

Inductive Verification of Hybrid Automata with Strongest Postcondition Calculus

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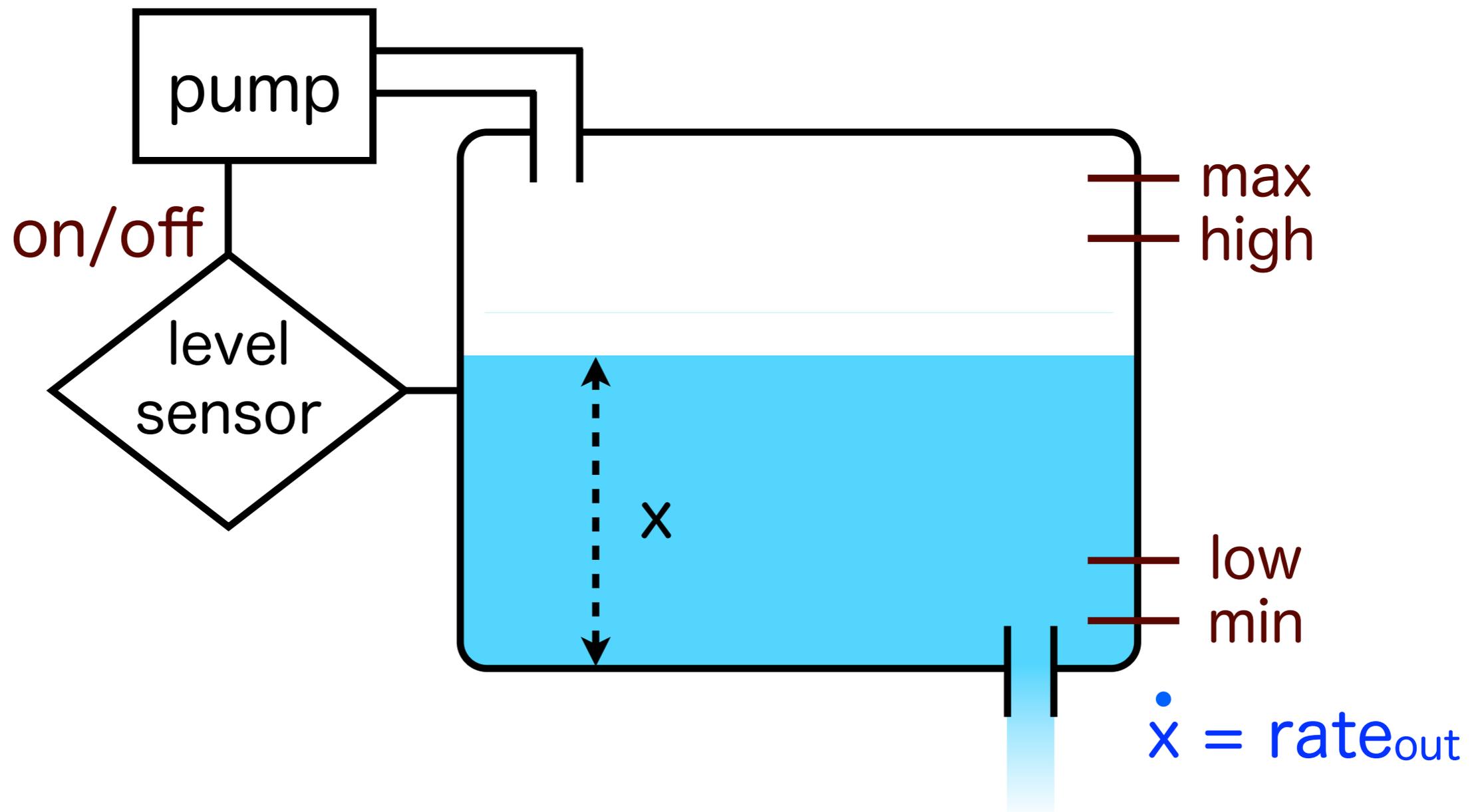
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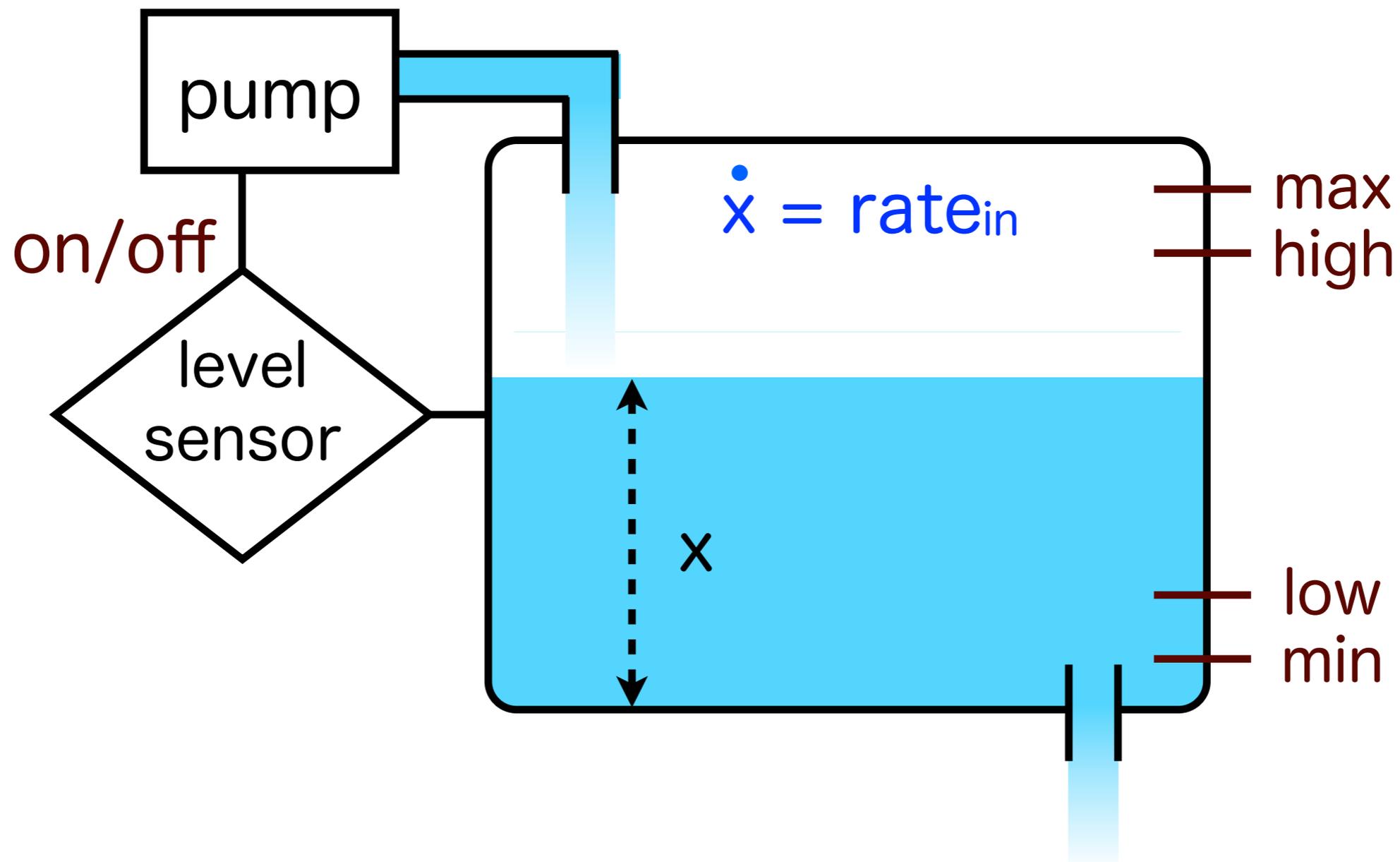
Hybrid Systems

- Systems whose states can make both continuous and discrete changes
- Example: [Water-level monitor](#)



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Verification of Hybrid Systems

- **Model-checking approach**
 - Based on **hybrid automata**
 - **Many practical automated tools:** e.g. HyTech [Henzinger+ 96] and PHAVer [Frehse 02]
 - **Tractable problems are limited:** small linear models, without uncertain parameters

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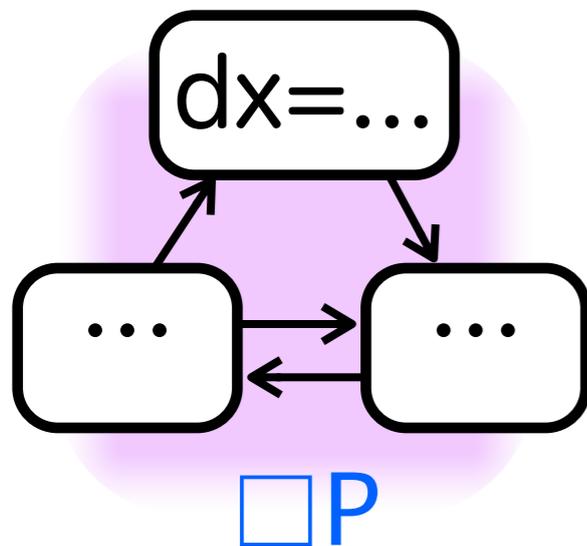
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- **Tractable problems are limited:** small linear models, without uncertain parameters

