

# A “Hybrid” Solver for Optimal Control of Hybrid Systems

Nicolò Giorgetti and Alberto Bemporad

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*Dept. of Information Engineering*  
*University of Siena, Italy*  
giorgetti@dii.unisi.it



## Limitations of MIP Approaches

Current approach for solving **optimal control** of hybrid systems: **Mixed-Integer** (linear/quadratic) **Programming**.

1) The discrete/logic part of the hybrid system must be converted into linear MI inequalities (e.g.: by HYSDEL)



**LOSS of the original discrete structure**  
**Introduction of auxiliary binary variables**

2) The efficiency of the MIP solver relies upon the tightness of the continuous LP/QP relaxations.



**Poor performance (= many LP/QPs) if relaxations are not tight**

**We need better solution techniques**

## “Hybrid” Solvers

Combine MIP and Constraint Logic Programming (CLP) to overcome the previous difficulties

### Why CLP ?

- More flexible modeling than MIP
- Structure is kept and exploited to direct the search.

### Why MIP ?

- Specialized techniques for highly structured problems (e.g. LP problems);
- A wide range of tight relaxations are available

### Why a combined approach ?

Performance increase already shown in other application domains

(Harjunkski, Jain, Grossmann, 2000)

## Constraint Logic Programming

CLP is a set of techniques for solving a **finite domain problem**

(= set of constraints over a set of integer finite domain variables)

Example:

$$X \neq 7 \quad Z \neq 2 \quad X = Z + 3Y \\ X \in [1\#8] \quad Y \in [1\#10] \quad Z \in [1\#10]$$

GOAL: Find all feasible assignments for X,Y, and Z

CLP alternates two techniques for solving the problem:

- Constraint Propagation: efficient inference mechanism used to reduce the domains of the variables.
- Constraint Distribution: splits a problem into complementary cases once constraint propagation cannot advance further.

# Constraint Logic Programming

Example (continued):

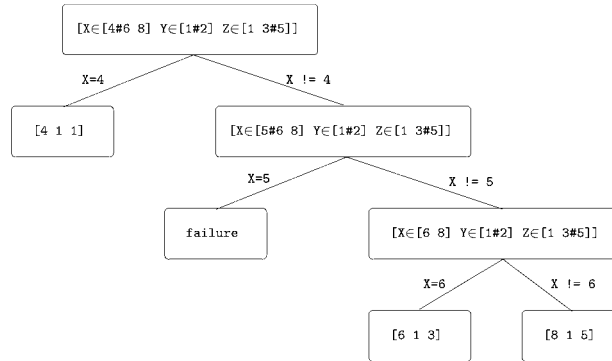
$$X \neq 7 \quad Z \neq 2 \quad X = Z + 3Y$$

$$X \in [1\#8] \quad Y \in [1\#10] \quad Z \in [1\#10]$$

A pre-phase of constraint propagation produces

$$X \in [4\#6 \ 8] \quad Y \in [1\#2] \quad Z \in [1 \ 3\#5]$$

By iterating between propagation and distribution the solutions of the problem are determined:



# Logic-Based Branch&Bound

The basic modeling framework has the following form:

(Bockmayr, Kasper, 1998)

