

Environment Modeling During Model Checking of Cyber-Physical Systems

Guangyao Chen¹[0000–0003–1367–8331] and Zhihao Jiang²[0000–0002–6730–6915]

¹ ShanghaiTech University, Shanghai, China
chengy2@shanghaitech.edu.cn

² ShanghaiTech University, Shanghai, China
jiangzh@shanghaitech.edu.cn

Abstract. Ensuring the safety and efficacy of Cyber-Physical Systems (CPSs) is challenging due to the large variability of their operating environment. Model checking has been proposed for validation of CPSs, but the models of the environment are either too specific to capture the variability of the environment, or too abstract to provide counter-examples interpretable by experts in the application domain. Domain-specific solutions to this problem require expertise in both formal methods and the application domain, which prevents effective application of model checking in CPSs validation. A domain-independent framework based on timed-automata is proposed for abstraction and refinement of environment models during model checking. The framework maintains an abstraction tree of environment models, which provides interpretable counter-examples while ensuring coverage of environment behaviors. With the framework, experts in the application domain can effectively use model checking without expertise in formal methods.

Keywords: Abstraction Tree · Timed Automata · Formal Methods · UPPAAL.

1 The Emergence of Cyber-Physical Systems (CPS)

With the development of technologies, it is now possible to develop software that can make real-time decisions under complex situations. As a result, problems in the physical world can be solved by software-controlled physical systems with little to none human intervention. These *Cyber-Physical Systems (CPSs)* are relieving human from tedious jobs and dangerous working environment, and have improved quality of lives and the overall efficiency of the society.

With human-operated systems, decisions are made by domain experts, who have been trained to deal with the complexity and variability of the environment, and are responsible for preventing safety hazards. For CPSs with increasing autonomy, malfunctions of the systems cannot receive timely human intervention, which can cause serious harm to people and properties in its operating environment, especially in safety-critical domains like medical devices [1]. Manufacturers of CPSs are required to demonstrate the safety and efficacy of the systems, especially their software components.

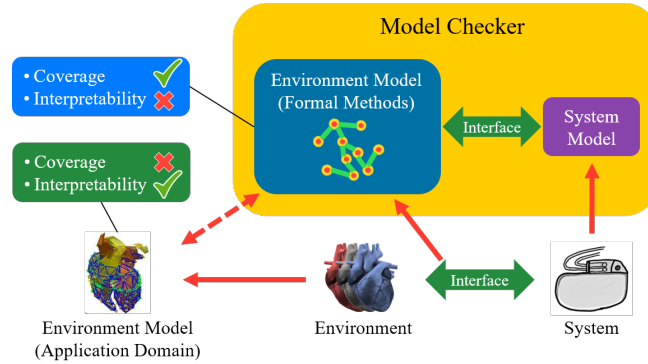


Fig. 1. Environment models are developed from two different perspectives, but neither satisfy the need for validating CPSs. A domain-independent framework that can balance the coverage and interpretability of environment models is needed for effective use of model checking in CPS domains.

1.1 Validation of CPS Using Model Checking

With increasing autonomy, CPS are required to make correct decisions under ALL possible environment conditions. CPSs cannot be exhaustively tested as the amount of environment conditions is infinite. Model checking exhaustively examines the reachable states of a model, which is suitable for validation of CPSs [2,3]. By modeling the CPS and its operating environment, model checking tools can prove that the CPS satisfies safety and efficacy requirements under conditions specified in the environment models, or provide counter-examples when requirements are violated. Development cost can be significantly reduced when bugs are found and safety guarantees are provided in the early development stage.

1.2 Environment Modeling for CPS

The environment models represent assumed environment conditions, and the results of the model checking can only support safety and efficacy of CPS under these conditions. Model(s) of the environment should satisfy the following requirements:

Coverage: Environment model(s) of CPSs should either 1) cover all environment conditions, or 2) cover all environment behaviors observable to the system. These two conditions are equivalent but the second one can be better defined and quantified.

Interpretability: The safety and efficacy of CPS are evaluated on the states of the environment. i.e. The patient's condition should be improved with a medical device compared to without the device. In order to judge whether environment conditions have been improved, the models of the environment should have states and executions that are interpretable by experts in the application domain.