- **Logical analytic approach**

- Based on annotated **imperative programs**
- Theoretically studied but **few practical tools**
- **KeYmaera** [Platzer+ 08] has been successful
- **Applicable to larger class of systems**
 - *systems with symbolic parameters, nonlinear systems
- **Verification process is not fully automatic**

Talk Overview

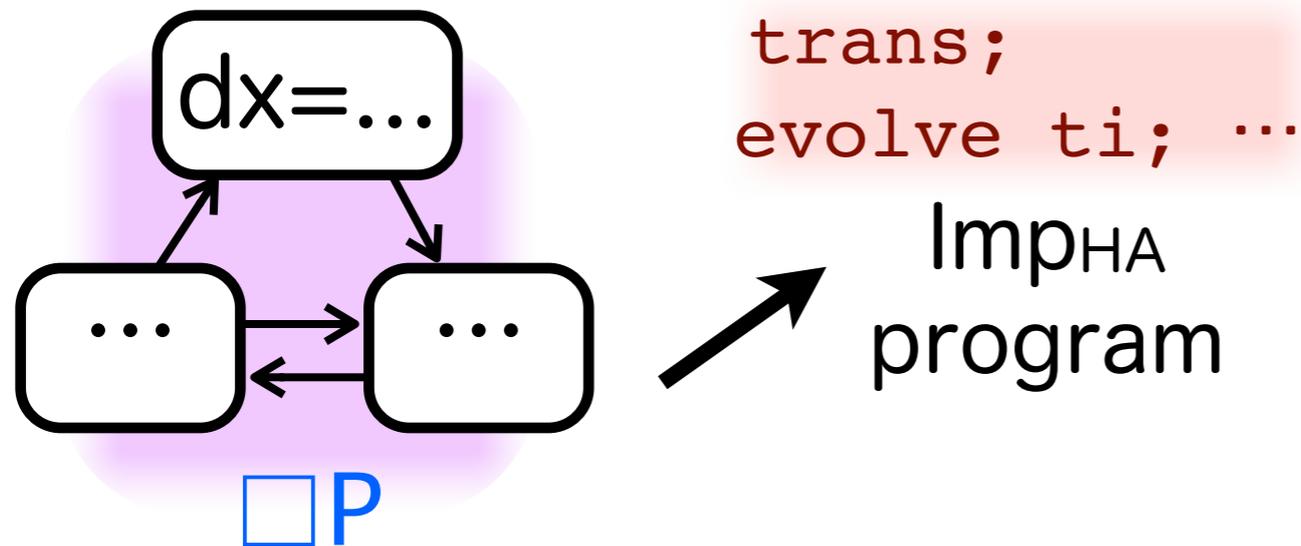
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 - **Clear and correct scheme:**
conjunction of simple transformations
 - **Applicable to a large class of hybrid automata**
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safety verification of
hybrid automaton

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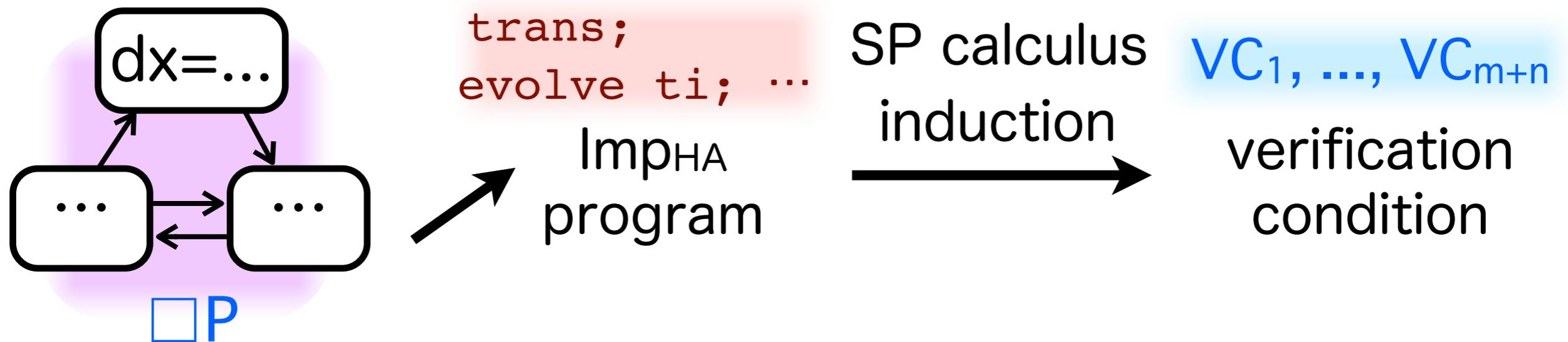
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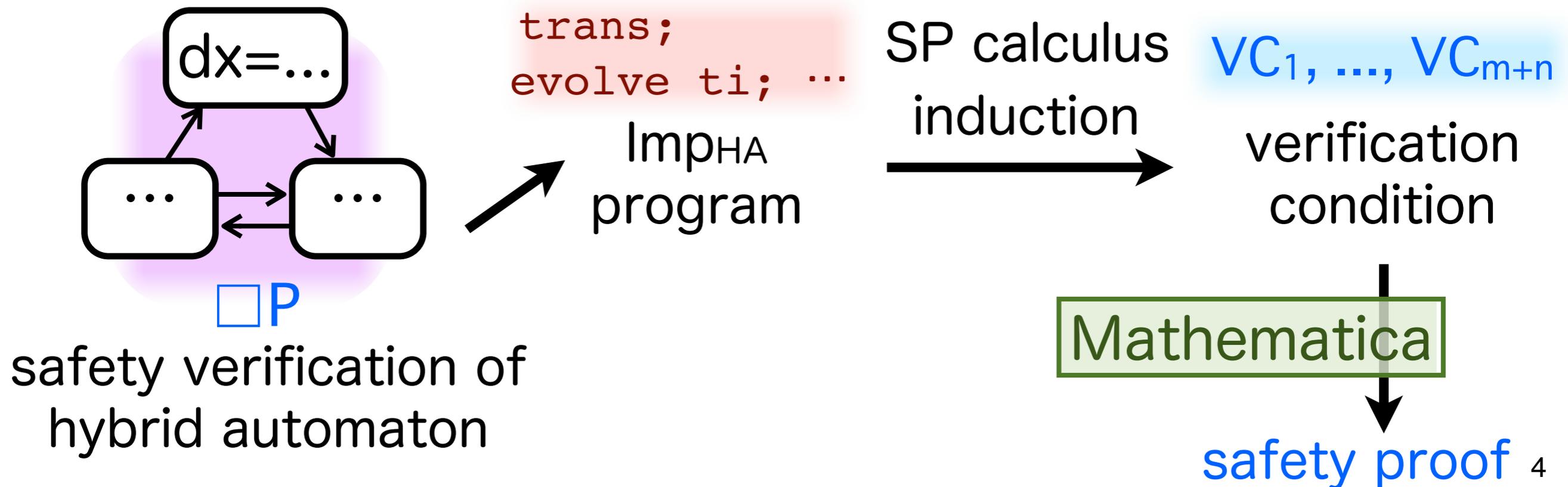
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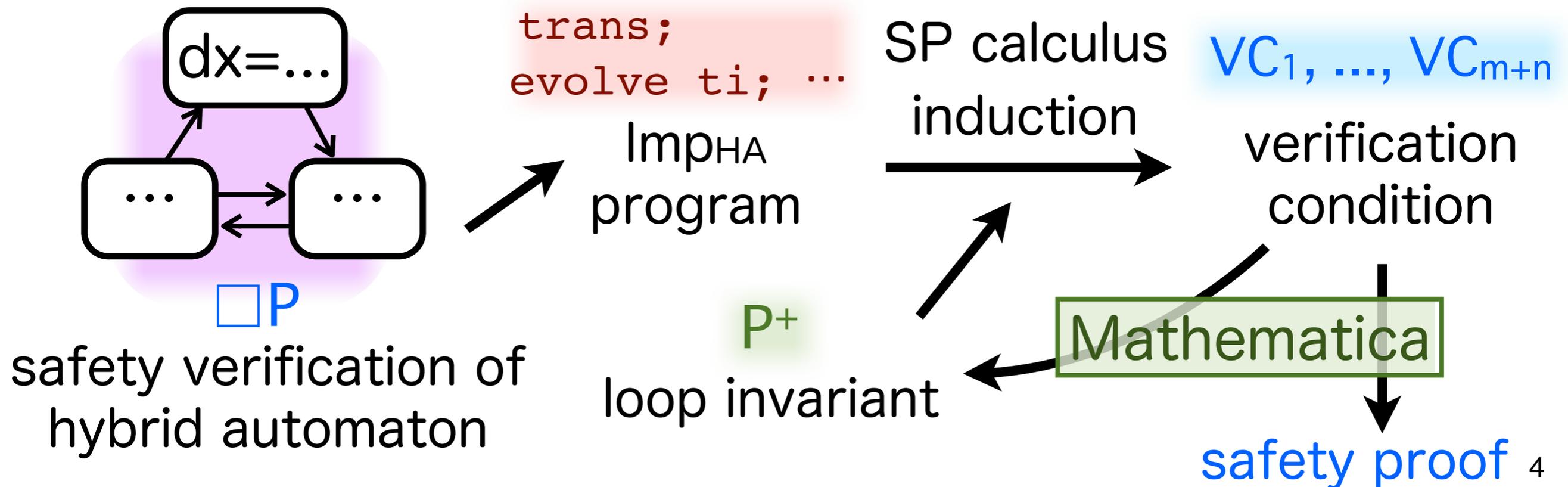
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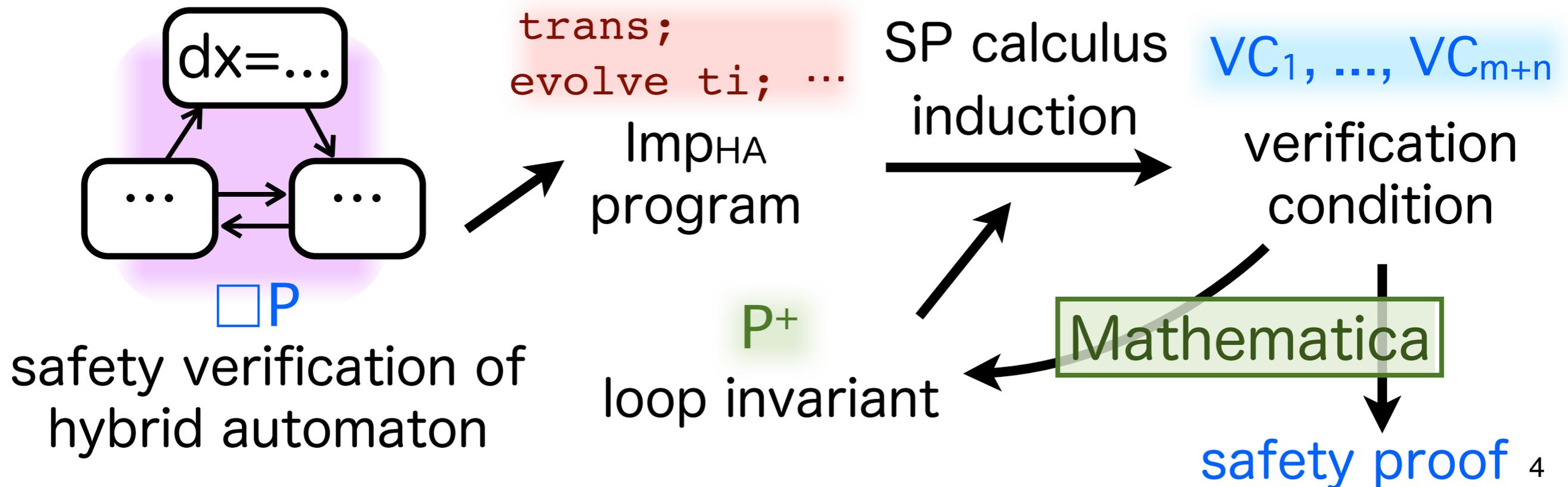
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Talk Outline

