



# Embedded Systems 2012/13

## Lecture 7 Construction of Symbolic Models



Basilica di Santa Maria di Collemaggio, 1287, L'Aquila

# Construction of symbolic models

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We now consider **digital control systems**, i.e. control systems where input signals are piecewise constant.

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Consider a nonlinear digital control system  $\Sigma$

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m$$

and  $\tau > 0$ , define:

$$T^*_{\tau}(\Sigma) = (Q, L, \xrightarrow{\quad}, O, H)$$

where:

- $Q = \mathbb{R}^n$
- $L$  is the collection of constant functions  $u: [0, \tau] \rightarrow \mathbb{R}^m$
- $p \xrightarrow{u} q$ , if  $x(\tau, p, u) = q$
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... summarizing

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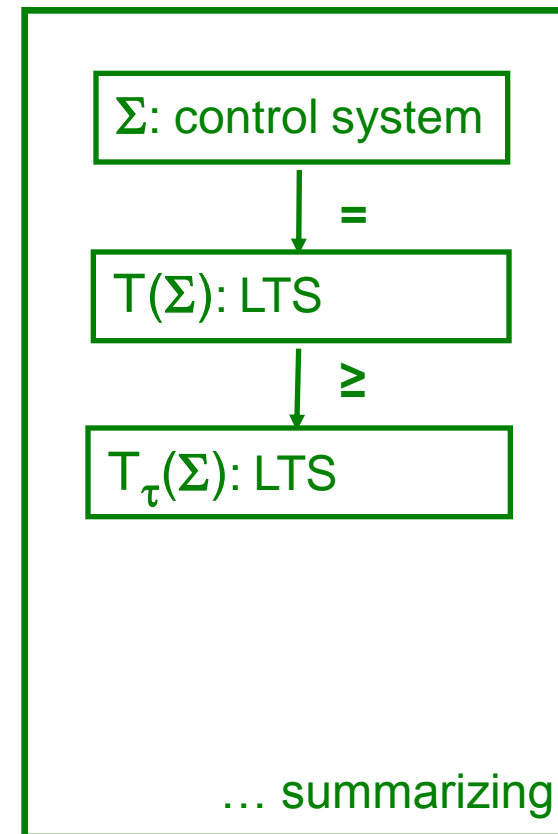
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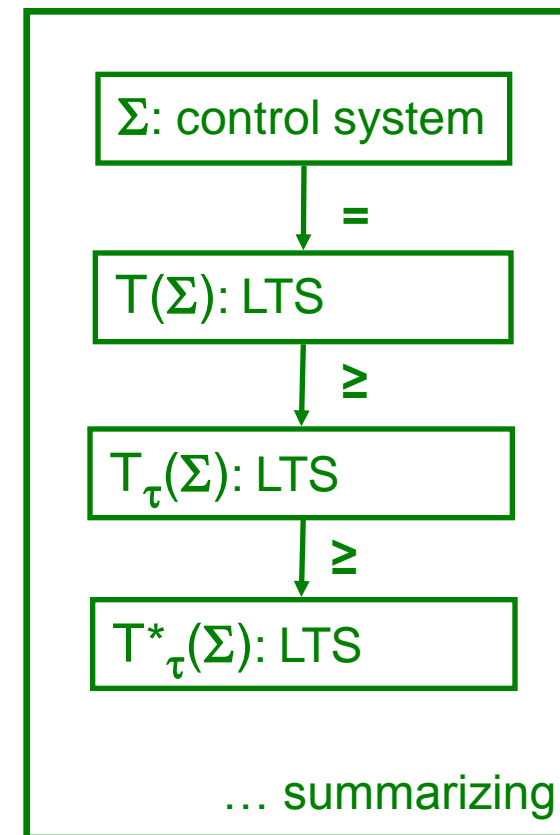
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# Construction of symbolic models

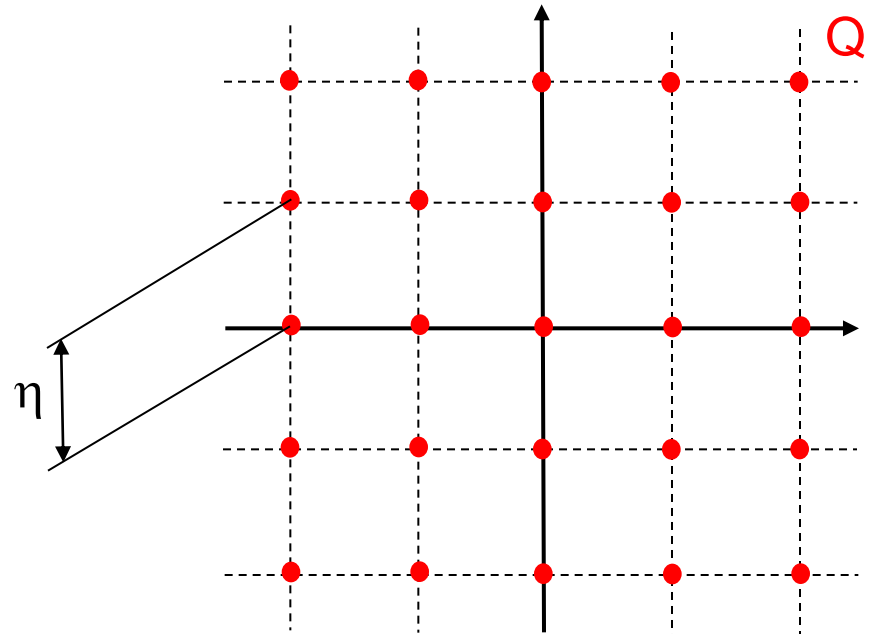
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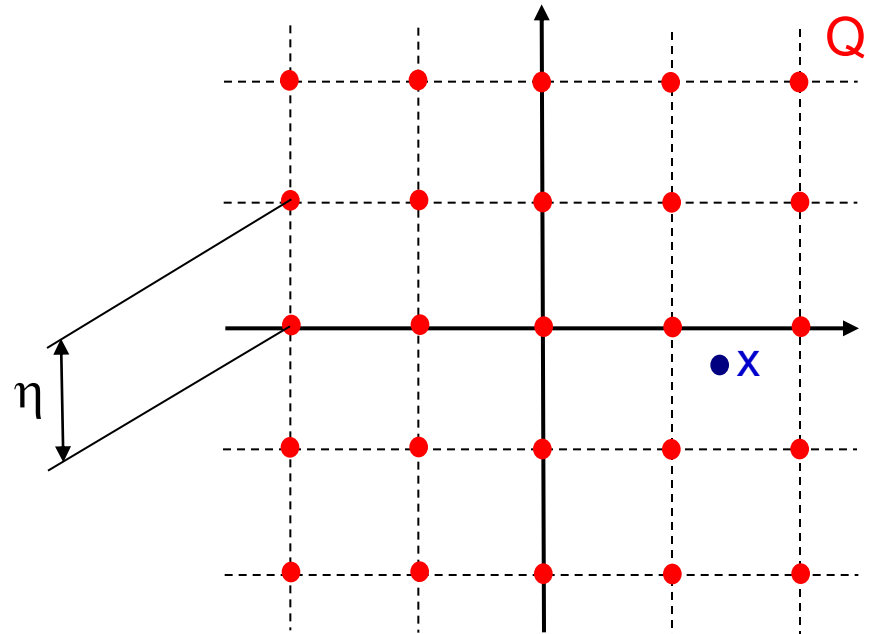
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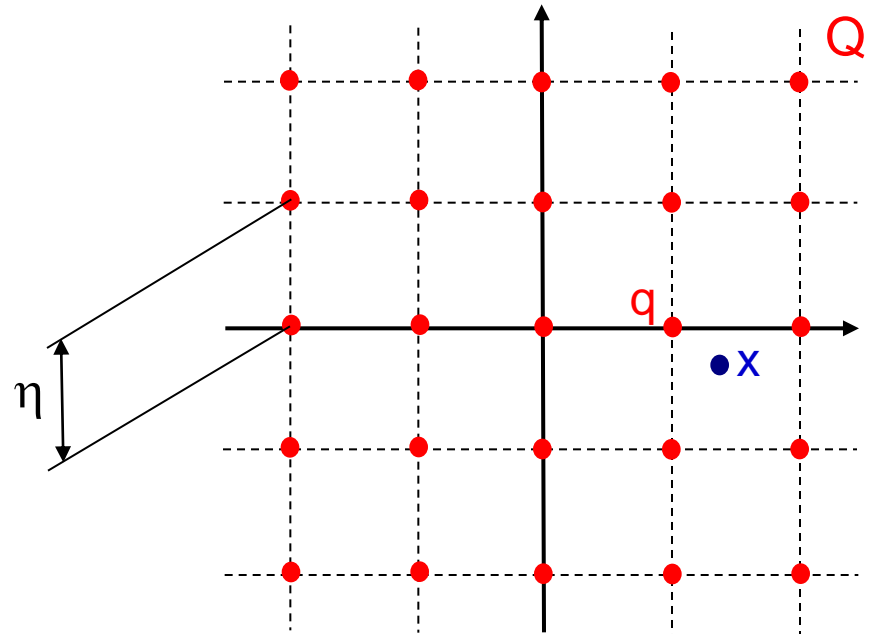
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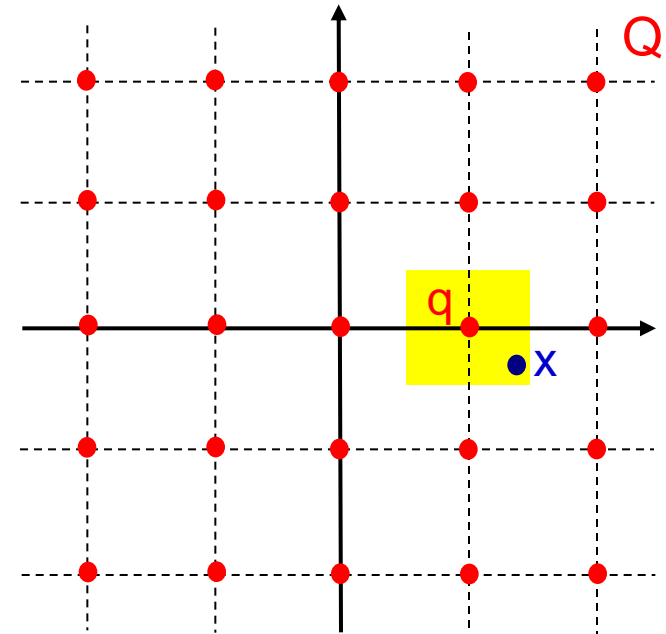
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$\forall x \in R^n \quad \exists q \in Q \text{ s.t. } x \in \mathcal{B}_{[\eta/2]}(q)$



# Construction of symbolic models

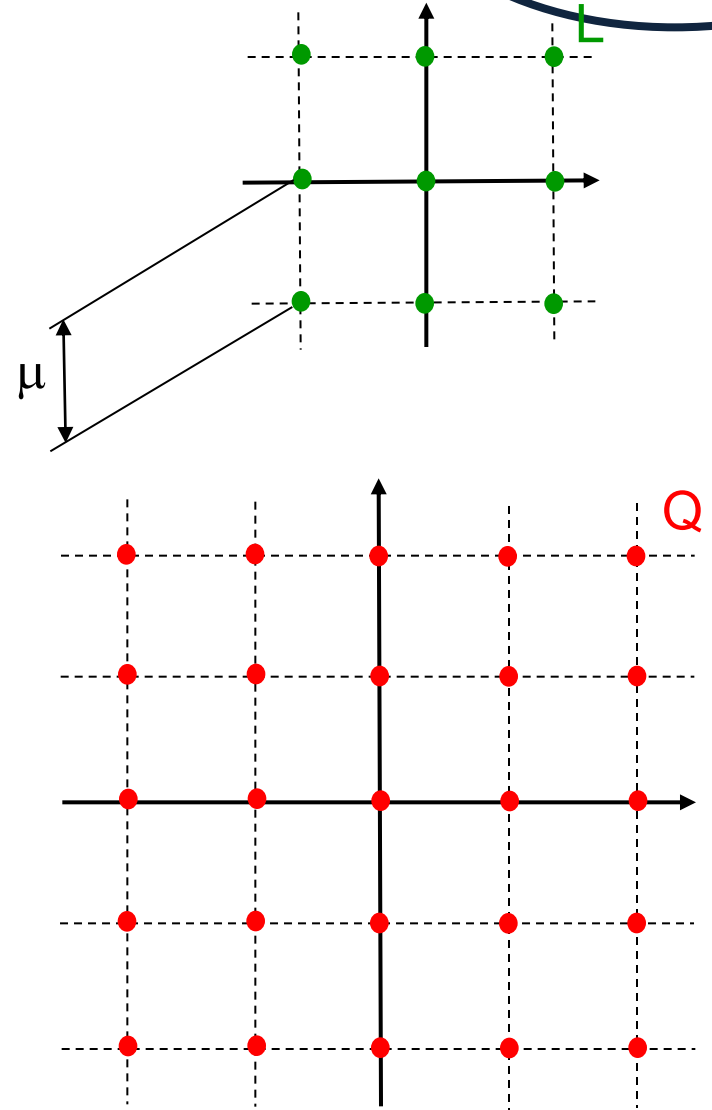
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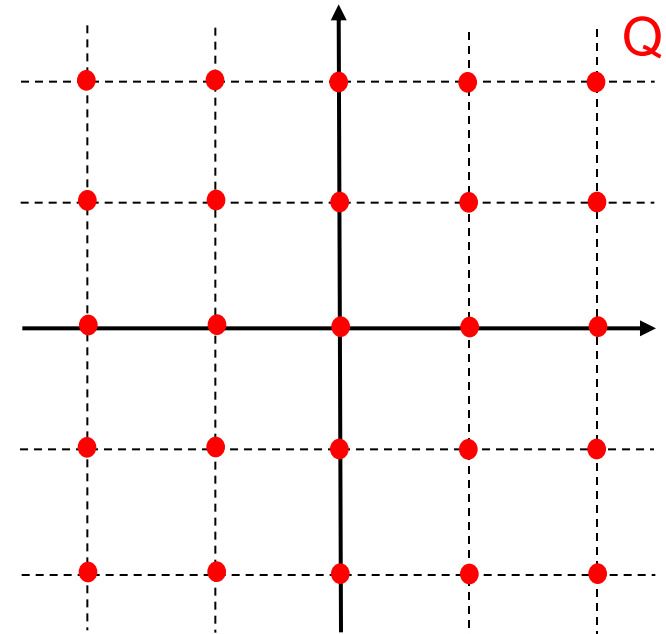
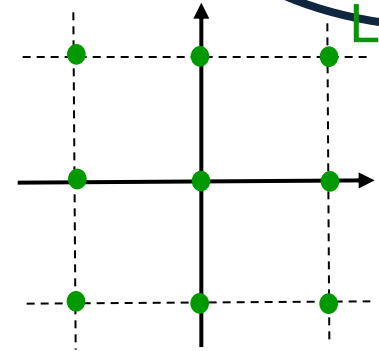
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Labelled transition system  $T_{\tau,\eta,\mu}(\Sigma)$  is countable