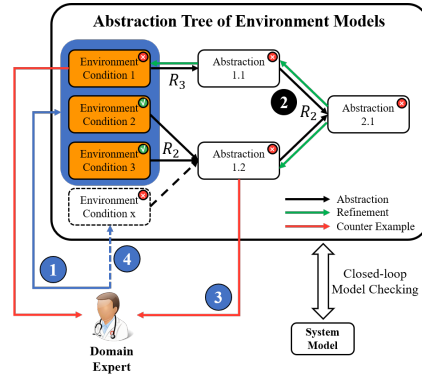


Fig. 2. Experts in the application domain provide a set of base environment models (1), and the framework returns a set of counter-examples with the most refined context (3). The experts can also provide additional base models after analyzing the returned counter-examples and their corresponding levels of abstraction (4). Inside the framework an abstraction tree is created by abstracting the set of base models using domain-independent abstraction rules (2), which is hidden from the domain expert.

Unfortunately, these two requirements conflict with each other in most cases, and no single model can satisfy both. Relaxing domain-specific constraints within the models may introduce new observable behaviors, which increases coverage at the cost of interpretability. As shown in Fig. 1, on one hand, experts in the application domain develop models to study the mechanisms of the problem. These models have great interpretability, but are not suitable for model checking due to their inadequate coverage. On the other hand, abstract models of the environment are created in the formal methods community to cover observable behaviors of the environment. These models are usually abstracted from the interface between the system and the environment, and coverage can be easily quantified.

Formal relationships between the formal models and the domain models are needed in order to balance coverage and interpretability. However, establishing connections require expertise in both formal methods and the application domain.

In [4], Jiang et. al proposed the use of over-approximation [5] to increase the coverage of environment models in closed-loop model checking of implantable cardiac devices, and refine the environment models to provide interpretability to counter-examples. Unfortunately, the proposed method requires abstraction rules based on extensive domain knowledge, which cannot be applied directly in other domains. A domain-independent framework for environment modeling is essential for model checking to be effectively adopted.

2 Domain-independent Model Checking Framework with Environment Abstraction & Refinement

In this project, we propose a model checking framework with environment abstraction & refinement which can balance the coverage and interpretability of environment models during model checking. The framework is also domain-independent such that the abstraction and refinement of environment models do not require domain-specific information.

The framework is illustrated in Fig. 2, which involves four main steps:

Step 1: Initial Set of Environment Models:

A set of base environment models are first provided by domain experts, which represent prior knowledge of environment conditions that the CPS may encounter. These base environment models does not provide adequate coverage, but their execution traces, including counter-examples returned from the model checker, are interpretable by domain experts.

Step 2: Construction of the abstraction tree:

Domain-independent abstraction rules are then applied to the base environment models so that the abstract model over-approximates the original model(s), covering more observable behaviors of the environment. By abstracting and combining models of the environment, an *abstraction tree* of environment models can be built. This step is hidden from the experts in the application domain, so expertise in formal methods is not required for using the framework.

Step 3: Model Checking and Counter-example Refinement:

The safety and efficacy of the system model can then be validated using the abstraction tree of environment models. The system model is first verified against the root environment model. If the requirements are satisfied, the system model is safe under all possible environment conditions. Otherwise, the system model is then verified against the environment model(s) that are children of the root environment model. The process traverse the abstraction tree in the Breath-First Search (BFS) manner until 1) the leaves of the abstraction tree is reached, or 2) all children of the current environment model satisfy the requirement. The counter-example(s) are attached to the abstraction tree, and returned to the domain experts for further analysis.

Step 4: Environment Model Refinement:

Depending on the "completeness" of the set of base models and the topology of the abstraction tree, the refined counter-examples returned may not correspond to the leaf nodes in the abstraction tree. In this case the violations of requirement happen in environment conditions that are not included in the set of base models. Moreover, the counter-examples returned may not have adequate context for comprehensive interpretation. Domain experts can create new base models, which are refinements of the model that returned the counter-examples. i.e. in Fig. 2, environment condition x can be created by "subtracting" environment condition 2 and 3 from Abstraction 1.2.

The framework hides domain knowledge in formal methods from experts in the application domain, so that model checking becomes a more friendly tool for validating CPSs.