1. Hybrid Automata

2. Imp_{HA} and Strongest Postcondition Calculus

3. Inductive Verification

4. Experimental Results

Hybrid Automata (HA)

- **Mathematical model of hybrid systems**
 - Discrete aspect is described by an automaton
 - Continuous dynamics is described by a differential equation indexed by locations of the automaton

Hybrid Automata (HA)

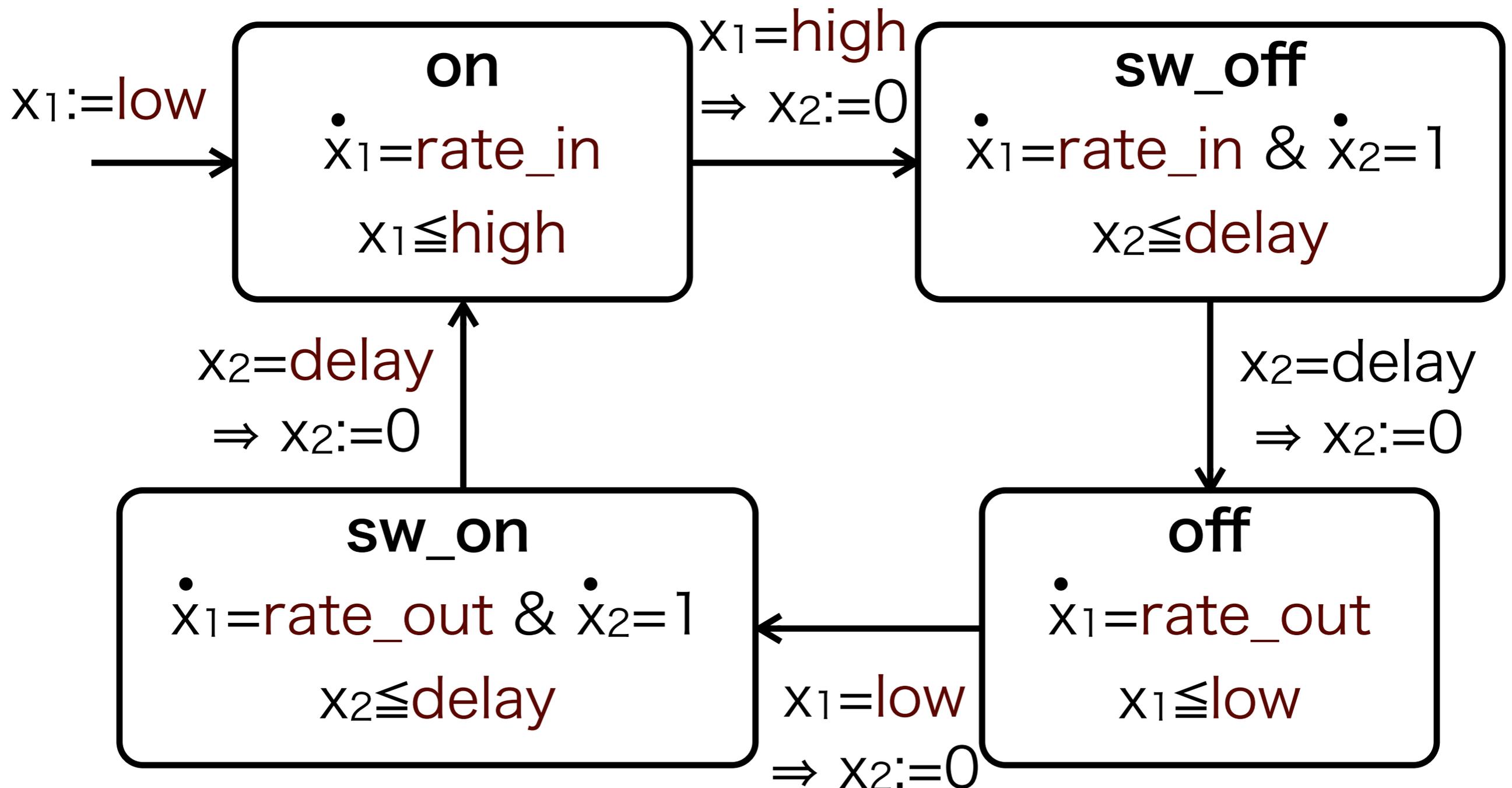
- A **hybrid automaton** is a septet

$$HA = \langle \text{Loc}, \text{Var}, \text{Init}, \text{Grd}, \text{Rst}, \text{Flow}, \text{Inv} \rangle$$

that consists of:

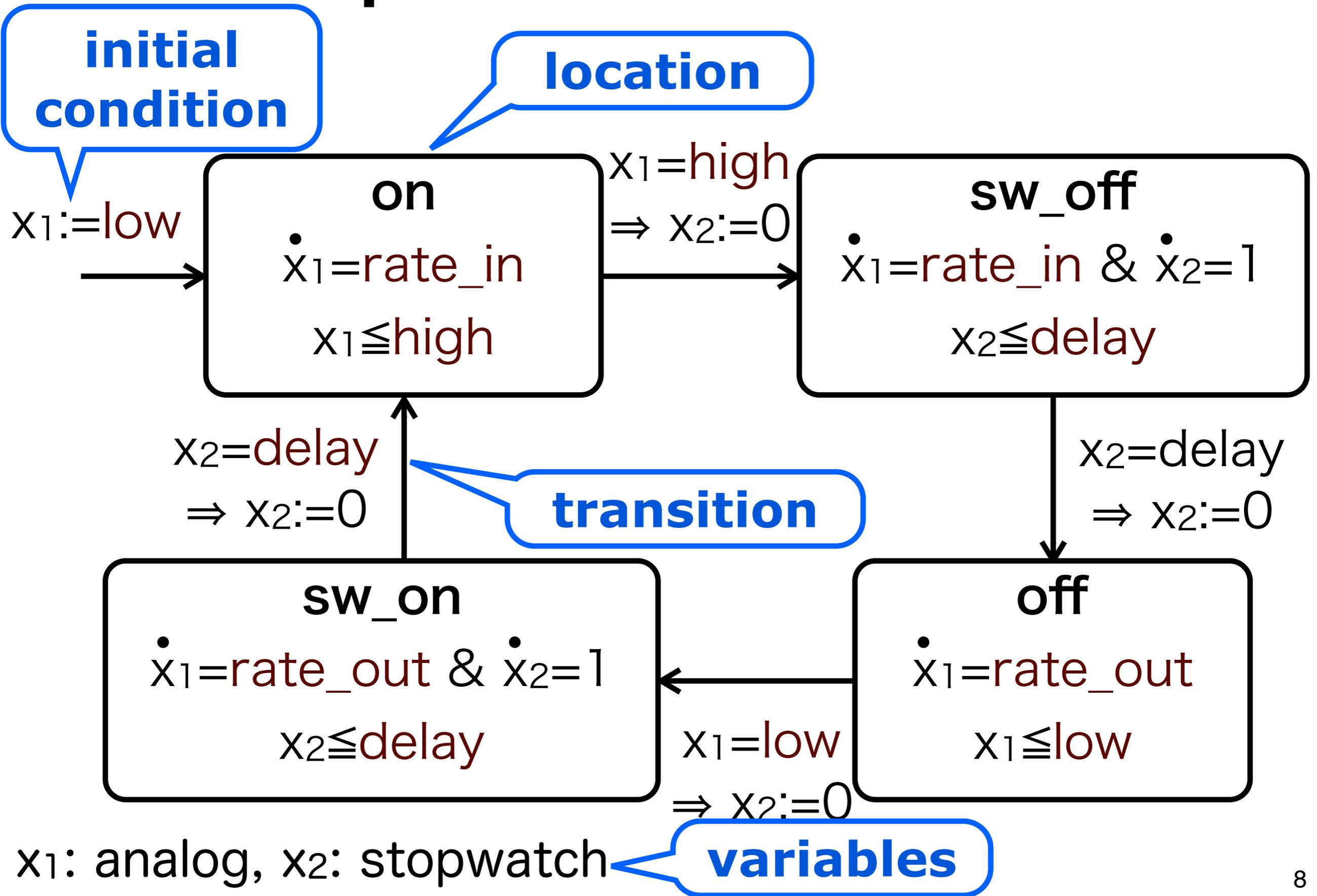
- Finite set $\text{Loc} = \{L_1, \dots, L_p\}$ of **locations**
- Finite set $\text{Var} = \{x_1, \dots, x_q\}$ of real-valued **variables**
- **Initial condition** Init in $L \times \mathbf{R}^{\text{Var}}$
- Family $\mathbf{Grd} = \{\text{Grd}_{L,L'}\}_{L \in \text{Loc}, L' \in \text{Loc}}$ of **guard conditions**
 $\text{Grd}_{L,L'}$ in \mathbf{R}^{Var}
- Family $\mathbf{Rst} = \{\text{Rst}_{L,L'}\}_{L \in \text{Loc}, L' \in \text{Loc}}$ of **reset functions**
 $\text{Rst}_{L,L'} : \mathbf{R}^{\text{Var}} \rightarrow \mathbf{R}^{\text{Var}}$
- Family $\mathbf{Flow} = \{\text{Flow}_L\}_{L \in \text{Loc}}$ of **vector fields**
 $\text{Flow}_L : \mathbf{R}^{\text{Var}} \rightarrow \mathbf{R}^{\text{Var}}$
- Family $\mathbf{Inv} = \{\text{Inv}_L\}_{L \in \text{Loc}}$ of **location invariants**
 Inv_L in \mathbf{R}^{Var}