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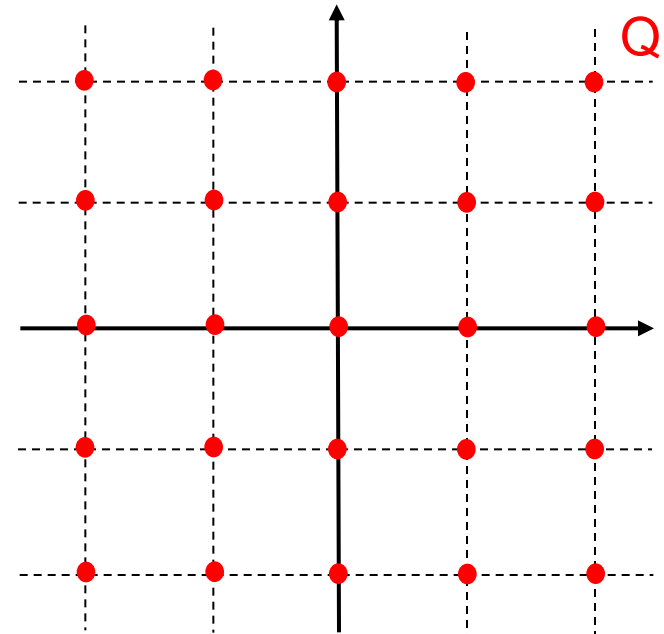
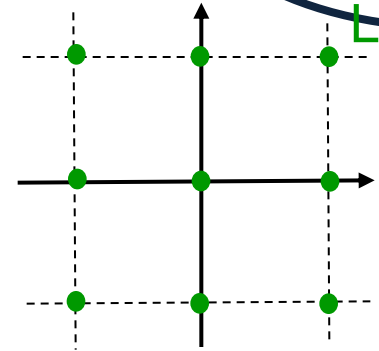
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How do I construct this symbolic model ?



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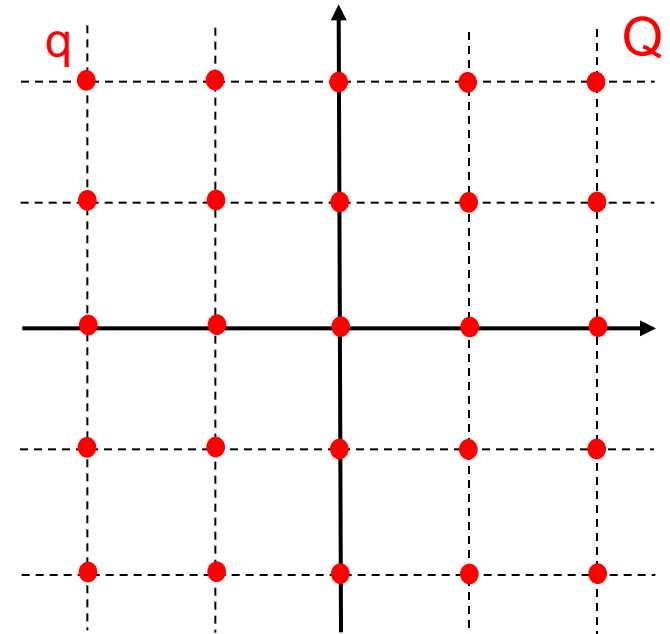
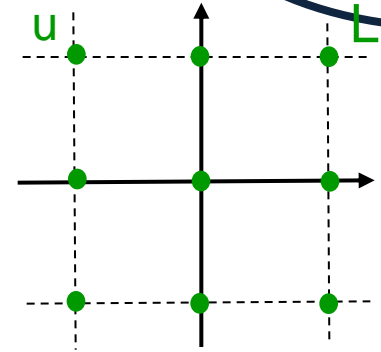
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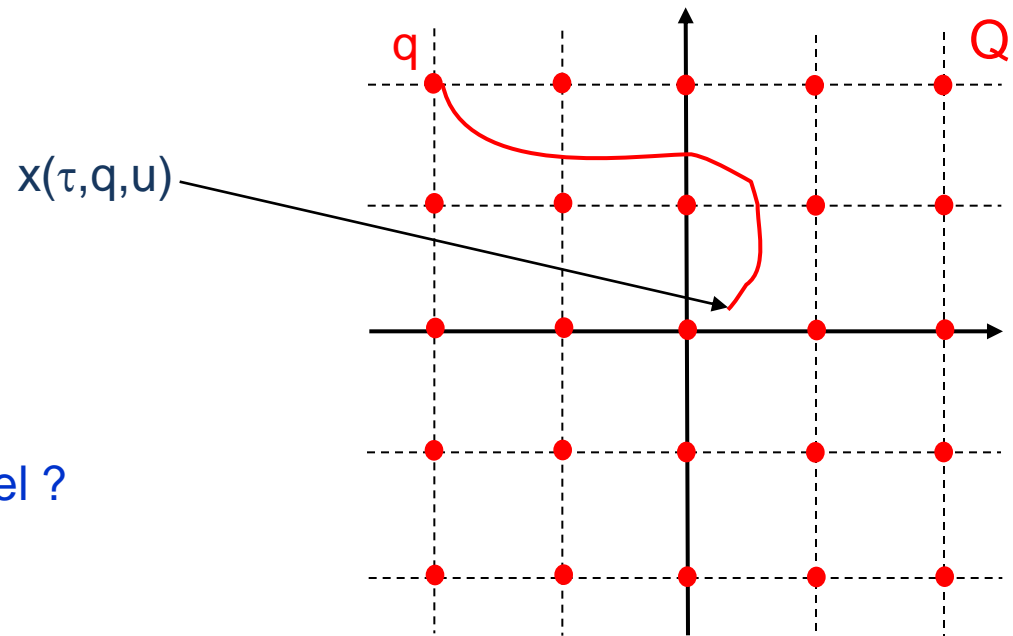
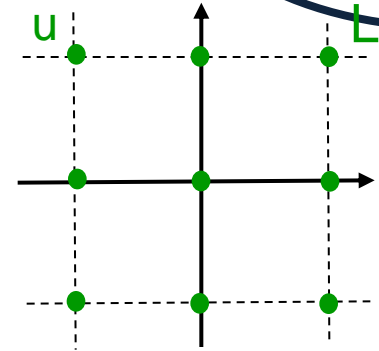
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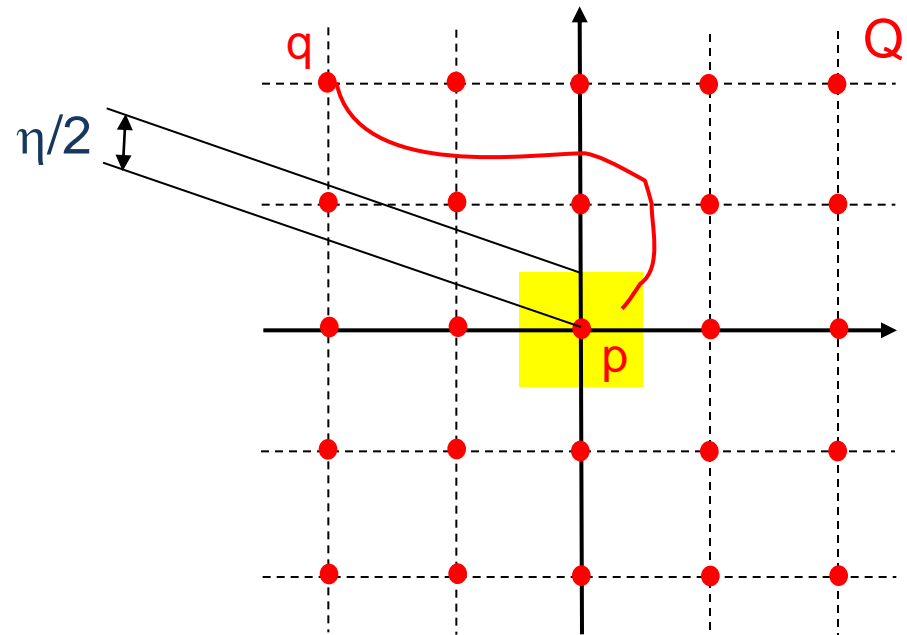
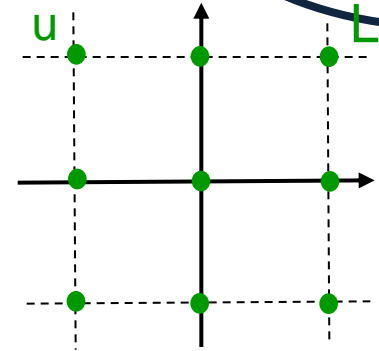
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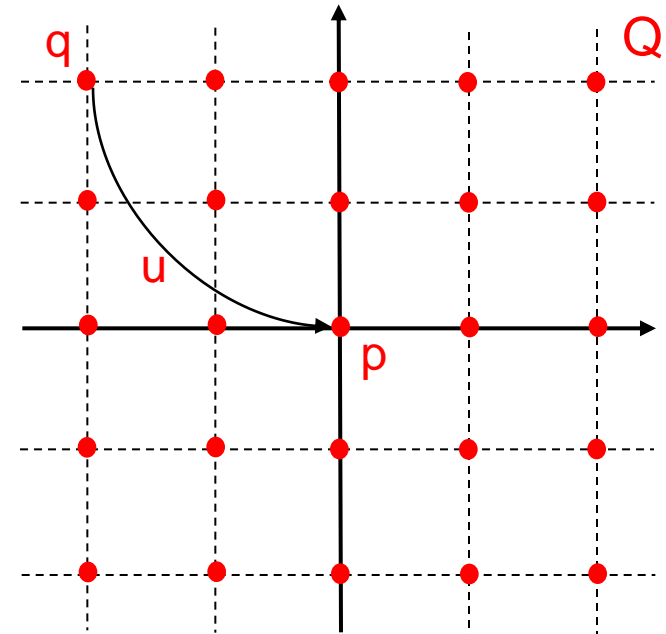
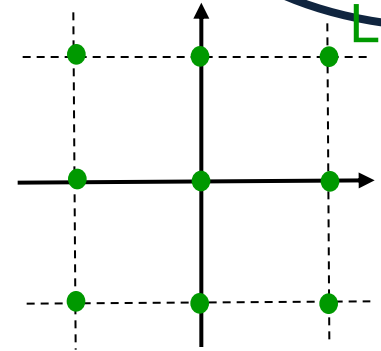
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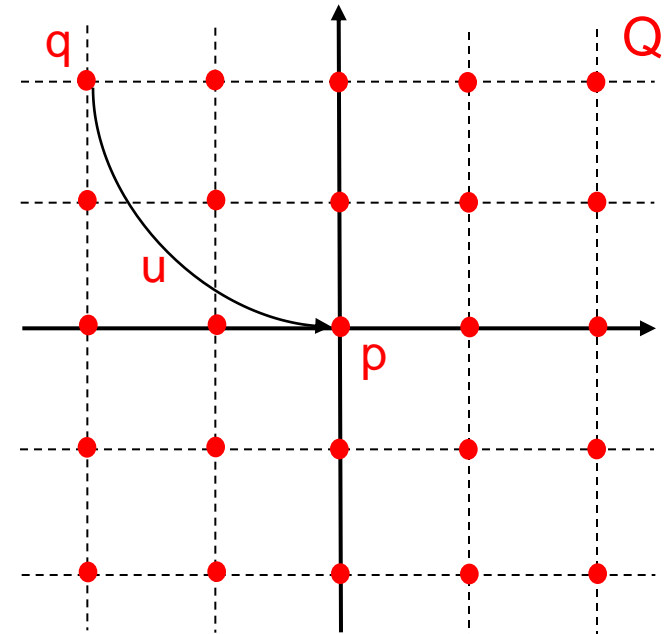
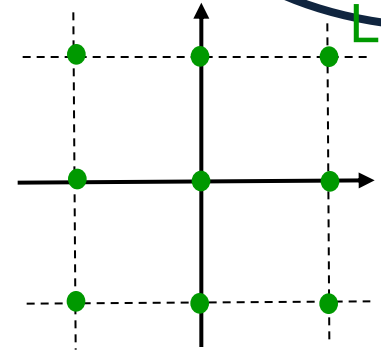
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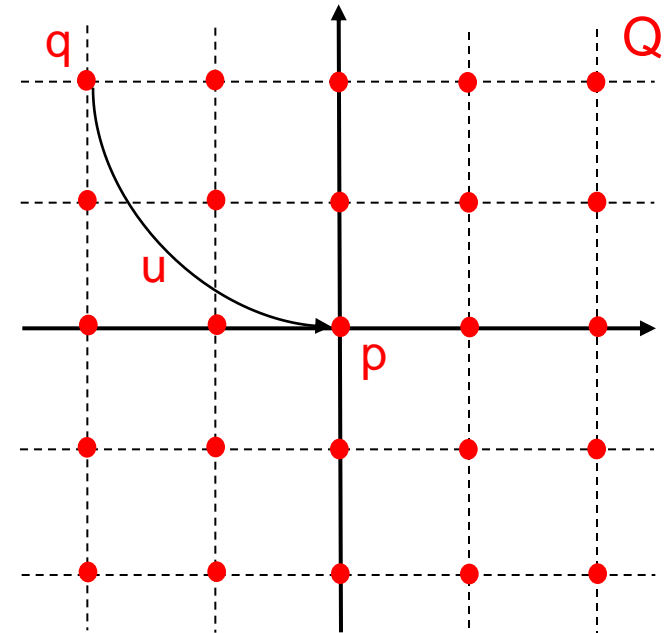
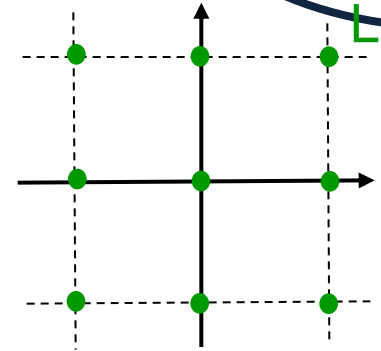
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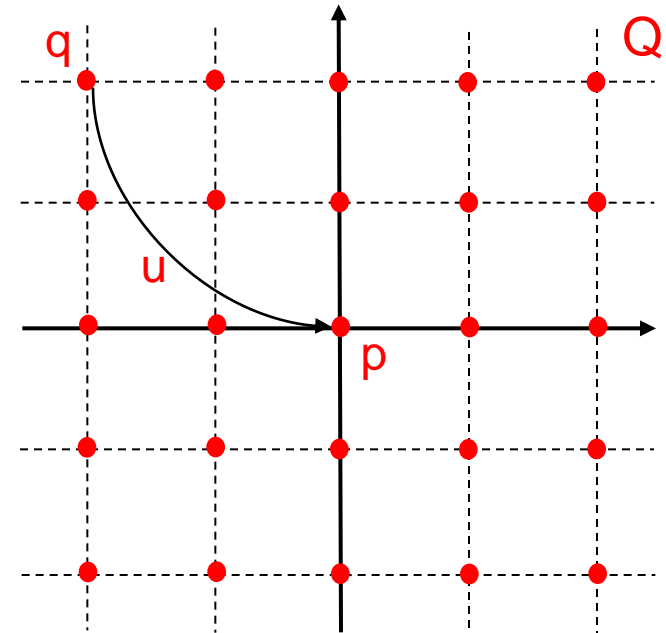
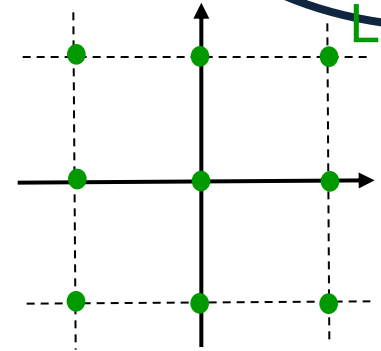
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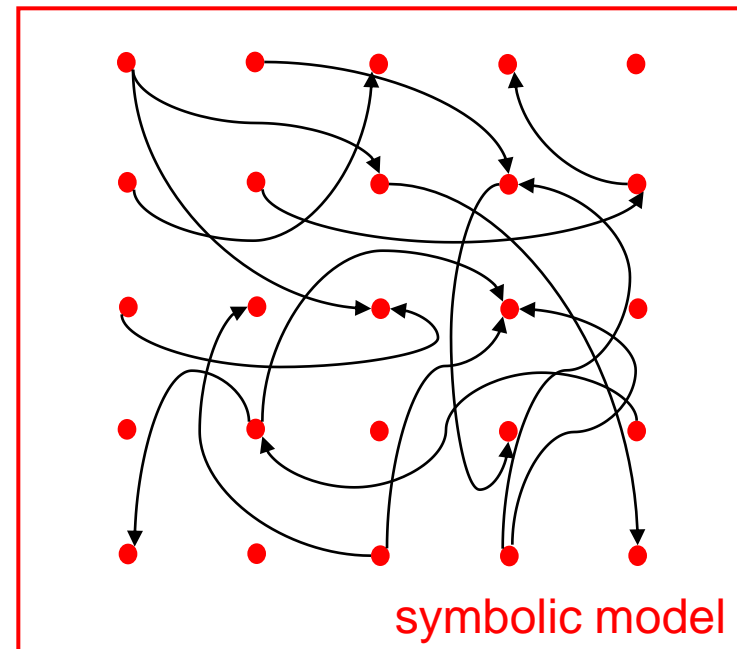
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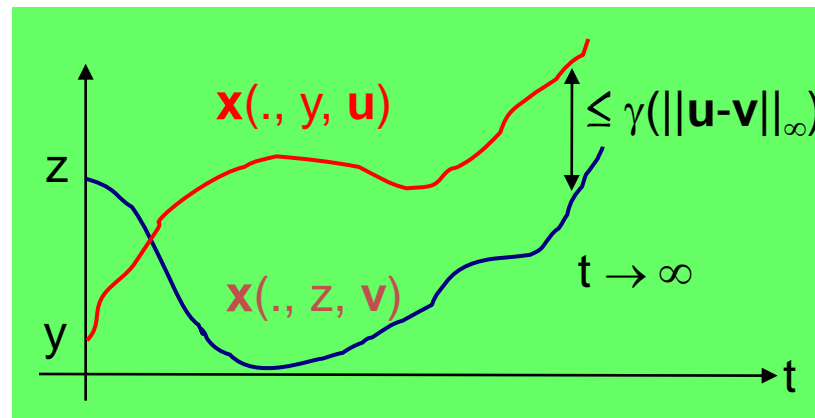
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A control system  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u})$  is Incrementally Input-to-State Stable ( $\delta$ -ISS) if there exist a KL function  $\beta$  and a  $K_\infty$  function  $\gamma$  so that for any  $t \geq 0$ ,  $y, z \in \mathbb{R}^n$  and  $\mathbf{u}, \mathbf{v}$