3 Abstraction Tree Construction with Timed Automata

In order for the framework to achieve domain-independence, the application and selection of the abstraction rules should not contain knowledge in the application domain. In this project, we use timed automata [6] as modeling formalism and UPPAAL [7] as model checker. Abstraction rules targeting the structure of timed-automata are proposed and their effect on observable environment behaviors are informally proved. Formal proofs of the Theorems can be found in [].

3.1 Timed Automata and Model Checker UPPAAL

Timed-automata [6] is a formalism developed to model real-time systems. It has the expressiveness for modeling complex system behaviors [8], and the simplicity for decidable reachability. Timed automata also supports non-determinism, which can be used to capture the uncertainty within the environment. The framework proposed in this project is applicable when both the system and the environment are modeled using timed automata.

A timed automaton is a tuple (L, l_0, X, A, E, G, I) , where

1. L is a set of locations.
2. $l_0 \in L$ is the initial location.
3. X is the set of clocks.
4. A is a set of actions, including sending actions (a!) and receiving actions (a?).
5. $E \subseteq L \times A \times 2^X \times L$.
An edge (transition) $e \in E$ is a tuple (l, a, r, l') , where l is the start location, a is the action, r is the set of clocks to be reset and l' is the target location.
6. $G : E \times 2^X \times 2^{\mathbb{N}} \rightarrow \Psi_G$ assigns guards to edges.
 G can be written as $G(E, X, N) = \{g_i(e_i, X_i, N_i) \mid i \in \mathbb{N} \text{ and } i \leq \text{len}(E)\}$.
Each g_i denotes the guard of the edge e_i , which constrains the set of clocks in X_i with the set of lower bounds N_i .
7. $I : L \times 2^X \times 2^{\mathbb{N}} \rightarrow \Psi_I$ assigns invariants to locations.
 I can be written as $I(L, X, M) = \{inv_i(l_i, X_i, M) \mid i \in \mathbb{N} \text{ and } i \leq \text{len}(E)\}$.
Each inv_i denotes the invariant of the location l_i , which constrains the set of clocks in X_i with the set of upper bounds N_i .
8. Ψ is the clock constraints for clock variables X .
 $\Psi := x \perp n \parallel \Psi_1 \wedge \Psi_2$, where $x \in X$, $\perp \in \{\leq, \geq\}$, and $n \in \mathbb{N}$.
For particular guard and invariant clock constraints, we have
 - $\Psi_G \in \Psi^X$ and $\Psi_G := x \geq n \parallel \Psi_1 \wedge \Psi_2$
 - $\Psi_I \in \Psi^X$ and $\Psi_I := x \leq n \parallel \Psi_1 \wedge \Psi_2$

Multiple timed automata can run in parallel and interact with each other via actions. i.e. the system model and the environment model form a closed-loop system. We use $\mathcal{A}_1 | \mathcal{A}_2$ to represent automata composition. The semantics [9] is defined as a labelled transition system $\langle S, s_0, \rightarrow \rangle$, where

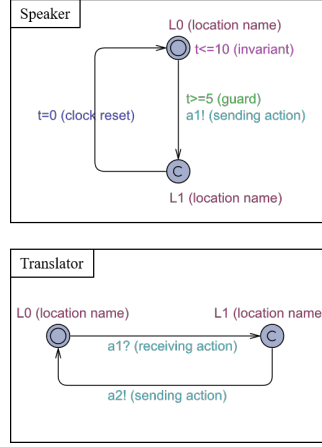


Fig. 3. The composed model *Speaker|Translator*. The time interval between two consecutive action $a1$ is larger than 5 unit time and shorter than 10 unit time. The *Translator* sends $a2$ immediately after receiving $a1$.

1. $S \subseteq L \times R^C$ is the set of states,
2. $s_0 = \langle l_0, u_0 \rangle$ is the initial state,
3. $u : C \rightarrow \mathbb{R}_{\geq 0}$ is the function of a clock valuation and
4. $\rightarrow \subseteq S \times (R_{\geq 0} \cup A) \times S$ is the transition relation such that:
 - $(l, u) \xrightarrow{d} (l, u + d)$ if $\forall d' : 0 \leq d' \leq d \implies u + d' \in I(l, x, n)$, where $x \in 2^X$ and $n \in 2^{\mathbb{N}}$
 - $(l, u) \xrightarrow{a} (l', u')$ if there exists $e = (l, a, r, l') \in E$ s.t. $u \in G(e, x', n')$, $u' = [r \mapsto 0]u$, and $u' \in I(l', n')$, where $x' \in 2^X$ and $n' \in \mathbb{N}$

UPPAAL [7] is a model checking tool using timed automata as formalism, and is very friendly to people with little programming experience. Users can model their system and its environment in a graphic interface, and counter-examples returned by the model checker are visualized in the simulator.

