots, n\}$

$$\begin{cases} 1 \leq \sum_{i \in P_1} \delta_i + \sum_{i \in N_1} (1 - \delta_i) \\ \vdots \\ 1 \leq \sum_{i \in P_m} \delta_i + \sum_{i \in N_m} (1 - \delta_i) \end{cases}$$

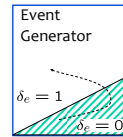
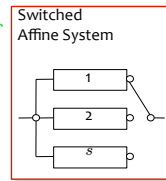
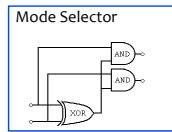
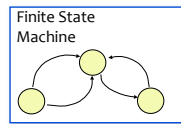
$$[\delta_e^i(k) = 1] \leftrightarrow [H^i x_c(k) \leq W^i]$$

$$\begin{cases} H^i x_c(k) - W^i \leq M^i (1 - \delta_e^i) \\ H^i x_c(k) - W^i > m^i \delta_e^i \end{cases}$$

$$\text{IF } [\delta = 1] \text{ THEN } z = a_1^T x + b_1^T u + f_1$$

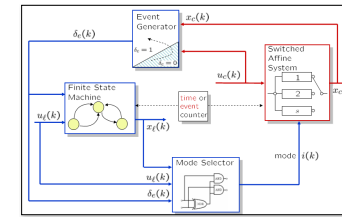
$$\text{ELSE } z = a_2^T x + b_2^T u + f_2$$

$$\begin{cases} (m_1 - M_2)(1 - \delta) + z \leq a_1 x + b_1 u + f_1 \\ (m_2 - M_1)(1 - \delta) - z \leq -a_1 x - b_1 u - f_1 \\ (m_2 - M_1)\delta + z \leq a_2 x + b_2 u + f_2 \\ (m_1 - M_2)\delta - z \leq -a_2 x - b_2 u - f_2 \end{cases}$$



Mixed Logical Dynamical (MLD) systems

Discrete Hybrid Automaton



HYSDEL
(Torrissi, Bemporad, 2004)

Mixed Logical Dynamical (MLD) systems

$$\begin{cases} x_{k+1} = Ax_k + B_1 u_k + B_2 \delta_k + B_3 z_k + B_5 \\ y_k = Cx_k + D_1 u_k + D_2 \delta_k + D_3 z_k + D_5 \\ E_2 \delta_k + E_3 z_k \leq E_4 x_k + E_1 u_k + E_5 \end{cases}$$

Continuous and binary variables $x \in \mathbb{R}^{n_c} \times \{0, 1\}^{n_b}$, $u \in \mathbb{R}^{m_c} \times \{0, 1\}^{m_b}$
 $y \in \mathbb{R}^{p_c} \times \{0, 1\}^{p_b}$, $\delta \in \{0, 1\}^{r_b}$, $z \in \mathbb{R}^{r_c}$

(Bemporad, Morari 1999)

- Computationally oriented (mixed-integer programming)
- Suitable for **controller** synthesis, **verification**, ...

Mixed-integer models in Operations Research

Translation of logical relations into linear inequalities is heavily used in **operations research (OR)** for solving complex decision problems by using **mixed-integer (linear) programming (MIP)**

Example: Timetable generation (for demanding professors ...)

	9-11	11-13	14-16	16-18
Lunedì	Red	Red	Red	Red
Martedì	Red	Green	Green	Green
Mercoledì	Red	Green	Green	Green
Giovedì	Green	Green	Green	Green
Venerdì	Red	Green	Green	Green
Sabato	Red	Red	Grey	Grey

■ decisamente no
■ preferibilmente no
■ neutro
■ preferibilmente si
■ decisamente si

Cost function:
sum of professors' preferences

Constraints:
professors [students] cannot teach [take] two courses at the same time, etc.

