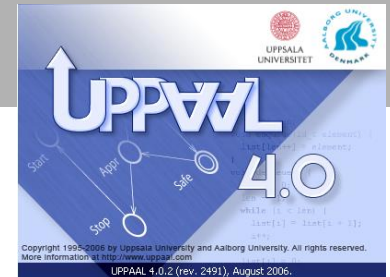


# Symbolic and Statistical Model Checking in UPPAAL

Alexandre David  
Kim G. Larsen

Marius Mikucionis, Peter Bulychev,  
Axel Legay, Dehui Du, Guangyuan Li,  
Danny B. Poulsen, Amélie Stainer,  
Zheng Wang



CAV11, FORMATS11, PDMC11,  
QAPL12, LPAR12, NFM12, iWIGP12,  
RV12, FORMATS12, HBS12, ISOLA12,  
SCIENCE China



# Overview

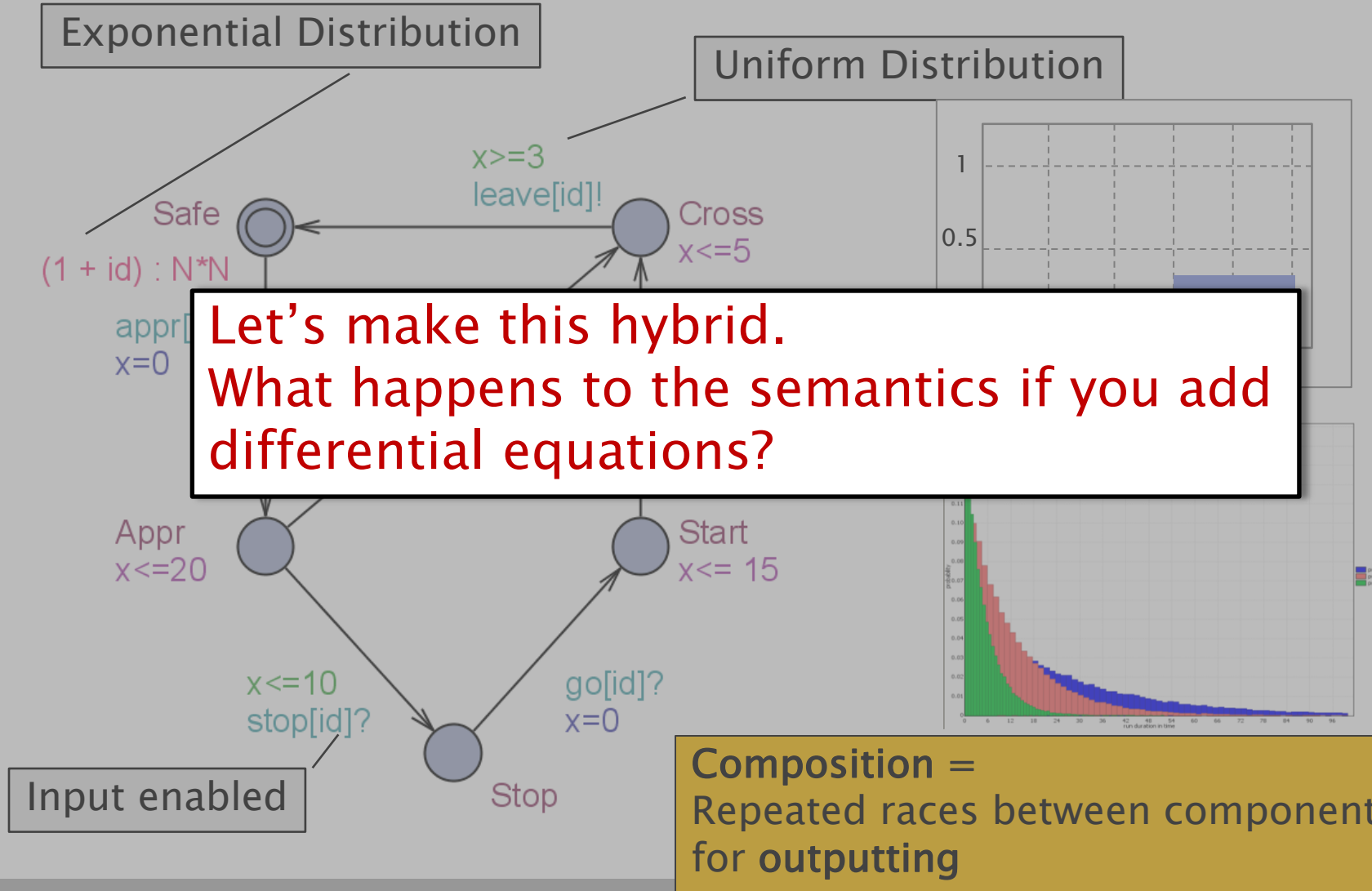
- Stochastic **Hybrid** Automata
- Biological Oscillator
  - Continuous vs. Stochastic Models
- Parameter Optimization – ANOVA
  - Energy Aware Building
- Controller Synthesis for Hybrid Systems



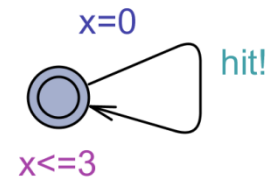
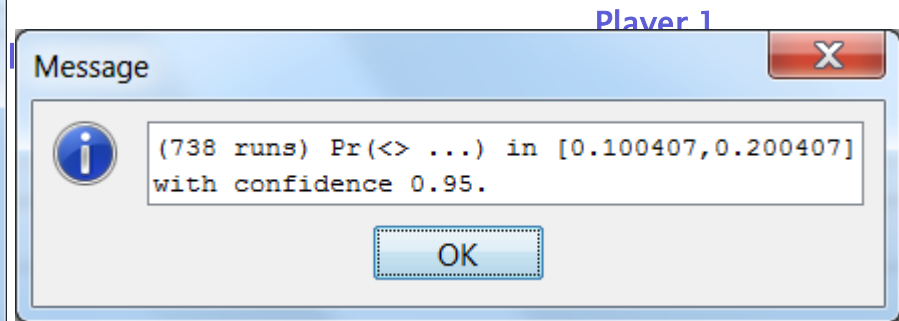
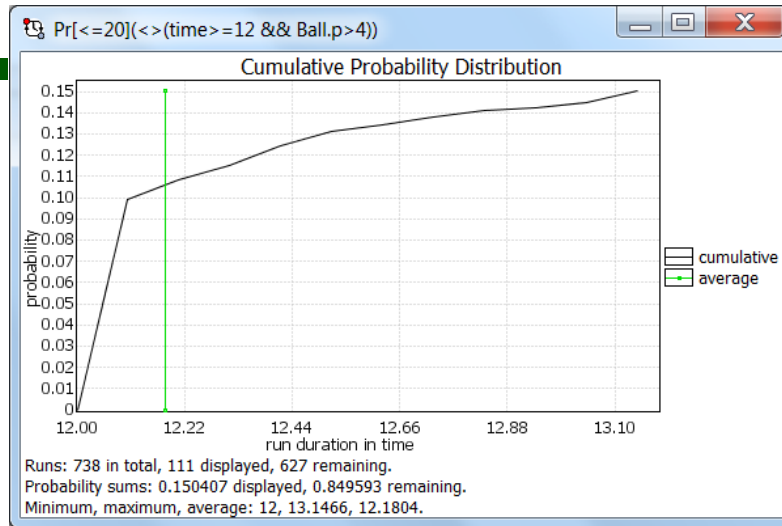
# Stochastic Hybrid Automata



# Stochastic Semantics of TA

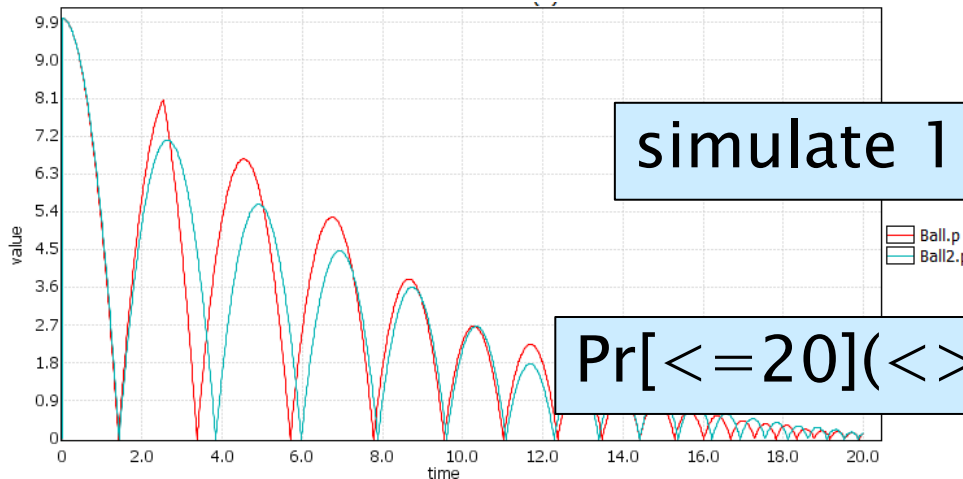


# Stochastic Hybrid Systems



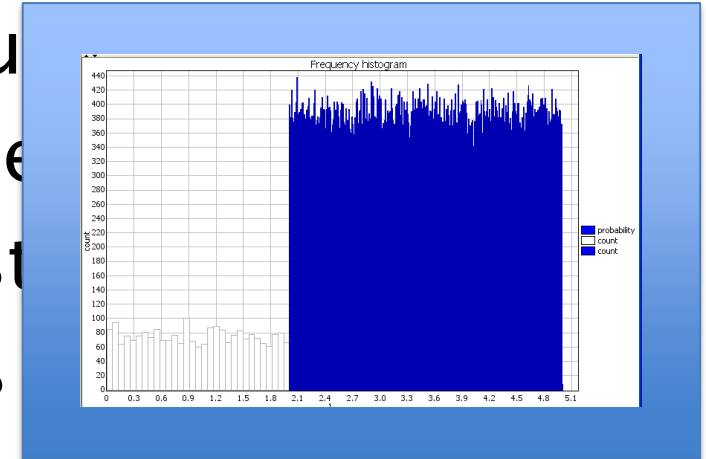
simulate 1 [ $\leq 20$ ]{Ball1.p, Ball2.p}

Pr[ $\leq 20$ ]( $\langle \rangle$ (time  $\geq 12$  && Ball.p  $> 4$ ))



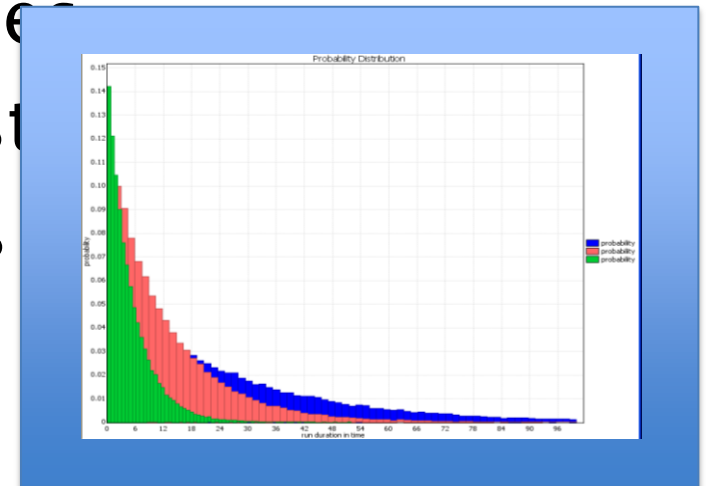
# UPPAAL SMC

- Uniform distributions (bounded delay)
- Exponential distributions (unbounded delay)
- Discrete probabilistic choices
- Distribution on successor states
- Hybrid flow by use of ODEs
- + usual UPPAAL features
- Logic: MITL support.



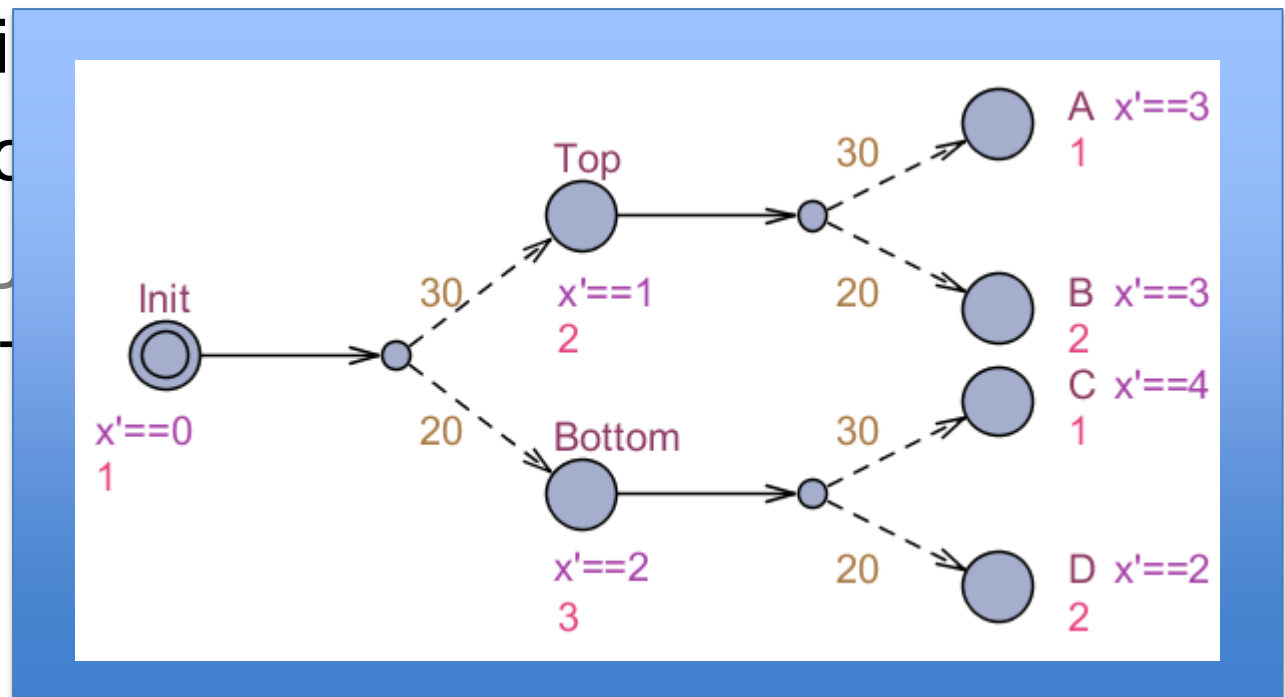
# UPPAAL SMC

- Uniform distributions (bounded delay)
- **Exponential distributions (unbounded delay)**
- Discrete probabilistic choices
- Distribution on successor states
- Hybrid flow by use of ODEs
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# UPPAAL SMC

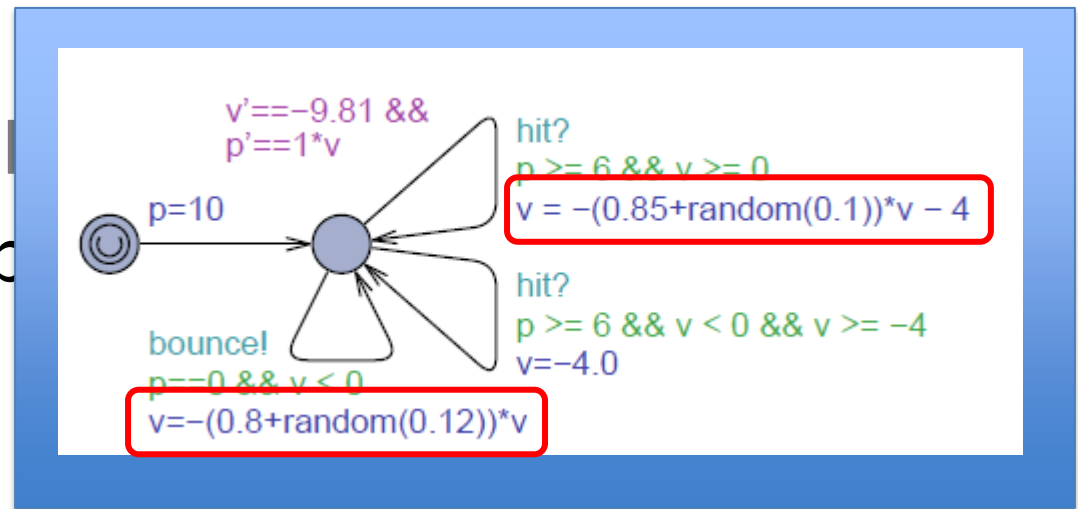
- Uniform distributions (bounded delay)
- Exponential distributions (unbounded delay)
- **Discrete probabilistic choices**
- Distribution
- Hybrid flow
- + usual UPPAAL
- Logic: MITL





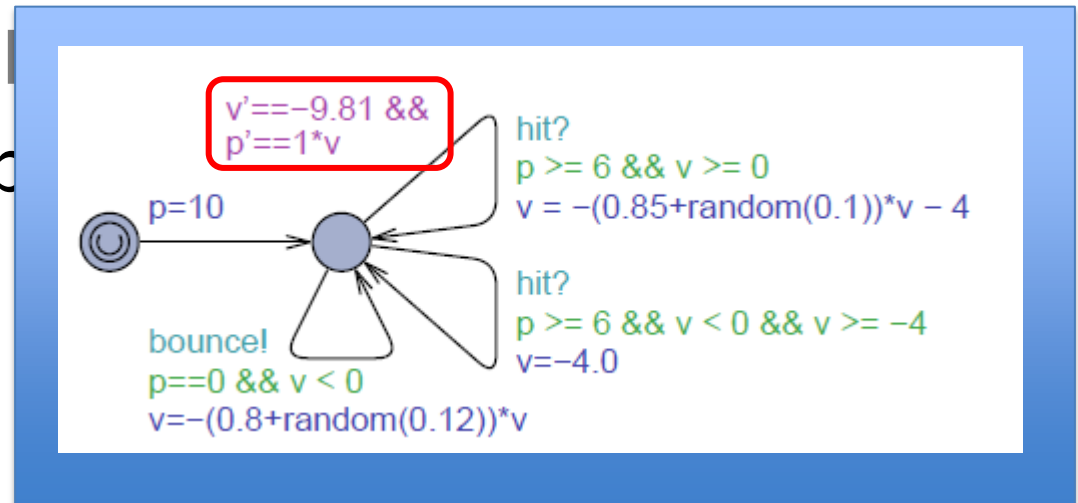
# UPPAAL SMC

- Uniform distributions (bounded delay)
- Exponential distributions (unbounded delay)
- Discrete probabilistic choices
- **Distribution on successor state – random**
- Hybrid flow by
- + usual UPPAAL
- Logic: MITL sup