$$\| \mathbf{x}(t, y, \mathbf{u}) - \mathbf{x}(t, z, \mathbf{v}) \| \leq \beta( \|y - z\|, t ) + \gamma( \|\mathbf{u} - \mathbf{v}\|_\infty )$$



Further details in D. Angeli, A Lyapunov approach to incremental stability properties, IEEE-TAC 02

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## Theorem:

A control system  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u})$  is  $\delta$ -ISS if there exists a smooth function  $V : \mathbb{R}^n \times \mathbb{R}^n \rightarrow \mathbb{R}^+$  and  $K_\infty$  functions  $\alpha_1, \alpha_2, \rho, \sigma$  such that:

- i)  $\alpha_1( \| \mathbf{x} - \mathbf{y} \| ) \leq V( \mathbf{x}, \mathbf{y} ) \leq \alpha_2( \| \mathbf{x} - \mathbf{y} \| )$  for all  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$
- ii)  $dV/d\mathbf{x} \mathbf{f}(\mathbf{x}, \mathbf{u}) + dV/d\mathbf{y} \mathbf{f}(\mathbf{y}, \mathbf{v}) < -\rho( \| \mathbf{x} - \mathbf{y} \| ) + \sigma( \| \mathbf{u} - \mathbf{v} \| )$  for all  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$  and  $\mathbf{u}, \mathbf{v} \in \mathbb{R}^m$

Further details from D. Angeli, A Lyapunov approach to incremental stability properties, IEEE-TAC 02



**Theorem** Consider a nonlinear digital control system  $\Sigma$

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m$$

If  $\Sigma$  is  $\delta$ -ISS then for any desired precision  $\varepsilon > 0$  and for any  $\tau, \eta, \mu > 0$  satisfying

$$\beta(\varepsilon, \tau) + \eta / 2 + \gamma(\mu) \leq \varepsilon$$

The labelled transition systems  $T_{\tau}^*(\Sigma)$  and  $T_{\tau, \eta, \mu}(\Sigma)$  are approximately bisimilar with precision  $\varepsilon$ .

# Construction of symbolic models

**Theorem 5.1.** Consider a control system  $\Sigma$  and any desired precision  $\varepsilon \in \mathbb{R}^+$ . If  $\Sigma$  is  $\delta$ -ISS then for any  $\tau \in \mathbb{R}^+$ ,  $\eta \in \mathbb{R}^+$ , and  $\mu \in \mathbb{R}^+$  satisfying the following inequality:

$$\beta(\varepsilon, \tau) + \gamma(\mu) + \eta/2 \leq \varepsilon, \quad (17)$$

the transition system  $T_{\mathcal{U}_\tau}(\Sigma)$  is  $\varepsilon$ -bisimilar to  $T_{\tau, \eta, \mu}(\Sigma)$ .

Before giving the proof of this result we point out that, analogously to condition (8) of Theorem 4.1, there always exist parameters  $\tau \in \mathbb{R}^+$ ,  $\eta \in \mathbb{R}^+$ , and  $\mu \in \mathbb{R}^+$  satisfying condition (17).

**Proof.** Consider the relation  $R \subseteq Q_1 \times Q_2$  defined by  $(x, q) \in R$  if and only if  $\|x - q\| \leq \varepsilon$ . By construction  $R(Q_1) = Q_2$ ; since  $Q_1 \subseteq \bigcup_{q_2 \in Q_2} \mathcal{B}_{\eta/2}(q_2)$  and by (17),  $\eta/2 < \varepsilon$ , we have that  $R^{-1}(Q_2) = Q_1$ . We now show that  $R$  is an  $\varepsilon$ -approximate bisimulation relation between  $T_{\mathcal{U}_\tau}(\Sigma)$  and  $T_{\tau, \eta, \mu}(\Sigma)$ . Consider any  $(x, q) \in R$ . Condition (i) in Definition 3.2 is satisfied by the definition of  $R$ . Let us now show that condition (ii) in Definition 3.2 holds. Consider any  $l_1 \in L_1$  and the transition  $x \xrightarrow{1}{l_1} y$  in  $T_{\mathcal{U}_\tau}(\Sigma)$ . Consider a label  $l_2 \in L_2$  such that:

$$\|l_1 - l_2\| \leq \mu, \quad (18)$$

and set  $z = \mathbf{x}(\tau, q, l_2)$ . (Notice that such label  $l_2 \in L_2$  exists because the assumptions on  $U$  make  $L_2 = [L_1]_\mu$  non-empty.) For later use notice that since  $l_1$  and  $l_2$  are constant functions, then  $\|l_1 - l_2\| = \|l_1 - l_2\|_\infty$ . Since  $Q_1 \subseteq \bigcup_{q_2 \in [\mathbb{R}^n]_\eta} \mathcal{B}_{\eta/2}(q_2)$ , there exists  $p \in Q_2$  such that:

$$\|z - p\| \leq \eta/2, \quad (19)$$

and therefore  $q \xrightarrow{2}{l_2} p$  in  $T_{\tau, \eta, \mu}(\Sigma)$ . Since  $\Sigma$  is  $\delta$ -ISS and by (18), (19) and (17), the following chain of inequalities holds:

$$\begin{aligned} \|y - p\| &= \|y - z + z - p\| \leq \|y - z\| + \|z - p\| \\ &\leq \beta(\|x - q\|, \tau) + \gamma(\|l_1 - l_2\|_\infty) + \eta/2 \\ &\leq \beta(\varepsilon, \tau) + \gamma(\mu) + \eta/2 \leq \varepsilon. \end{aligned} \quad (20)$$

Hence  $(y, p) \in R$  and condition (ii) in Definition 3.2 holds. We now show that also condition (iii) holds. Consider any  $(x, q) \in R$ ,  $l_2 \in L_2$  and the transition  $q \xrightarrow{2}{l_2} p$  in  $T_{\tau, \eta, \mu}(\Sigma)$ . By definition of  $T_{\tau, \eta, \mu}(\Sigma)$

$$\|z - p\| \leq \eta/2, \quad (21)$$

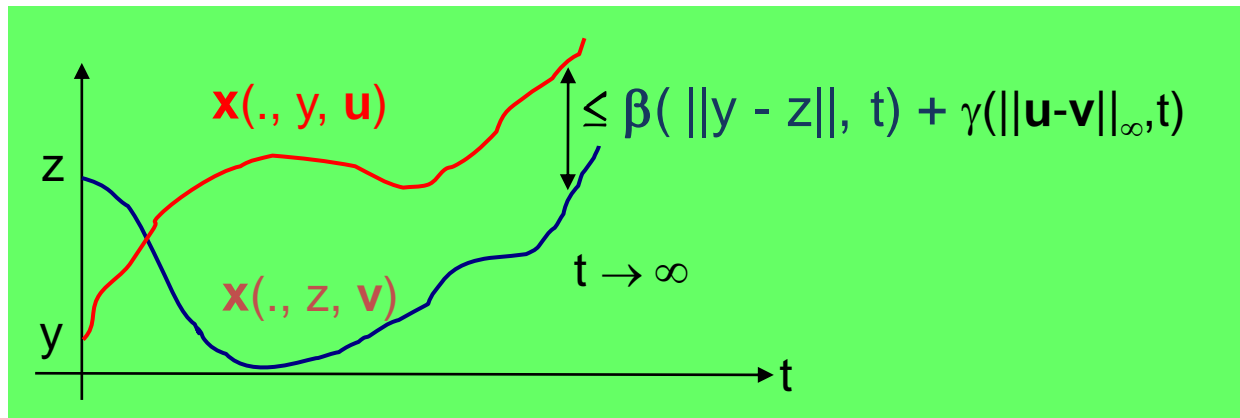
where  $z = \mathbf{x}(\tau, q, l_2) \in Q_1$ . Choose  $l_1 = l_2 \in L_1$  and consider now the transition  $x \xrightarrow{1}{l_1} y$  in  $T_{\mathcal{U}_\tau}(\Sigma)$ . Since  $\Sigma$  is  $\delta$ -ISS and by (21) and (17), the chain of inequalities in (20) holds. Thus  $(y, p) \in R$ , which completes the proof. ■

... taken from:

G. Pola, A. Girard, P. Tabuada  
Approximately bisimilar symbolic models  
for nonlinear control systems  
Automatica 44 (2008) 2508-2516

A control system  $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u})$  is **Incrementally Forward Complete ( $\delta$ -FC)** if it is forward complete and there exist continuous functions  $\beta$  and  $\gamma$  such that  $\beta(\cdot, s)$  and  $\gamma(\cdot, s)$  are  $K_\infty$  functions for any  $s \in \mathbb{R}^+$ , and for any  $t \geq 0$ , for any  $y, z \in \mathbb{R}^n$  and any  $\mathbf{u}, \mathbf{v}$ , the following condition is satisfied:

$$\|\mathbf{x}(t, y, \mathbf{u}) - \mathbf{x}(t, z, \mathbf{v})\| \leq \beta(\|y - z\|, t) + \gamma(\|\mathbf{u} - \mathbf{v}\|_\infty, t)$$



Further details in Zamani et al., Symbolic Models for Nonlinear Control Systems Without Stability Assumptions, IEEE TAC 12.