Mixed-integer models in Operations Research

Translation of logical relations into linear inequalities is heavily used in **operations research (OR)** for solving complex decision problems by using **mixed-integer (linear) programming (MIP)**

Example: Timetable generation (for demanding professors ...)



CPU time: 0.2 s

	8	9	10	11	12	13	14	15	16	17	18	19
lun			Sistemi Operativi (*18)				Misure per la Automazione (*7)		Ingegneria del Software (*16)			
mar			Basi di Dati (*3)	Sistemi Operativi (*3)			Robotica ed Automazione di Processo (*18)		Basi di Dati (*18)			
mer			Robotica ed Automazione di Processo (*9)		Misure per la Automazione (*7)				Laboratorio di Robotica e Realtà Virtuale (*15)			
gio							Ingegneria del Software (*18)					
ven							Basi di Dati (*3)		Sistemi Operativi (*6)			
							Laboratorio di Robotica e Realtà Virtuale (*15)					
ven							Robotica ed Automazione di Processo (*6)				Misure per la Automazione (*7)	
							Ingegneria del Software (*18)					
sab												

Effort: 5% mathematical problem setup (MILP model)
 35% database & web interfaces
 60% deal with professors' complaints, complaints, complaints ...

Mixed-integer models in Operations Research

Translation of logical relations into linear inequalities is heavily used in **operations research (OR)** for solving complex decision problems by using **mixed-integer (linear) programming (MIP)**

Example: Optimal multi-period investments for maintenance and upgrade of electrical energy distribution networks
(Bemporad, Muñoz, Piazzesi, 2006)



A simple example

• System:

$$x(k+1) = \begin{cases} 0.8x(k) + u(k) & \text{if } x(k) \geq 0 \\ -0.8x(k) + u(k) & \text{if } x(k) < 0 \end{cases}$$

$$-10 \leq x(k) \leq 10$$

• Associate $[\delta(k) = 1] \leftrightarrow [x(k) \geq 0]$ and transform

$$\begin{aligned} &\longrightarrow x(k) \geq m(1 - \delta(k)) & M = -m = 10 \\ &x(k) \leq -\epsilon + (M + \epsilon)\delta(k) & \epsilon > 0 \text{ "small"} \end{aligned}$$

• Then $x(k+1) = 1.6\delta(k)x(k) - 0.8x(k) + u(k)$

$$\begin{aligned} z(k) = \delta(k)x(k) &\longrightarrow z(k) \leq M\delta(k) & \delta(k) \in \{0, 1\} \\ &\longrightarrow z(k) \geq m\delta(k) \\ &z(k) \leq x(k) - m(1 - \delta(k)) \\ &z(k) \geq x(k) - M(1 - \delta(k)) \end{aligned}$$

• Rewrite as a **linear** equation

$$\longrightarrow x(k+1) = 1.6z(k) - 0.8x(k) + u(k)$$

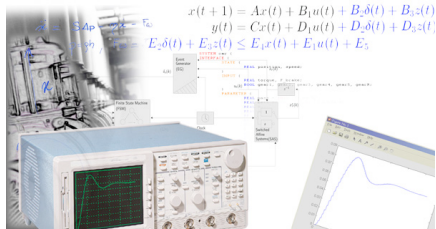
(Note: this is the nonlinear system $x(k+1) = 0.8|x(k)| + u(k)$)

HYSDEL

(HYbrid Systems Description Language)

• Describe **hybrid systems**:

- Automata
- Logic
- Lin. Dynamics
- Interfaces
- Constraints



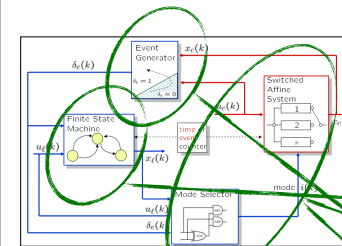
(Torrissi, Bemporad, 2004)