# UPPAAL SMC

- Uniform distributions (bounded delay)
- Exponential distributions (unbounded delay)
- Discrete probabilistic choices
- Distribution on successor state – **random**
- **Hybrid flow by use of ODEs**
- + usual UPPAAL
- Logic: MITL sup



# UPPAAL SMC

- Uniform distributions (bounded delay)
- Exponential distributions (unbounded delay)
- Discrete probabilistic choices
- Distribution on successor state – **random**
- Hybrid flow by use of ODEs
- + usual UPPAAL features
- **Logic: MITL support.**

$$\phi = \phi \vee \phi \mid \phi \wedge \phi \mid \neg \phi \mid \top \mid \perp \mid \phi U_{[a;b]} \phi \mid \phi R_{[a;b]} \phi \mid X\phi \mid \alpha$$

$$\Box_{\leq 1000}(\phi_{peakN} \implies \Diamond_{\leq p} \phi_{peakN})$$



# Hybrid Automata

$H = (L, l_0, \Sigma, X, E, F, \text{Inv})$

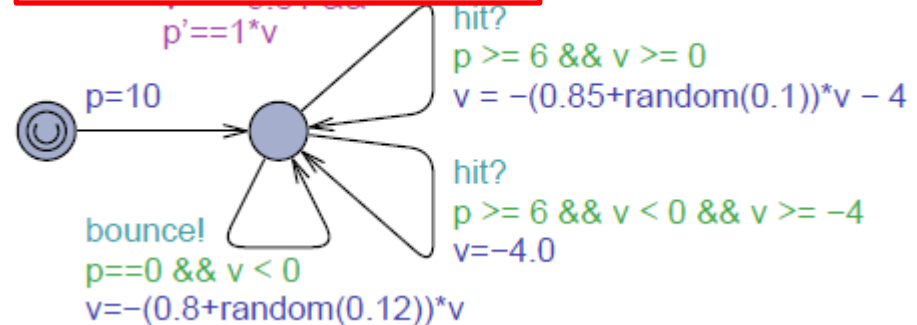
where

- $L$  set of locations
- $l_0$  initial location
- $\Sigma = \Sigma_i \cup \Sigma_o$  set of **actions**
- $X$  set of **continuous variables**

valuation  $\nu: X \rightarrow \mathbb{R}$   
( $= \mathbb{R}^X$ )

- $E$  set of **edges**  $(l, g, a, \phi, l')$   
with  $g \subseteq \mathbb{R}^X$  and  
 $\phi \subseteq \mathbb{R}^X \times \mathbb{R}^X$  and  $a \in \Sigma$
- For each  $l$  a **delay function**  
 $F(l): \mathbb{R}_{>0} \times \mathbb{R}^X \rightarrow \mathbb{R}^X$
- For each  $l$  an **invariant**  
 $\text{Inv}(l) \subseteq \mathbb{R}^X$

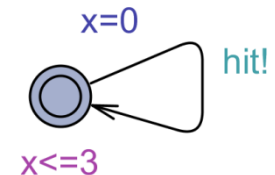
I/O - broadcast sync  
 $\Rightarrow$  input-enabled



Player 1



Player 2



# Hybrid Automata

$H=(L, l_0, \Sigma, X, E, F, \text{Inv})$

where

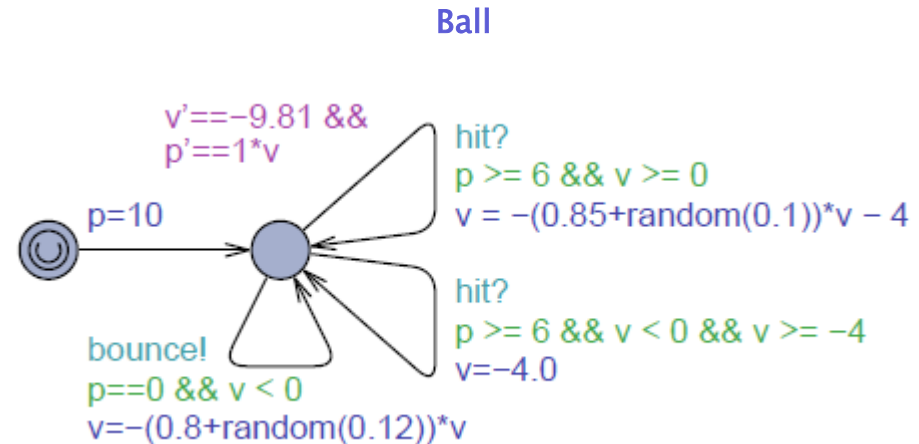
- $L$  set of locations
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- $X$  set of **continuous variables**

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- For each  $l$  a **delay function**  
 $F(l): \mathbb{R}_{>0} \times \mathbb{R}^X \rightarrow \mathbb{R}^X$

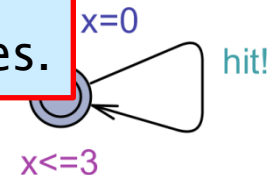
- For each  $l$  an **invariant**  
 $\text{Inv}(l) \subseteq \mathbb{R}^X$



Player 1

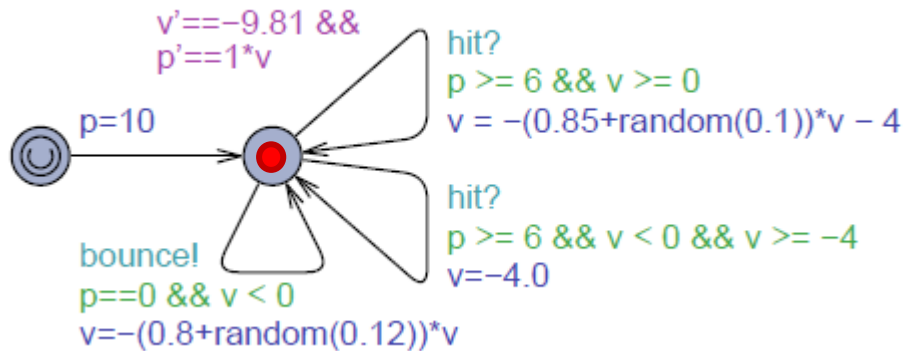
Player 2

General “delay”.  
Handles clock rates.



# Hybrid Automata

Ball



$$\begin{aligned}
 (p = 10, v = 0) &\xrightarrow{d} (p = 10 - 9.81/2d^2, v = -9.81d) \\
 &\xrightarrow{\text{bounce!}} (p = 0, v = 14.02 \cdot 0.83) \text{ at } d = 1.43 \\
 &\xrightarrow{d} (p = 6.92, v = 0) \text{ at } d = 1.18 \\
 &\xrightarrow{d} (p = 0, v = 11.51) \text{ at } d = 1.18 \\
 &\xrightarrow{\text{bounce!}} \dots
 \end{aligned}$$

## Semantics

### States

$(l, \nu)$  where  $\nu \in \mathbb{R}^X$

### Transitions

$(l, \nu) \rightarrow_d (l', \nu')$  where

$\nu' = F(l)(d, \nu)$

provided  $\nu' \in \text{Inv}(l)$

$(l, \nu) \rightarrow_a (l', \nu')$  if

there exists  $(l, g, a, \phi, l') \in E$

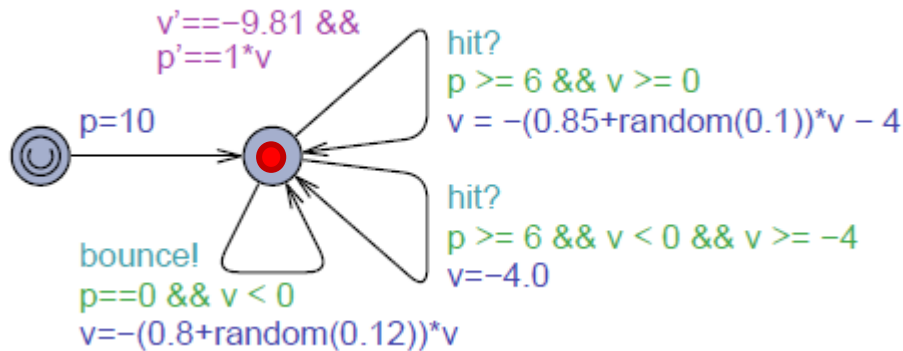
with  $\nu \in g$  and

$(\nu, \nu') \in \phi$  and

$\nu' \in \text{Inv}(l')$

# Stochastic Hybrid Automata

Ball



$$(p = 10, v = 0) \xrightarrow{d} (p = 10 - 9.81/2d^2, v = -9.81d)$$

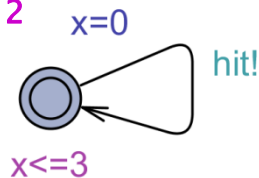
bounce!

$$\xrightarrow{\quad} (p = 0, v = 14.02 \cdot 0.83) \text{ at } d = 1.43$$

Player 1



Player 2



$$Pr_1[\text{hit! bounce!}] = \int_{t=0}^{t=1.43} 2.5 e^{-2.5t} dt$$

$$= [-e^{-2.5t}]_0^{1.43} = 0.97$$

$$Pr_2[\text{hit! bounce!}] = \int_{t=0}^{t=1.43} 1/3 dt$$

$$= [1/3 t]_0^{1.43} = 0.48$$

## Stochastic Semantics

For each state  $s=(l, \nu)$

Delay density function\*

$$\mu_s: \mathbb{R}_{>0} \rightarrow \mathbb{R}$$

Output Probability Function

$$\gamma_s: \Sigma_o \rightarrow [0, 1]$$

Next-state density function\*

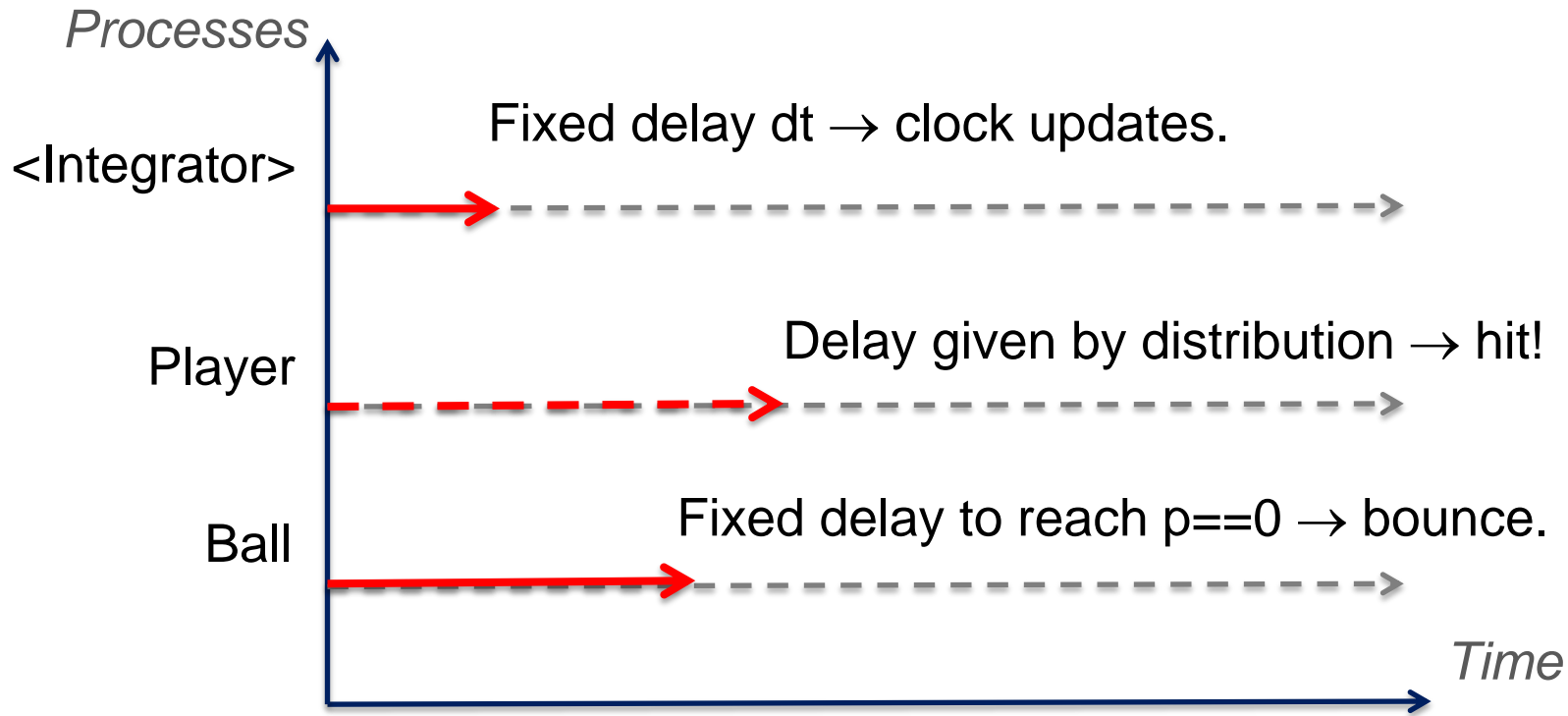
$$\eta_{a \ s}: \text{St} \rightarrow \mathbb{R}$$

where  $a \in \Sigma$ .

\* Dirac's delta functions for deterministic delays / next state



# Solving ODEs / Stochastic Semantics

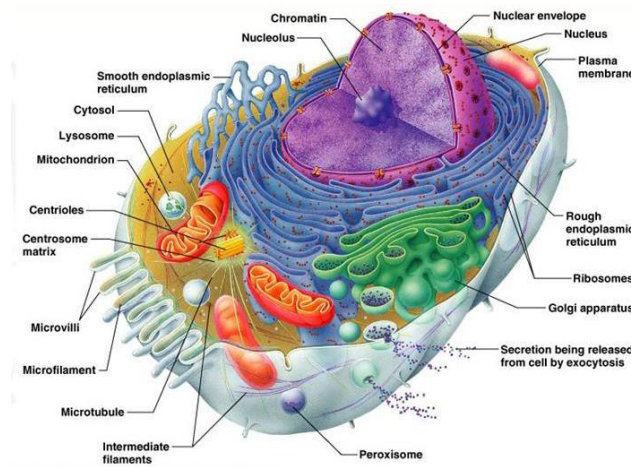


Race between processes.

Choice of  $dt$  and clock updates can be changed (solver).



# Biological Oscillator



# A Biological Oscillator

- Circadian oscillator.

*N. Barkai and S. Leibler. Biological rhythms: Circadian clocks limited by noise. Nature, 403:267–268, 2000*

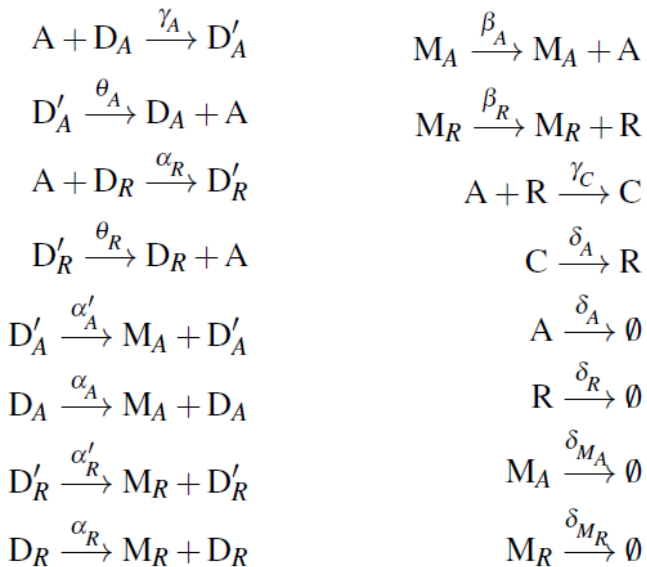
- Two ways to model:

1. Stochastic model that follow the reactions.
2. Continuous model solving the ODEs.

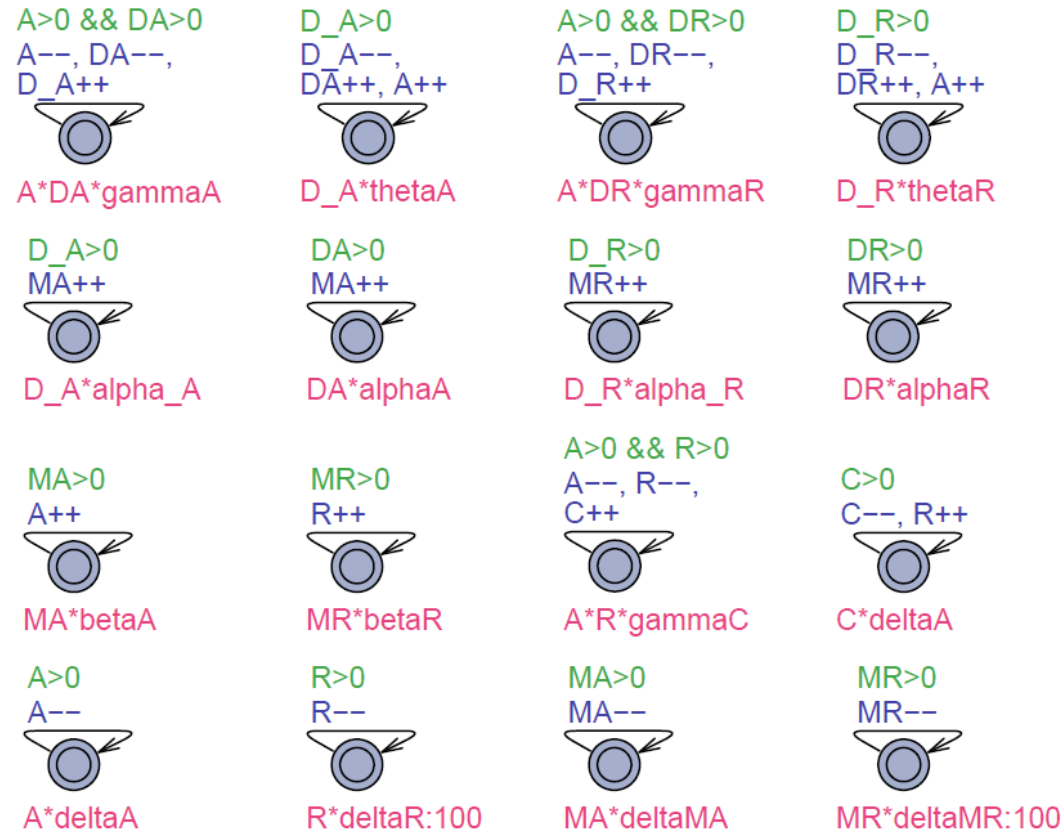
- Analysis:

- Evaluate time between peaks.
- The continuous model is the limit behavior of the stochastic model.
- Use frequency analysis for comparison.