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## Properties

- Incremental FC requires the distance between two arbitrary trajectories to be bounded by a continuous function that captures the mismatch between initial conditions and control inputs.
- The class of delta-FC control systems is quite large and includes unstable linear systems.
- Any linear system is delta-FC
- $\delta$ -FC is implied by  $\delta$ -ISS

Further details in Zamani et al., Symbolic Models for Nonlinear Control Systems Without Stability Assumptions, IEEE TAC 12.

# Bisimulation equivalence

Given  $T_1 = (Q_1, L_1, \longrightarrow_1, O_1, H_1)$  and  $T_2 = (Q_2, L_2, \longrightarrow_2, O_2, H_2)$  with  $O_1 = O_2$ , a relation

$$R \subseteq Q_1 \times Q_2$$

is a **bisimulation relation** between  $T_1$  and  $T_2$  if for all  $(q_1, q_2) \in R$

- $H_1(q_1) = H_2(q_2)$
- $q_1 \xrightarrow{l_1}_1 p_1$  in  $T_1$  implies existence of  $q_2 \xrightarrow{l_2}_2 p_2$  in  $T_2$  so that  $(p_1, p_2) \in R$
- $q_2 \xrightarrow{l_2}_2 p_2$  in  $T_2$  implies existence of  $q_1 \xrightarrow{l_1}_1 p_1$  in  $T_1$  so that  $(p_1, p_2) \in R$

LTSs  $T_1$  and  $T_2$  are **bisimilar** if  $\pi|_{Q_1}(R) = Q_1$  and  $\pi|_{Q_2}(R) = Q_2$

# Approximate bisimulation equivalence

Given  $T_1 = (Q_1, L_1, \longrightarrow_1, O_1, H_1)$  and  $T_2 = (Q_2, L_2, \longrightarrow_2, O_2, H_2)$  with  $O_1 = O_2$ , and a precision  $\varepsilon > 0$ , a relation

$$R \subseteq Q_1 \times Q_2$$

is an **approximate bisimulation relation** between  $T_1$  and  $T_2$  if for all  $(q_1, q_2) \in R$

- ~~$H_1(q_1) = H_2(q_2)$~~   $d(H_1(q_1), H_2(q_2)) \leq \varepsilon$
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LTSs  $T_1$  and  $T_2$  are **bisimilar with precision  $\varepsilon$**  if  $\pi|_{Q_1}(R) = Q_1$  and  $\pi|_{Q_2}(R) = Q_2$

We here consider a modified version of bisimulation where outputs need not to coincide but to be close, up to a given precision  $\varepsilon$

*... is approximate bisimulation equivalence an equivalence relation?*

# Approximate bisimulation equivalence

Given  $T_1 = (Q_1, L_1, \longrightarrow_1, O_1, H_1)$  and  $T_2 = (Q_2, L_2, \longrightarrow_2, O_2, H_2)$  with  $O_1 = O_2$ , and a precision  $\varepsilon > 0$ , a relation

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LTSs  $T_1$  and  $T_2$  are  **$\varepsilon$ -bisimilar** if  $\pi|_{Q_1}(R) = Q_1$  and  $\pi|_{Q_2}(R) = Q_2$

**Drawback:** in the presence of non-determinism, this notion can be shown not to capture correctly the robustness requirement of control strategies!

# Alternating approximate bisimulation

Given  $T_1 = (Q_1, L_1, \longrightarrow_1, O_1, H_1)$  and  $T_2 = (Q_2, L_2, \longrightarrow_2, O_2, H_2)$  with  $O_1 = O_2$ , and a precision  $\varepsilon > 0$ , a relation

$$R \subseteq Q_1 \times Q_2$$

is an **alternating approximate bisimulation relation** between  $T_1$  and  $T_2$  if for all  $(q_1, q_2) \in R$

- $d(H_1(q_1), H_2(q_2)) \leq \varepsilon$
- for any  $l_1 \in L_1$ , there exists  $l_2 \in L_2$  s.t. the existence of  $q_2 \xrightarrow{l_2}_2 p_2$  in  $T_2$  implies the existence of  $q_1 \xrightarrow{l_1}_1 p_1$  in  $T_1$  with  $(p_1, p_2) \in R$
- for any  $l_2 \in L_2$ , there exists  $l_1 \in L_1$  s.t. the existence of  $q_1 \xrightarrow{l_1}_1 p_1$  in  $T_1$  implies the existence of  $q_2 \xrightarrow{l_2}_2 p_2$  in  $T_2$  with  $(p_1, p_2) \in R$

LTSs  $T_1$  and  $T_2$  are **A&A-bisimilar** if  $\pi|_{Q_1}(R) = Q_1$  and  $\pi|_{Q_2}(R) = Q_2$

*From [Alur et al., 1998] symbolic control strategies designed for  $S_1$  can be appropriately transferred to  $S_2$  if the systems are **A&A-bisimilar***



Consider the following parameters:

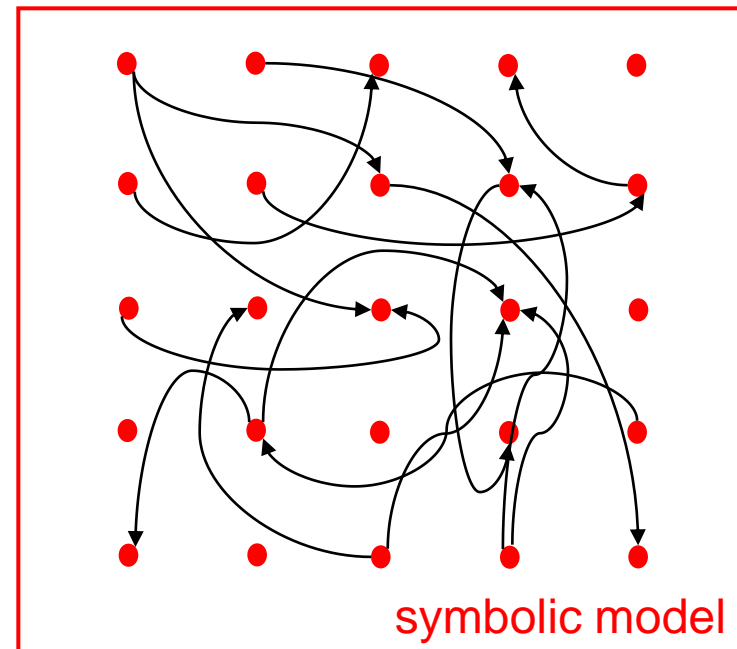
- $\tau > 0$  sampling time
- $\eta > 0$  state space quantization
- $\mu > 0$  input space quantization
- $\theta > 0$  a design parameter

and define  $T_{\tau,\eta,\mu,\theta}(\Sigma) = (Q, L, \xrightarrow{\quad}, O, H)$

where:

- $Q = [R^n]_{\eta} = \eta \mathbb{Z}^n$
- $L = [R^m]_{\mu} = \mu \mathbb{Z}^m$
- $q \xrightarrow{u} p$ , if  $\|x(\tau, q, u) - p\| \leq \beta(\theta, \tau) + \gamma(\mu, \tau) + \eta$
- $O = R^n$
- $H$  is the identity function

How do I construct this symbolic model ?



**Theorem** Consider a nonlinear digital control system  $\Sigma$

$$\dot{x} = f(x, u), \quad x \in \mathbb{R}^n, u \in \mathbb{R}^m$$

If  $\Sigma$  is  $\delta$ -FC then for any desired precision  $\varepsilon > 0$  and for any  $\tau, \eta, \mu, \theta > 0$  satisfying

$$\eta \leq \varepsilon \leq \theta$$

one has:

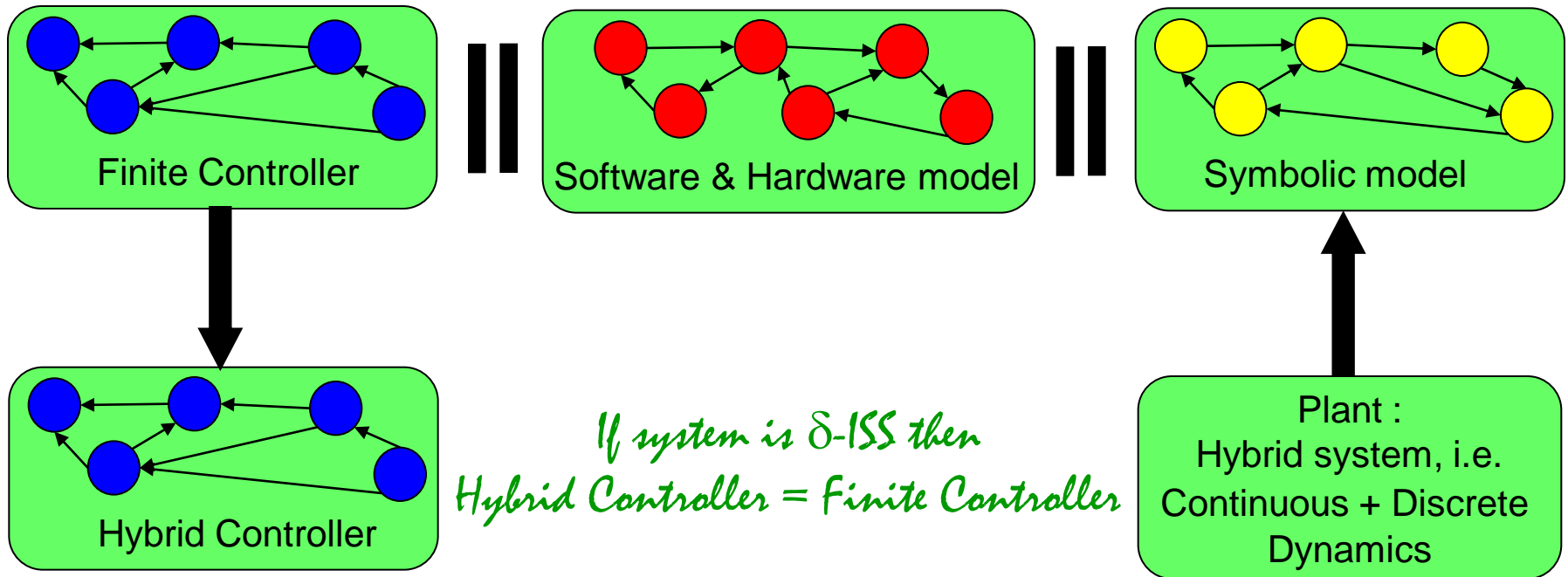
$$T_{\tau, \eta, \mu, \theta}(\Sigma) \preceq_{\varepsilon} T_{\tau}^*(\Sigma) \preceq_{\varepsilon}^{alt} T_{\tau, \eta, \mu, \theta}(\Sigma)$$

## Remark

The previous result implies that any controller synthesized for the finite model  $T_{\tau, \eta, \mu, \theta}(\Sigma)$  can be refined to a controller enforcing the same specification on  $T_{\tau}^*(\Sigma)$ .