• Automatically generate MLD models in MATLAB

Download: <http://cse.lab.imtlucca.it/~bemporad/hybrid/toolbox>

Reference: <http://control.ethz.ch/~hybrid/hysdel>

DHA and HYSDEL models



```

SYSTEM name {
INTERFACE {
STATE {
REAL xc [xmin,xmax];
BOOL xl; }
INPUT {
REAL uc [umin,umax];
BOOL ul; }
PARAMETER {
REAL param1 = 1; }
} /* end of interface */

IMPLEMENTATION {
AUX { BOOL d;
REAL z; }

AUTOMATA { xl = xl & ~ul; }
DA { z = { IF d THEN 2*xc ELSE -xc; } }
AD { d = xc - 1 <= 0; }
CONTINUOUS {
xc = z; }

MUST {
xc + uc <= 2;
~(xl & ul); }
} /* end implementation */
} /* end system */
    
```

Additional relations
constraining system's
variables

Example 1: Definition of event variables



$$[\delta = 1] \leftrightarrow [h \geq h_{\max}] \quad \begin{array}{l} \delta \in \{0, 1\} \\ h \in \mathbb{R} \end{array}$$



```
AD { delta = hmax - h <= 0; }
```

Example 2: Nonlinear (PWA) functions

Nonlinear amplification unit

$$u_{\text{NL}}(k) = \begin{cases} u(k) & \text{if } u(k) < u_t \\ 2.3u(k) - 1.3u_t & \text{if } u(k) \geq u_t \end{cases}$$

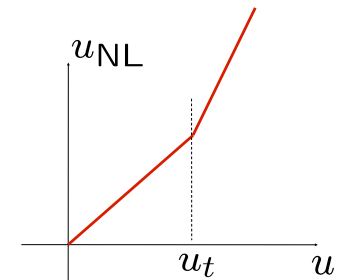


```
DA { un1 = { IF th THEN 2.3*u - 1.3*ut
              ELSE u }; }
```

```
AD { th = ut - u <= 0; }
```



$$[t_h = 1] \leftrightarrow [u \geq u_t]$$



Example 3: Logical relations



Rule: brake if there is an alarm signal, but only if the train is not on fire in a tunnel

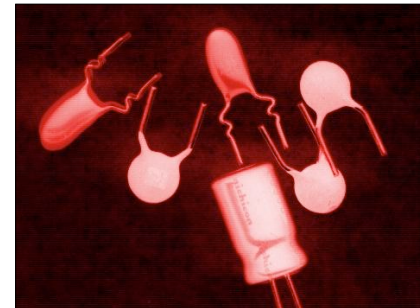
$$\delta_{\text{brake}}, \delta_{\text{alarm}}, \delta_{\text{tunnel}}, \delta_{\text{fire}} \in \{0, 1\}$$

$$\delta_{\text{brake}} = \delta_{\text{alarm}} \wedge \neg(\delta_{\text{tunnel}} \wedge \delta_{\text{fire}})$$

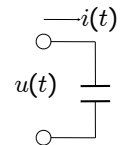


```
LOGIC {
  brake = alarm & ~(tunnel & fire);
}
```

Example 4: Continuous dynamics



$$i(t) = C \frac{du(t)}{dt}$$



Apply forward difference rule:

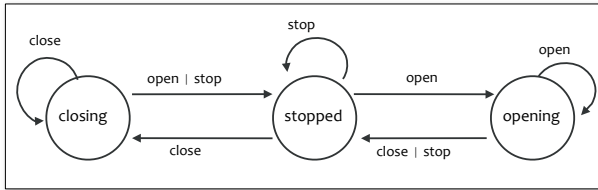
$$u((k+1)T_s) = u(kT_s) + \frac{T}{C} i(kT_s)$$



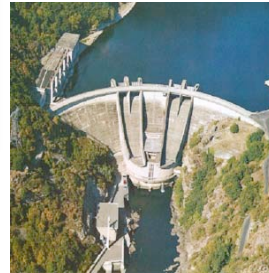
```
CONTINUOUS {
  u = u + Ts*iC*i;
}
```