# Stochastic Model



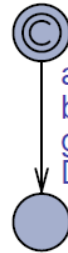
(a) Reactions.



(b) UPPAAL model representation.

# Continuous Model

$$\begin{aligned}
 dD_A/dt &= \theta_A D'_A - \gamma_A D_A A \\
 dD_R/dt &= \theta_R D'_R - \gamma_R D_R A \\
 dD'_A/dt &= \gamma_A D_A A - \theta_A D'_A \\
 dD'_R/dt &= \gamma_R D_R A - \theta_R D'_R \\
 dM_A/dt &= \alpha'_A D'_A + \alpha_A D_A - \delta_{M_A} M_A \\
 dM_R/dt &= \alpha'_R D'_R + \alpha_R D_R - \delta_{M_R} M_R \\
 dA/dt &= \beta_A M_A + \theta_A D'_A + \theta_R D'_R \\
 &\quad - A(\gamma_A D_A + \gamma_R D_R + \gamma_C R + \delta_A) \\
 dR/dt &= \beta_R M_R - \gamma_C A R + \delta_A C - \delta_R R \\
 dC/dt &= \gamma_C A R - \delta_A C
 \end{aligned}$$



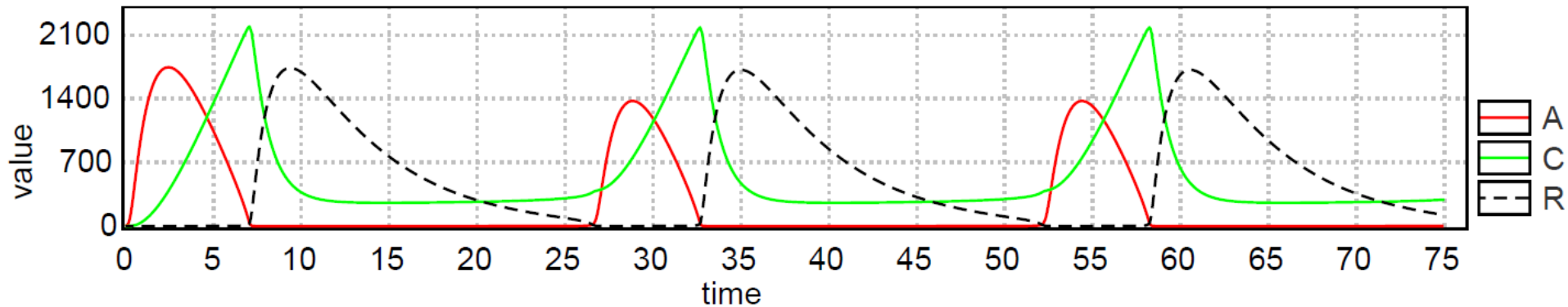
alphaA=50, alpha\_A=500, alphaR=0.01, alpha\_R=50,  
 betaA=50, betaR=5, deltaMA=10, deltaMR=0.5, deltaA=1, deltaR=0.2,  
 gammaA=1, gammaR=1, gammaC=2, thetaA=50, thetaR=100,  
 DA=1, DR=1, D\_A=0, D\_R=0, MA=0, MR=0, A=0, R=0, C=0

alphaA'==0 && alpha\_A'==0 && alphaR'==0 && alpha\_R'==0 &&  
 betaA'==0 && betaR'==0 && deltaA'==0 && deltaR'==0 &&  
 deltaMA'==0 && deltaMR'==0 && gammaA'==0 &&  
 gammaR'==0 && gammaC'==0 && thetaA'==0 && thetaR'==0 &&  
 DA'== thetaA\*D\_A-gammaA\*DA\*A &&  
 DR'== thetaR\*D\_R-gammaR\*DR\*A &&  
 D\_A'== gammaA\*DA\*A-thetaA\*D\_A &&  
 D\_R'== gammaR\*DR\*A-thetaR\*D\_R &&  
 MA'== alpha\_A\*D\_A+alphaA\*DA-deltaMA\*MA &&  
 MR'== alpha\_R\*D\_R+alphaR\*DR-deltaMR\*MR &&  
 A'== betaA\*MA+thetaA\*D\_A+thetaR\*D\_R  
 -A\*(gammaA\*DA+gammaR\*DR+gammaC\*R+deltaA) &&  
 R'== betaR\*MR-gammaC\*A\*R+deltaA\*C-deltaR\*R &&  
 C'== gammaC\*A\*R-deltaA\*C

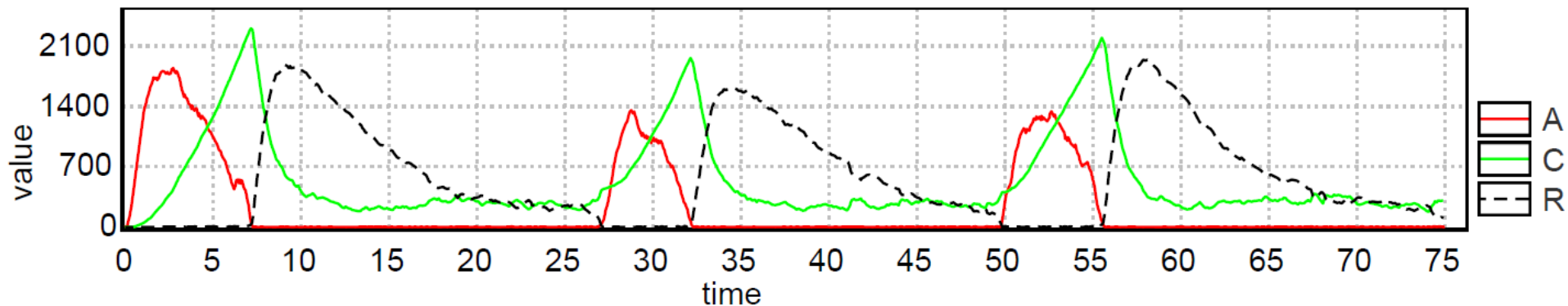
(a) Ordinary differential equations.

(b) UPPAAL automaton representation.

# Results of Simulations



(a) ODE model simulation plot.



(b) Stochastic model simulation plot.

# Frequency Domain Analysis (Fourier Transform)

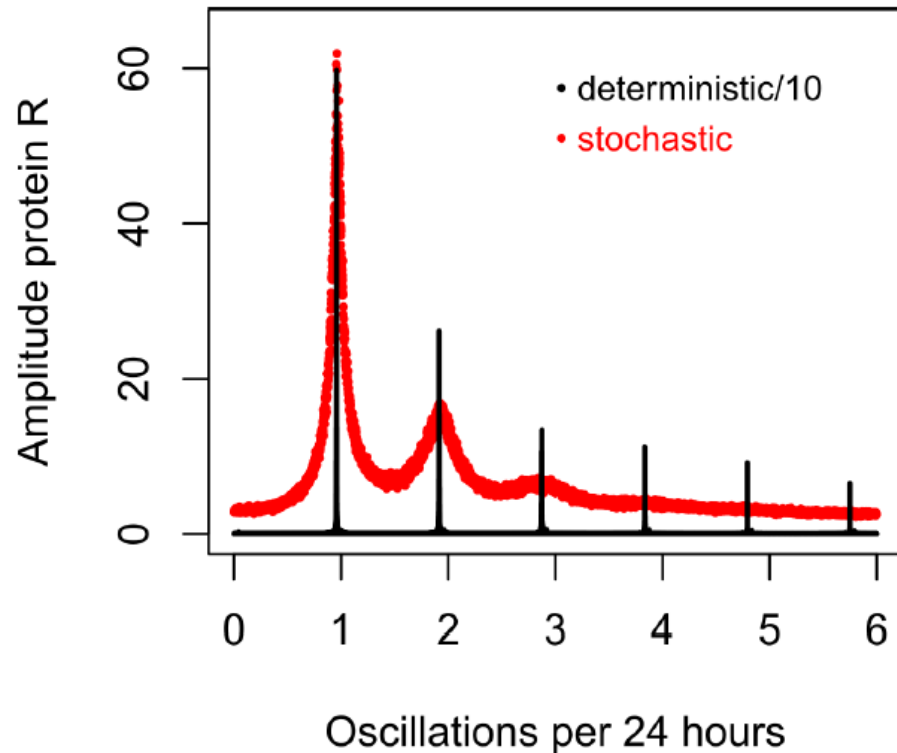


Fig.11: Average frequency spectra of protein R.  $\delta t = 2h$ ,  $K = 12500$ .  $N = 100$  ( $N = 1$ ) for the stochastic (deterministic) model.

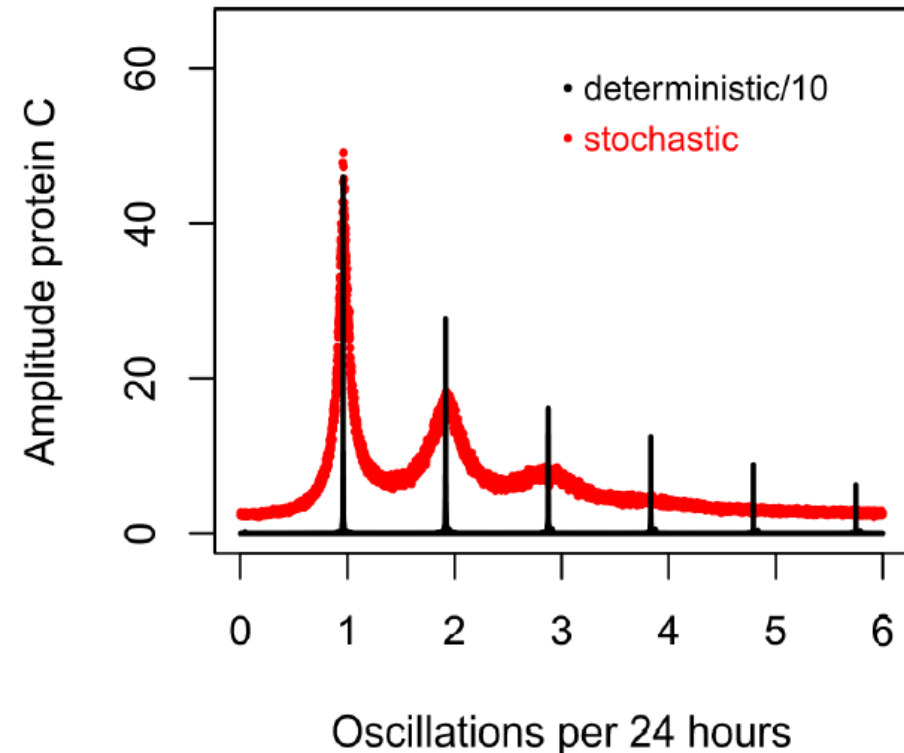
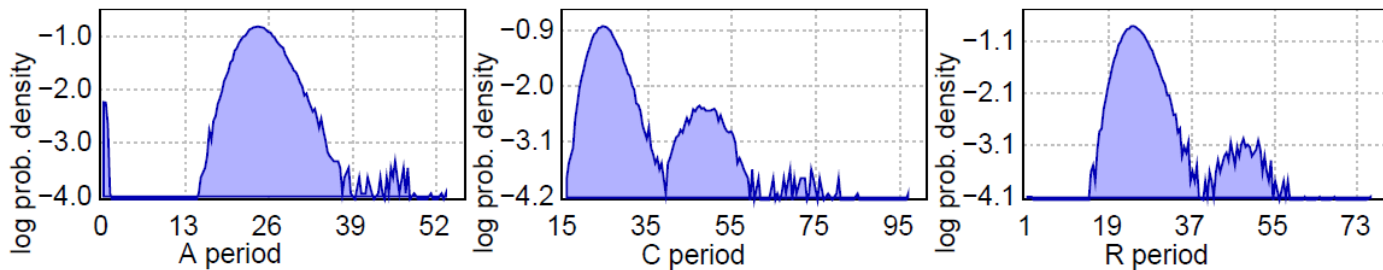
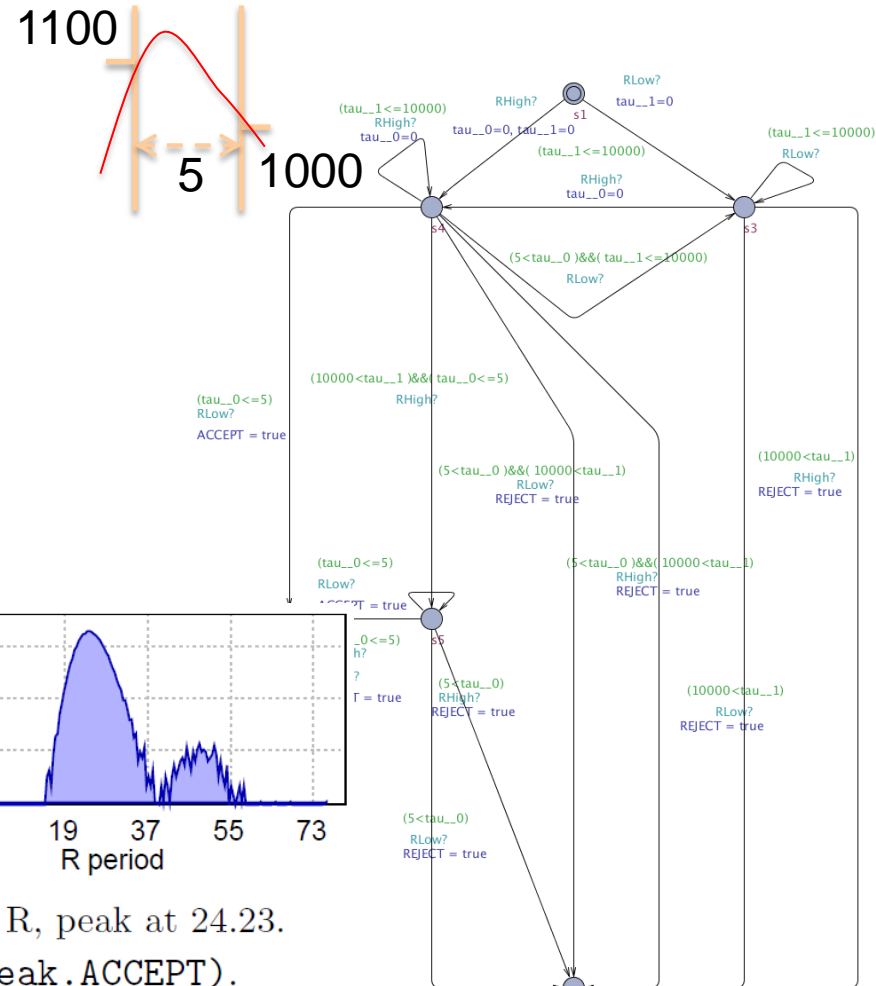


Fig.12: Average frequency spectra of protein C.  $\delta t = 2h$ ,  $K = 12500$ .  $N = 100$  ( $N = 1$ ) for the stochastic (deterministic) model.

# Time Between Peaks

- Use the MITL formula  
 $\text{true } U[\leq 1000] \ (A > 1100 \ \& \ \text{true } U[\leq 5] \ A \leq 1000)$ .
- Generate monitors (one shown).
- Run SMC.



(a) A, peak at 24.22. (b) C, peak at 24.21. (c) R, peak at 24.23.

Fig. 9: Estimated periods:  $\text{Pr}[x \leq 100] \ (\Leftrightarrow \text{secondPeak.ACCEPT})$ .

# Energy Aware Buildings





# What This Work is About

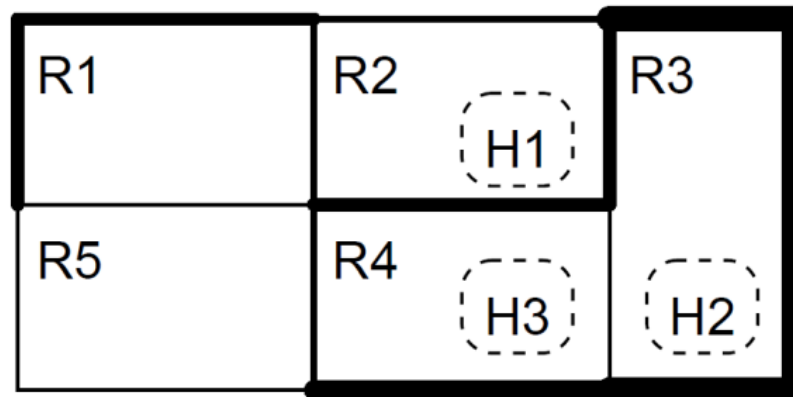
- Find optimal parameters for, e.g., a controller.
  - Applied to stochastic hybrid systems.
  - Suitable for different domains: biology, avionics...
- Technique: statistical model-checking.
  - This work: Apply ANOVA to reduce the number of needed simulations.

# Overview

- Energy aware buildings
  - The case-study in a nutshell
- Choosing the parameters
  - Naïve approach
- Efficiently choosing the (best) parameters
  - ANOVA

# Energy Aware Buildings

- The case:
  - Building with **rooms** separated by doors or walls.
  - Contact with the **environment** by windows or walls.
  - Few transportable **heat sources** between the rooms.
  - Objective: **maintain the temperature** within range.