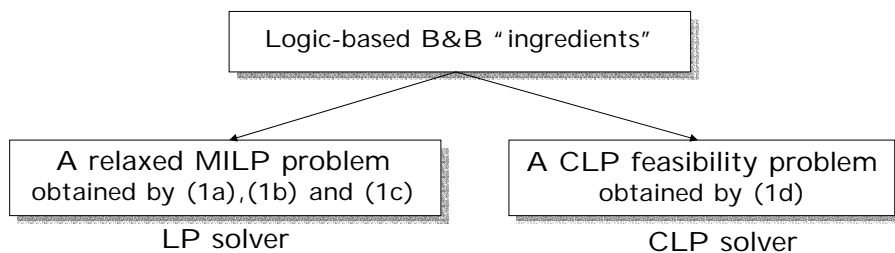
$$\min_v f(v) \quad (1a)$$

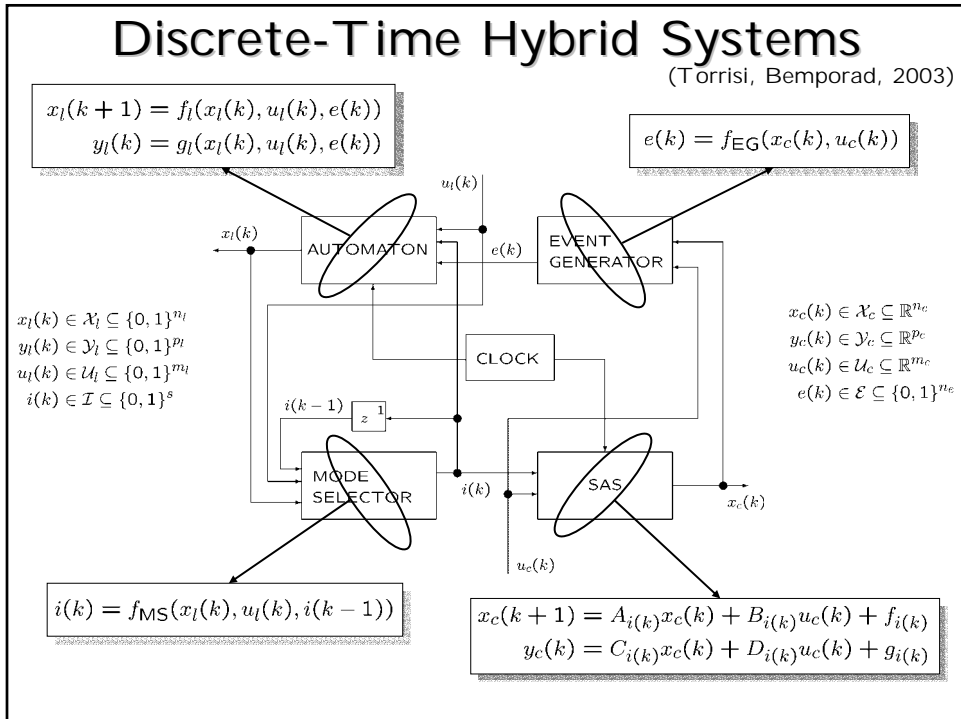
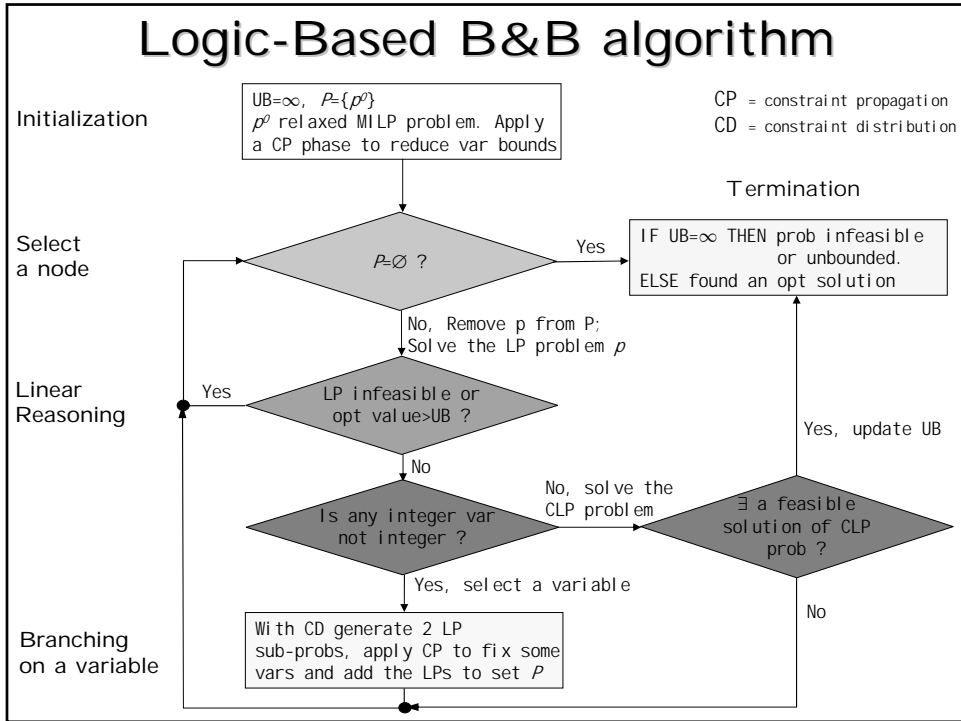
$$\text{s.t. } Gv \leq d, \quad \overline{G}v = \overline{d} \quad (1b) \text{ Continuous constraints}$$