Note: $iC = 1/C$ is used due to a bug in HYSDEL, that cannot handle division by a scalar parameter

Example 5: Automaton



Flow control through a dam



binary inputs: $u_{open}, u_{close}, u_{stop} \in \{0, 1\}$

binary states: $x_{opening}, x_{closing}, x_{stopped} \in \{0, 1\}$



```
AUTOMATA {
  xclosing = (uclose & xclosing) | (uclose & xstopped);
  xstopped = (ustop | (uopen & xclosing) | (uclose & xopening);
  xopening = (uopen & xstopped) | (uopen & xopening);
}
```

Example 6: Impose a constraint



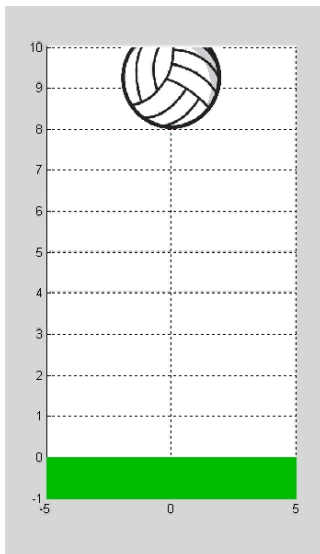
$$0 \leq h(k) \leq h_{max}$$



```
MUST {
  h - hmax <= 0;
  -h <= 0;
}
```

Arbitrary logic constraints are also supported

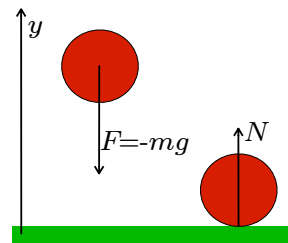
Bouncing ball example



$$\ddot{y} = -g$$

$$y \leq 0 \Rightarrow \dot{y}(t^+) = -(1 - \alpha)\dot{y}(t^-)$$

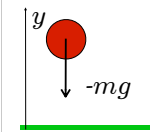
$$\alpha \in [0, 1]$$



How to model the bouncing ball as a discrete-time hybrid system?

Bouncing ball – Time discretization

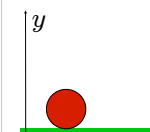
$$y(t) > 0 \quad v(t) \approx \frac{y(t) - y(t-1)}{T_s} \quad -g = \dot{v}(t) \approx \frac{v(t) - v(t-1)}{T_s}$$



$$\begin{cases} v(t+1) = v(t) - T_s g \\ y(t+1) = y(t) + T_s v(t+1) \\ = y(t) + T_s v(t) - T_s^2 g \end{cases}$$

$$y(t) \leq 0$$

$$v(t) = -(1 - \alpha)v(t-1) \quad y(t+1) = y(t-1) = y(t) - T_s v(t)$$



$$\begin{cases} v(t+1) = -(1 - \alpha)v(t) \\ y(t+1) = y(t) - T_s v(t) \end{cases}$$

go to demo `/demos/hybrid/bball.m`

Bouncing ball - HYSDEL model

```

SYSTEM bouncing_ball {
INTERFACE {
/* Description of variables and constants */
STATE { REAL height [-10,10];
        REAL velocity [-100,100]; }

PARAMETER {
        REAL g;
        REAL alpha; /* 0=elastic, 1=completely anelastic */
        REAL Ts; }
}
IMPLEMENTATION {
AUX { BOOL negative;
      REAL hnext;
      REAL vnext; }

AD { negative = height <= 0; }

DA { hnext = { IF negative THEN height-Ts*velocity
              ELSE height+Ts*velocity-Ts*Ts*g};
      vnext = { IF negative THEN -(1-alpha)*velocity
              ELSE velocity-Ts*g}; }

CONTINUOUS {
        height = hnext;
        velocity = vnext; }
}}
    
```