Example: Water-level Monitor

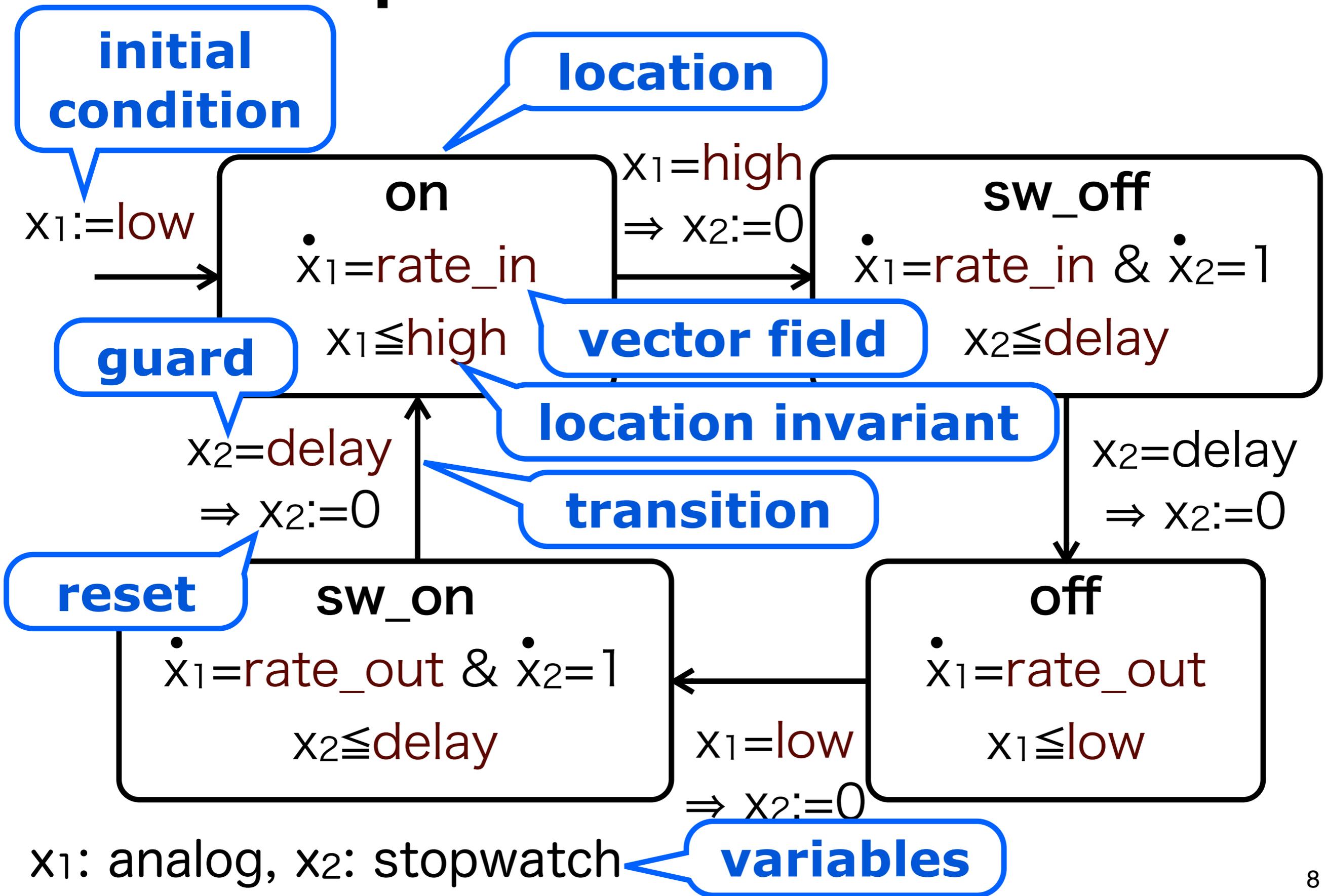


x_1 : analog, x_2 : stopwatch

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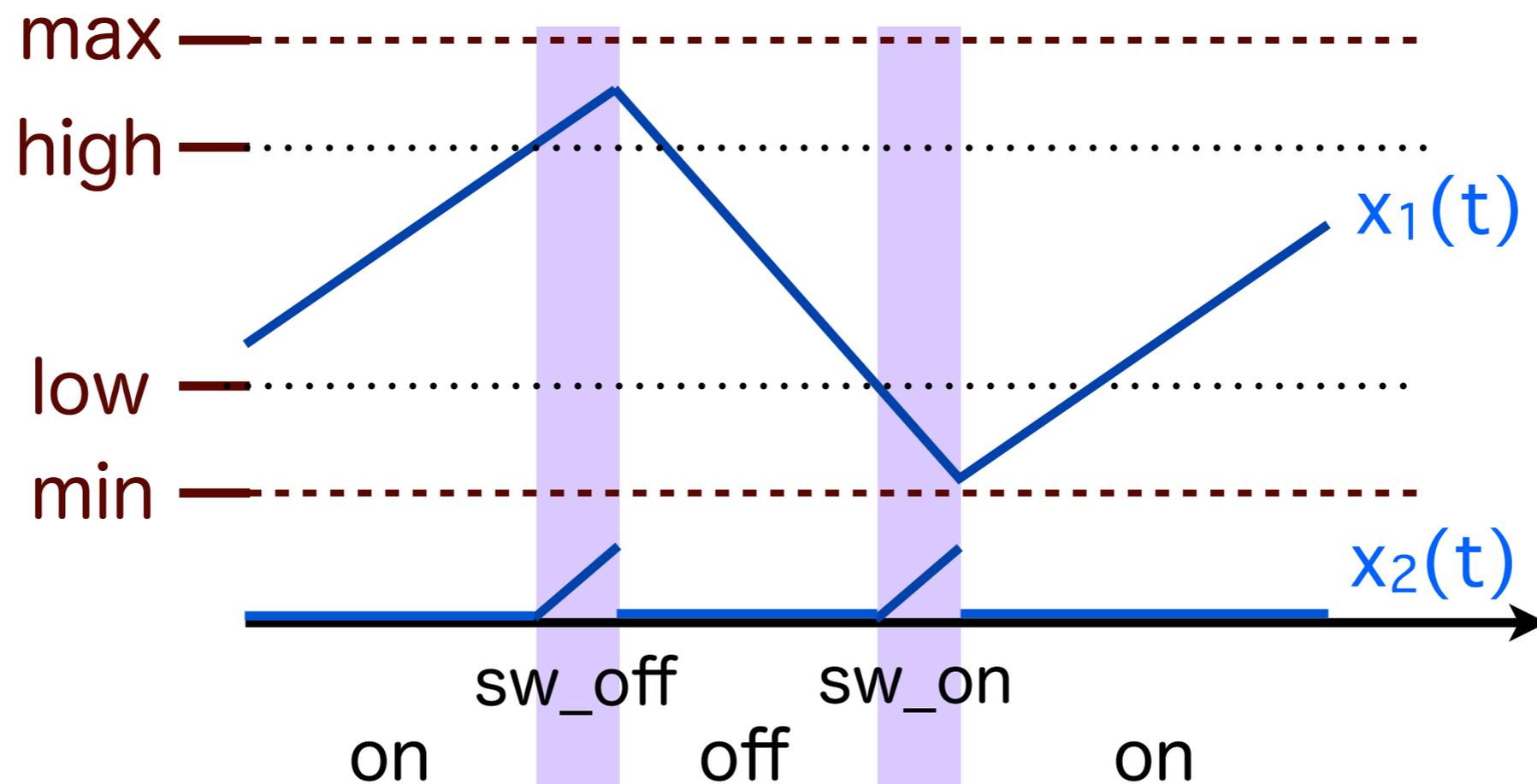


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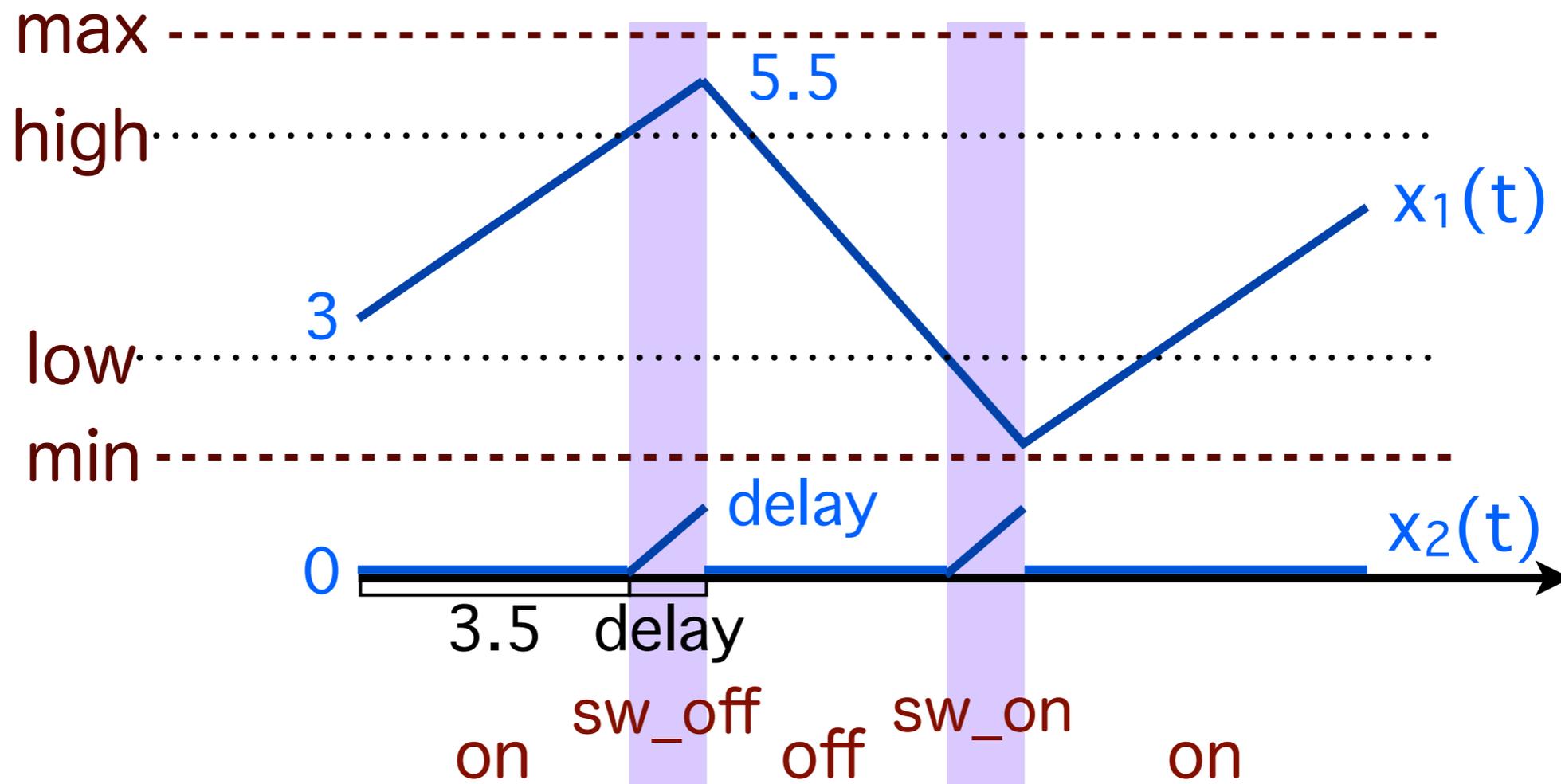
Execution of Water-level Monitor

- Two rates of the water flow: **rate_in** and **rate_out**
- The controller tries to turn on/off the pump when the water level reaches **low/high**
- It takes **delay** seconds to turn on/off the pump



Execution of HA

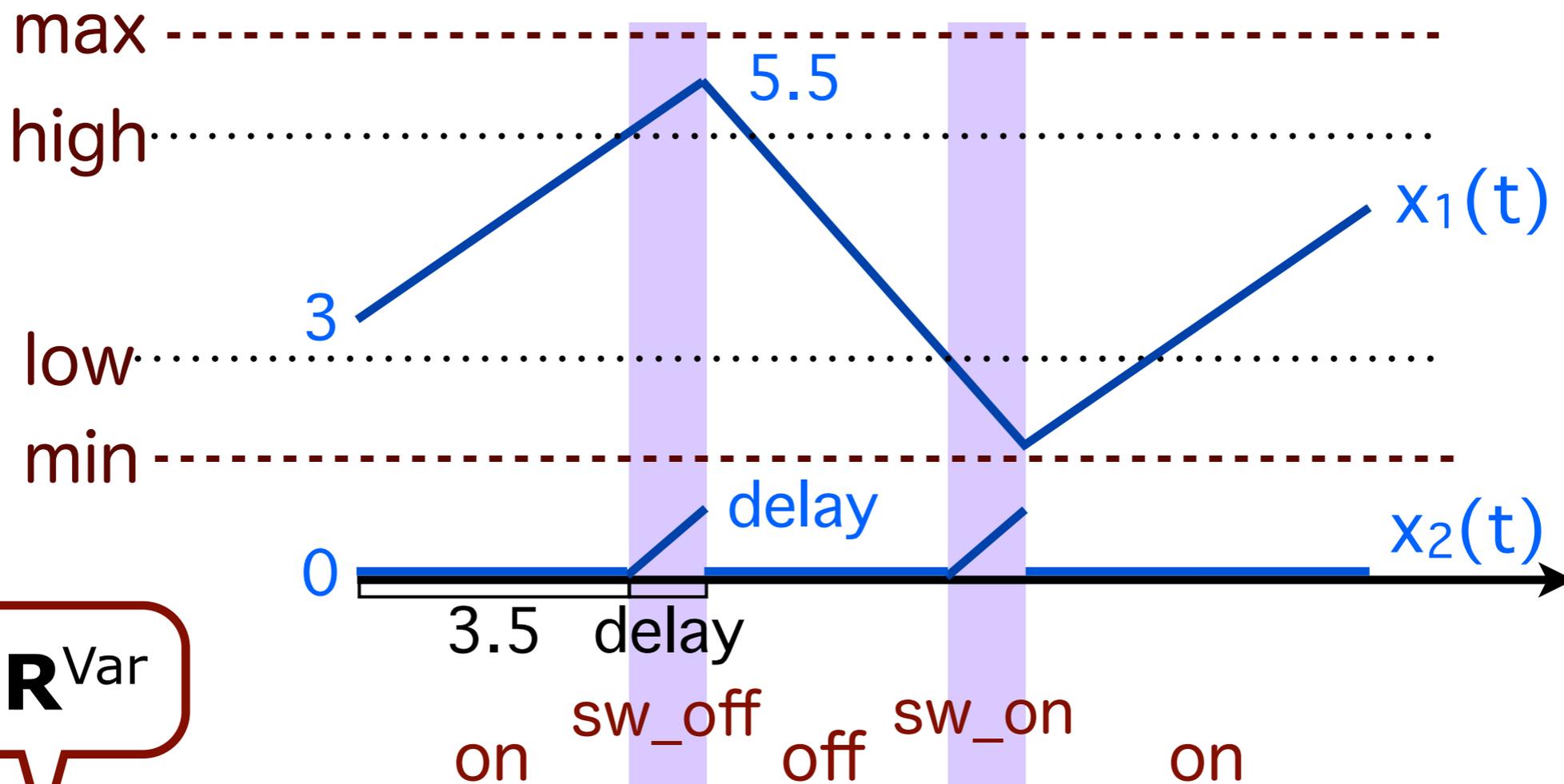
- (Finite- or infinite-length) **execution** of a HA is a valuation of variables as functions over time