Synthesis through a three phase process:

1. Construct the finite model  $T$  of the plant system  $\Sigma$
2. Synthesize a finite controller  $C$  solving the specification  $S$  on  $T$
3. Synthesize a controller  $C'$  for  $\Sigma$  on the basis of  $C$



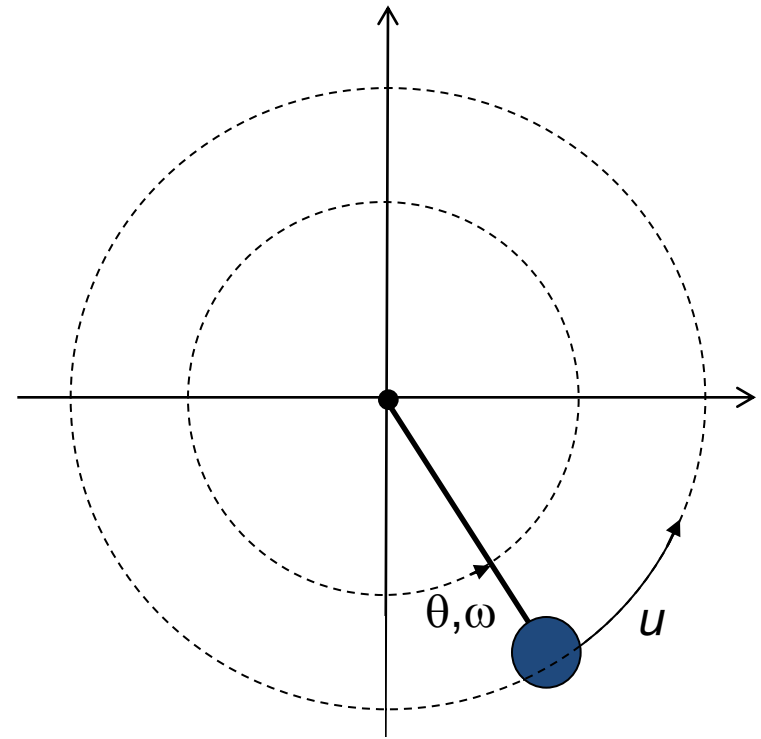
# Example 1: pendulum

One of the simplest mechanical systems studied in the literature is the pendulum:

$$\Sigma: \begin{cases} \dot{\theta} = \omega \\ \dot{\omega} = -\frac{g}{l} \sin \theta - \frac{k}{m} \omega + \frac{1}{ml^2} u \end{cases}$$

We suppose:

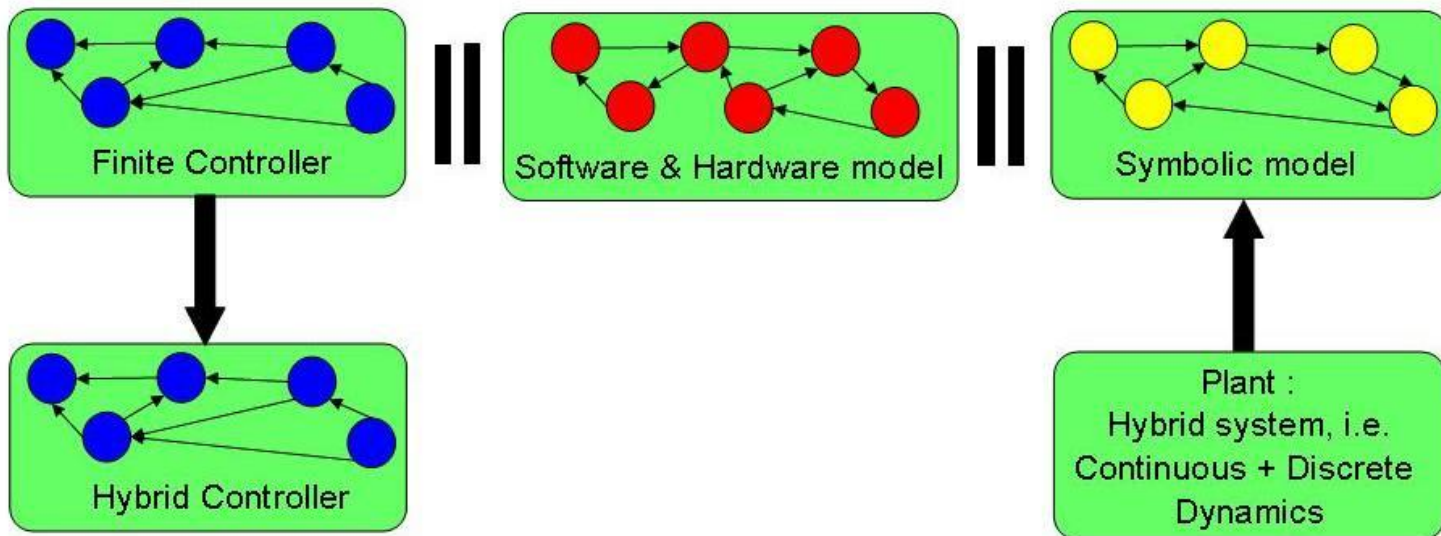
- $g = 9.8, l = 5, m = 0.5, k = 3$
- $u \in U = [-1.5, 1.5]$
- Control signals are piecewise constant
- $(\theta, \omega) \in X = [-1, 1] \times [-1, 1]$



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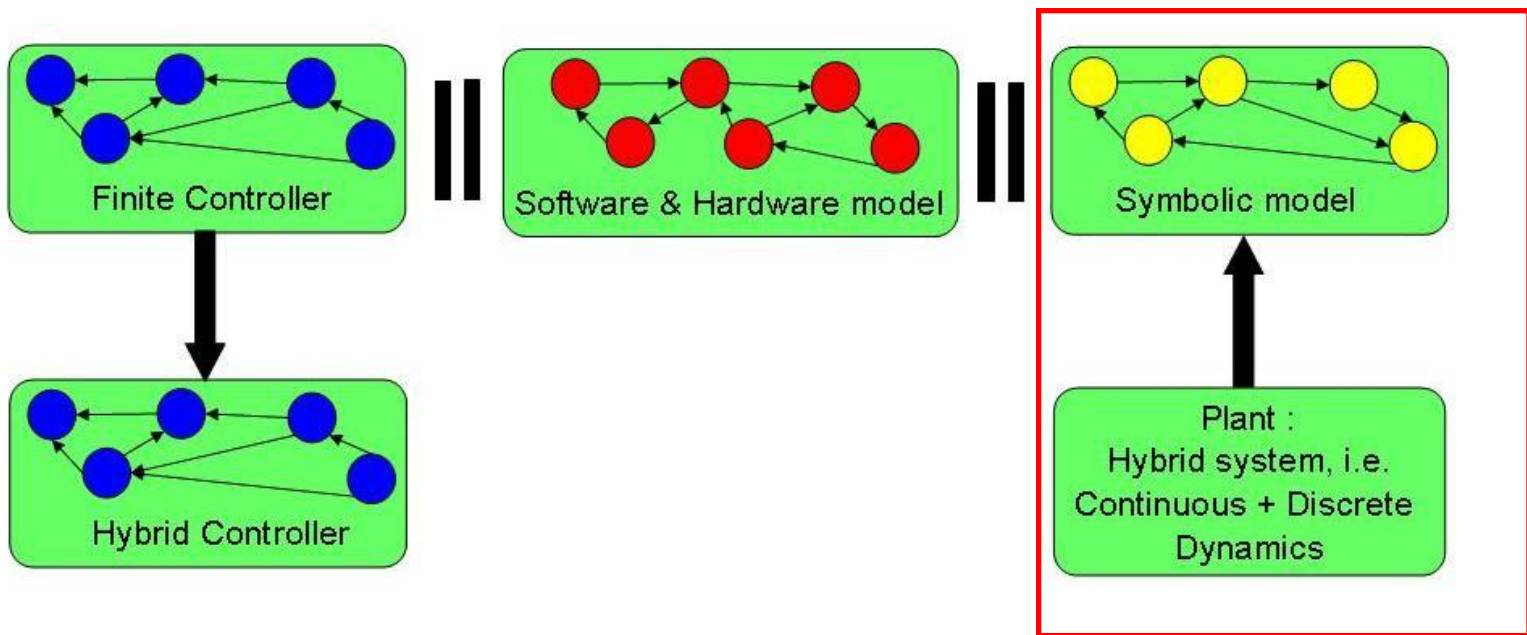
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# Example 1: pendulum

Let us construct a symbolic model for the pendulum control system:

- Check  $\delta$ -ISS property (directly or by Lyapunov functions)
- Find functions  $\beta$  and  $\gamma$  functions which satisfy

A control system  $\dot{x} = f(x, u)$  is

**incrementally input-to-state stable** ( $\delta$ -ISS)

if there exist a KL function  $\beta$  and a  $K_\infty$  function  $\gamma$  so that for any  $t \geq 0$ ,  $y, z \in \mathbb{R}^n$  and  $u, v$

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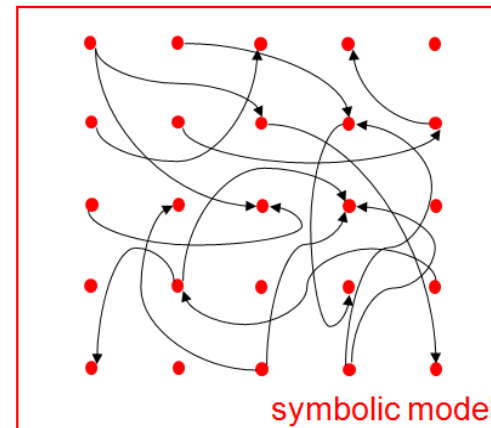
- Choose parameters  $\varepsilon, \tau, \eta, \mu$
- Construct the symbolic model

define  $T_{\tau, \eta, \mu}(\Sigma) = (Q, L, \longrightarrow, O, H)$

where:

- $Q = [R^n]_\eta = \eta Z^n$
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- $q \xrightarrow{u} p$ , if  $x(\tau, q, u) \in \mathcal{B}_{[\eta/2]}(p)$
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How do I construct this symbolic model ?



# Example 1: pendulum

Let us construct a symbolic model for the pendulum control system:

- Check  $\delta$ -ISS property (directly or by Lyapunov functions)

$$V(x, y) = \frac{1}{2}(x - y)' \begin{bmatrix} \frac{1}{2} \left( \frac{k}{m} \right)^2 & \frac{1}{2} \frac{k}{m} \\ \frac{1}{2} \frac{k}{m} & 1 \end{bmatrix} (x - y).$$

It is possible to show that  $V$  satisfies condition (i) of Definition 2.4 with  $\alpha_1(r) = 0.49 r^2$  and  $\alpha_2(r) = 18.51 r^2$ . Moreover, by defining for any  $z_1, z_2 \in \mathbb{R}$ ,

$$\zeta(z_1, z_2) = (\sin(z_1) - \sin(z_2)) / (z_1 - z_2),$$

one obtains  $\zeta_{\min} = \min_{z_1, z_2 \in [-1, 1]} \zeta(z_1, z_2) = 0.84$  and  $\zeta_{\max} = \max_{z_1, z_2 \in [-1, 1]} \zeta(z_1, z_2) = 1$  and hence:

$$\begin{aligned} \frac{\partial V}{\partial x} f(x, u) + \frac{\partial V}{\partial y} f(y, v) &= -\frac{1}{2} \frac{k}{m} \frac{g}{l} \zeta(x_1, y_1) (x_1 - y_1)^2 \\ &\quad - \frac{g}{l} \zeta(x_1, y_1) (x_1 - y_1) (x_2 - y_2) - \frac{1}{2} \frac{k}{m} (x_2 - y_2)^2 \\ &\quad + \left( \frac{1}{2} \frac{k}{m} (x_1 - y_1) + x_2 - y_2 \right) (u - v) \\ &\leq -\frac{1}{2} a \|x - y\|_2^2 + b |u - v|, \end{aligned} \tag{23}$$

*... pendulum is  $\delta$ -ISS !*



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Let us construct a symbolic model for the pendulum control system:

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$$\|x(t, y, u) - x(t, z, v)\| \leq \beta(\|y - z\|, t) + \gamma(\|u - v\|_\infty)$$

Using inequality (23), the definition of  $V$  and the comparison lemma (Khalil, 1996), it is possible to show that for any  $x, y \in X$ , any  $u, v \in \mathcal{U}$  and any time  $t \in \mathbb{R}_0^+$ :

$$\|x(t, x, u) - x(t, y, v)\| \leq \beta(\|x - y\|, t) + \gamma(\|u - v\|_\infty),$$

where  $\beta(r, s) := 6.17 e^{-2.08s} r$  and  $\gamma(r) := \sqrt{3.96 r}$  for any  $r, s \in \mathbb{R}$ .

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- Choose parameters  $\varepsilon, \tau, \eta, \mu$

$$6.17e^{-2.08\tau} \varepsilon + \sqrt{3.96\mu} + \eta/2 \leq \varepsilon$$

a possible choice is  $\varepsilon = 0.25$ ,  $\tau = 2$ ,  $\eta = 0.4$ ,  $\mu = 1.5 \cdot 10^{-4}$

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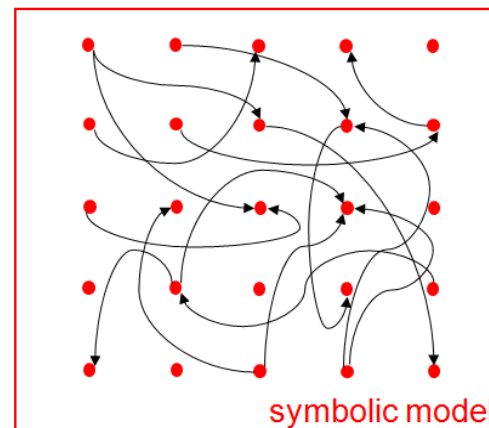
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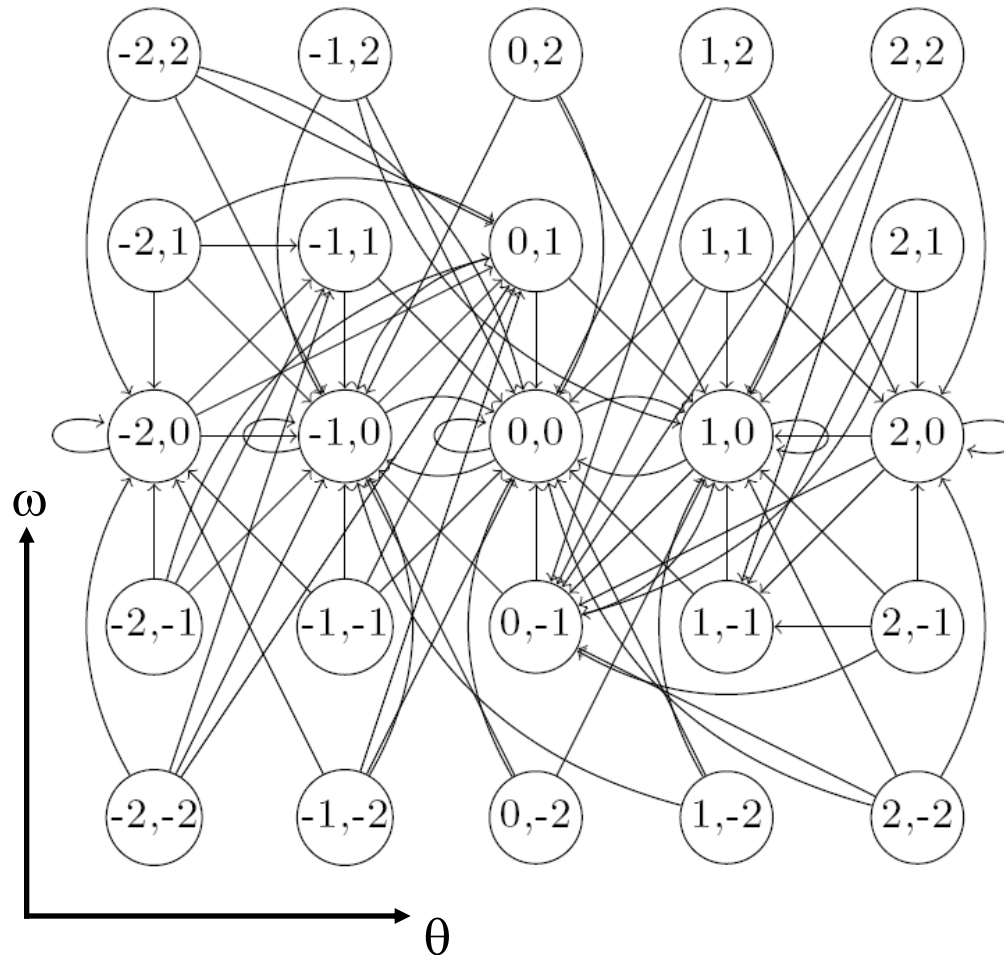
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How do I construct this symbolic model ?