(a) Rooms  $R_i$  with heaters  $H_k$ .

# Energy Aware Buildings

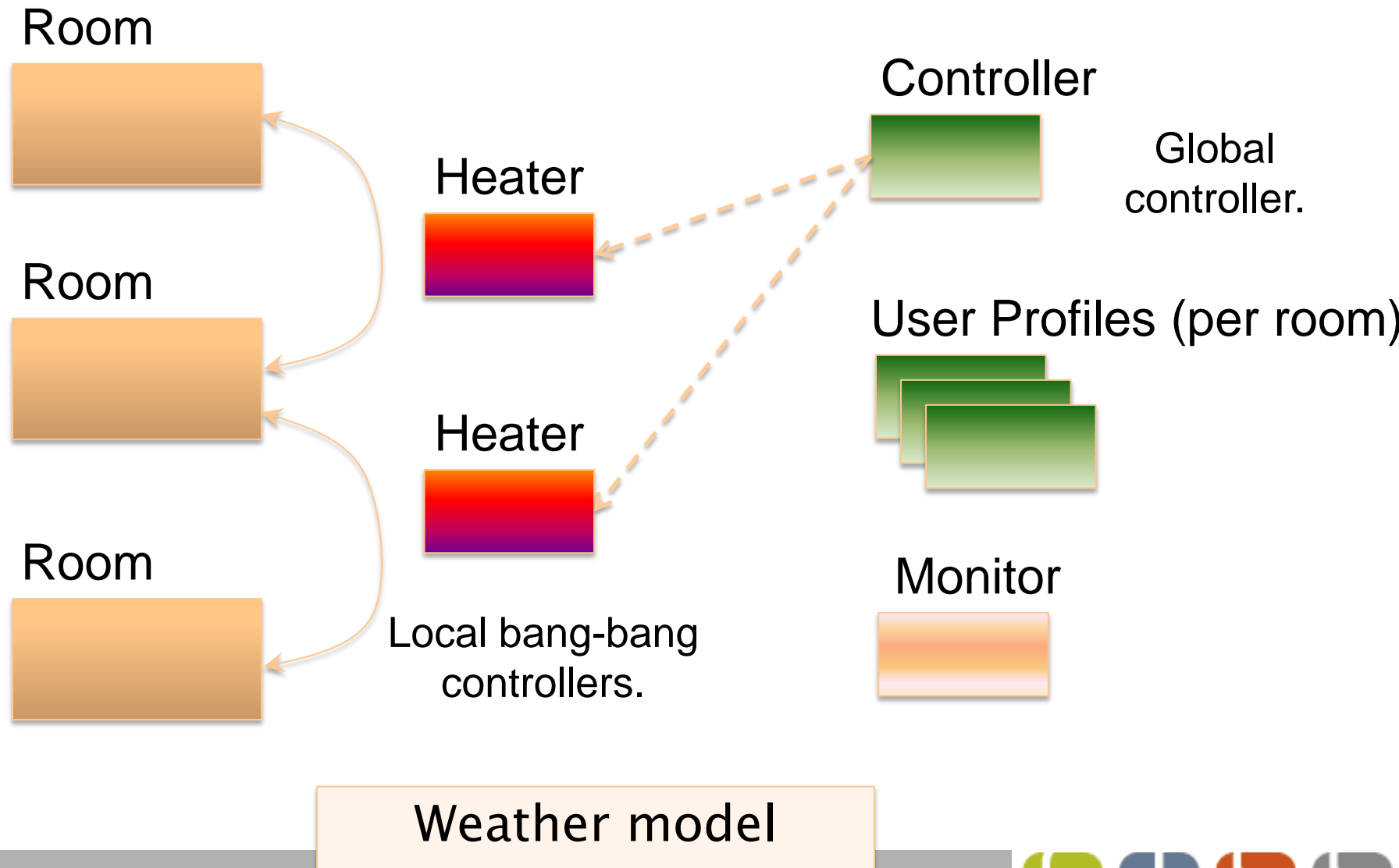
- Model:
  - Matrix of coefficients for heat transfer between *rooms*.

$$T'_i = \sum_{j \neq i} a_{i,j} (T_j - T_i) + b_i (u - T_i) + c_i h_i$$

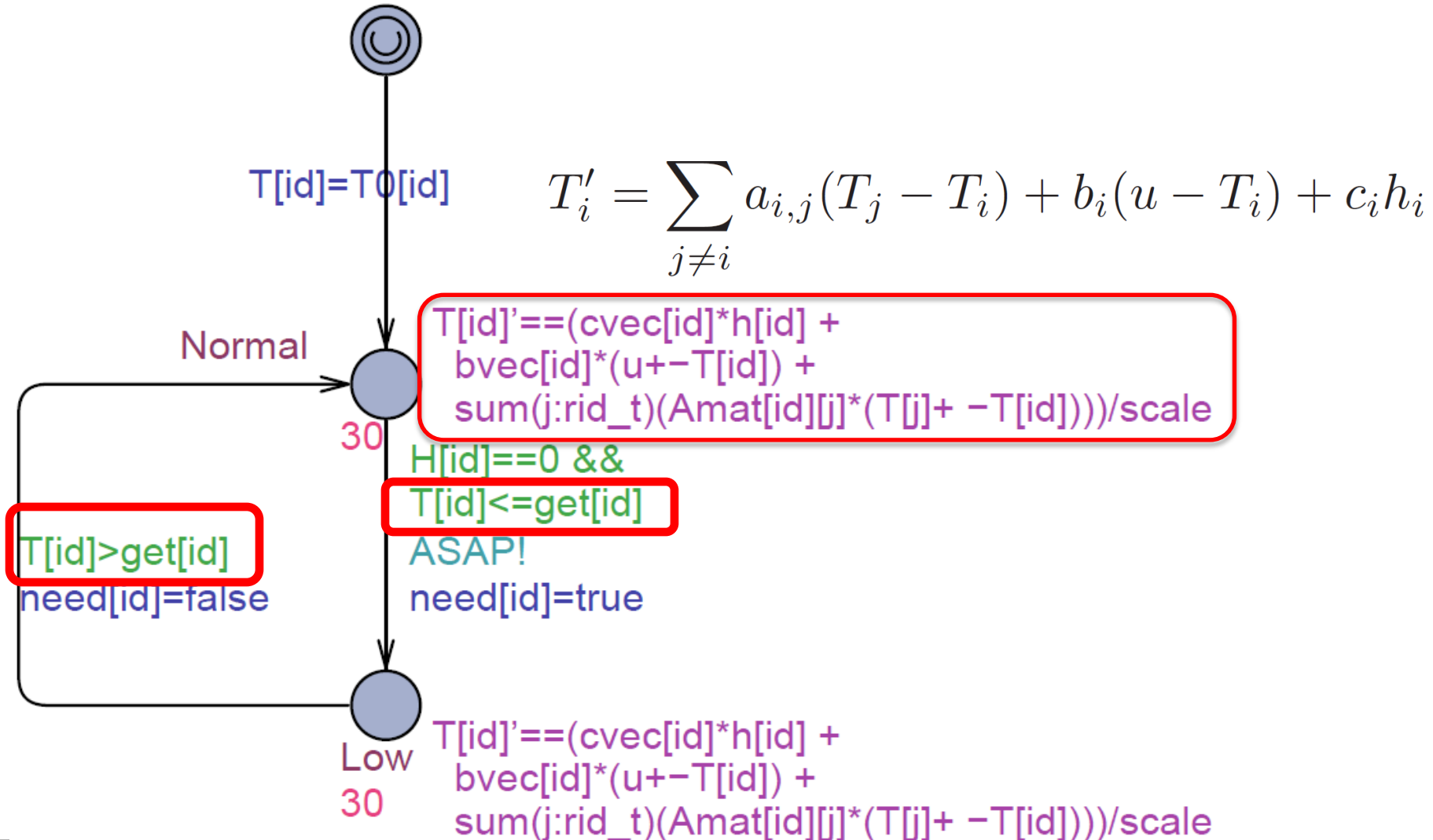
- Environment temperature → *weather* model.
  - Different controllers → *user* profiles.
- Goal:
  - *Optimize the controller.*

Science China  
2012

# Model Overview

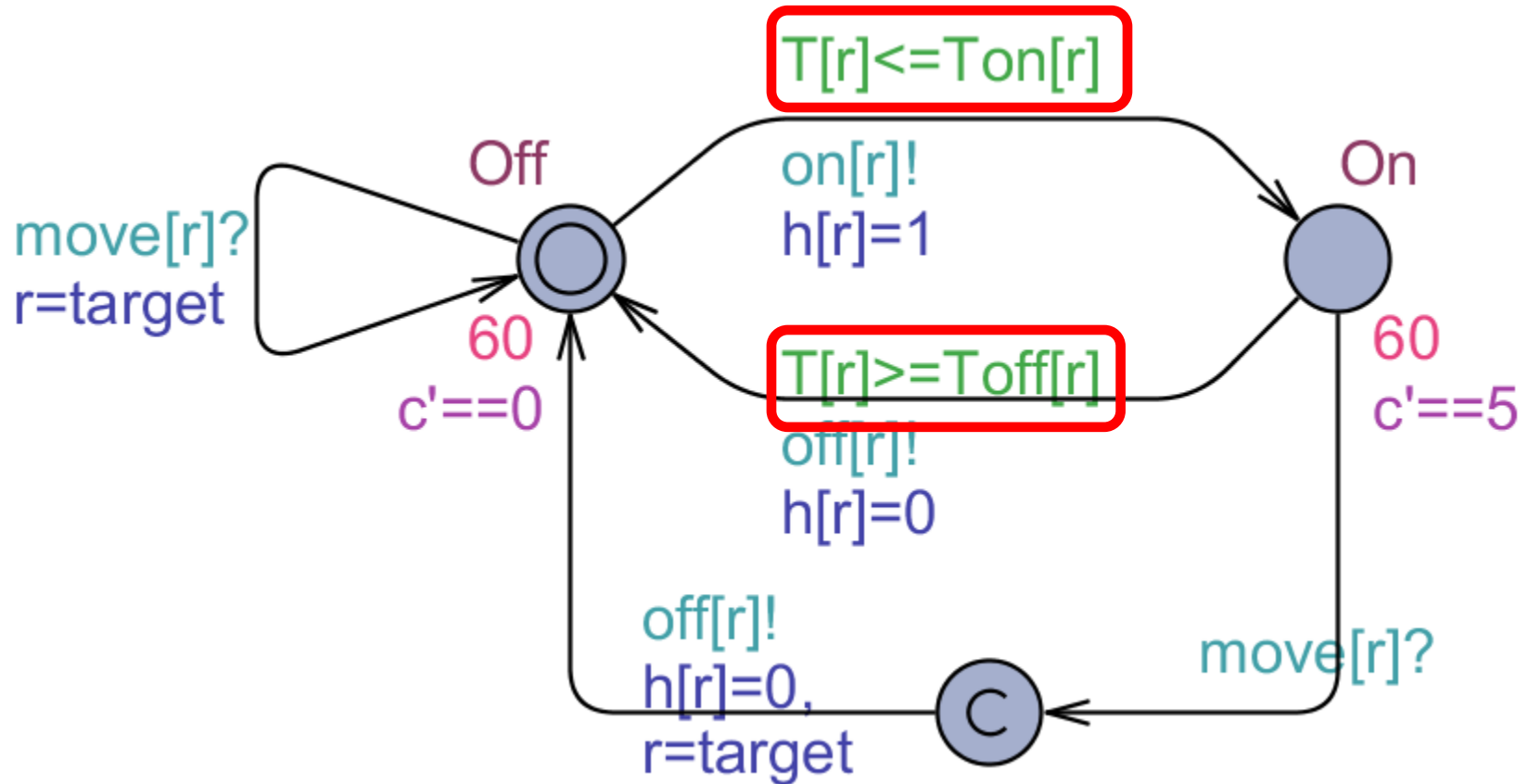


# Stochastic Hybrid Model of the Room



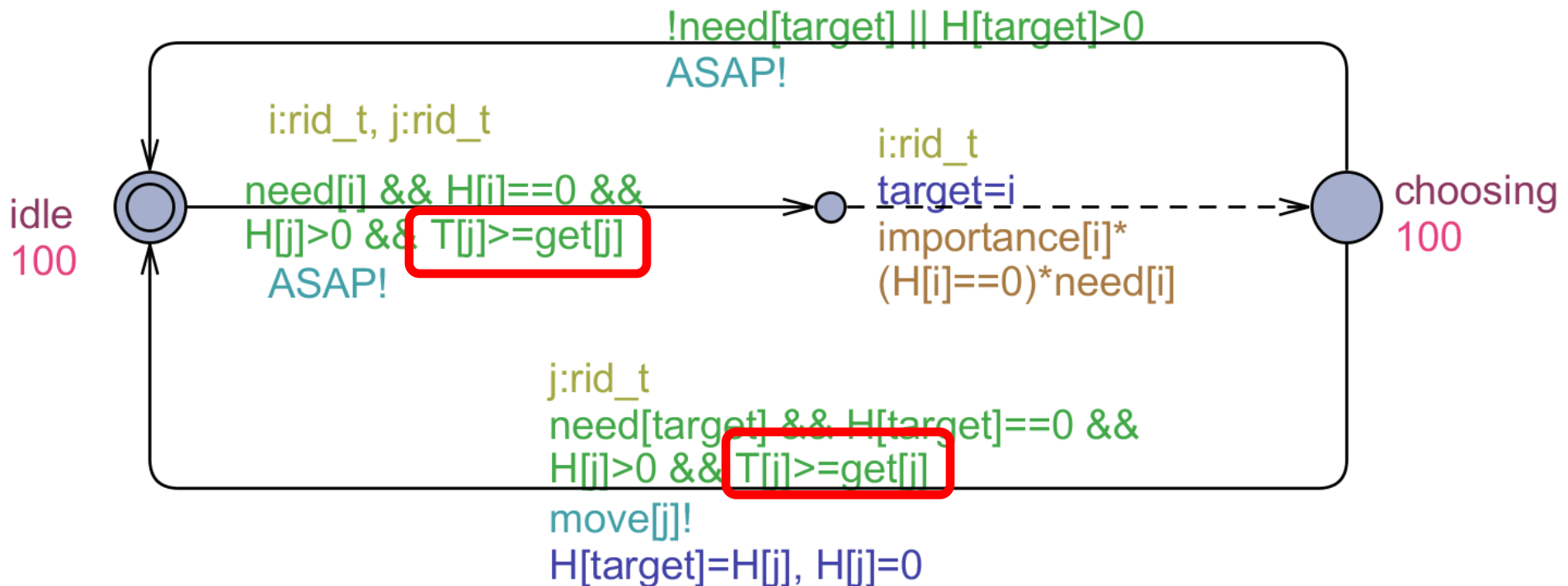
(a) Template for Room temperature.

# Model of the Heater



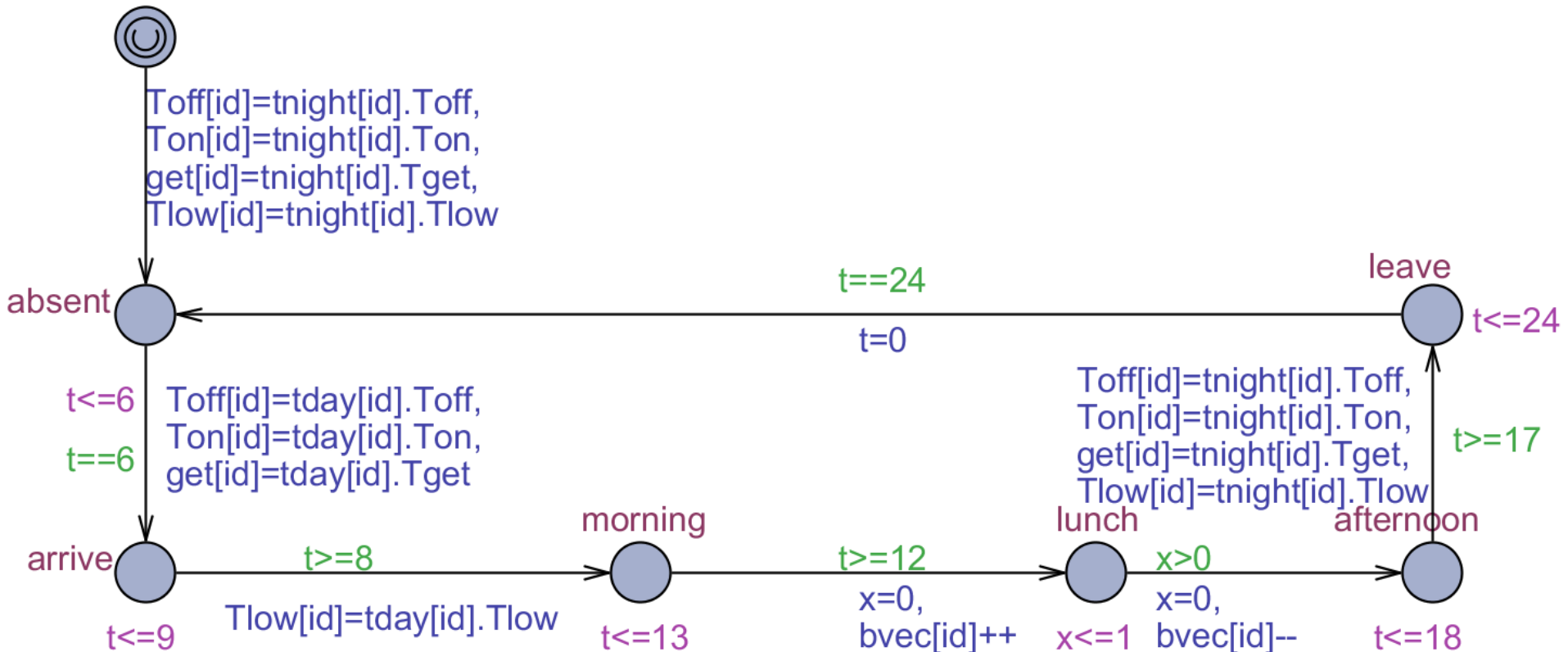
Local “bang-bang” controller.