Fig. 3 shows the composed timed automaton *Speaker|Translator* that interact with each other via action $a1$ in UPPAAL. The sending action $a1!$ in *Speaker* is confined by guard $t \geq 5$ and invariant $t \leq 10$, which represents the uncertainty in behaviors $a1!$.

3.2 Prerequisite of the Framework

Currently the framework is suitable for problems with the following constraints:

1. The environment contains multiple independent agents interacting with each other via events.
2. The system also interacts with the environment via events. Only a subset of events in the environment are observable to the system, and the system operates base on the timing and patterns of these events.

3. Both the system and the environment are modeled using timed automata.
4. At a particular state of the environment, the observable events can occur within a timing interval $[T_{min}, T_{max}]$
5. The differences among base environment models are parameters-only.

3.3 Coverage of Environment Behaviors

The system can only observe a subset of actions $A_o \subseteq A$ in the environment. Environment behavior is defined as *timed word* [6] over observable actions $A_o \subseteq A$, which is a pair (Σ, T) where $\Sigma = \sigma_1\sigma_2\dots$, $\sigma_i \in A_o$ represents the sequence of actions, and $T = \tau_1\tau_2\dots$, $\tau_i \in \mathbb{R}$ represents the global time the actions happened. The timed language of a timed automaton \mathcal{A} is the set of all the possible timed words of \mathcal{A} , which is represented as $\mathcal{L}(\mathcal{A})$. The coverage of environment behaviors is then measured on the "size" of the language.

Followings are the formal definition of timed sequence, timed word and timed language.

Definition: Timed sequence

A timed sequence [10] $\tau = \tau_1\tau_2\dots$ is an infinite sequence of time values $\tau_i \in R$ with $\tau_i > 0$, satisfying the following constraints:

1. Monotonicity
 - τ increases strictly monotonically, i.e. for all $i \geq 1$, we have $\tau_{i+1} > \tau_i$;
2. Progress
 - For every $t \in R$, there is some $i \geq 1$ such that $\tau_i > t$.

Definition: Timed word

A timed word [10] over observable actions $A_o \subseteq A$ is a pair (σ, τ) where $\sigma = \sigma_1\sigma_2\dots$ where each σ_i indicates whether an observable action is observed.

For example, let $A = \{a_{o1}, a_{o2}, a_{u1}, a_{o3}, a_{u2}\}$ is the set of all actions, $A_o = \{a_{o1}, a_{o2}, a_{o3}\}$ is the set of observable actions, and $A_u = \{a_{u1}, a_{u2}\}$ is the set of unobservable actions.

If at time τ_1 , no observable actions is observed, then $\sigma_1 = \langle 0, 0, 0 \rangle$.

If at time τ_2 , a_{o1} and a_{o3} are observed, then $\sigma_2 = \langle 1, 0, 1 \rangle$.

Definition: Timed language

For a timed automaton $\mathcal{A} = (L, l_0, X, A, E, G, I)$, where A is the set of actions. The timed language [10] of \mathcal{A} is the set of all the possible timed words of \mathcal{A} .

3.4 Domain-independent Abstraction Rules

A set of abstraction rules on timed automata is defined that can increase the coverage of observable behaviors of the environment. The correctness of the abstraction rules are informally proved to provide intuition for the audience of this paper. Interested audience can find formal proofs in the appendix.

\mathcal{R}_1 : Increase Transition Uncertainty **Intuition:** A transition $e = (l, a, r, l')$ is *enabled* when the guard assigned to the transition $g(e, X_g, N)$ evaluates to true, and it has to be taken when the invariant of its source location $inv(l, X_i, M)$ is about to be violated due to the increase of X_i . The interval $[N, M]$ represents the uncertainty when event a can occur. If the interval is expanded, intuitively the constraints on sending the event a are relaxed, and the new model covers more behaviors.