$\langle \text{on}, (3, 0) \rangle \xrightarrow{3.5} \langle \text{on}, (\text{high}, 0) \rangle \xrightarrow{0} \langle \text{sw_off}, (\text{high}, 0) \rangle \xrightarrow{\text{delay}}$
 $\langle \text{sw_off}, (5.5, \text{delay}) \rangle \xrightarrow{0} \dots$

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$\text{Loc} \times \mathbf{R}^{\text{Var}}$

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continuous evolution for duration 3.5

discrete transition

Operational Semantics of HA

- Continuous evolution

$$\frac{t > 0 \quad \phi(0) = v \quad \forall \tilde{t} \in [0, t] \quad \frac{d\phi}{d\tilde{t}} = \text{Flow}_L(\phi(\tilde{t})) \wedge \text{Inv}_L[\phi(\tilde{t})]}{\langle L, v \rangle \xrightarrow{t} \langle L, \phi(t) \rangle}$$

- Discrete transition

$$\frac{\text{Grd}_{L_1, L_2}[v_1] \quad v_2 = \text{Rst}_{L_1, L_2}(v_1) \quad \text{Inv}_{L_2}[v_2]}{\langle L_1, v_1 \rangle \xrightarrow{0} \langle L_2, v_2 \rangle}$$

Safety Verification of HA

- **Safety property** $\square P$ means:
 - P holds initially, and is preserved by every continuous evolution and discrete transition
 - i.e. invariance
 - Example: $\square(\min \leq x_1 \wedge x_1 \leq \max)$

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Modeling HA Executions with Programs

- **Given a HA, we consider a **simple imperative language Imp_{HA}****
- Program's states correspond to the states in the HA executions
- Motivation: reuse of traditional program verification tools

Modeling HA Executions with Programs

- **Given a HA, we consider a **simple imperative language Imp_{HA}****
- Program's states correspond to the states in the HA executions
- Motivation: reuse of traditional program verification tools
- **We provide a notion of **strongest-postcondition (SP)** for each Imp_{HA} statements**
- Cf. forward reachability analysis

Imp_{HA} Language

- Imp_{HA} is a simple **imperative language** for sketching an execution of the HA

- $S ::= \text{skip} \mid S; S \mid \text{evolve } t \mid \text{trans}$

- Consider **implicit variables**

$$X_S = \langle X_L, X_V \rangle = \langle X_L, (X_1, \dots, X_q) \rangle$$

that express the “current state” in $\text{Loc} \times \mathbf{R}^{\text{Var}}$ of the HA execution

- **Program state** S is a map from variable names to program values

- Example: $S = \{x_L \mapsto \text{on}, x_1 \mapsto 3.1, x_2 \mapsto 0\}$

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Strongest Postcondition Calculus

- **Lemma 3.** For any program s in Imp_{HA} , if the initial state satisfies P , the final state satisfies $SP(P, s)$ with SP defined as follows:

$$SP(P, \text{skip}) := P$$

$$SP(P, s_1 ; s_2) := SP(SP(P, s_1), s_2)$$

$$SP(P, \text{evolve } t) := \exists \phi P[x_v \leftarrow \phi(0)] \wedge \\ \phi(t) = x_v \wedge (\forall \tilde{t} \in [0, t] \frac{d\phi}{d\tilde{t}} = \text{Flow}_{x_L}(\phi(\tilde{t})) \wedge \text{Inv}_{x_L}[\phi(\tilde{t})])$$

$$SP(P, \text{trans}) := \exists \langle L', x_v' \rangle P[x_s \leftarrow \langle L', x_v' \rangle] \wedge \\ \text{Grd}_{L', x_L}[x_v'] \wedge x_v = \text{Rst}_{L', x_L}(x_v') \wedge \text{Inv}_{x_L}[x_v]$$

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derived from the operational semantics of HA

Example: Water-level Monitor

1. Consider

$$SP((x_L = \text{on} \wedge x_1 = \text{low}), \text{evolve } t) \Rightarrow x_1 \leq \text{max}$$

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2. From the definition of SP

$$(\exists \phi (x_L = \text{on} \wedge \phi_{x_1}(0) = \text{low}) \wedge$$

$$\phi(t) = (x_1, x_2) \wedge (\forall \tilde{t} \in [0, t] \frac{d\phi}{d\tilde{t}} = \text{Flow}_{x_L}(\phi(\tilde{t})) \wedge \text{Inv}_{x_L}[\phi(\tilde{t})])) \\ \Rightarrow x_1 \leq \text{max}$$

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3. We can solve the ODE into a closed form and simplify as

$$\forall \tilde{t} \in [0, t] \text{Inv}_{x_L}[(\text{low} + \text{rate}_{in} \tilde{t}, x_2)]$$

$$\Rightarrow \text{low} + \text{rate}_{in} t \leq \text{max}$$

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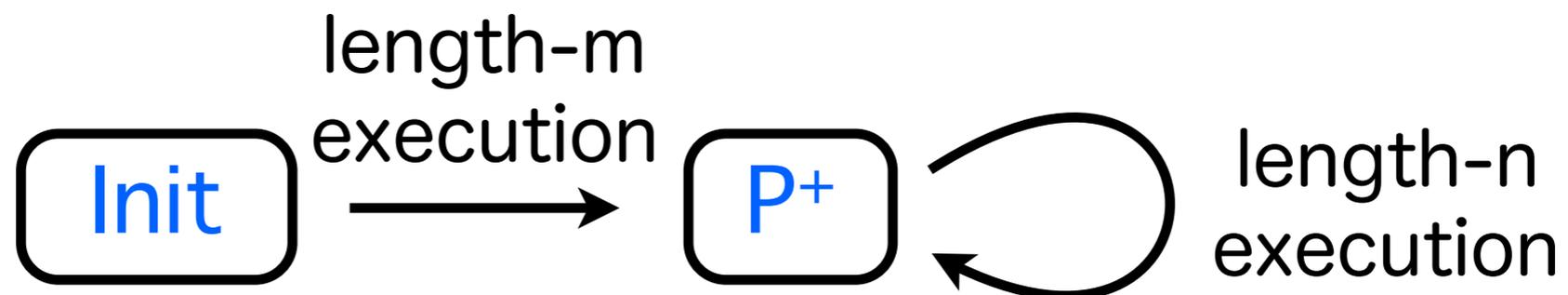
Inductive Verification

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- Encoding **safety obligation for all the** (infinitely-many) **executions** of the HA into a **bounded number of verification conditions**
- **Induction strategy**
 - Inspired by deductive program verification technique for handling loops
 - **Construction of a *lasso-shaped structure***

*Assume a loop invariant P^+



Induction Strategy (Simplest Case)

- **Theorem 1.** Given a predicate P^+ such that $P^+ \Rightarrow P$, the following inference rule is correct

$$\begin{array}{l} \text{VC}_0: \quad \text{Init} \Rightarrow P^+ \\ \text{VC}_1: \quad \forall t \geq 0 \text{ SP}(P^+, \text{evolve } t) \Rightarrow P \\ \text{VC}_{-1}: \quad \forall t \geq 0 \text{ SP}(P^+, \text{evolve } t; \text{trans}) \Rightarrow P^+ \\ \hline \text{HA} \models \Box P \end{array}$$