go to demo /demos/hybrid/bball.m

Bouncing ball - Simulation

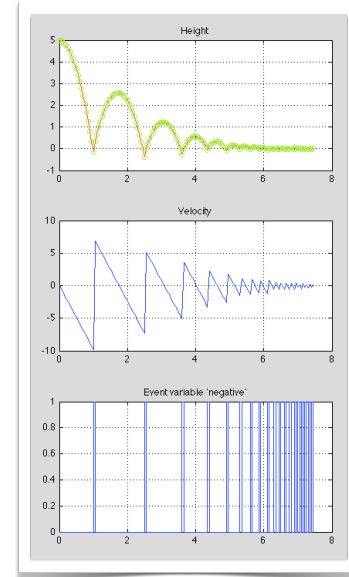
```

>>Ts=0.05;
>>g=9.8;
>>alpha=0.3;

>>S=mld('bouncing_ball',Ts);

>>N=150;
>>U=zeros(N,0);
>>x0=[5 0]';

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

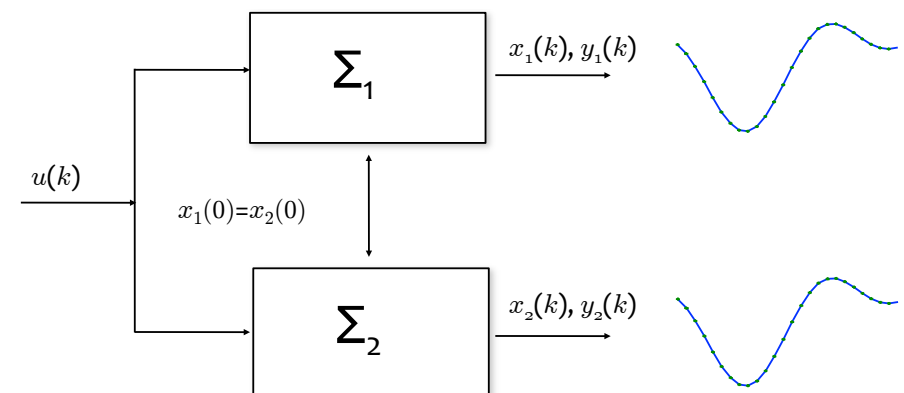


Note: no Zeno effect in discrete time !

Realization and Transformations (State-Space Hybrid Models)

Equivalences of hybrid models

Definition 1 Two hybrid systems Σ_1, Σ_2 are **equivalent** if for all initial conditions $x_1(0) = x_2(0)$ and input $\{u_1(k)\}_{k \in \mathbb{Z}_+} = \{u_2(k)\}_{k \in \mathbb{Z}_+}$ then $x_1(k) = x_2(k)$ and $y_1(k) = y_2(k)$, for all $k \in \mathbb{Z}_+$.



MLD and PWA Systems

Theorem MLD systems and PWA systems are equivalent

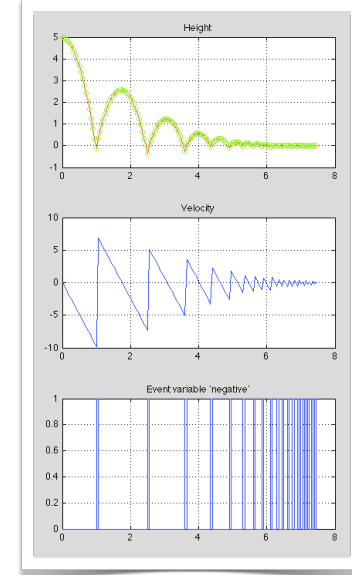
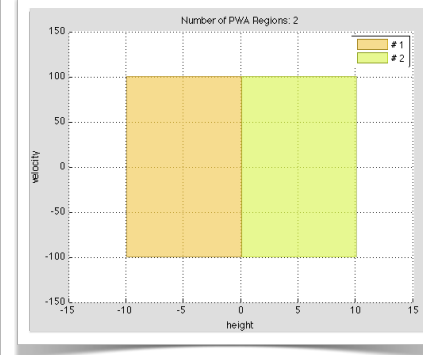
(Bemporad, Ferrari-Trecate, Morari, IEEE TAC, 2000)

- Proof is constructive: given an MLD system it returns its equivalent PWA form
- Drawback: it needs the **enumeration** of all possible combinations of binary states, binary inputs, and δ variables
- Most of such combinations lead to empty regions
- Efficient algorithms are available for converting MLD models into PWA models avoiding such an enumeration:
 - A. Bemporad, "Efficient Algorithms for Converting Mixed Logical Dynamical Systems into an Equivalent Piecewise Affine Form", IEEE Trans. Autom. Contr., 2004.
 - T. Geyer, F.D. Torrisi and M. Morari, "Efficient Mode Enumeration of Compositional Hybrid Models", HSCC'03

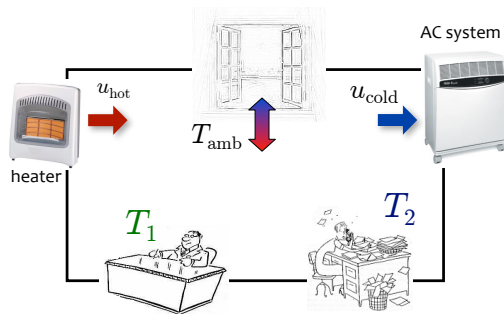
Bouncing ball - PWA equivalent

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

>>[X,T,I]=sim(P,x0,U);
```