Prerequisite: None.

Rule: Given a timed automaton $\mathcal{A}_1 = (L, l_0, X, A, E, G, I)$, and two non-negative vector $\Delta_G = \{\delta_i^G \mid i \in \mathbb{N} \text{ and } i \leq \text{len}(G)\}$ and $\Delta_I = \{\delta_i^I \mid i \in \mathbb{N} \text{ and } i \leq \text{len}(I)\}$, create another timed automaton $\mathcal{A}_2 = \mathcal{R}_1(\mathcal{A}_1, \Delta_G, \Delta_I) = (L, l_0, X, A, E, G^R, I^R)$ s.t.

- $\forall g_i(e_i, X_i, N_i^1) \in G, N_i^R = N_i^1 - \delta_i^G$ for all $g_i^R(e_i, X_i, N_i^R) \in G^R$
- $\forall inv_i(l_i, X_i, M_i^1) \in I, M_i^R = M_i^1 - \delta_i^I$ for all $inv_i^R(l_i, X_i, M_i^R) \in I^R$.

Theorem 1. $\mathcal{L}(\mathcal{A}) \subseteq \mathcal{L}(\mathcal{R}_1(\mathcal{A}, \Delta_G, \Delta_I))$ for non-negative Δ_G, Δ_I .

Informal Proof: First we prove that $\mathcal{R}_1(\mathcal{A}, \Delta_G, \Delta_I)$ is a *timed-simulation* of \mathcal{A} , which further implies that $\mathcal{L}(\mathcal{A}) \subseteq \mathcal{L}(\mathcal{R}_1(\mathcal{A}, \Delta_G, \Delta_I))$. True subset can then be proved by construction, as there always exists timed words in $\mathcal{R}_1(\mathcal{A}, \Delta_G, \Delta_I)$ which are not in \mathcal{A} .

\mathcal{R}_2 : Merge Models with the Same Structure **Intuition:** When the differences between two timed automata are confined to N in guards $G(E, X, N)$ and M in invariants $I(L, X, M)$, creating a timed automaton with the minimum of N and the maximum of M covers the behaviors of both models, as well as additional behaviors. **Prerequisite:** The differences between \mathcal{A}_1 and \mathcal{A}_2 should be confined to the N of guards $g(e, X, N) \in G$ and the M of invariants $inv(l, X, M)$.

Rule: Given two timed automata $\mathcal{A}_1 = (L, l_0, X, A, E, G^1, I^1)$ and $\mathcal{A}_2 = (L, l_0, X, A, E, G^2, I^2)$, create another timed automaton $\mathcal{A}_3 = \mathcal{R}_2(\mathcal{A}_1, \mathcal{A}_2) = (L, l_0, X, A, E, G^3, I^3)$ such that

- $\forall g_i^1(e_i, X_i, N_i^1) \in G^1$ and $\forall g_i^2(e_i, X_i, N_i^2) \in G^2, N_i^3 = \text{elm_min}(N_i^1, N_i^2)$ for all $g_i^3(e_i, X_i, N_i^3) \in G^3$
- $\forall inv_i^1(l_i, X_i, M_i^1) \in I^1$ and $\forall inv_i^2(l_i, X_i, M_i^2) \in I^2, M_i^3 = \text{elm_max}(M_i^1, M_i^2)$ for all $inv_i^3(l_i, X_i, M_i^3) \in I^3$

where $\text{elm_min}()$ and $\text{elm_max}()$ calculate element-wise minimum and maximum of vectors.

Theorem 2. $\mathcal{L}(\mathcal{A}_1) \cup \mathcal{L}(\mathcal{A}_2) \subseteq \mathcal{L}(\mathcal{A}_3)$

Proof: If we define $\Delta_G = \mathcal{A}_1.G.N - \text{elm_min}(\mathcal{A}_1.G.N, \mathcal{A}_2.G.N)$ and $\Delta_I = \text{elm_max}(\mathcal{A}_1.I.M, \mathcal{A}_2.I.M) - \mathcal{A}_1.I.M$, we have $\mathcal{A}_3 = \mathcal{R}_1(\mathcal{A}_1, \Delta_G, \Delta_I)$. Then according to Theorem 1, $\mathcal{L}(\mathcal{A}_1) \subseteq \mathcal{L}(\mathcal{A}_3)$. Similarly we have $\mathcal{L}(\mathcal{A}_2) \subseteq \mathcal{L}(\mathcal{A}_3)$, therefore the theorem holds.