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- **Proof sketch**

- VC_0 checks that all the initial states satisfy P^+
- VC_{-1} verifies that all the two step executions

$$\sigma_i \xrightarrow{t_{i+1}} \sigma_{i+1} \xrightarrow{0} \sigma_{i+2}$$

from a state σ_i that satisfies P^+ evolve for a duration t_{i+1} to a state σ_{i+2} that again satisfies P^+

- VC_1 ensures that P was not broken during the continuous evolution

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- Verification is possible with the inference rule where the loop invariant is set as

$$P^+ \equiv x_L = \text{sw_off} \Rightarrow x_1 = \text{high} \wedge x_L = \text{sw_on} \Rightarrow x_1 = \text{low}$$

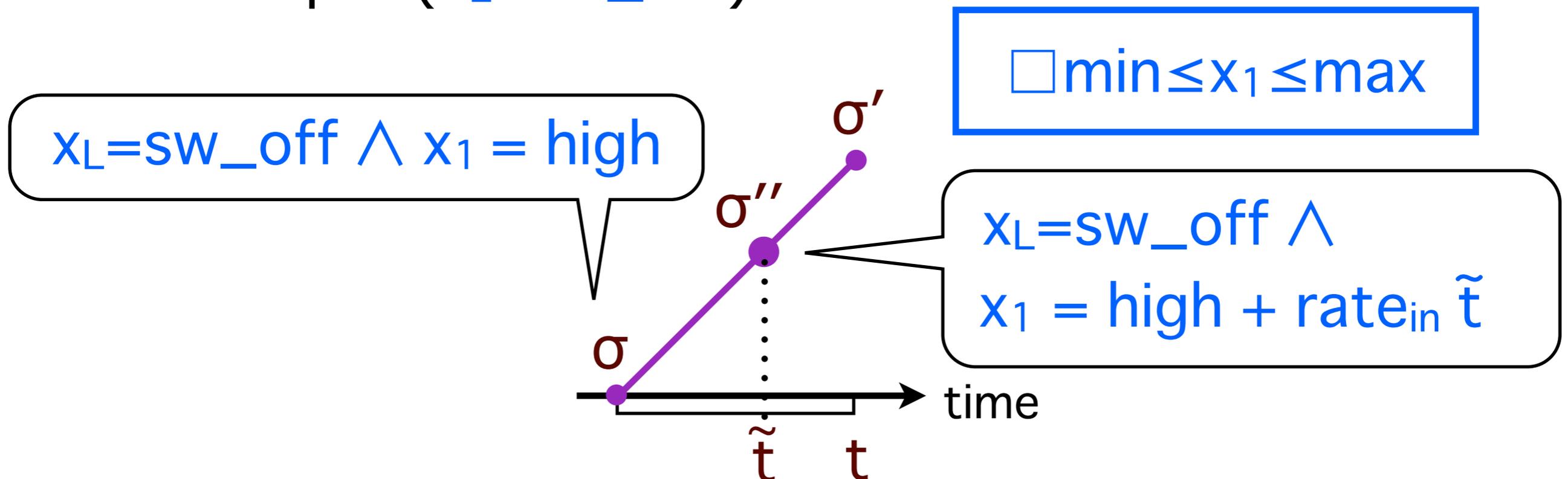
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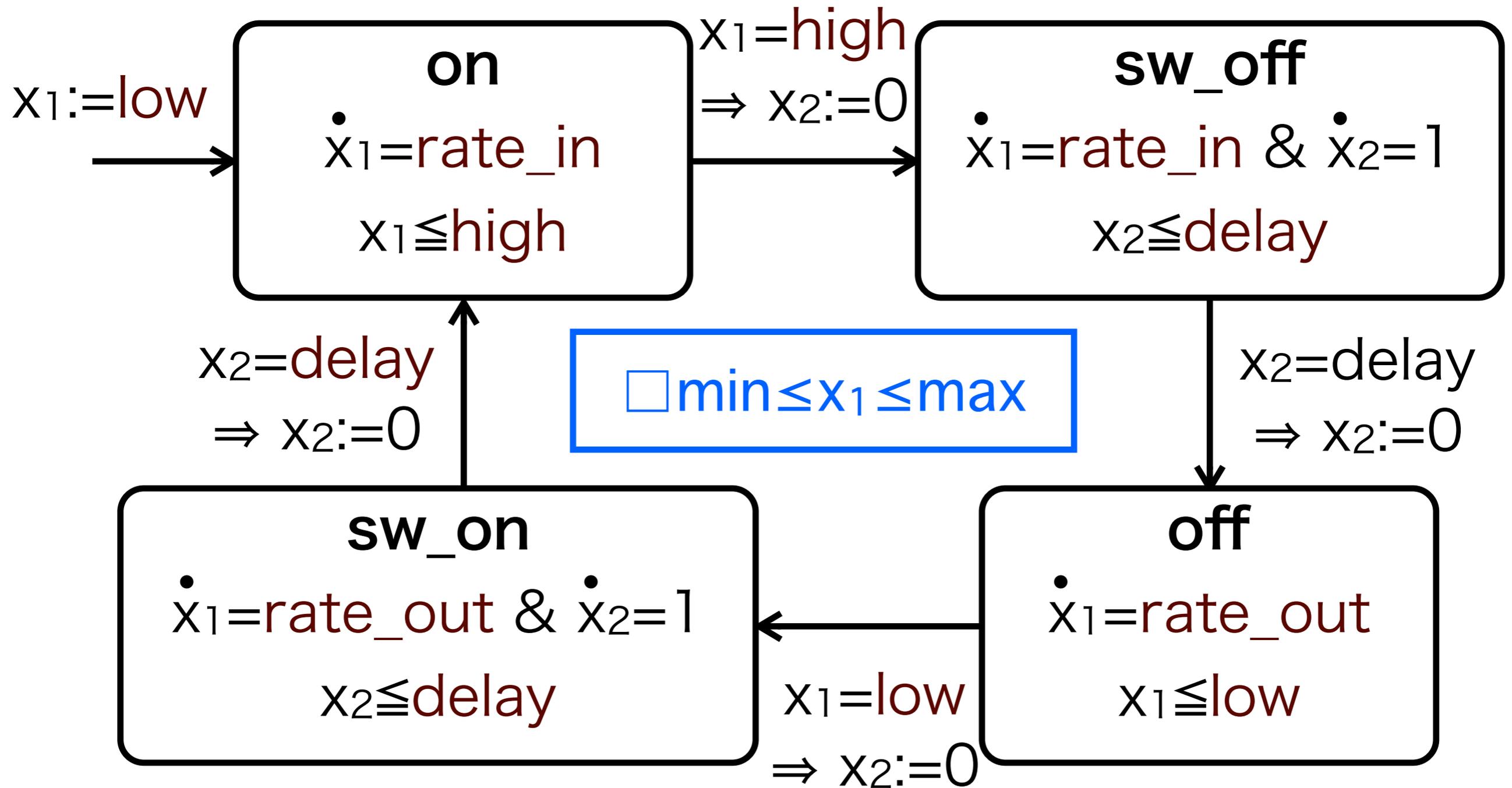
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- This loop invariant is necessary to verify the safety during the continuous evolution

- Example ($x_L = \text{sw_off}$):



Example: Water-level Monitor



Induction Strategy (Unrolled Case)

- **Theorem 2.** Given a predicate P^+ such that $P^+ \Rightarrow P$, consider the following $m+n+2$ SPs...

$$SP_1 \equiv SP(\text{Init}, \text{evolve } t_1)$$

$$SP_2 \equiv SP(SP(SP_1, \text{trans}), \text{evolve } t_2)$$

$$\vdots$$

$$SP_m \equiv SP(SP(SP_{m-1}, \text{trans}), \text{evolve } t_m)$$

$$SP_0 \equiv SP(SP_m, \text{trans})$$

$$SP_{m+1} \equiv SP(P^+, \text{trans}), \text{evolve } t_1)$$

$$SP_{m+2} \equiv SP(SP(SP_{m+1}, \text{trans}), \text{evolve } t_2)$$

$$\vdots$$

$$SP_{m+n} \equiv SP(SP(SP_{m+n-1}, \text{trans}), \text{evolve } t_n)$$

$$SP_{-1} \equiv SP(SP_{m+n}, \text{trans})$$

Induction Strategy (Unrolled Case)

- **Theorem 2 (cont).** Then, the following inference rule is correct

$$\begin{array}{l} VC_1: \quad \forall t_1 \geq 0 \ SP_1 \Rightarrow P \\ VC_2: \quad \forall t_1, t_2 \geq 0 \ SP_2 \Rightarrow P \\ \quad \quad \quad \vdots \\ VC_m: \quad \forall t_1..t_m \geq 0 \ SP_m \Rightarrow P \\ VC_0: \quad \forall t_1..t_m \geq 0 \ SP_0 \Rightarrow P^+ \\ \hline VC_{m+1}: \quad \forall t_1 \geq 0 \ SP_{m+1} \Rightarrow P \\ \quad \quad \quad \vdots \\ VC_{m+n}: \quad \forall t_1..t_n \geq 0 \ SP_{m+n} \Rightarrow P \\ VC_{-1}: \quad \forall t_1..t_n \geq 0 \ SP_{-1} \Rightarrow P^+ \\ \hline HA \models \square P \end{array}$$