# Example 1: pendulum

The obtained symbolic model is:

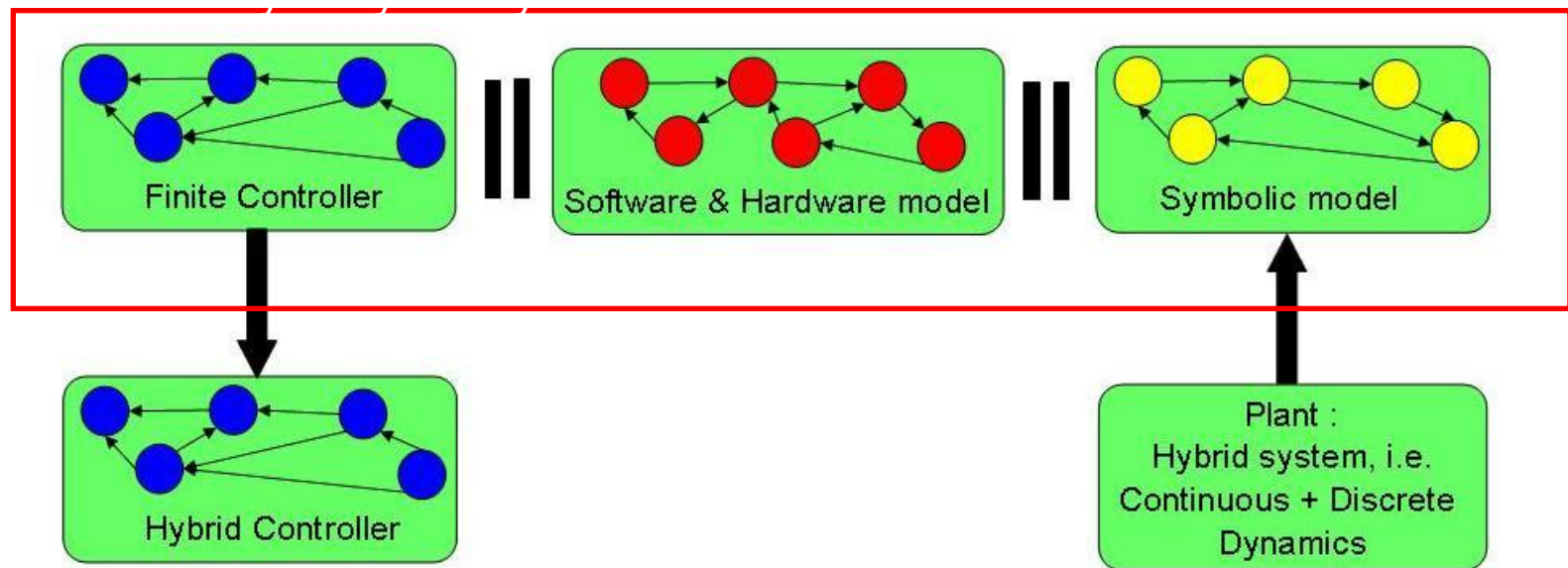


where discrete states  $(\eta_i, \eta_j)$  have been labelled by  $(i, j)$

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# Example 1: pendulum

Let us use the symbolic model for controller synthesis:

Consider a specification given by the concatenation of tasks  $P_1$  and  $P_2$  as follows:

$$P_1, P_1, P_2, P_1, P_1$$

where:

- $P_1$  requires a periodic orbit from  $(\theta, \omega) = (-1, 0)$  to  $(0, 0)$
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This type of specification is a paradigm for illustrating more complex controller synthesis problems where a task is given by the coordination of smaller tasks.

Any digital controller enforcing  $P_1, P_1, P_2, P_1, P_1$  in system  $\Sigma$  will result in a hybrid system combining the discrete memory, needed to describe the current task, with the continuous control law enforcing the specification in each task.



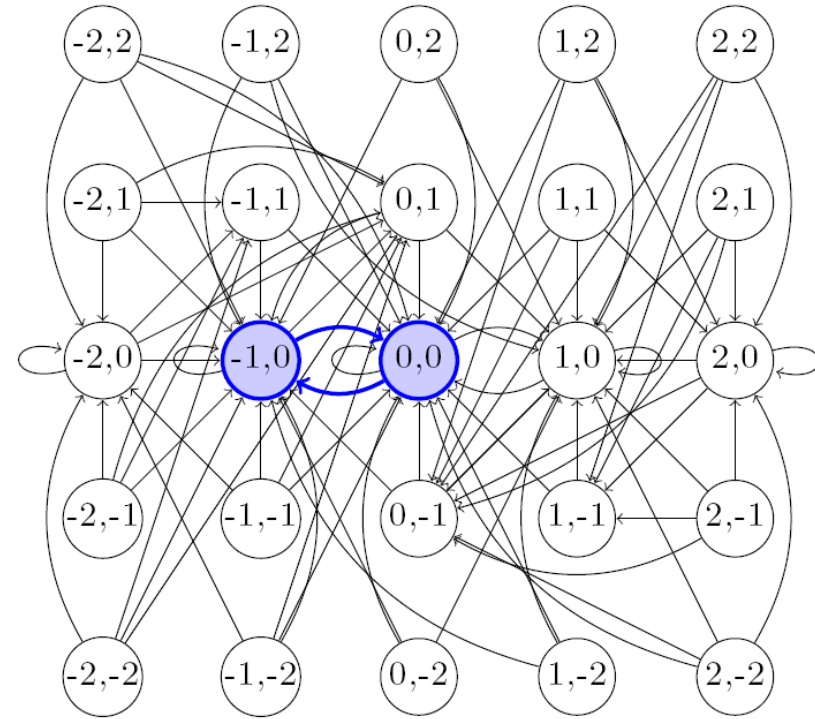
# Example 1: pendulum

By exploiting the symbolic model:

One obtains:

- Specification  $P_1$ :

$$(-1,0) \xrightarrow{1.38} (0,0) \xrightarrow{-1.5} (-1,0)$$



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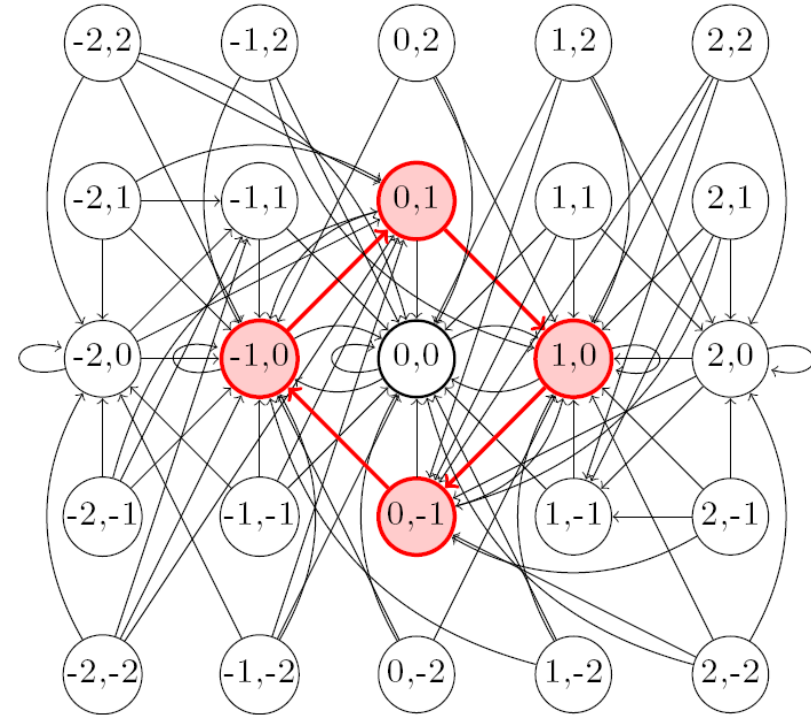
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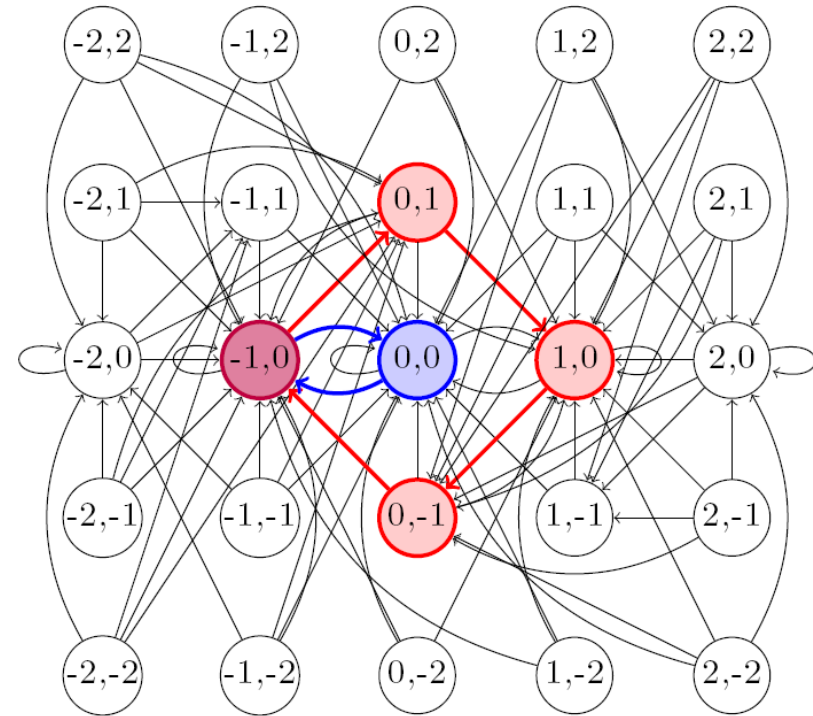
$$(-1,0) \xrightarrow{1.38} (0,0) \xrightarrow{-1.5} (-1,0)$$

- Specification  $P_2$ :

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- Overall specification:

$$\begin{aligned} &(-1,0) \xrightarrow{1.38} (0,0) \xrightarrow{-1.5} (-1,0) \xrightarrow{1.38} (0,0) \xrightarrow{-1.5} (-1,0) \xrightarrow{1.5} (0,1) \xrightarrow{1.5} (1,0) \xrightarrow{-1.5} (0,-1) \\ &\xrightarrow{-0.71} (-1,0) \xrightarrow{1.38} (0,0) \xrightarrow{-1.5} (-1,0) \xrightarrow{1.38} (0,0) \xrightarrow{-1.5} (-1,0) \end{aligned}$$