\mathcal{R}_3 : Remove Internal Receiving Actions Intuition: Edges with receiving actions can be taken only when the action is sent. Therefore receiving actions are equivalent to guards on edges. If receiving actions are removed and the action is not observable to the system, it is equivalent to setting the guard to True, or setting the $[N, M]$ interval to $[0, \infty]$, therefore increase behavior coverage.

Prerequisite: There exists $a \notin A^O$ and a is a broadcast channel. There is also no guard on transition $e = (l, a, r, l')$

Rule: For a timed automaton $\mathcal{A} = \mathcal{A}_1 | \mathcal{A}_2 | \dots | \mathcal{A}_N$, if there exists $\mathcal{A}_i = (L^i, l_0^i, X^i, A^i, E^i, G^i, I^i)$ and $\mathcal{A}_j = (L^j, l_0^j, X^j, A^j, E^j, G^j, I^j)$, $i, j \in [1, N]$ such that a_m^i is a sending action, a_n^j is a receiving action and $a_m^i, a_n^j \notin A^O$, create a new timed automaton $\mathcal{A}' = \mathcal{R}_3(\mathcal{A}) = \mathcal{A}_1 | \mathcal{A}_2 | \dots | \mathcal{A}_n$ such that $a_n^j = \emptyset$ for \mathcal{A}_j .

Theorem 3. $\mathcal{L}(\mathcal{A}') \subseteq \mathcal{L}(\mathcal{R}_3(\mathcal{A}))$

Proof: We first prove that $\mathcal{R}_3(\mathcal{A})$ is a timed simulation of \mathcal{A} . Since a receiving action is removed from a transition, that transition is always enabled. Therefore when the sending action is taken, the new transition is enabled and can be taken at the same time, which satisfies the timed simulation requirement. Timed simulation ensures $\mathcal{L}(\mathcal{A}') \subseteq \mathcal{L}(\mathcal{R}_3(\mathcal{A}))$. We can then use prove by construction to show that \mathcal{A}' has timed words that are not in \mathcal{A} , therefore the theorem holds.

In the next section, we use a simple case study to demonstrate the application of the proposed framework.

4 Case Study: Ensuring Pedestrian Safety in Autonomous Driving

Autonomous vehicles are CPSs which are required to safely operate within complex environment with large variabilities. The environment consists of multiple agents with different states and parameters. In this case study, we focus on a simple scenario in which an autonomous vehicle is crossing an intersection with traffic lights, with one pedestrian who may cross the road in front of the car (Fig. 4.(a)). The environment for the autonomous vehicle contains two components: the traffic lights and the pedestrian. The autonomous vehicle can observe the color of the active light as well as the pedestrian's crossing and finishing actions. The safety property is to prevent collision with the pedestrian, such that the car and the pedestrian cannot cross at the same time.

4.1 Step 1: Base Environment Models

Although traffic light is also part of the environment, due to its lack of variability, only the pedestrian model will be abstracted. Domain experts can also decide to exclude certain components based on prior knowledge. As shown in Fig. 5, we start with two base pedestrian models: *Pedestrian0_2* who complies to traffic rules, and *Pedestrian0_1* who may cross the road when the traffic light is red.