Example: Room temperature



Hybrid dynamics

- #1 turns the heater (A/C) on whenever he is cold (hot)
- If #2 is cold he turns the heater on, unless #1 is hot
- If #2 is hot he turns A/C on, unless #2 is cold
- Otherwise, heater and A/C are off

• $\dot{T}_1 = -\alpha_1(T_1 - T_{amb}) + k_1(u_{hot} - u_{cold})$ (body temperature dynamics of #1)

• $\dot{T}_2 = -\alpha_2(T_2 - T_{amb}) + k_2(u_{hot} - u_{cold})$ (body temperature dynamics of #2)

go to demo /demos/hybrid/heatcool.m

HYSDEL model

```
SYSTEM heatcool (
INTERFACE {
STATE { REAL T1 [-10,50];
        REAL T2 [-10,50];
}
INPUT { REAL Tamb [-10,50];
}
PARAMETER {
REAL Ts, alpha1, alpha2, k1, k2;
REAL Thot1, Tcold1, Thot2, Tcold2, Uc, Uh;
}
IMPLEMENTATION {
AUX { REAL uhot, ucold;
      BOOL hot1, hot2, cold1, cold2;
}
AD { hot1 = T1>=Thot1;
     hot2 = T2>=Thot2;
     cold1 = T1<=Tcold1;
     cold2 = T2<=Tcold2;
}
DA { uhot = (IF cold1 | (cold2 & ~hot1) THEN Uh ELSE 0);
     ucold = (IF hot1 | (hot2 & ~cold1) THEN Uc ELSE 0);
}
CONTINUOUS { T1 = T1+Ts*(-alpha1*(T1-Tamb)+k1*(uhot-ucold));
             T2 = T2+Ts*(-alpha2*(T2-Tamb)+k2*(uhot-ucold));
}
}
```

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

get the MLD model in MATLAB

```
>>[XX,TT]=sim(S,x0,U);
```

simulate the MLD model

Hybrid MLD model

- MLD model

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

- 2 continuous states: (temperatures T_1, T_2)
- 1 continuous input: (room temperature T_{amb})
- 2 auxiliary continuous vars: (power flows u_{hot}, u_{cold})
- 6 auxiliary binary vars: (4 thresholds + 2 for OR condition)
- 20 mixed-integer inequalities