Induction Strategy (Unrolled Case)

- **Proof sketch**

- VC_0 checks that all the initial states σ_0 reach a state σ_{2m} that satisfy P^+ after m steps

$$\sigma_0 \xrightarrow{t_1} \dots \xrightarrow{0} \sigma_{2m}$$

- VC_{-1} verifies that all the n step executions

$$\sigma_{2m} \xrightarrow{t_1} \dots \xrightarrow{0} \sigma_{2m+2n}$$

from a state σ_{2m} that satisfies P^+ reaches a state σ_{2m+2n} that again satisfies P^+

- Other VCs i.e. VC_1, \dots, VC_{m+n} ensure that P was not broken before/after a discrete transition and during a continuous evolution

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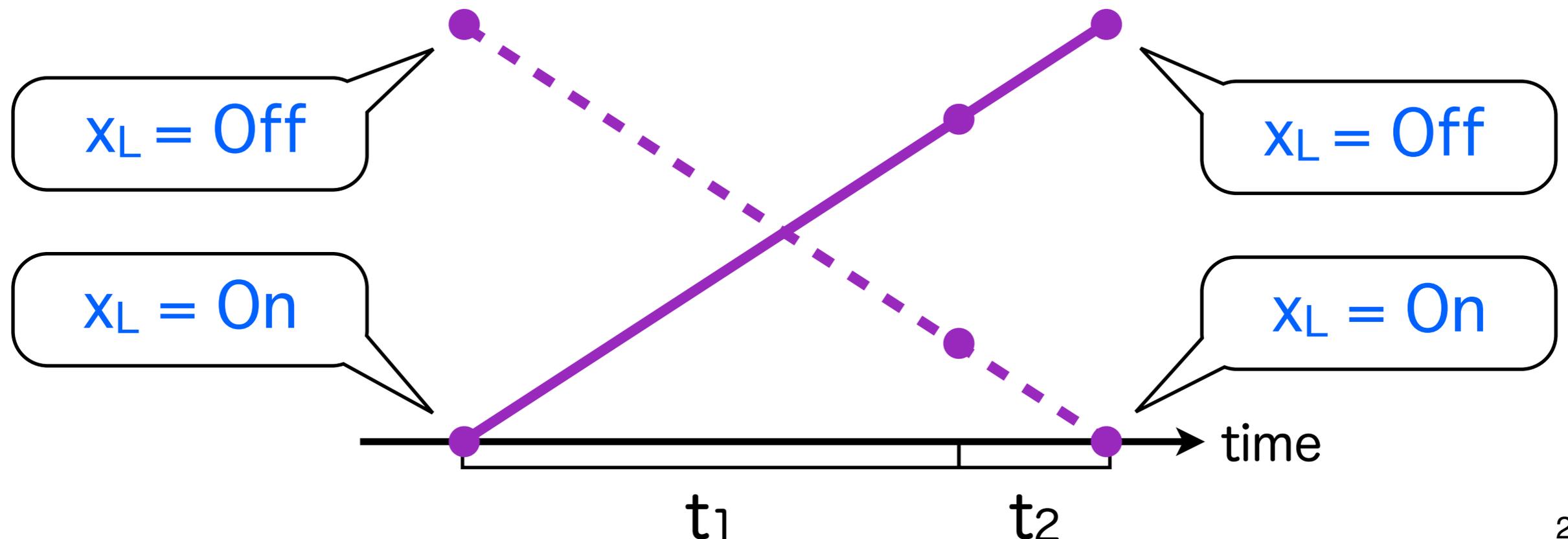
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Example: Water-level Monitor

- Verification is also possible with the inference rule for $m=0$ and $n=2$ where the loop invariant is set as

$$P^+ \equiv x_L = \text{on} \vee x_L = \text{off}$$



Algorithm for Inductive Verification

- Following algorithm generates VCs for $(m,n) \in [0, m_{max}] \times [1, n_{max}]$ and discharges the VCs

Input: $HA; P; m_{max} \in \mathbb{N}_{\geq 0}; n_{max} \in \mathbb{N}_{> 0}$

Output: *true*: $HA \models \Box P$; *false*: cannot decide $\Box P$ within $m_{max} + n_{max}$ steps

```
1: for  $m \in \{0, \dots, m_{max}\}; n \in \{1, \dots, n_{max}\}$  do
2:    $P^+ := P$ 
3:   while  $P^+ \neq false$  do
4:     if  $\neg \forall i \in \{0, \dots, m\} \text{ Validate}(VC_i)$  then
5:       break
6:     end if
7:     if  $\exists j \in \{m + 1, \dots, m + n, -1\} \neg \text{Validate}(VC_j)$  then
8:        $P^+ := P^+ \wedge \text{Learn}(VC_j)$ 
9:     else
10:      return true
11:    end if
12:  end while
13: end for
14: return false
```

Algorithm for Inductive Verification

- Following algorithm generates VCs for $(m,n) \in [0, m_{max}] \times [1, n_{max}]$ and discharges the VCs

Input: $HA; P; m_{max} \in \mathbb{N}_{\geq 0}; n_{max} \in \mathbb{N}_{> 0}$

Output: *true*: $HA \models \Box P$; *false*: cannot decide $\Box P$ within $m_{max} + n_{max}$ steps

```
1: for  $m \in \{0, \dots, m_{max}\}; n \in \{1, \dots, n_{max}\}$  do
2:    $P^+ := P$ 
3:   while  $P^+ \neq false$  do
4:     if  $\neg \forall i \in \{0, \dots, m\}$  Validate( $VC_i$ ) then
5:       break
6:     end if
7:     if  $\exists j \in \{m + 1, \dots, m + n, -1\}$   $\neg$ Validate( $VC_j$ ) then
8:        $P^+ := P^+ \wedge$  Learn( $VC_j$ )
9:     else
10:      return true
11:    end if
12:  end while
13: end for
14: return false
```

initialize P^+

loop invariant generation

Loop Invariant Generation

- When verification of $VC_i : \forall t_1..t_i \geq 0 (SP(P^+ \wedge s) \Rightarrow P)$ fails, we can generate a loop invariant using a **quantifier elimination (QE)** method

$$QE(\forall x_s \forall t_1..t_i VC_i)[x_0 \leftarrow x_s]$$

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- This process is hard in general
 - VC_i should be simplified by assuming a class of problems
 - * Otherwise QE might not succeed
 - Generated invariant might be too large. They should be simplified manually
 - * Two strategies for the simplification

Implementation

- **We have implemented the verification algorithm using **Mathematica**** (i.e. algebraic formula manipulation system)
 - ODEs are solved into closed forms with **DSolve**
 - **Validate** is implemented using built-in functions, **FullSimplify**, **Reduce**, and **FindInstance**
 - **Learn** is implemented using a built-in function **Resolve** that performs QE

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- **Several optimizations**
 - Separation of formulas wrt HA locations
 - Reutilization of simplification process of sub-formulas

Talk Outline

1. Hybrid Automata

**2. Imp_{HA} and
Strongest Postcondition Calculus**

3. Inductive Verification

4. Experimental Results

Experimental Results

Problem	#Loc #Var	m/n	Our method	MC tools	KeYma era
WLM	4/2	0/1	0.85s	N/A	1.8s
Gas burner	2/3	4/2	2.22s	0.004s	N/A
Temp. ctrl	4/3	1/1	2.82s	0.012s	N/A
Bouncing ball	1/2	0/1	0.49s	N/A	0.9s
ETCS	2/3	0/1	4.48s	N/A	3.1s
Highway 9	10/9	0/2	0.22s	0.22s	N/A
Highway 19	20/19	0/2	3.64s	N/A	N/A

Comparison with Other Tools

- **MC tools**

- **HyTech** [Henzinger+ 96] and **PHAVer** [Frahse 02]
- Solve three problems quite efficiently (Ex.2,3,6)
- Cannot handle instances with uncertain parameters (Ex.1,4)
- Some scaling issues (Ex.6,7)

- **KeYmaera** [Platzer+ 08]

- Handles various *hybrid programs* automatically
- However, does not succeed on most of programs translated from HA (Ex.2,3,6,7)
 - *Models should be annotated manually
 - *Otherwise, users need to interact with underlying theorem prover

Conclusion

- **Automated logical analytic method for a large class of linear and nonlinear HA**

- Algorithmic verification with SP calculus and limited derivation rules, i.e., induction and loop unrolling
- Loop invariant generation guided by the response from the decision process
- Promising experimental results with several HA

- **Future work**

- Automation of generation process of efficient loop invariants
- Support for larger class of HA, e.g., with unsolvable ODEs, parallel composition