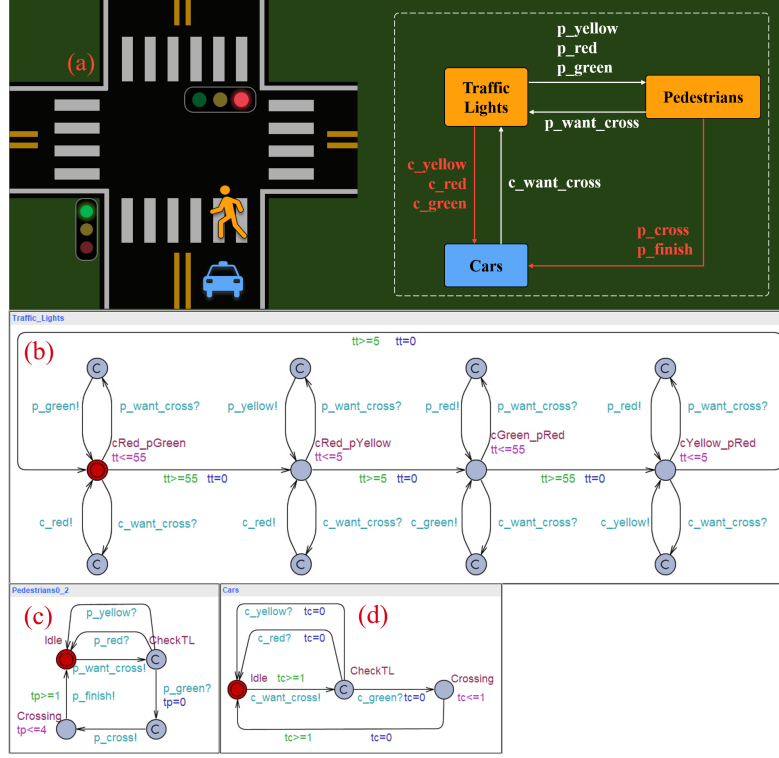


Fig. 4. (a) The environment of the car includes a pedestrian and a traffic light. Actions in red represent observable actions. (b) The controller of the traffic lights sends the color of the light after receiving cross requests. (c) The model of a regular pedestrian who crosses only during green light within $[1,4]$ seconds. (d) A simple and faulty car controller which crosses whenever the light is green.

4.2 Step 2: Construction of the Abstraction Tree

Although the base environment models already covers uncertain behaviors of the pedestrian (i.e. the timing of the cross intention and the duration of the cross action), the set of base environment models does not cover all possible observable behaviors of a pedestrian. Abstraction rules were applied to the base models to construct the abstraction tree:

\mathcal{R}_1 was applied to both *Pedestrian0.1* and *Pedestrian0.2*, increasing the time range for crossing the road from $[1,4]$ to $[1,15]$ and $[0,10]$ respectively, resulting in abstract models *Pedestrian1.1* and *Pedestrian1.2*. \mathcal{R}_3 was then applied to *Pedestrian1.2*, removing interactions between the traffic light and the pedestrian, resulting in *Pedestrian2.1*. *Pedestrian2.1* and *Pedestrian1.1* now have the same structure, and therefore can be merged by \mathcal{R}_2 , resulting in *Pedestrian3.1*.

In this example we use the abstraction tree with *Pedestrian3_1* as root, although the behavior coverage of *Pedestrian3_1* can still be improved by applying \mathcal{R}_1 .