Possible combination of integer variables: $2^6 = 64$

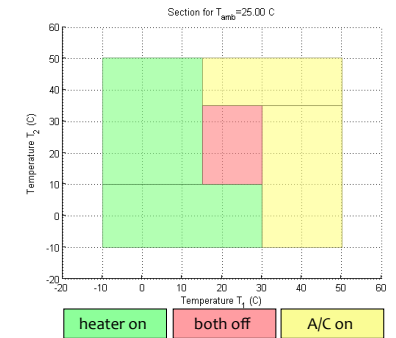
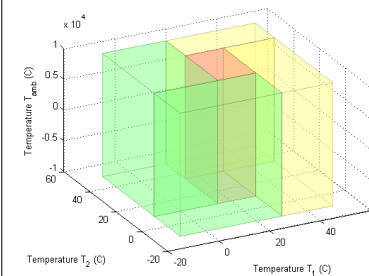
Hybrid PWA model

- PWA model

$$\begin{aligned} x(k+1) &= A_{i(k)}x(k) + B_{i(k)}u(k) + f_{i(k)} \\ y(k) &= C_{i(k)}x(k) + D_{i(k)}u(k) + g_{i(k)} \\ i(k) &\text{ s.t. } H_{i(k)}x(k) + J_{i(k)}u(k) \leq K_{i(k)} \end{aligned}$$

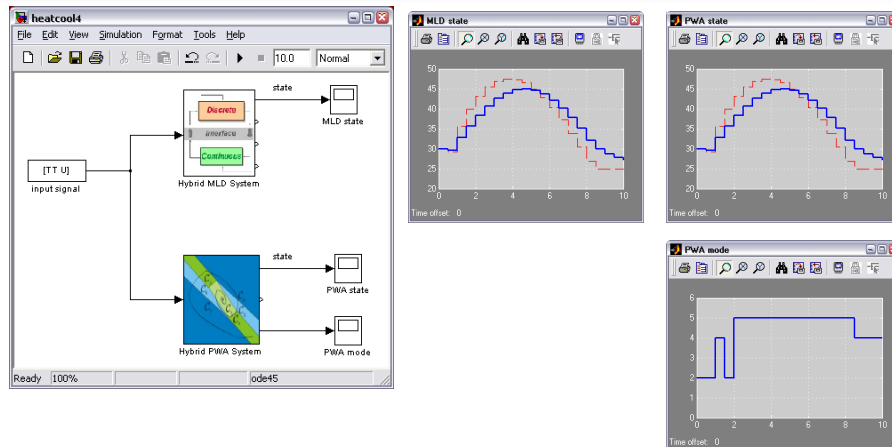
>>P=pwa(S);

- 2 continuous states: (temperatures T_1, T_2)
- 1 continuous input: (room temperature T_{amb})



- 5 polyhedral regions (partition does not depend on input)

Simulation in Simulink



MLD and PWA models are equivalent

Using MLD to PWA for Model Checking

- Assume plant + controller can be modeled as DHA:
 - **Plant** = PWA approximation (e.g.: nonlinear switched model)
 - **Controller** = switched linear controller (e.g: a combination of threshold conditions, logic, linear feedback laws, ...)
- Write HYSDEL model, convert to MLD, then to PWA
- The resulting PWA map tells how the closed-loop system behaves in different regions of the state-space

Identification of Hybrid Systems

Hybrid system identification

- Sometimes a *hybrid model* of the process (or of a part of it) cannot be derived manually from available knowledge.
- Therefore, a model must be either
 - Estimated from data (**model unknown**)
 - or **hybridized** before it can be used for control/analysis (**model known but nonlinear**)
- If a linear model is enough, no problem: several algorithms are available (e.g.: use Ljung's ID TBX)
- If switching modes are known and data can be generated for each mode, no problem: we identify one linear model per mode (e.g.: use Ljung's ID TBX)
- If modes & dynamics must be identified together, we need