# Main Controller

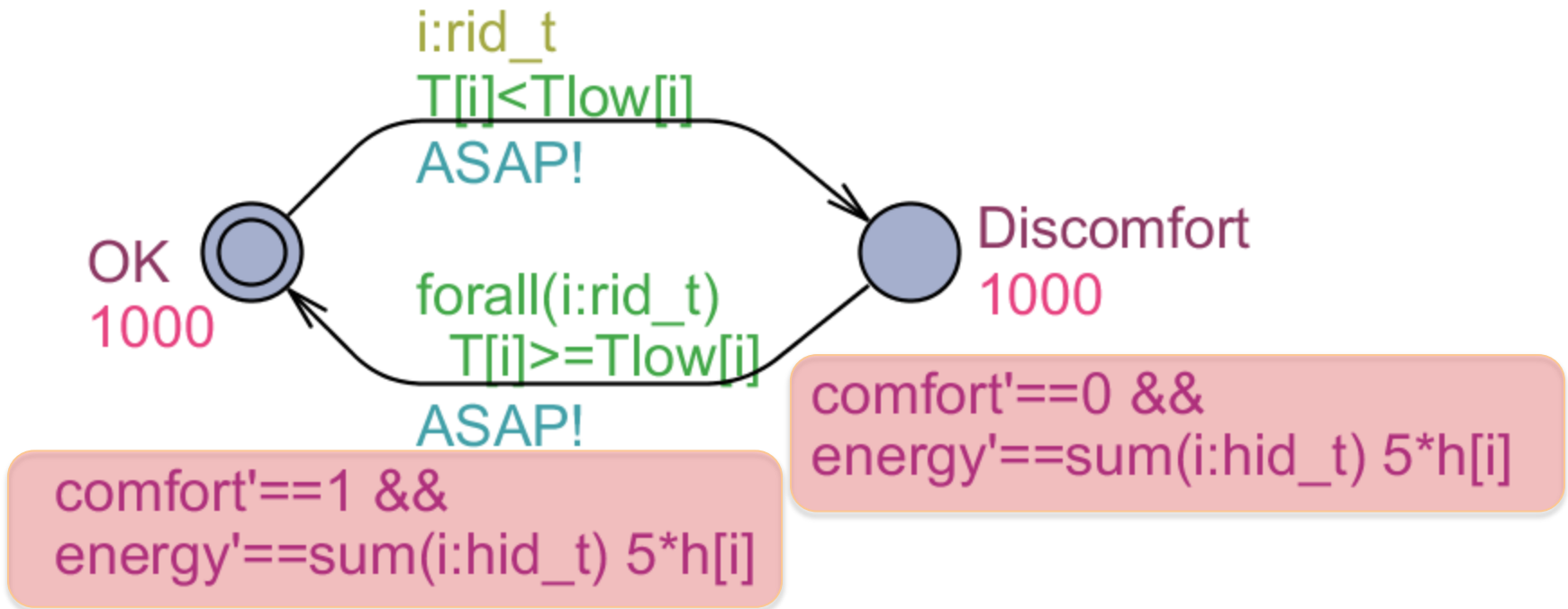




# Dynamic User Profile

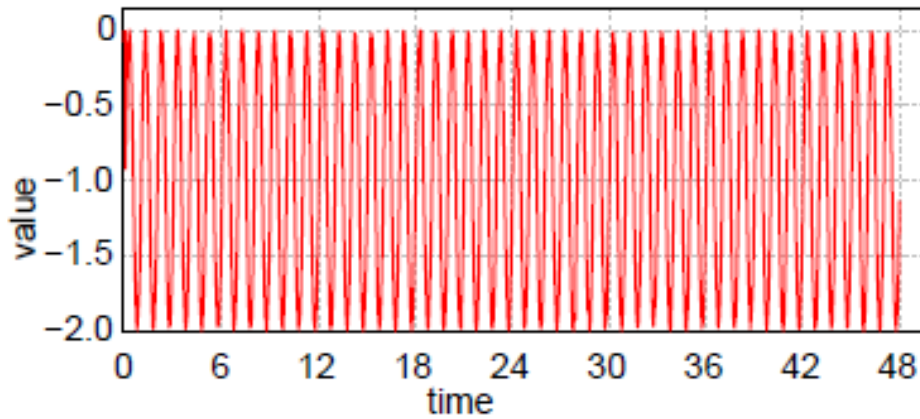


# Global Monitoring

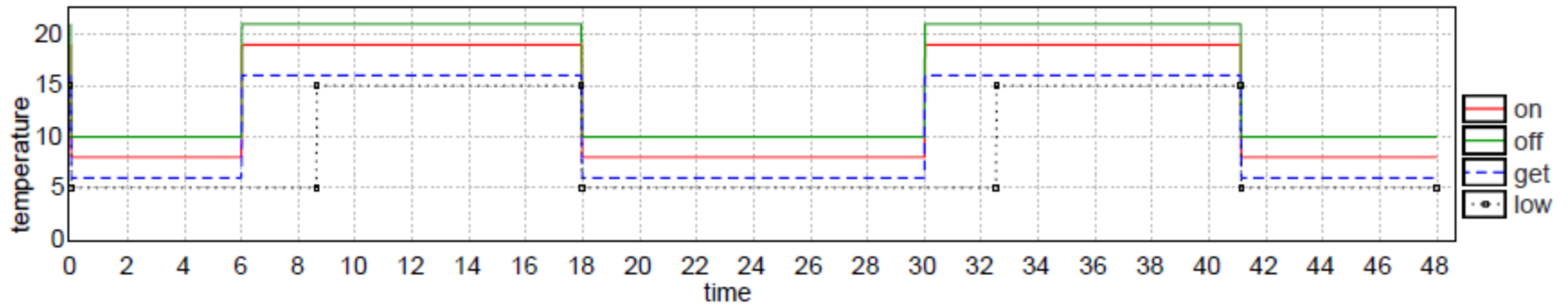


- + Maximize comfort.
  - Minimize energy.
  - ? Play with Ton and Tget.
- (Possible with Toff but not here).

# Simulations

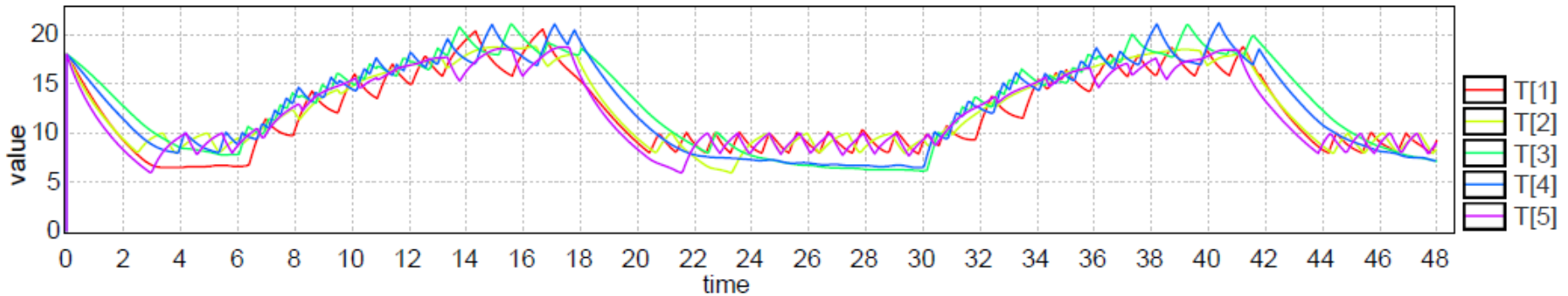


Weather Model

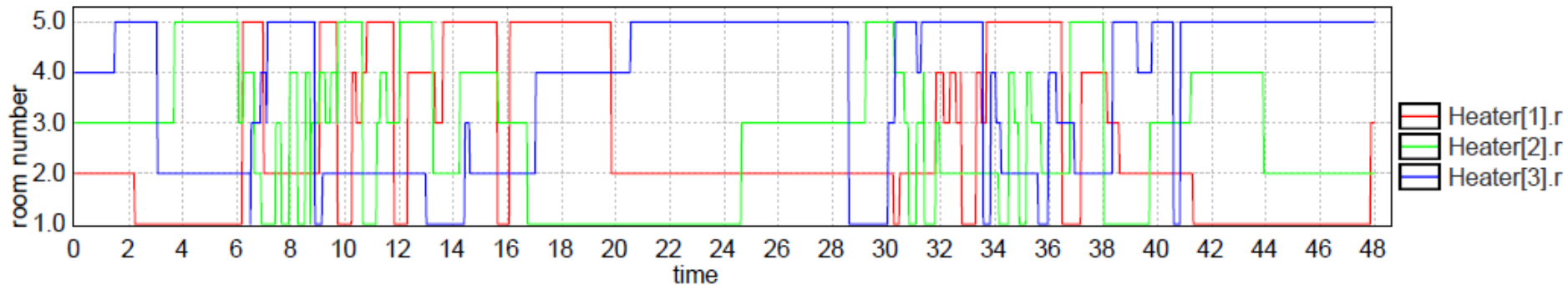


User Profile

# Simulations



```
simulate 1 [<=2*day]{ T[1], T[2], T[3], T[4], T[5] }
```



```
simulate 1 [<=2*day]{ Heater(1).r, Heater(2).r, Heater(3).r }
```

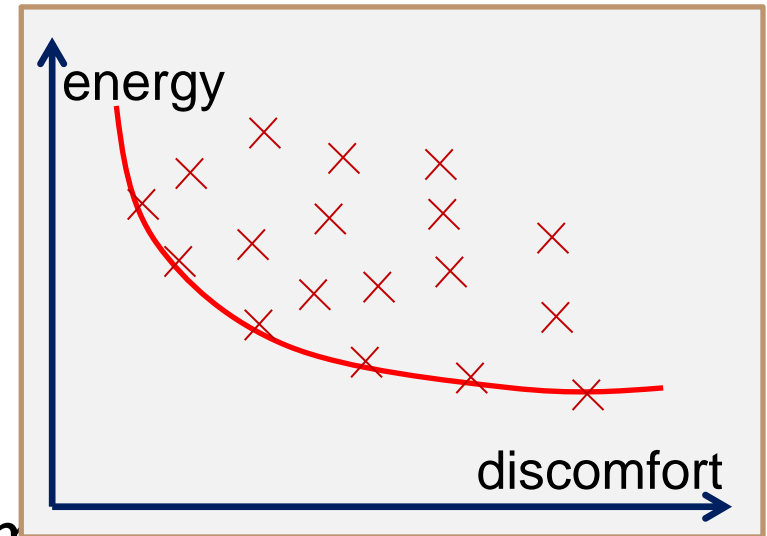
# How to Pick the Parameter Values?

- $T_{\text{on}}, T_{\text{get}} \in [16, 22] \rightarrow 49$  *discrete choices*.  
More if considering other parameters.
- Stochastic simulations.
  - Weather not deterministic.
  - User not deterministic (present, absent...)
- How to decide that one combination is better?
  - Probabilistic comparisons?  
49\*48 comparisons \* number of runs.
  - To optimize what? Discomfort or energy?



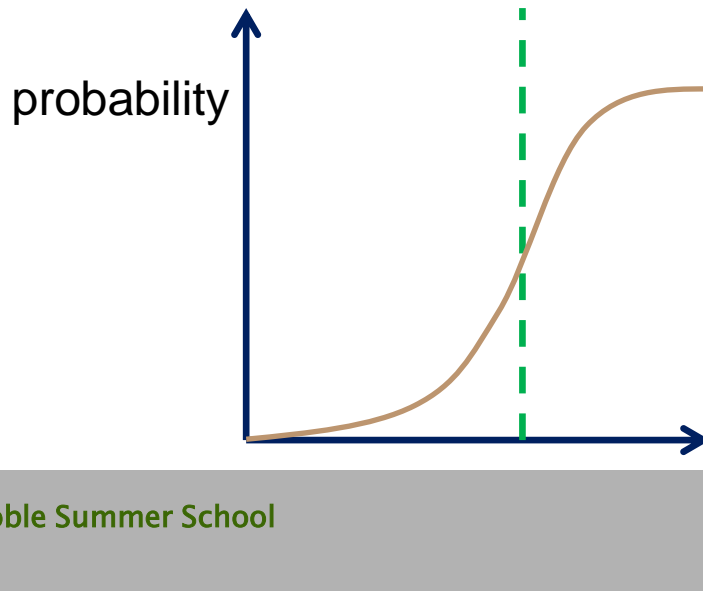
# How to Pick the Parameter Values?

- Remark:
  - Stochastic hybrid system  
 $\Rightarrow$  SMC
- Idea:
  - Generate runs.
  - Plot the result energy/discomfort.
  - Pick the Pareto frontier of the means.
- How many runs do you need?
  - What's the significance of the results?

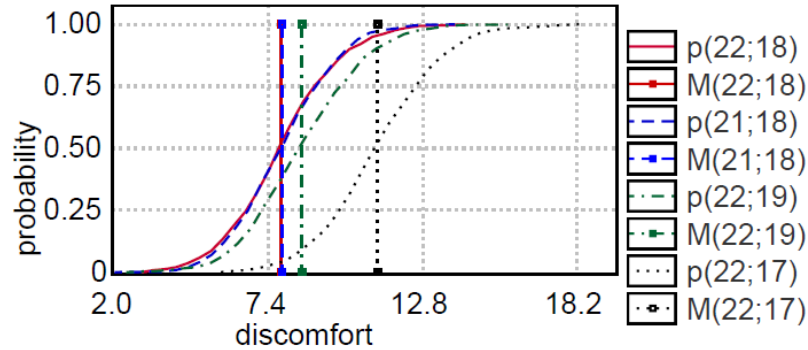


# “Naïve” Solution

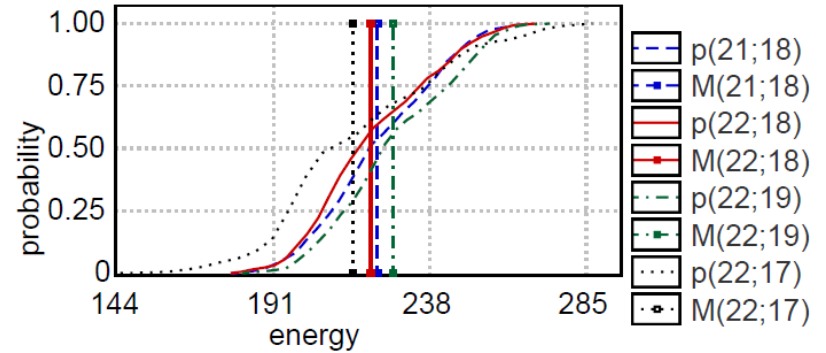
- Estimate the probabilities  
 $\Pr[\text{discomfort} \leq 100](\langle \rangle \text{ time} \geq 2 * \text{day})$   
 $\Pr[\text{energy} \leq 1000](\langle \rangle \text{ time} \geq 2 * \text{day})$
- From the obtained distributions (confidence known), compute the means.
- Pick the Pareto frontier of the means.



# “Naïve” Approach

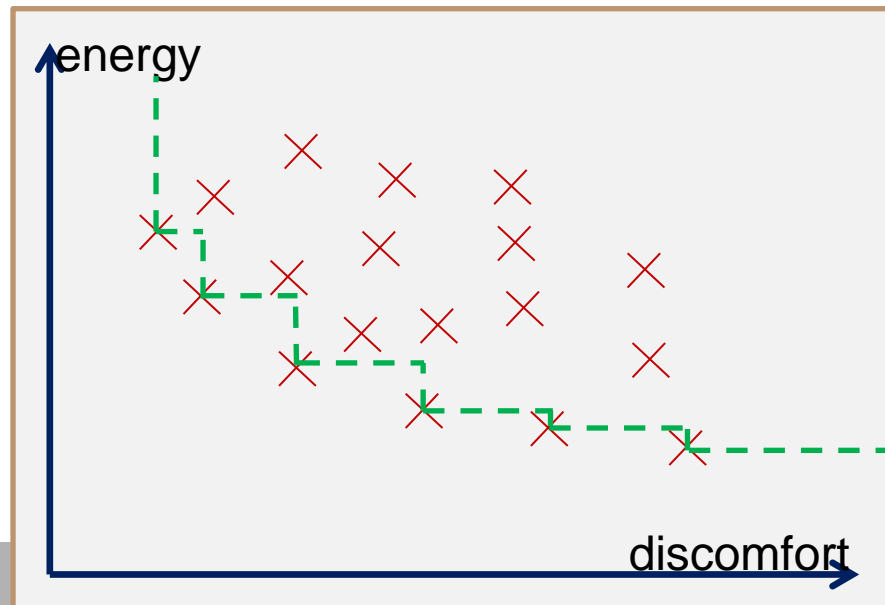


(a) Over discomfort.



(b) Over energy.

For each  $(T_{on}, T_{get})$





# ANOVA Method

- Compare several distributions.
  - Evaluate influence of each factor on the outcome.
- Generalization of Student's t-test.
  - Compare 2 distributions using the mean of their difference.
  - If confidence interval does not include zero, distributions are significantly different.
  - Cheaper than evaluating 2 means + on-the-fly possible.



# ANOVA Method

- 2-factor factorial experiment design
  - Ton, Tget are our 2 factors.
  - Each combination gives a distribution to compare.
  - Measure cost outcome (discomfort or energy).
- ANOVA estimates a linear model and computes the influence of each factor.
  - The measure of the influence is the F-statistic.
  - This is translated into P-value, the factor significance.
  - Assume balanced experiments.



# ANOVA Method

**Fewer runs, more efficient than before.**

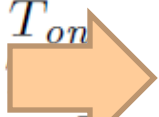
- Generate balanced measurements for each configuration to compare.
- Apply ANOVA on the data (used the tool R).
- If the factors are not significant, goto 1.
- Reuse the data and compute the confidence intervals of the means for each comparison.
- Compute the Pareto frontier.