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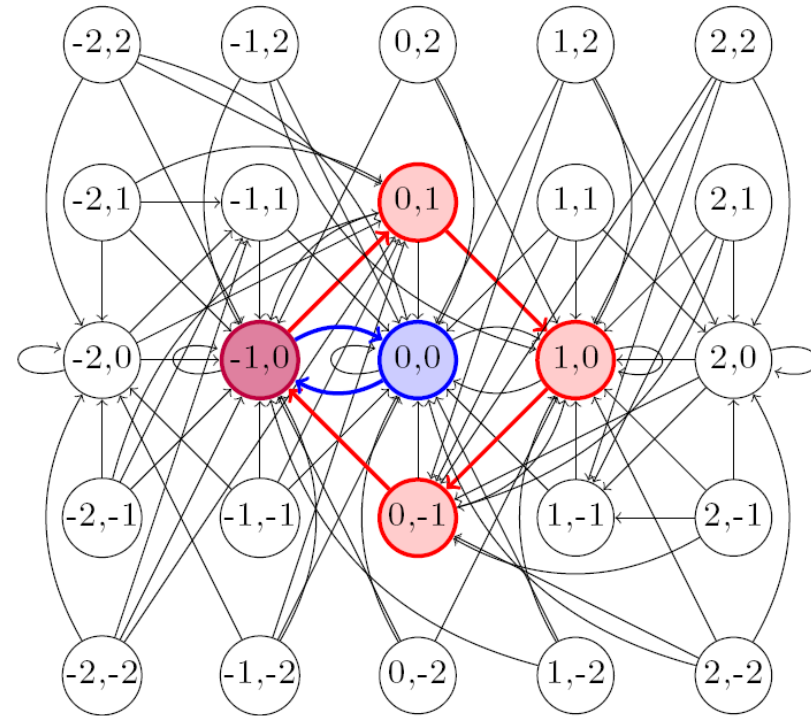
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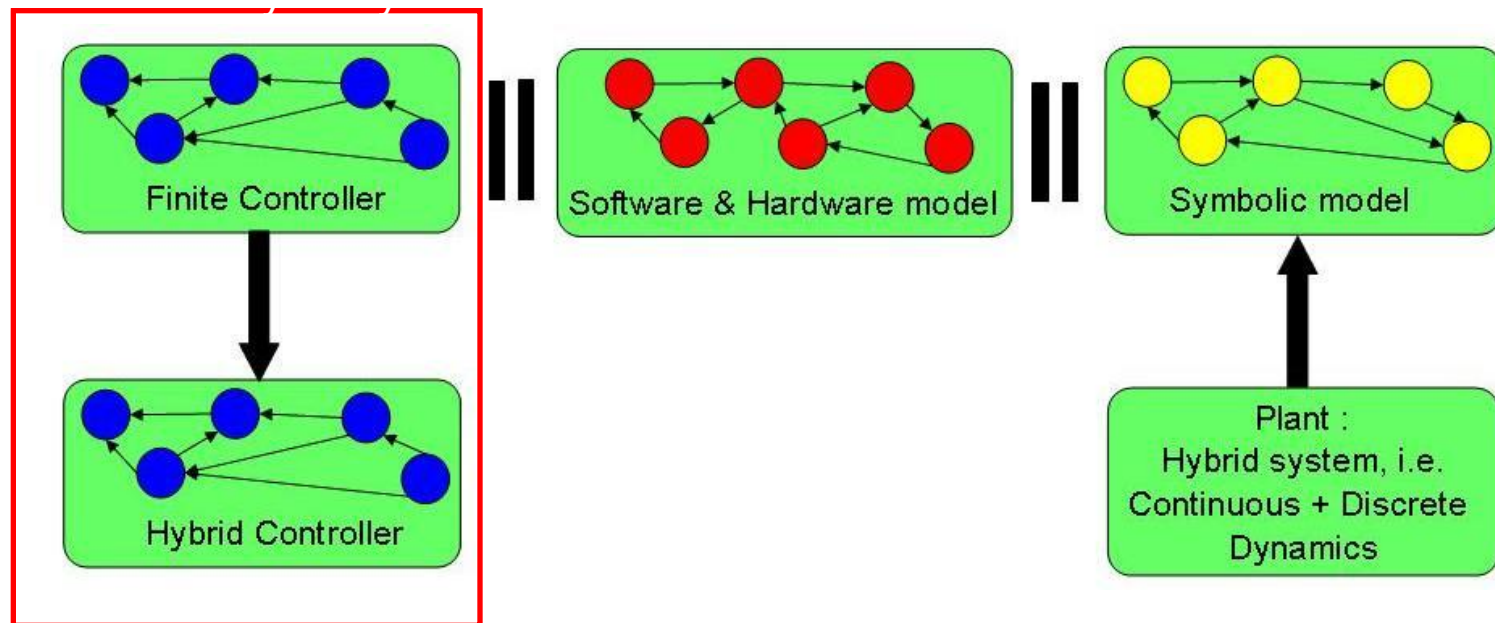


What does it happen if I apply this control strategy to the original control system?

# Example 1: pendulum

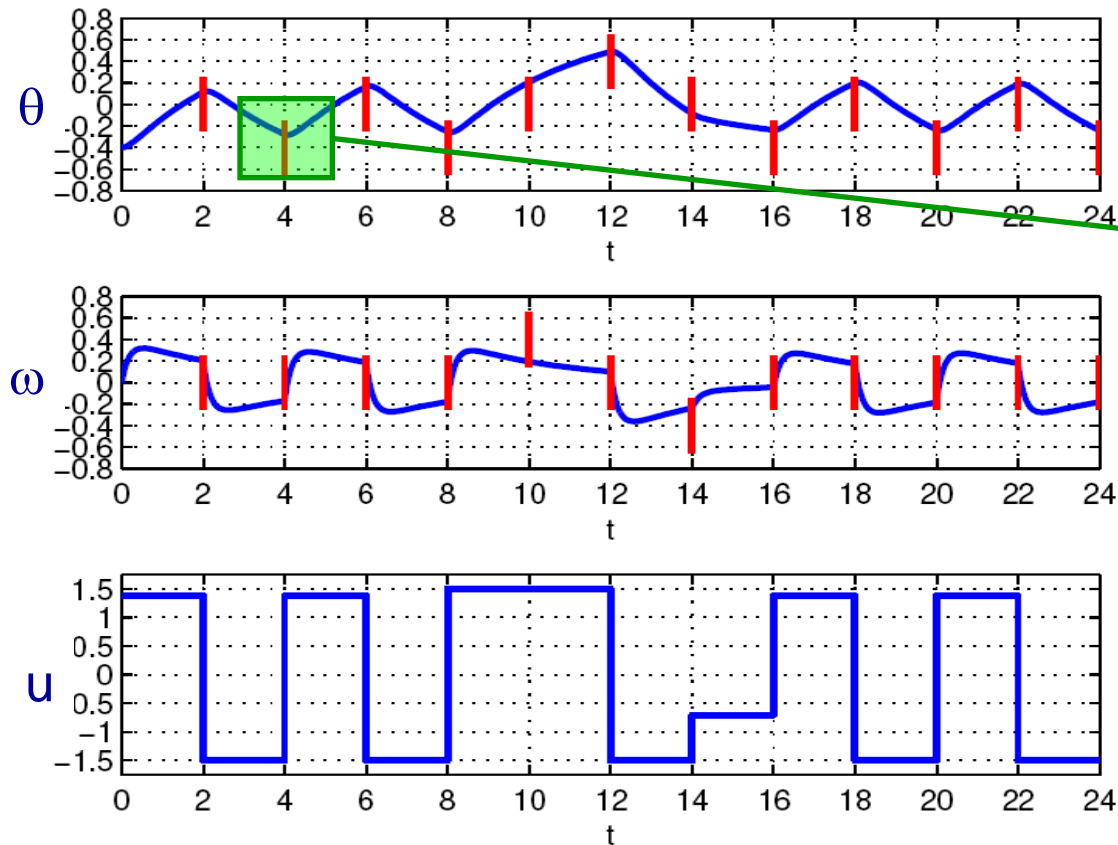
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# Example 1: pendulum

Let us apply the control strategy synthesized on the symbolic model  $T_{\tau,\eta,\mu}(\Sigma)$  to the continuous system  $\Sigma$ :



$$= -\eta + \varepsilon = -0.15$$

$$= -\eta = -0.40$$

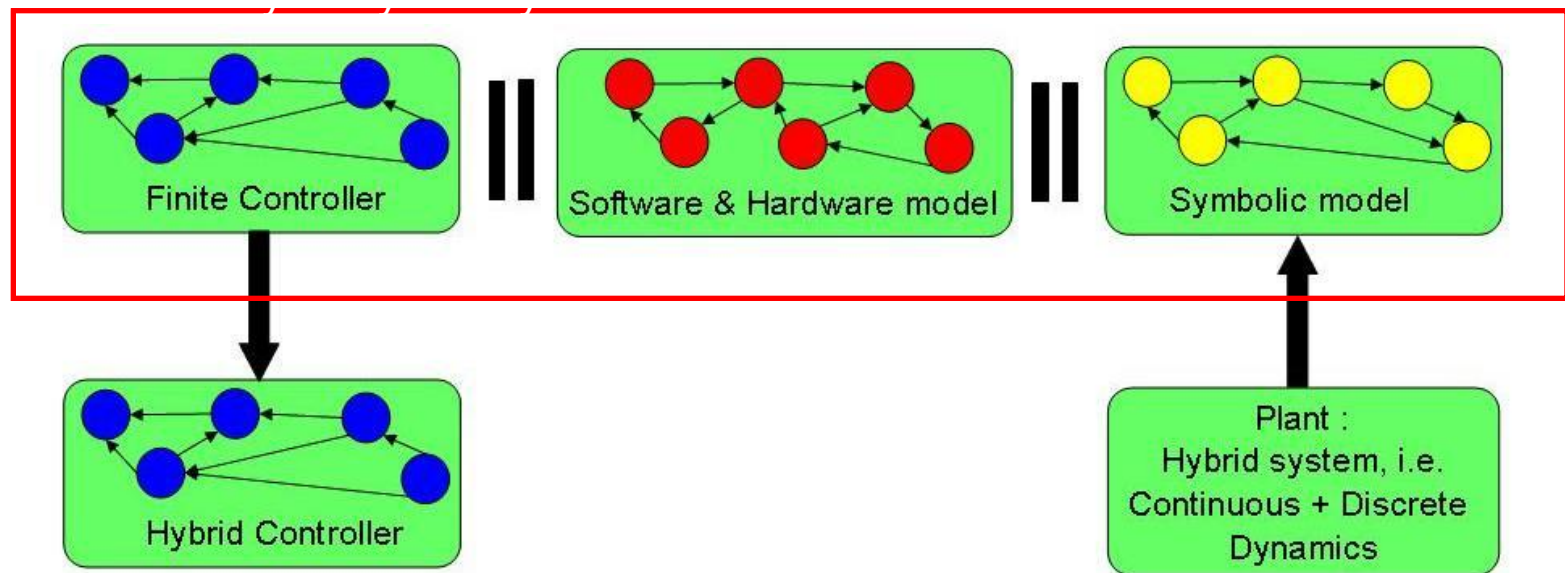
$$= -\eta - \varepsilon = -0.65$$

... as it is  
required  
by specification  $P_1$   
and precision  $\varepsilon$  !

# Example 1: pendulum

One of the simplest mechanical systems studied in the literature is the pendulum:

$$\Sigma: \begin{cases} \dot{\theta} = \omega \\ \dot{\omega} = -\frac{g}{l} \sin \theta - \frac{k}{m} \omega + \frac{1}{ml^2} u \end{cases}$$



*... traditional controller synthesis design?*



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Disadvantages:

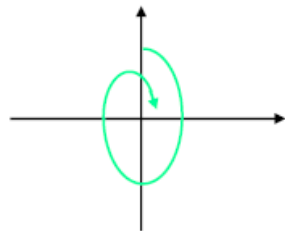
- Feedback linearization is not robust

*... traditional controller synthesis design?*

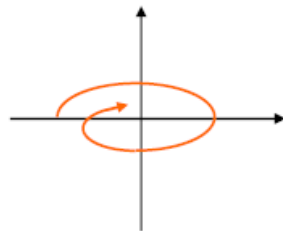
# Example 1: pendulum

## SWITCHING BETWEEN ASYMPTOTICALLY STABLE SYSTEMS

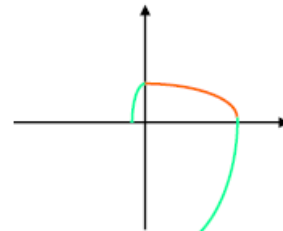
$$\dot{x} = A_{\sigma}x$$



$$\dot{x} = A_1x$$



$$\dot{x} = A_2x$$



unstable!

Disadvantages:

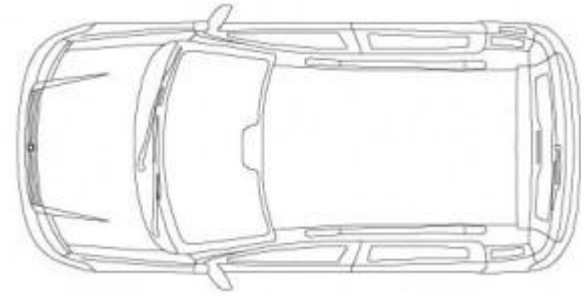
- Feedback linearization is not robust
- Dealing with stability of hybrid systems ...

*... traditional controller synthesis design?*

## Example 2: vehicle

A widely used model in the literature is the non-holonomic vehicle model:

$$\Sigma : \begin{cases} \dot{x} = v_0 \frac{\cos(\alpha + \theta)}{\cos(\alpha)} \\ \dot{y} = v_0 \frac{\sin(\alpha + \theta)}{\cos(\alpha)} \\ \dot{\theta} = \frac{v_0}{b} \tan(\delta) \end{cases}$$



with  $\alpha = \arctan(a \cdot \tan(\delta)/b)$ .