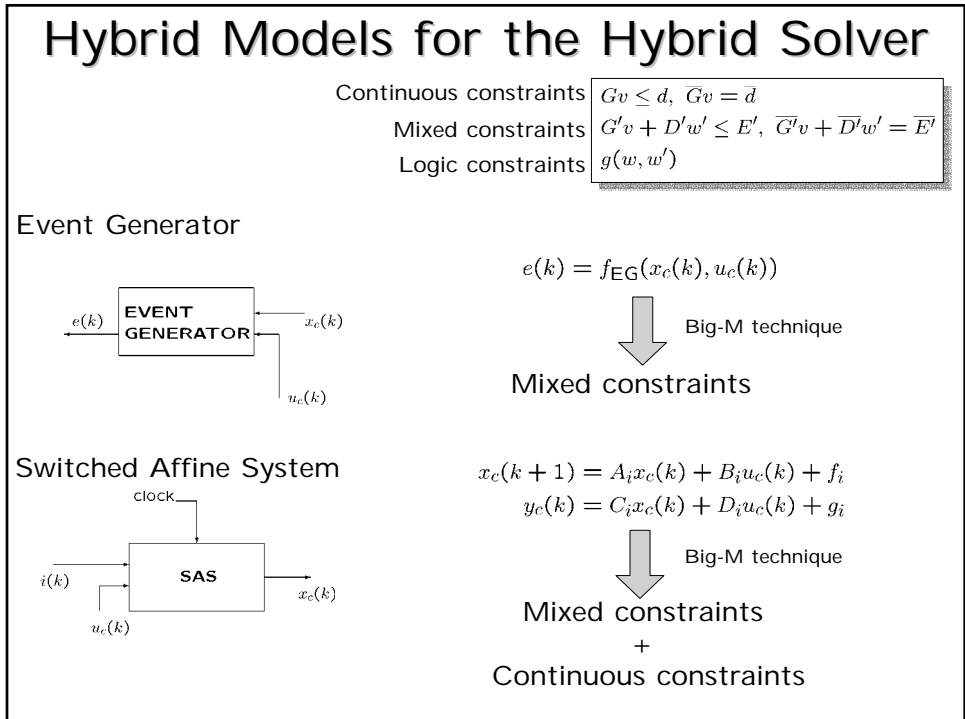
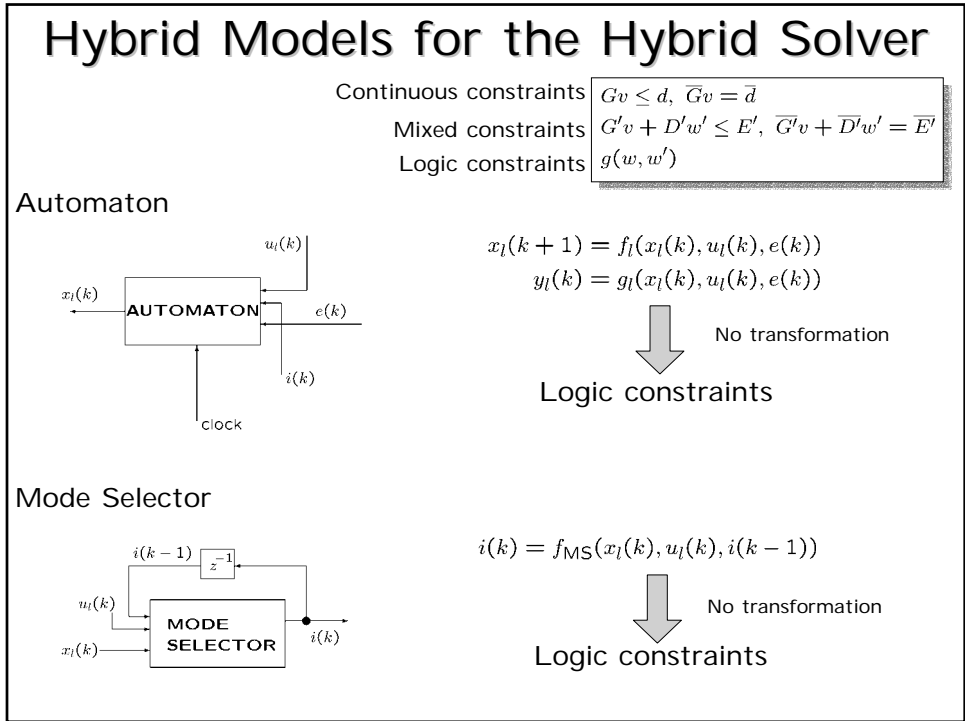
$$G'v + D'w' \leq E', \quad \overline{G}'v + \overline{D}'w' = \overline{E}' \quad (1c) \text{ Mixed constraints}$$

$$g(w, w') \quad (1d) \text{ Logic constraints}$$

$$v \in \mathcal{R}^{n_v}, \quad w \in \{0, 1\}^{n_w}, \quad w' \in \{0, 1\}^{n'_w}$$