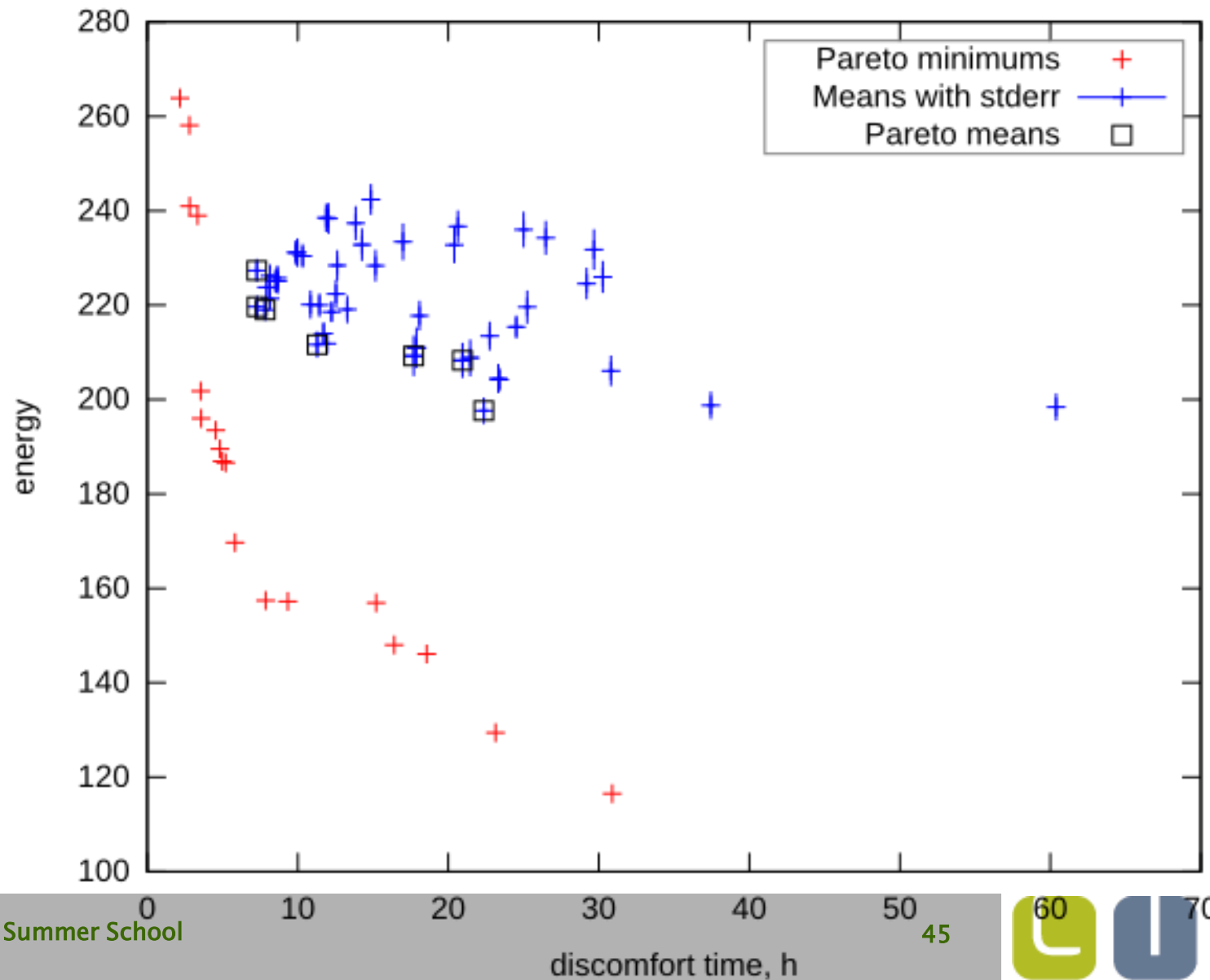
# ANOVA Results

$P < 0.05 \Rightarrow$  significant

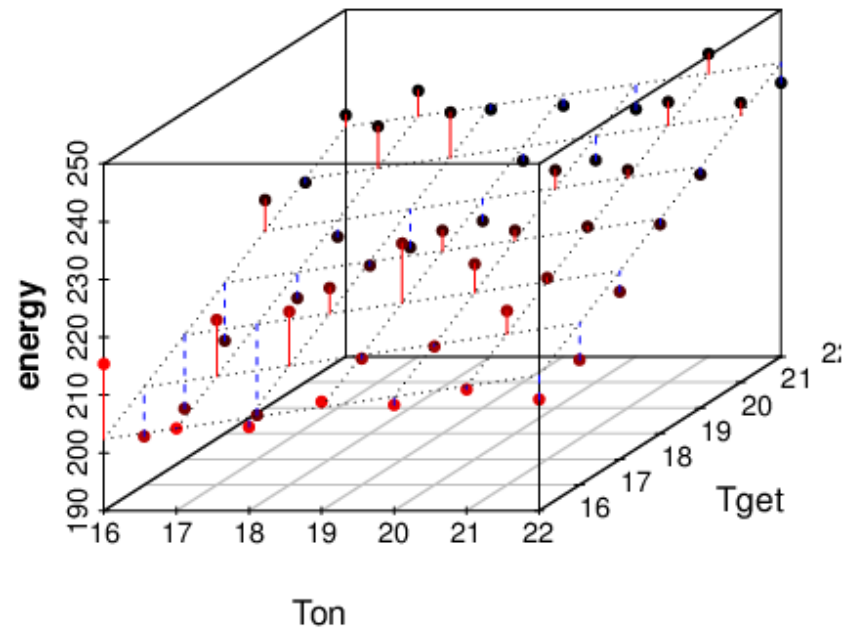
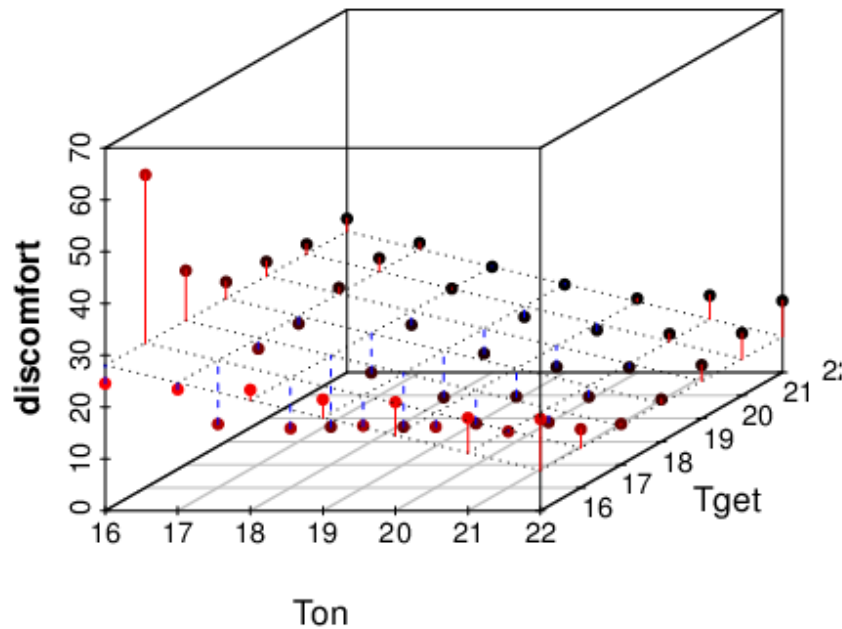
Number of runs	Factor	Discomfort time		Energy consumption	
		<i>F</i> value	P-value	<i>F</i> value	P-value
2 · 49	$T_{on}$	63.8874	<b>3.30e-12</b>	0.7147	0.4000
	$T_{get}$	0.0063	0.9369	17.5777	<b>6.24e-05</b>
	$T_{on} : T_{get}$	0.0629	0.8026	0.7181	0.3989
4 · 49	$T_{on}$	136.1676	<b>&lt;2e-16</b>	1.1647	0.2818
	$T_{get}$	0.1537	0.6955	17.9283	<b>3.55e-05</b>
	$T_{on} : T_{get}$	0.0003	0.9869	0.0582	0.8096
8 · 49	$T_{on}$	315.7978	<b>&lt;2e-16</b>	2.4425	0.1189
	$T_{get}$	0.1202	0.7290	35.8938	<b>4.76e-09</b>
	$T_{on} : T_{get}$	0.0096	0.9218	0.8253	0.3642
16 · 49	$T_{on}$	629.1384	<b>&lt;2e-16</b>	6.5909	0.01044
	$T_{get}$	0.5895	0.4429	90.9612	<b>&lt;2e-16</b>
	$T_{on} : T_{get}$	0.2852	0.5935	5.3053	0.02152
32 · 49	$T_{on}$	1263.5390	<b>&lt;2e-16</b>	27.9527	<b>1.42e-07</b>
	$T_{get}$	1.0840	0.2980	172.3296	<b>&lt;2.2e-16</b>
	$T_{on} : T_{get}$	0.5401	0.4625	3.2632	0.07104
64 · 49	$T_{on}$	2575.3208	<b>&lt;2e-16</b>	65.6245	<b>7.74e-16</b>
	$T_{get}$	4.6682	0.0308	405.4892	<b>&lt;2.2e-16</b>
	$T_{on} : T_{get}$	0.5949	0.4406	0.1926	0.6608



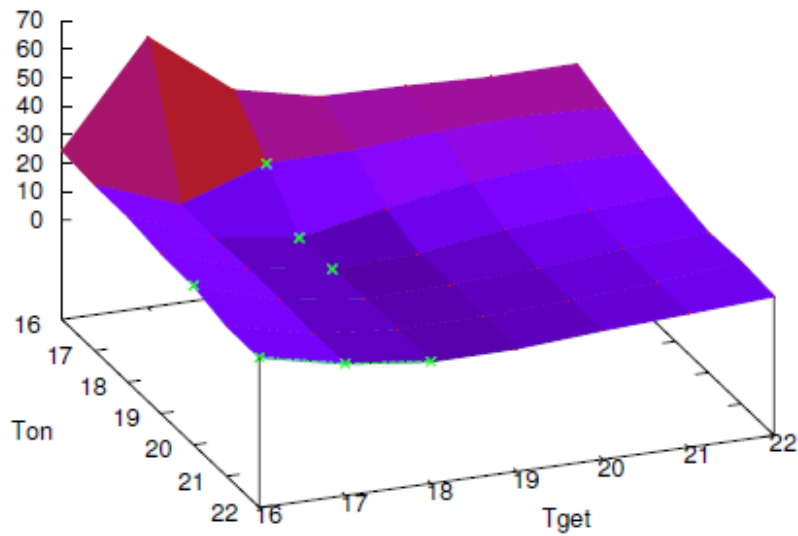
# Results



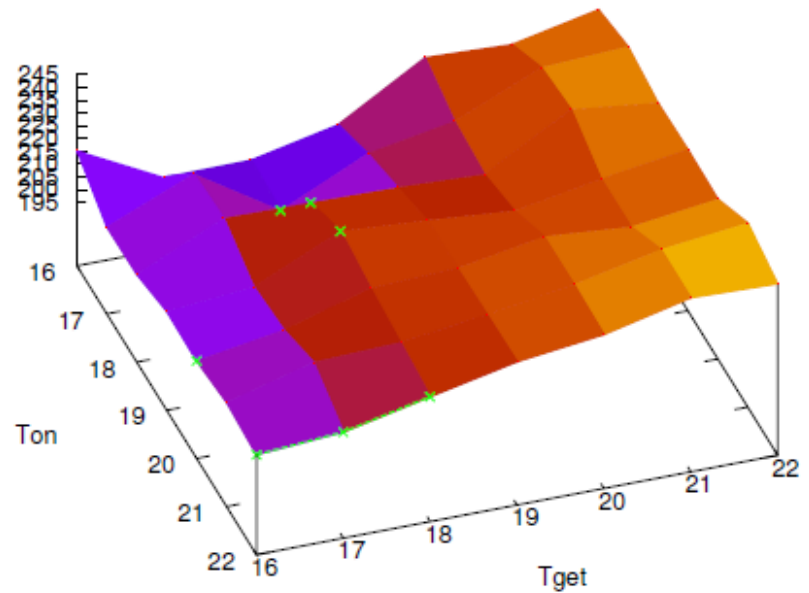
# Visualization of the Cost Model



# Results



(a) Discomfort.



(b) Energy.

# Comparison

- Naïve approach:  
738 runs per evaluation per cost  
\*2 (energy & discomfort) \*49  
(configurations).  
⇒ 1 h 5min
- ANOVA:  
3136 runs ⇒ 6min 6s.
- Core i7 2600



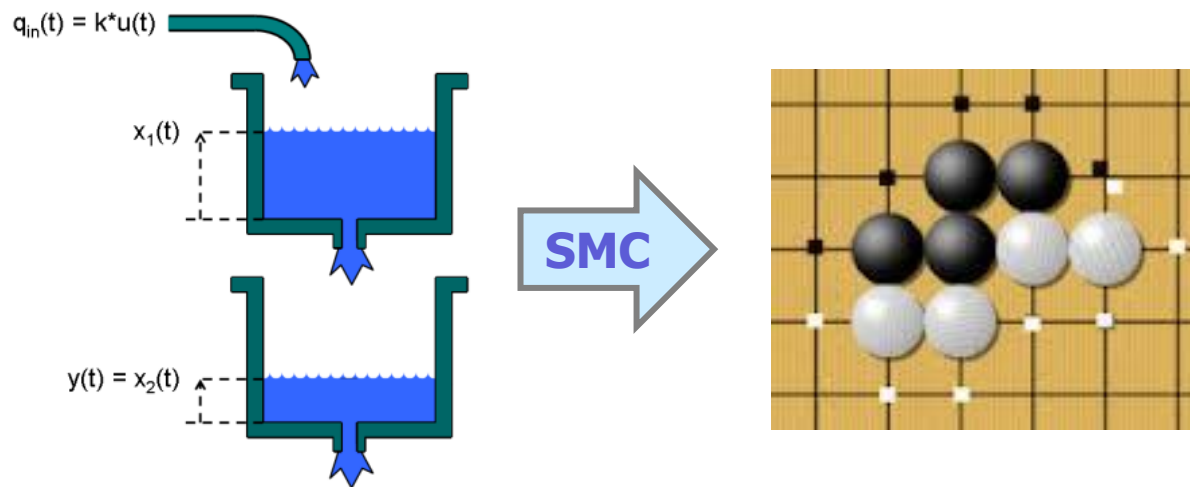


# Discussion

- Analysis of variance used sequentially to decide when there is enough data to distinguish the effect of 2 factors.
  - Efficient use of SMC.
- What if the factor has no influence?
  - Need an alternative test.
- Possible to distribute.
- Future work: Integrate ANOVA in UPPAAL



# Hybrid Controller Synthesis



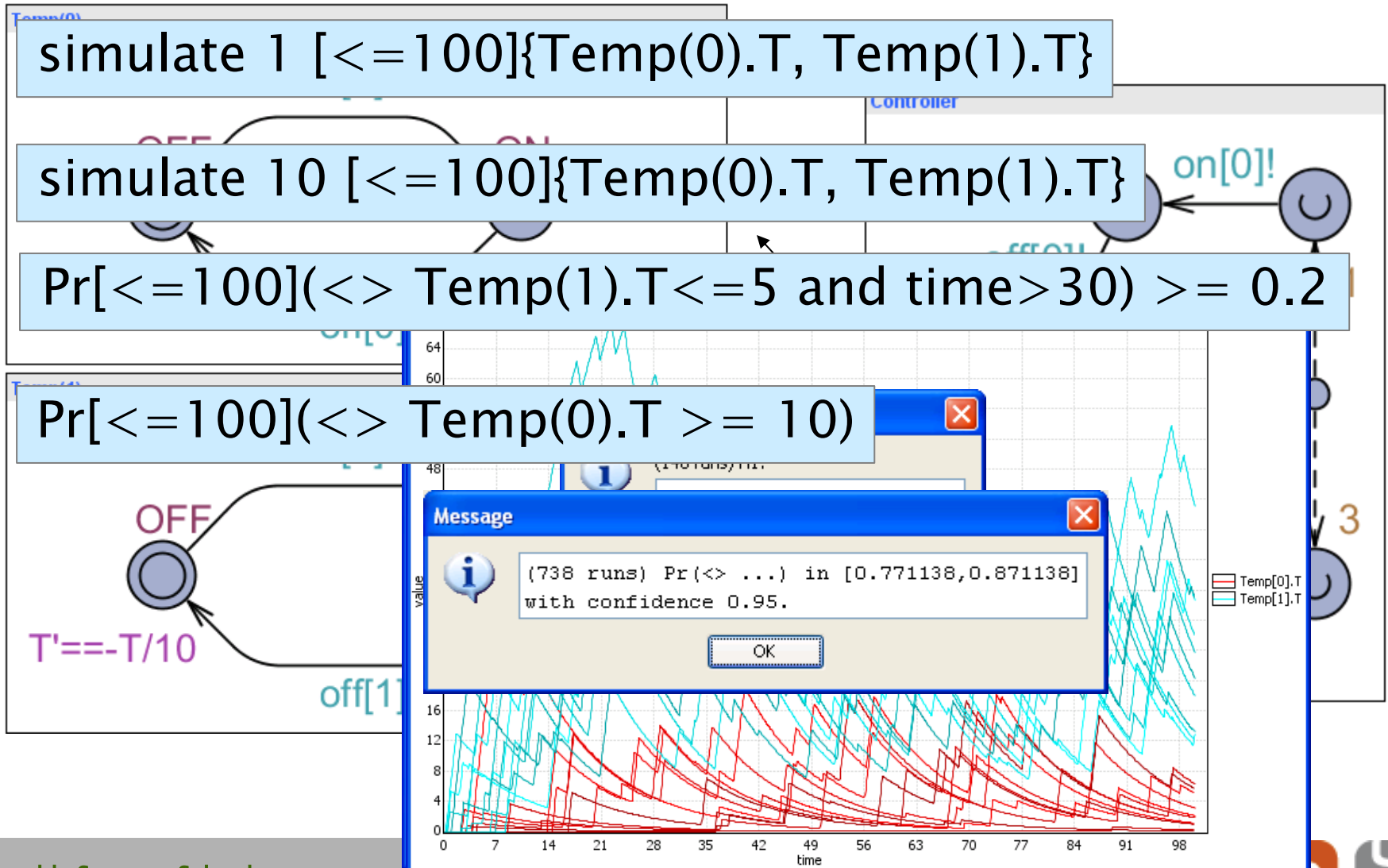
# Stochastic Hybrid Systems

simulate 1 [ $\leq 100$ ]{Temp(0).T, Temp(1).T}

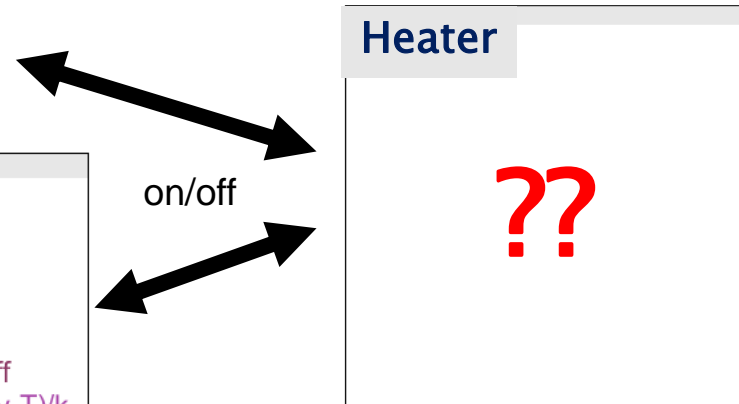
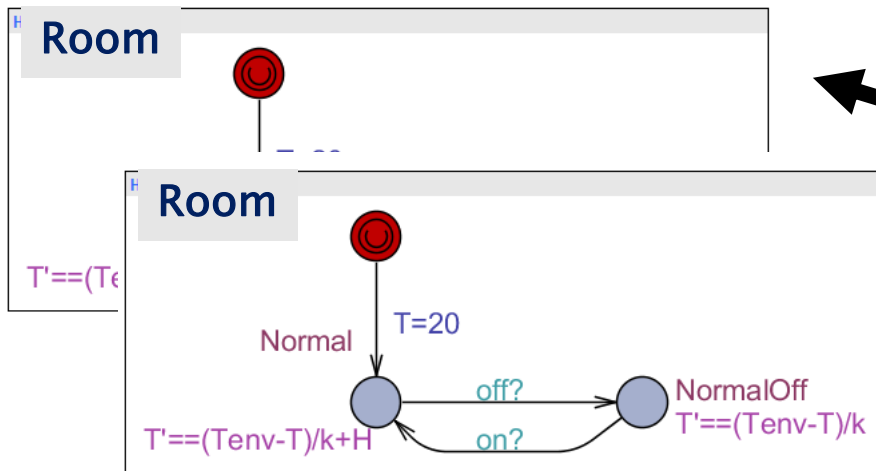
simulate 10 [ $\leq 100$ ]{Temp(0).T, Temp(1).T}