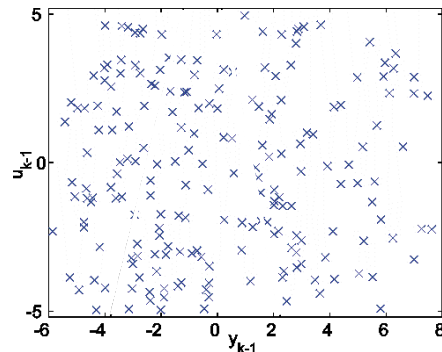
hybrid system identification

PWA identification problem

Estimate from data **both** the parameters of the affine submodels **and** the partition of the PWA map

Example Let the data be generated by the PWARX system

$$y_k = \begin{cases} \begin{bmatrix} -0.4 & 1 & 1.5 \end{bmatrix} \varphi_k + \varepsilon_k \\ \text{if } \begin{bmatrix} 4 & -1 & 10 \end{bmatrix} \varphi_k < 0 \\ \\ \begin{bmatrix} 0.5 & -1 & -0.5 \end{bmatrix} \varphi_k + \varepsilon_k \\ \text{if } \begin{bmatrix} -4 & 1 & -10 \\ 5 & 1 & -6 \end{bmatrix} \varphi_k \leq 0 \\ \\ \begin{bmatrix} -0.3 & 0.5 & -1.7 \end{bmatrix} \varphi_k + \varepsilon_k \\ \text{if } \begin{bmatrix} -5 & -1 & 6 \end{bmatrix} \varphi_k < 0 \end{cases}$$



with $\varphi_k = [y_{k-1} \ u_{k-1} \ 1]'$, $|u_k| \leq 5$
and $|\varepsilon_k| \leq 0.1$

PWA identification problem

$$\min_{\theta_j, H_j} \frac{1}{2N} \sum_{t=1}^N \left(\sum_{j=1}^s \|y_t - \varphi_t' \theta_j\|_{\mathcal{J}_j(\varphi_t)} \right)$$

subj. to $\mathcal{J}_j(\varphi_t) = \begin{cases} 1 & \text{if } H_j \varphi_t \leq 0 \\ 0 & \text{otherwise} \end{cases}$
+ linear bounds over θ_j, H_j

$$\begin{aligned} \|v\|_2^2 &= v'v && \text{Euclidean norm} \\ \|v\|_\infty &= \max |v_i| && \infty\text{-norm} \\ \|v\|_1 &= \sum |v_i| && 1\text{-norm} \end{aligned}$$

A. Known guardlines (partition H_j known, θ_j unknown): ordinary least-squares problem (or linear/quadratic program if linear bounds over θ_j are given)
EASY PROBLEM ...

B. Unknown guardlines (partition H_j and θ_j unknown): generally non-convex, local minima **HARD PROBLEM!**

Approaches to PWA Identification

- **Mixed-integer linear or quadratic programming**

J. Roll, A. Bemporad and L. Ljung, "Identification of hybrid systems via mixed-integer programming", *Automatica*, 2004

- **Bounded error & partition of infeasible set of inequalities**

A. Bemporad, A. Garulli, S. Paoletti and A. Vicino, "A Greedy Approach to Identification of Piecewise Affine Models", *HSCC'03 / IEEE TAC 2005*

- **K-means clustering in a feature space**

G. Ferrari-Trecate, M. Muselli, D. Liberati, and M. Morari, "A clustering technique for the identification of piecewise affine systems", *Automatica*, 2003

- **Bayesian approach**

Juloski A, Wieland S, Heemels WPMH, "A Bayesian approach to identification of hybrid systems. CDC 2004

- **Other approaches:**

- **Polynomial factorization (algebraic approach)** (R. Vidal, S. Soatto, S. Sastry, 2003)

- **Hyperplane clustering in data space** (E. Münz, V. Krebs, IFAC 2002)