# Optimal Control of Hybrid Systems

We define the following optimal control problem:

$$\min_{\{x(k), u(k)\}_{k=0}^{T-1}} \sum_{k=0}^{T-1} \|Q_x(x(k) - x_e(k))\|_\infty + \|Q_u(u(k) - u_e(k))\|_\infty + \|Q_y(y(k) - y_e(k))\|_\infty$$

s.to

$x_l(k+1) = f_l(x_l(k), u_l(k), e(k))$	(1)
$y_l(k) = g_l(x_l(k), u_l(k), e(k))$	(2)
$x_c(k+1) = A_{i(k)}x_c(k) + B_{i(k)}u_c(k) + f_{i(k)}$	(3)
$y_c(k) = C_{i(k)}x_c(k) + D_{i(k)}u_c(k) + g_{i(k)}$	(4)
$e(k) = f_{EG}(x_c(k), u_c(k))$	(5)
$i(k) = f_{MS}(x_l(k), u_l(k), i(k-1))$	(6)
$h_D(k)(\{x, u, e, i\}_0^{T-1}) \leq 0$	(7)
$h_A(k)(\{x, u, e, i\}_0^{T-1}) \leq 0$	(8)

where:

(1-6) are dynamical constraints

(7) are design constraints (e.g.: input/state/logic constraints)

(8) are ancillary constraints (don't change the solution, only help the solver. E.g.: reachability constraints)

**This problem can be solved by the hybrid solver**

## Motorbike Example

**MODEL:** a simplified hybrid model of a motorbike with three "semi-automatic" gears.

**GOAL:** solve an optimal control problem in order to track a desired speed profile



**SAS** - Discrete-time continuous dynamics

Speed (Km/h)  $v(k+1) = (1 - \alpha_i)v(k) + \beta_i\omega(k) - cu_b$   
 Engine speed (rpm)  $\omega(k+1) = \omega(k) + du_t - fu_b$

2 continuous inputs:

$u_t$  engine torque (Nm)  
 $u_b$  brake force (N)

2 continuous states:  
 $v$  and  $\omega$

**Event Generator** – Thresholds events for automatic gear shift

$$[d_1(k) = 1] \iff [\omega(k) \leq t_1]$$

$$[d_2(k) = 1] \iff [\omega(k) \leq t_2]$$

where  $t_1$  and  $t_2$ ,  $t_1 \leq t_2$ , are constant thresholds.

## Motorbike Example

**Automaton - Gear automaton**

3 binary inputs:  
**auto:** automatic gear shifts enabled  
**moveup, movedown:** manual gear shift commands

4 binary states:  
**idle, gear1, gear2, gear3**

**Mode selector** – Current gear of the motorbike ( $\equiv$  logic state)

Optimal Control Problem: minimize  $\sum_{k=0}^T |v(k) - v_e|$  subject to the hybrid dynamics and the following additional constraints:

Continuous Constraints

 $0 \leq u_t(k) \leq 50$   
 $0 \leq u_b(k) \leq 50.$

Logic condition on torque and brakes

 $[d_t = 0] \leftrightarrow [u_t(k) \leq 0]$   
 $[d_b = 0] \leftrightarrow [u_b(k) \leq 0]$   
 $\neg(d_t \wedge d_b)$

An exclusive-or condition on the gear commands

 $\text{moveup}(k) \oplus \text{movedown}(k)$

## Motorbike Example

Initial conditions:  
 $v(0) = 0$  Km/h  
 $\omega(0) = 0$  rad/s  
 gear = idle

$v_e = 200$  Km/h  
 $T = 20$  s  
 $t_1 = 100$  rpm  
 $t_2 = 400$  rpm

**Same results both with MILP and with lb-B&B**

CPU time	Horizon	Integer vars	MILP	lb-B&B	Reduction(%)
(Pentium IV 1.8GHz Ilog CPLEX 8.1 + Solver 5.3)	10	110	0.221	0.12	45.70
	20	220	8.091	1.992	75.38
	30	330	26.318	7.798	70.37
	40	440	69.961	20.128	71.23

LPs solved	Horizon	MILP	lb-B&B
	10	309	6
	20	30645	49
	30	71070	126
	40	146101	229

## Conclusions

### Current state:

- ✓ A unifying framework for MIP and CLP techniques for solving optimal control problems for hybrid systems.
- ✓ A more expressive modeling language for hybrid systems (due to CLP).
- ✓ A superior computation time in comparison to more standard mixed-integer programming techniques.

## Ongoing and Future Research

- MIQP logic-based solvers for quadratic performance indices
- Embed MILP solvers in CLP as logic constraints, and solve satisfiability problems
- Alternative relaxations tighter than big-M (particularly for SAS and EG parts)
- Applications



## References

- [1] A. Bemporad and N. Giorgetti. Logic-based hybrid solvers for optimal control of hybrid systems. 2003. Submitted CDC 2003.

Paper download: <http://www.dii.unisi.it/~bemporad>