$\Pr[\leq 100](\langle \rangle \text{Temp(1).T} \leq 5 \text{ and time} > 30) \geq 0.2$

$\Pr[\leq 100](\langle \rangle \text{Temp(0).T} \geq 10)$

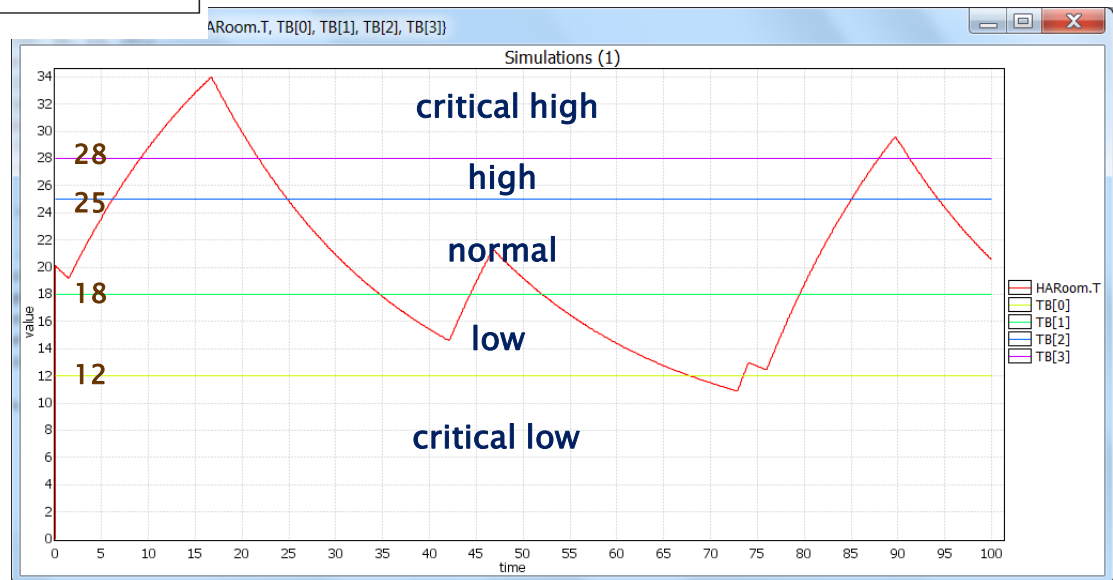


# Controller Synthesis

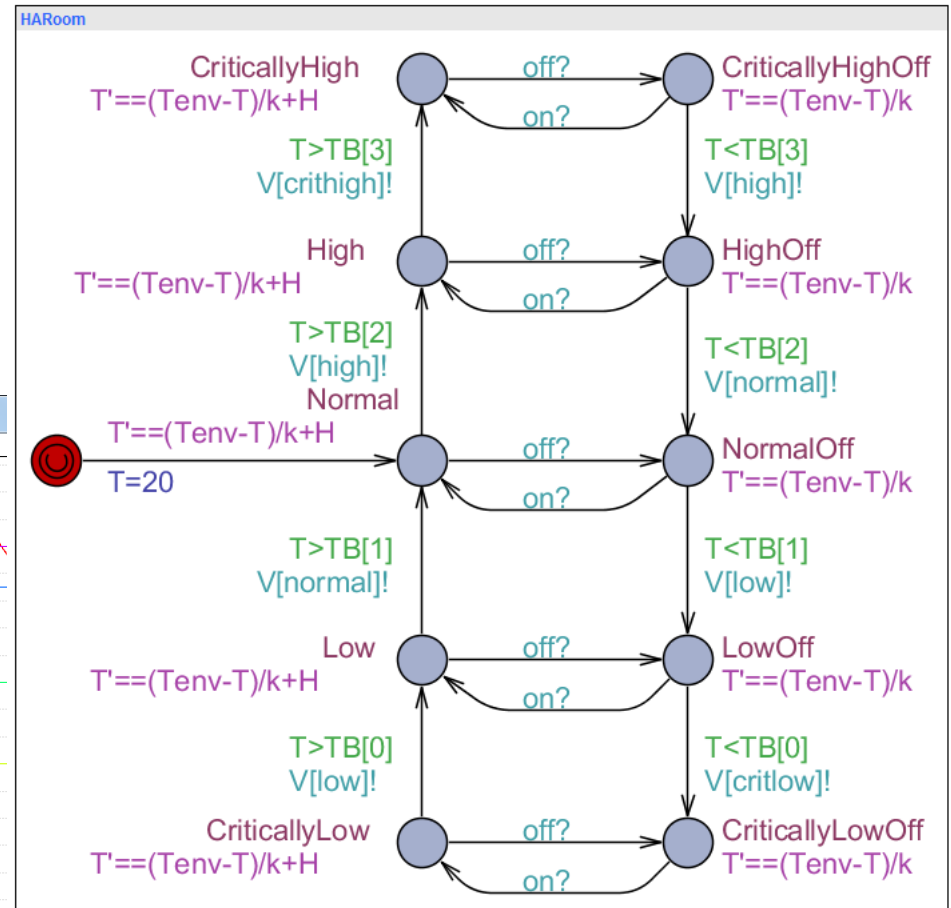
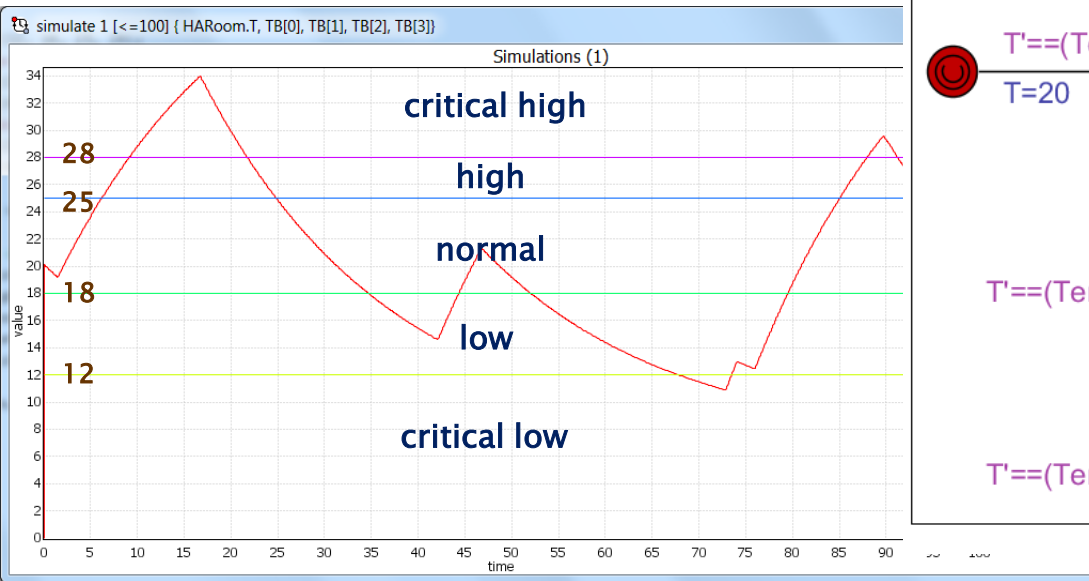
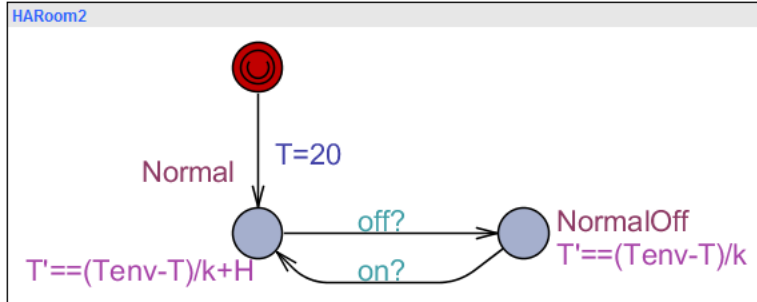


```

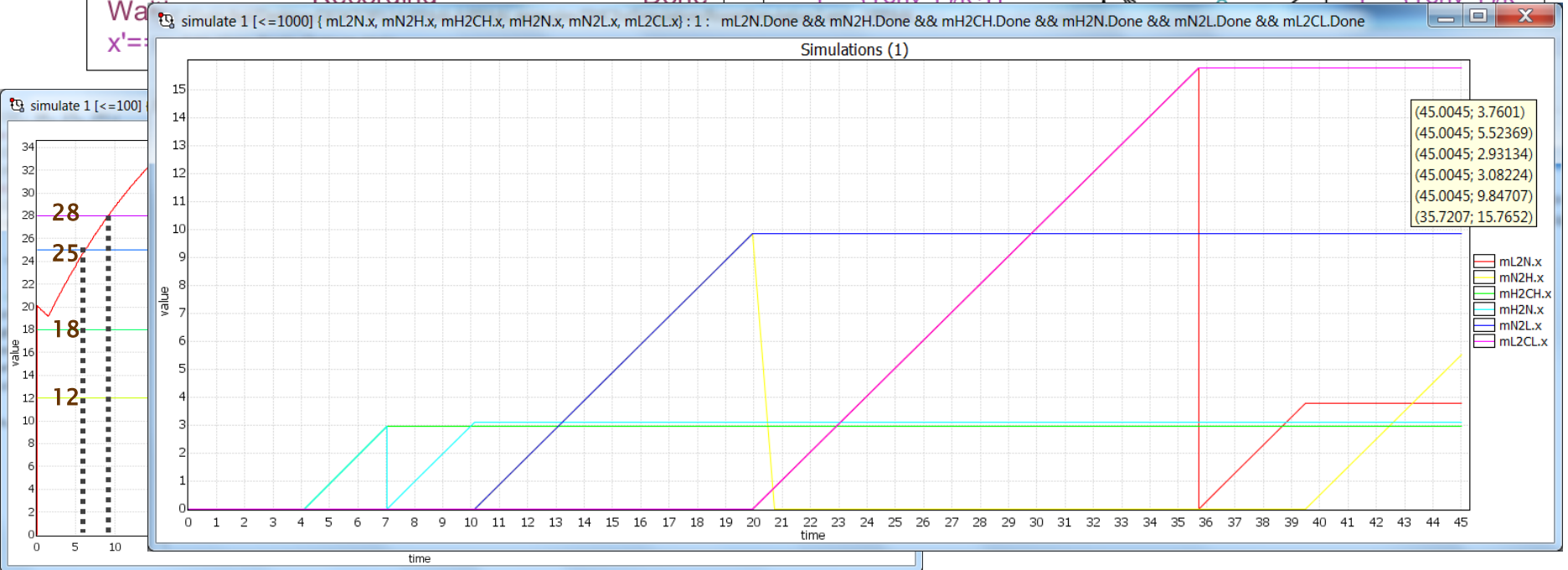
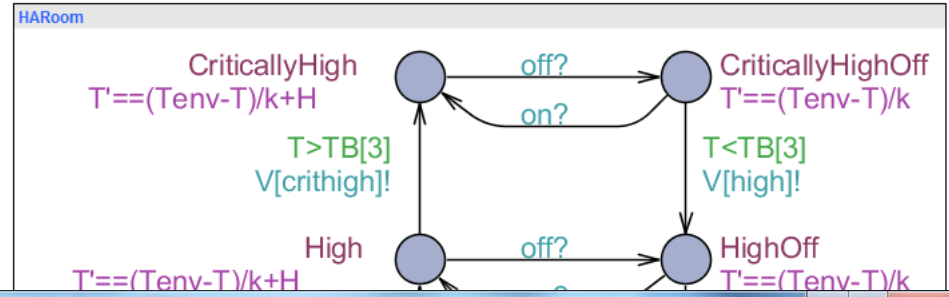
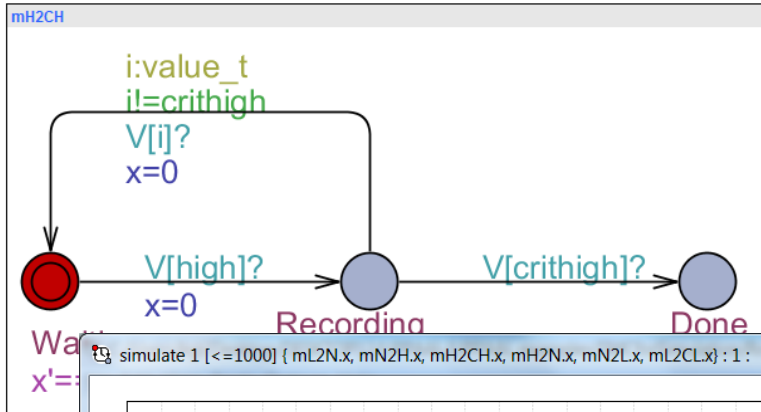
const int Tenv=7;
const int k=2;
const int H=20;
const int TB[4]=
    {12, 18, 25, 28};
    
```



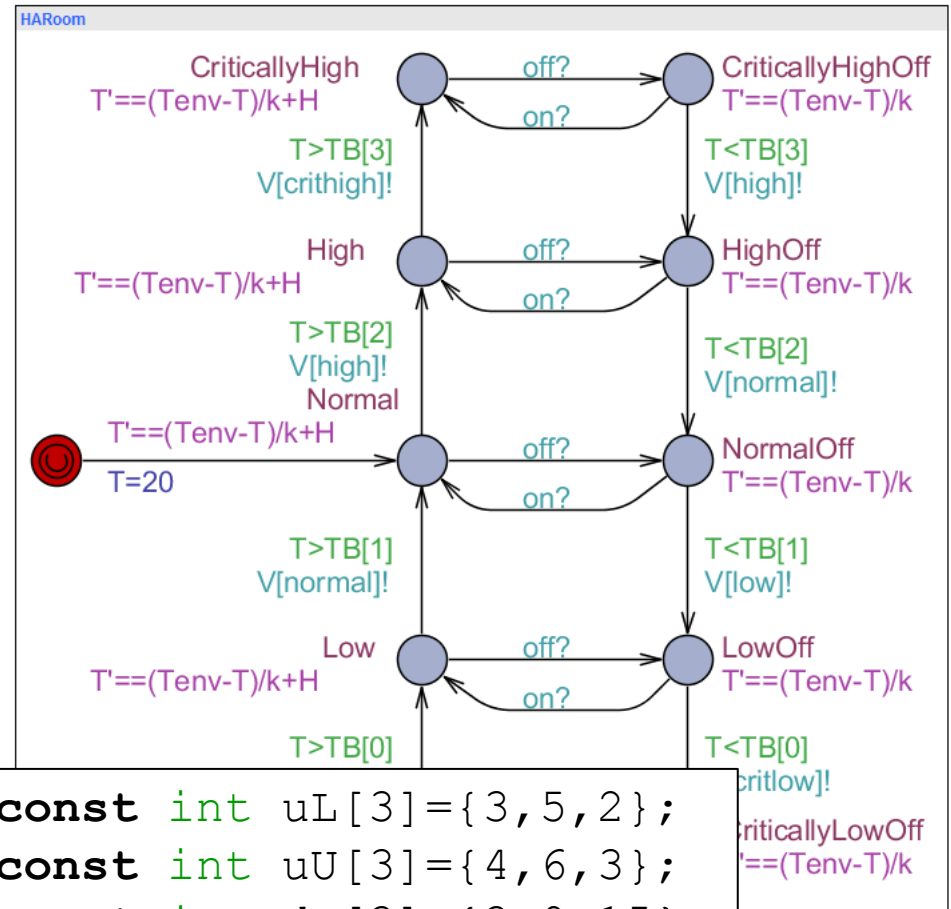
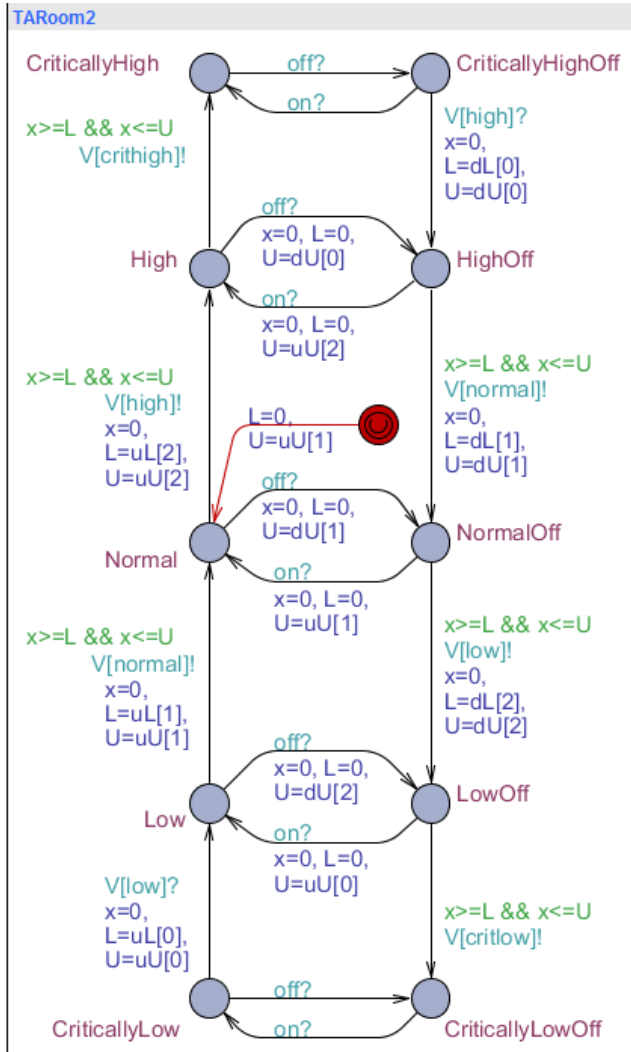
# Unfolding