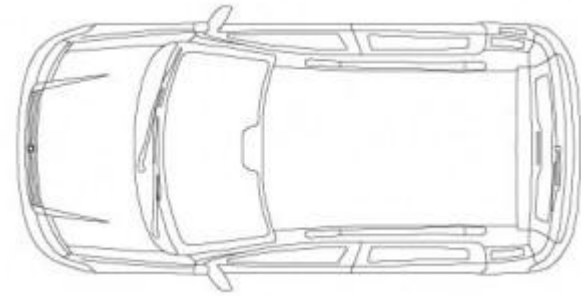
We suppose:

- $a = 0.5, b = 1$ .
- $u = (v_0, \delta) \in U = [-1, 1] \times [-1, 1]$
- Control signals are piecewise constant
- $(\theta, \omega) \in X = [-1, 1] \times [-1, 1]$

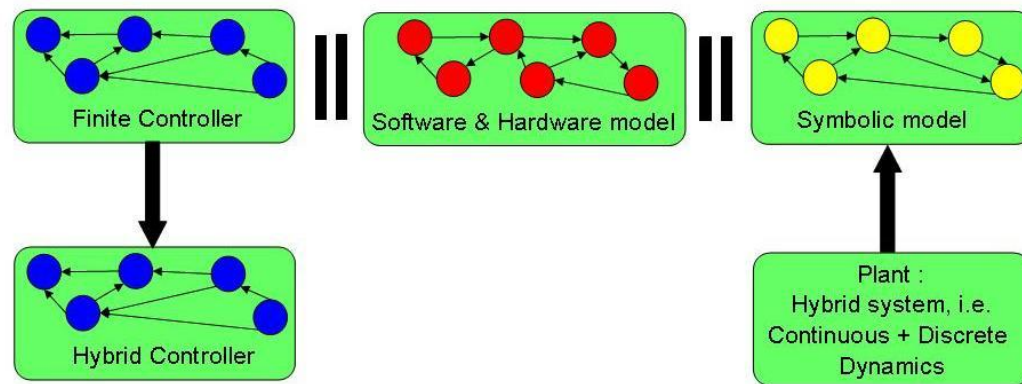
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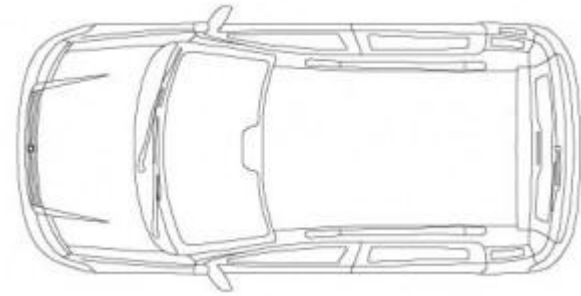
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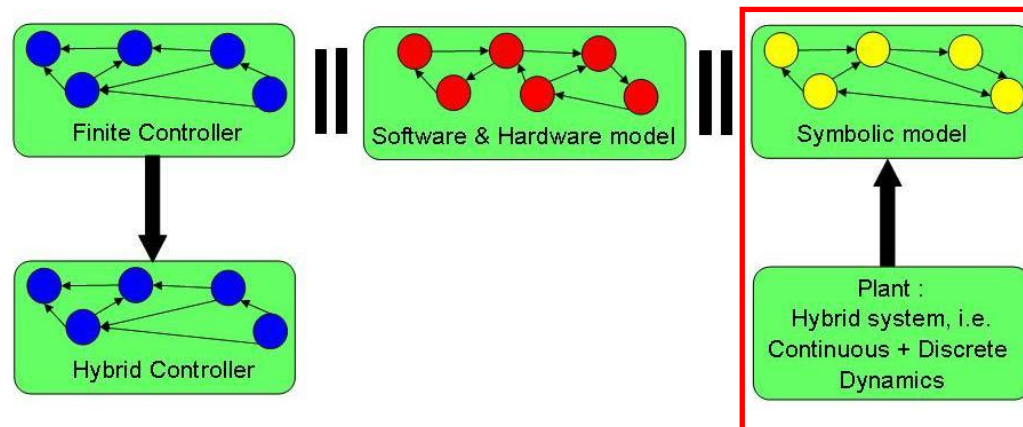
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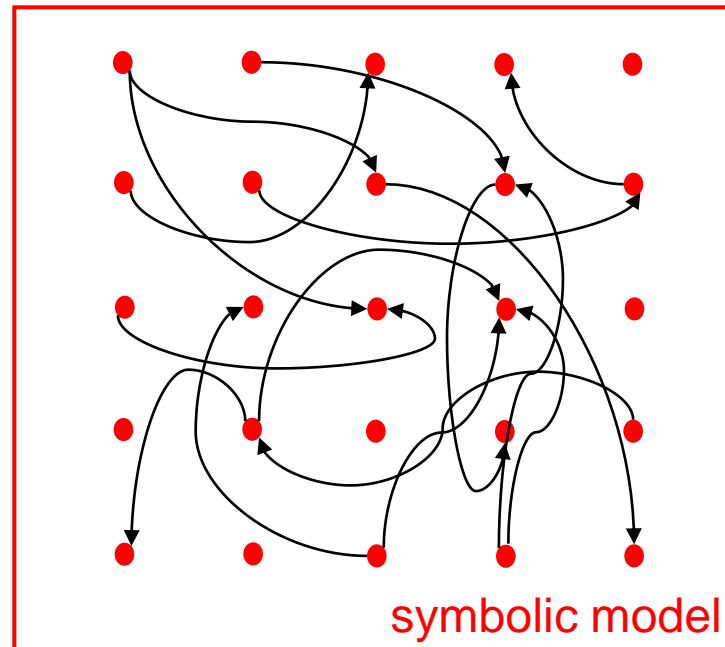
with  $\alpha = \arctan(a \cdot \tan(\delta)/b)$ .



## Example 2: vehicle

Let us construct a symbolic model for the vehicle control system:

- We check  $\delta$ -FC property (the vehicle is not  $\delta$ -ISS !)
- Find functions  $\beta$  and  $\gamma$  functions which satisfy the definition of  $\delta$ -FC.
- Choose parameters  $\varepsilon, \tau, \eta, \mu, \theta$ .
- Construct the symbolic model





## Example 2: vehicle

Let us construct a symbolic model for the vehicle control system:

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- Construct the symbolic model

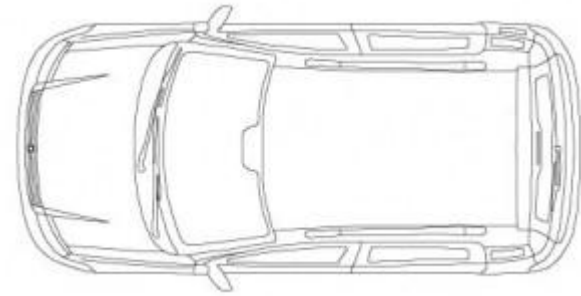
The system is  $\delta$ -FC with  $\beta(r, t) = 1 + 1.267t$

A possible choice is  $\varepsilon = 0.2, \tau = 0.3, \eta = 0.2, \mu = 0.3, \theta = 0.2$

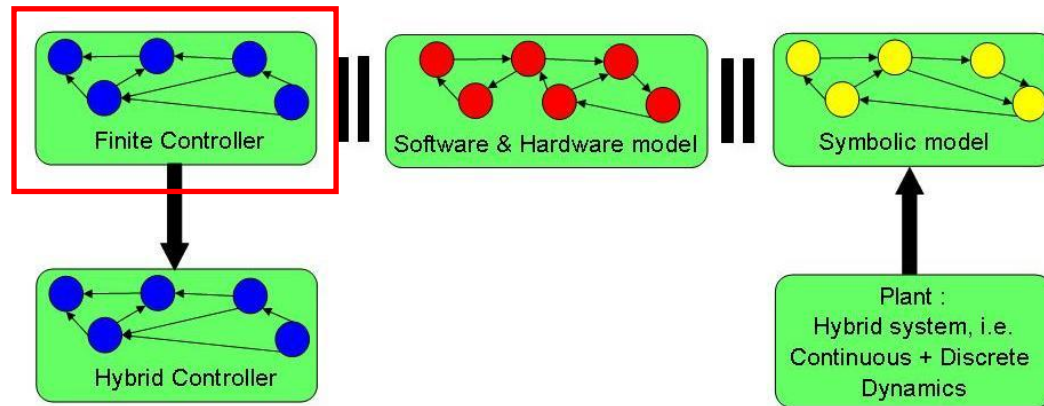
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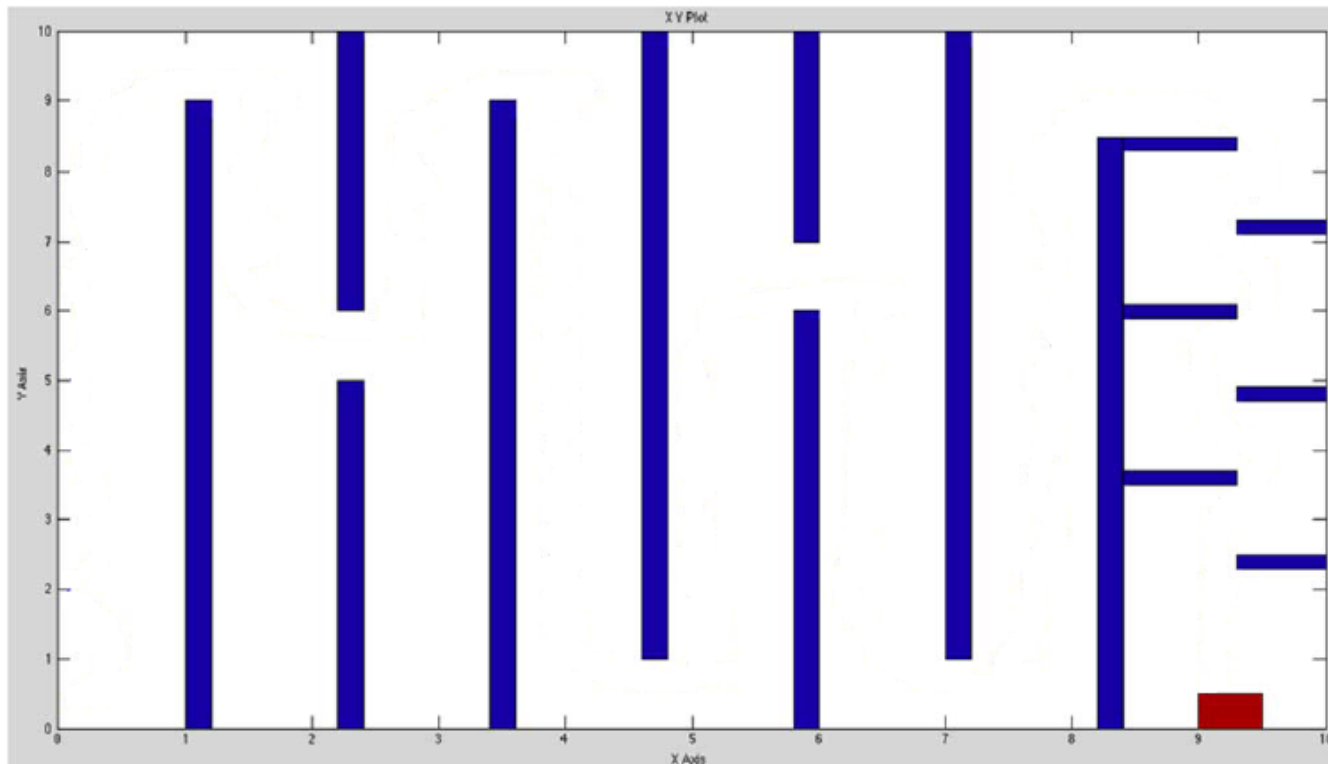
with  $\alpha = \arctan(a \cdot \tan(\delta)/b)$ .



## Example 2: vehicle

Let us use the symbolic model for controller synthesis:

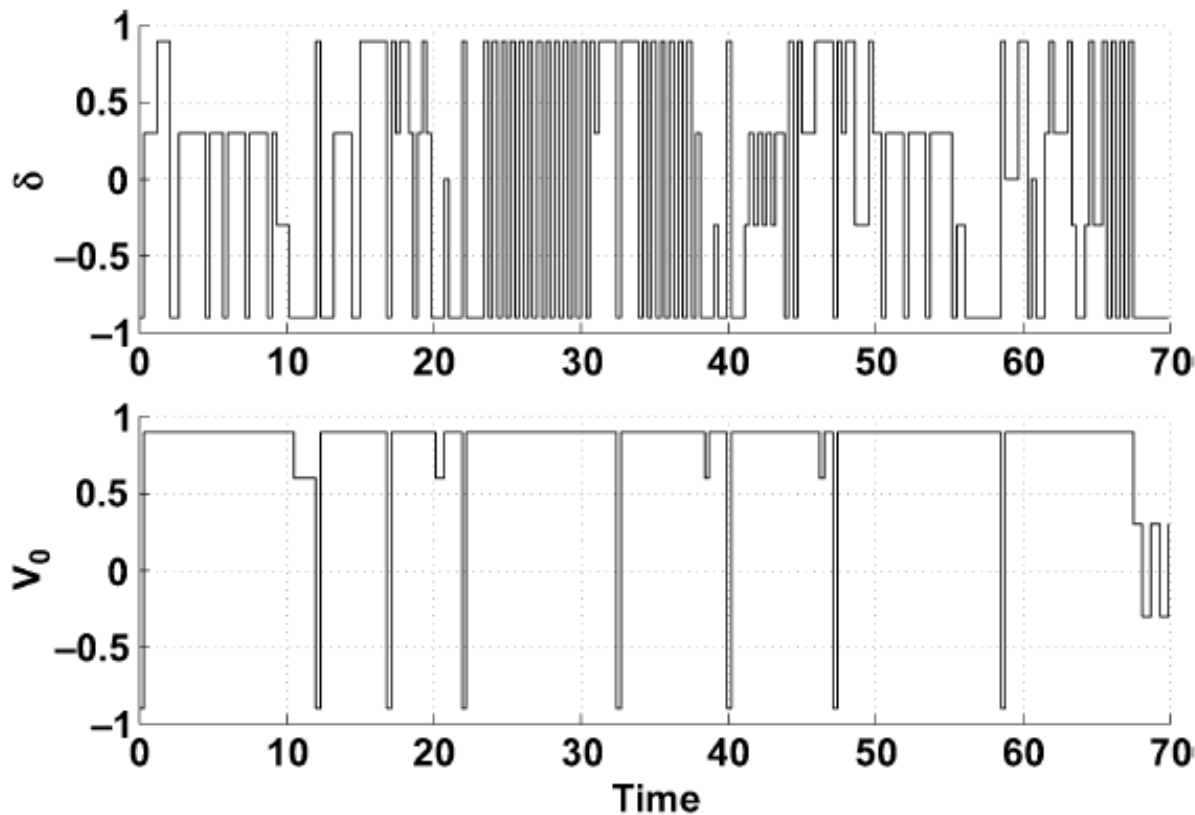
Our objective is to design a controller navigating the vehicle to reach the **target set** (indicated with a red box), while avoiding the **obstacles**, indicated as blue boxes.



## Example 2: vehicle

Let us use the symbolic model for controller synthesis:

A controller enforcing the specification was found by exploiting the symbolic model and by using standard algorithms.



## Example 2: vehicle

Let us use the symbolic model for controller synthesis:

A controller enforcing the specification was found by exploiting the symbolic model and by using standard algorithms. **The specification is fulfilled!**