# Timing



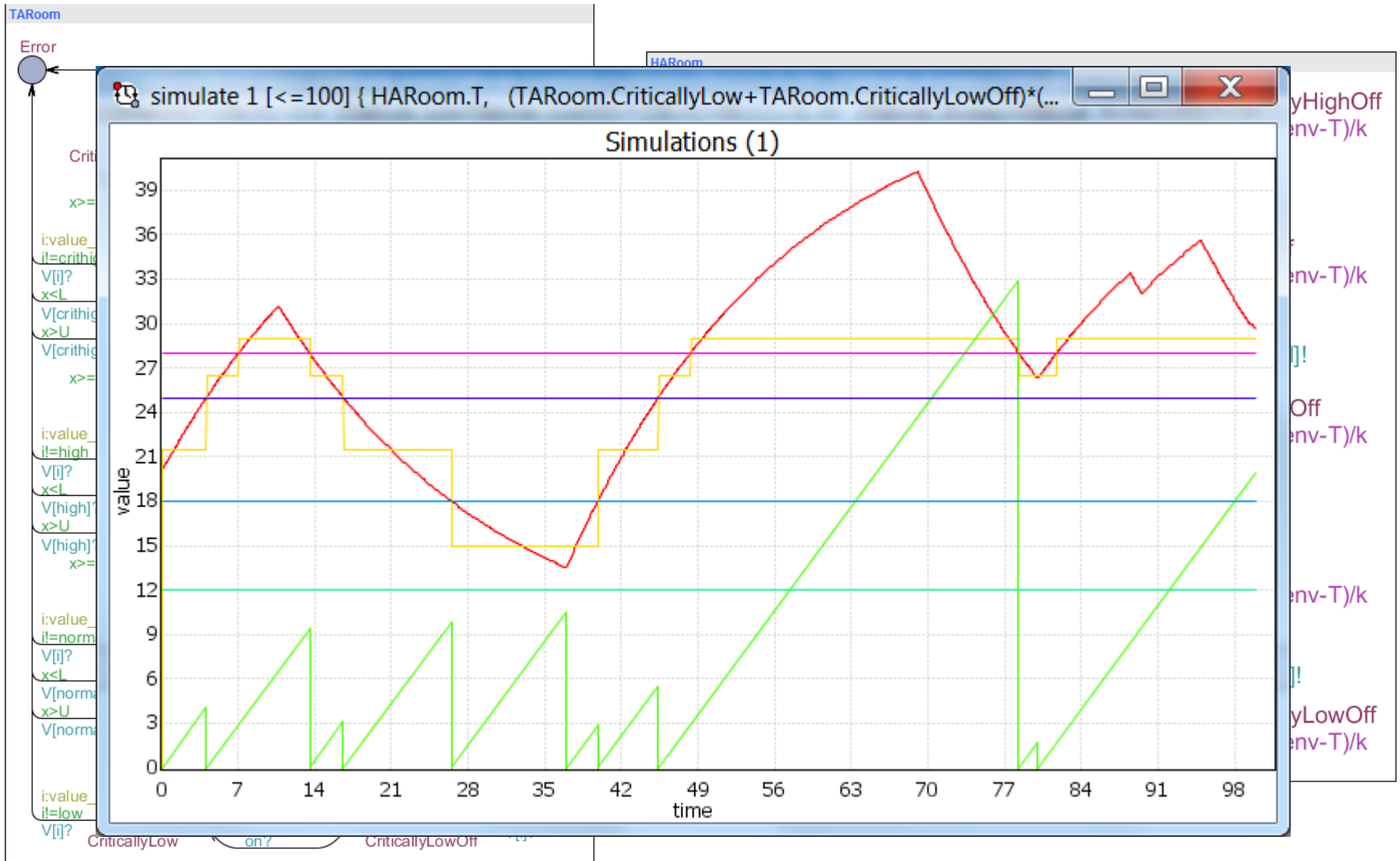
# TA Abstraction



```
const int uL[3] = {3, 5, 2};
const int uU[3] = {4, 6, 3};
const int dL[3] = {3, 9, 15};
const int dU[3] = {4, 10, 16};
```

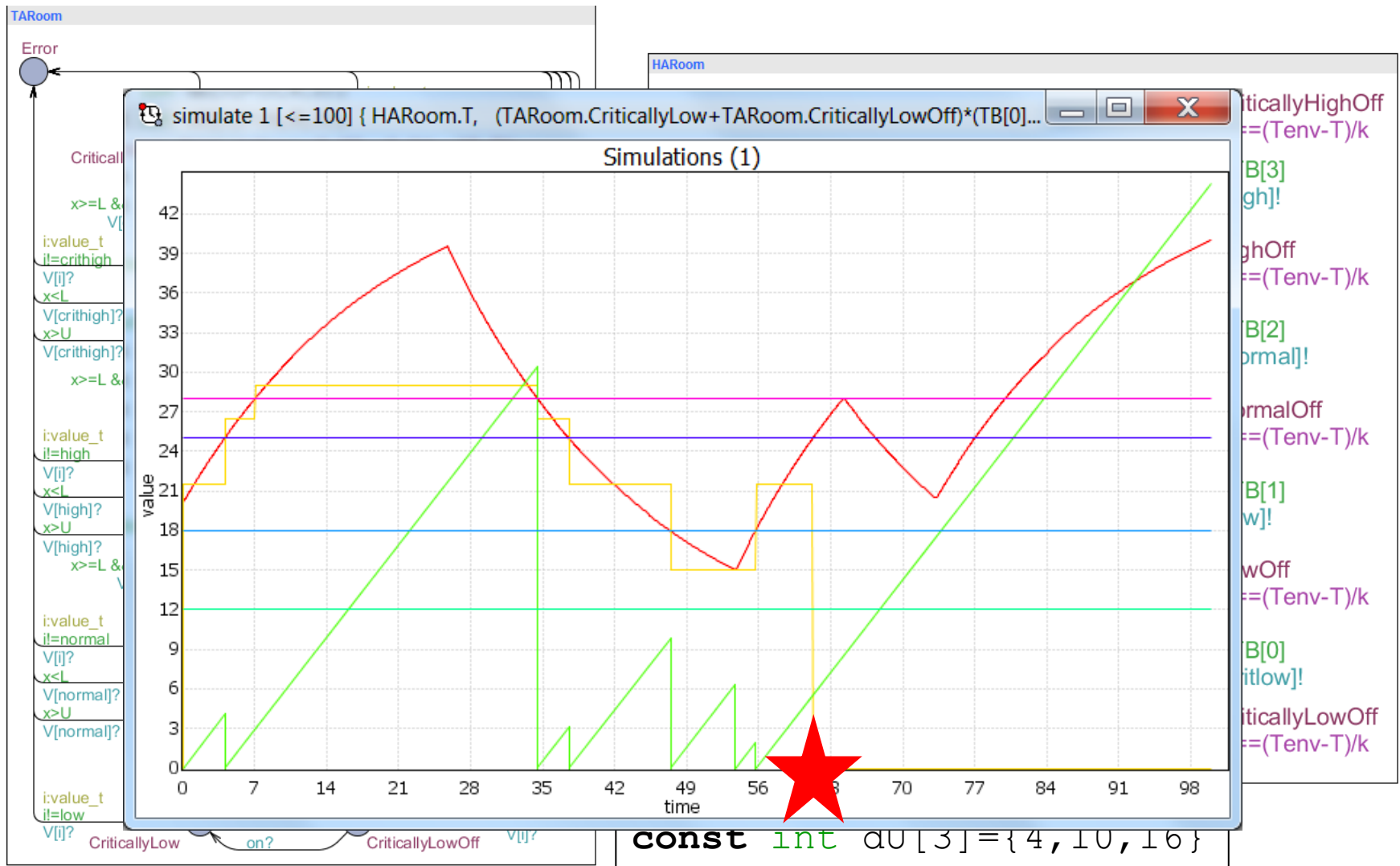


# Validation by co-Simulation



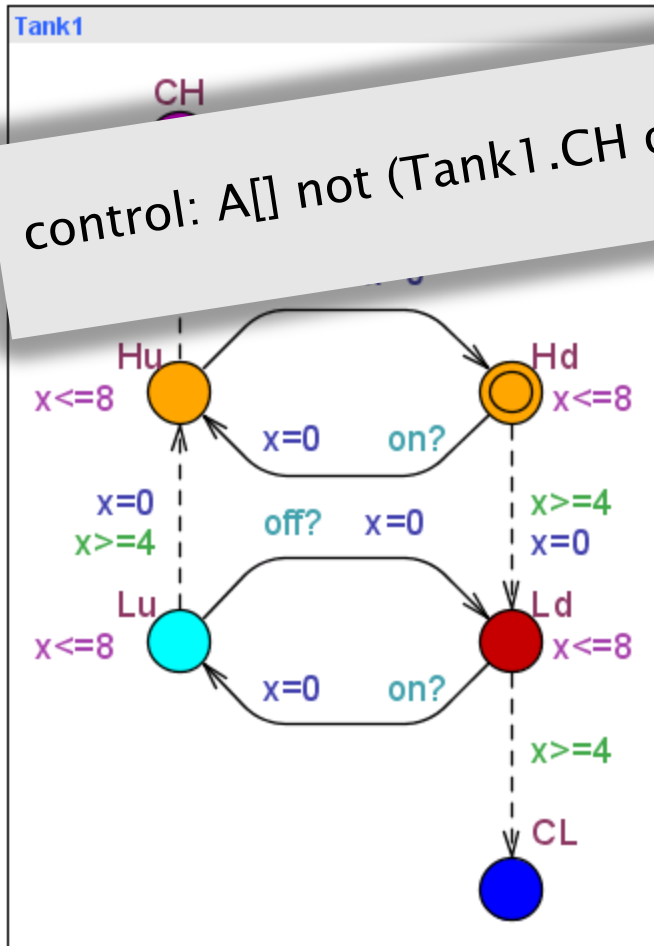


# Validation by co-Simulation



# Synthesis using TIGA

control: A[] not (Tank1.CH or Tank1.CL or Tank2.CH or Tank2.CL)

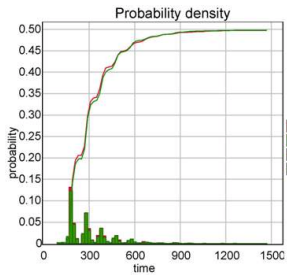


```

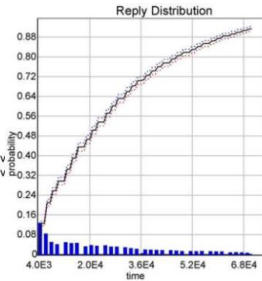
when you are in
    Controller.On->Controller.On { z == 1, tau, z := 0 }
    (Tank1.x==7 && Tank1.x-Tank2.x==5 && Tank2.x<=1 && Controller.z==1) ||
    (Tank1.x==8 && Tank1.x-Tank2.x==5 && Tank2.x-Controller.z==2 && Controller.z==1) ||
    (Tank1.x==5 && Tank1.x-Tank2.x==5 && Tank2.x-Controller.z==1 && Controller.z==1) ||
    (Tank1.x==7 && Tank1.x-Tank2.x==4 && Tank2.x-Controller.z==2 && Controller.z==1) ||
    (Tank1.x==6 && Tank1.x-Tank2.x==3 && Tank2.x-Controller.z==2 && Controller.z==1) ||
    (Tank1.x==6 && Tank1.x-Tank2.x==4 && Tank2.x-Controller.z==1 && Controller.z==1) ||
    (Tank1.x==5 && Tank1.x-Tank2.x==2 && Tank2.x-Controller.z==2 && Controller.z==1) ||
    (Tank1.x==5 && Tank1.x-Tank2.x==3 && Tank2.x-Controller.z==1 && Controller.z==1) ||
    (Tank1.x==5 && Tank1.x-Tank2.x==4 && Tank2.x==Controller.z && Controller.z==1) ||
    (Tank1.x==4 && Tank1.x-Tank2.x==1 && Tank2.x-Controller.z==2 && Controller.z==1) ||
    (Tank1.x==4 && Tank1.x-Tank2.x==2 && Tank2.x-Controller.z==1 && Controller.z==1) ||
    (Tank1.x==4 && Tank1.x-Tank2.x==3 && Tank2.x==Controller.z && Controller.z==1) ||
    (Tank1.x==3 && Tank1.x-Tank2.x==1 && Tank2.x-Controller.z==1 && Controller.z==1) ||
    (Tank1.x==3 && Tank1.x-Tank2.x==2 && Tank2.x==Controller.z && Controller.z==1) ||
    (Tank1.x==2 && Tank1.x-Tank2.x==1 && Tank2.x==Controller.z && Controller.z==1),
    take transition
    Controller.On->Controller.Off { z == 1, off!, z := 0 }
    Tank2.Hu->Tank2.Hd { 1, off?, x := 0 }
    ----
    State: ( Tank1.Lu Tank2.Ld Controller.on )
    when you are in
        (Tank1.x==8 && 1==Tank2.x && Tank1.x-Controller.z==7 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==7 && Tank1.x-Controller.z==6 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==6 && Tank1.x-Controller.z==5 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==5 && Tank1.x-Controller.z==4 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==6 && Tank1.x-Controller.z==3 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==3 && Tank1.x-Controller.z==2 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==2 && Tank1.x-Controller.z==1 && Tank2.x<3 && Controller.z==1) ||
        (Tank1.x==1 && Tank1.x==Controller.z && Tank2.x<3 && Controller.z==1),
        take transition
        Controller.On->Controller.Off { z == 1, off!, z := 0 }
        Tank1.Lu->Tank1.Ld { 1, off?, x := 0 }
        ----
        State: ( Tank1.Ld Tank2.Ld Controller.off )
        when you are in
            (Controller.z==1 && Tank1.x<2 && Tank2.x<4),
            take transition
    
```



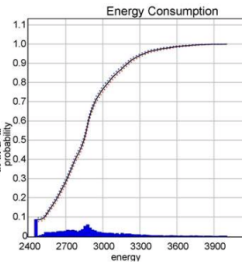
# Other Case Studies



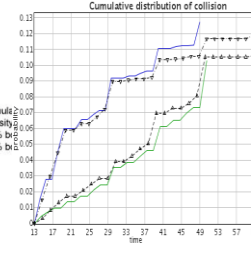
FIREWIRE



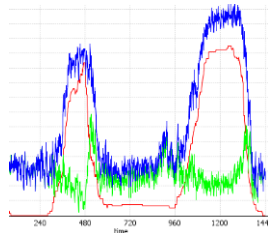
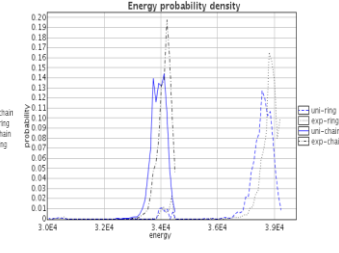
BLUETOOTH



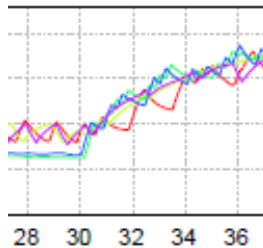
10 node LMAC



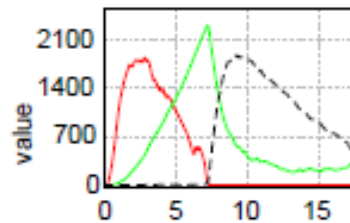
Schedulability  
Analysis for  
Mix Cr Sys



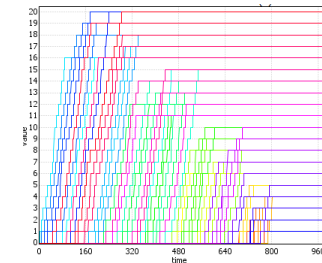
Smart Grid  
Demand /  
Response



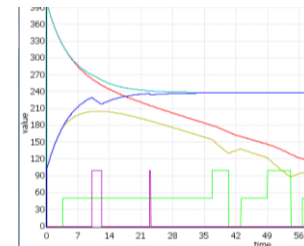
Energy Aware  
Buildings



Genetic Oscillator  
(HBS)



Passenger  
Seating in  
Aircraft



Battery  
Scheduling