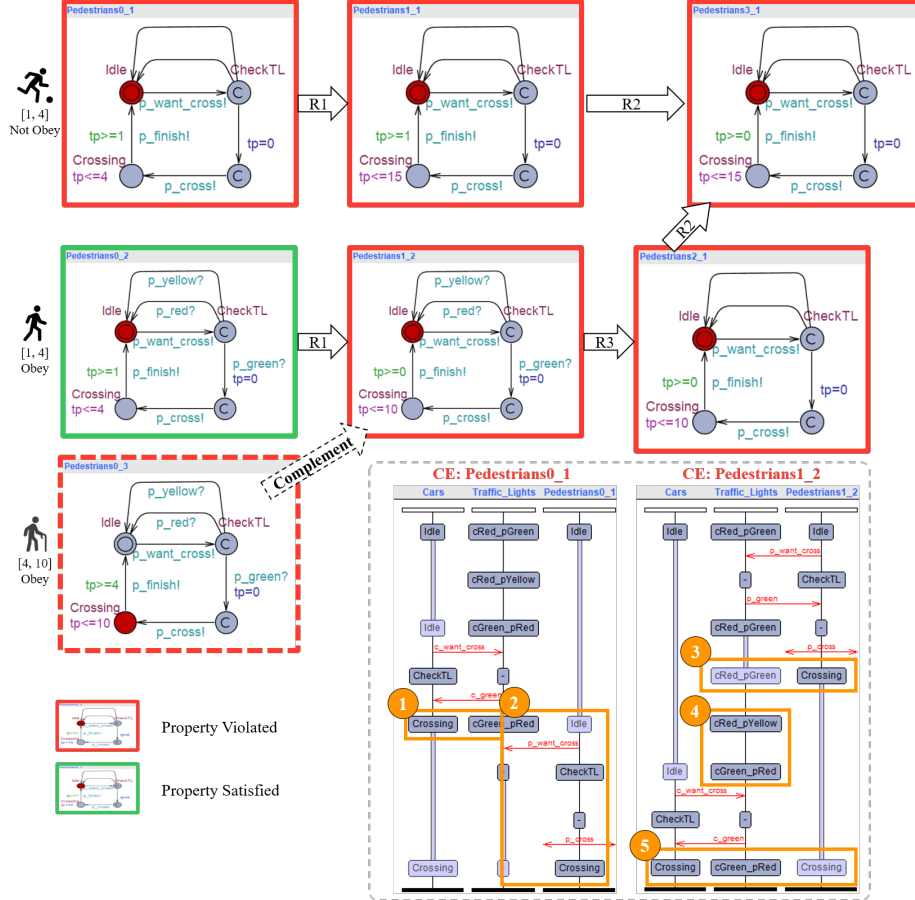


Fig. 5. Abstraction tree of the environment (pedestrian) and two counter-examples returned from *Pedestrian0_1* and *Pedestrian1_2*.

4.3 Step 3: Model Checking and Counter-example Refinement

The abstraction tree of pedestrian models can then be used for closed-loop model checking of control algorithms of the autonomous vehicle. In order to demonstrate the advantage of using the abstraction tree, we use a simple and faulty controller that crosses the road whenever the traffic light on its side is green (Fig. 4.(d)). This safety property is specified using TCTL language as $A[]$ not

(P.Crossing and C.Crossing). After traversing the abstraction tree, two refined counter-examples were returned which correspond to node *Pedestrian0_1* and *Pedestrian1_2* in the abstraction tree (Fig. 5).

The counter-example from *Pedestrian0_1* is as expected since the pedestrian may cross the road when the traffic light is red ((2) in Fig. 5), while the car is already crossing the road ((1) in Fig. 5).

The counter-example from *Pedestrian1_2* shows a different mechanism. Both the pedestrian and the car started to cross the road when the traffic light on their corresponding side was green ((3) and (5) in Fig. 5). The traffic light switched when the pedestrian was still crossing the road ((4) in Fig. 5), triggering the crossing of the car and collision with the pedestrian.

4.4 Step 4: Environment Model Refinement

The safety property was violated in *Pedestrian1_2*, but was satisfied in its child *Pedestrian0_2*. In order to pinpoint the environment condition in which the counter-example occurred, a new base model *Pedestrian0_3* can be obtained by "subtracting" *Pedestrian0_2* from *Pedestrian1_2* (Fig. 5), and the property was also violated with the same counter-example mechanism. From *Pedestrian0_3* we can see that the collision happened due to the long crossing time of the pedestrian, which provided more interpretation to the counter-example.

5 CONCLUSION

Model checking of CPSs requires environment models that not only cover all possible environment conditions, but also provide interpretability to the counter-examples. Balancing these conflicting requirements requires expertise in both formal methods and the application domain, which prevents model checking from being effectively adopted for validation of CPSs. In this project, a set of domain-independent abstraction rules for timed automata were developed to increase the coverage of environment models. A domain-independent framework for abstraction and refinement of environment models was proposed for model checking of CPSs. The framework balances coverage and interpretability in environment models, and experts in the application domain can use model checking effectively without expertise in formal methods.

Currently the base environment models are required to have the same model structure. The next step is developing new abstraction rules that can remove locations, even components from environment models, so that this constraint can be relaxed. The sequence for abstraction rule application may affect the completeness and abstraction level of counter-examples. The next step is identifying the optimal sequence for abstraction rule application, and quantification of coverage.

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A Formal proofs of abstractions rules

First of all, we need to refer to the definitions of timed simulation and transition enabled interval.

Definition: Timed simulation

For two timed automata $\mathcal{A}_1 = (L^1, l_0^1, C^1, A^1, E^1, G^1, I^1)$ and $\mathcal{A}_2 = (L^2, l_0^2, C^2, A^2, E^2, G^2, I^2)$, a timed simulation relation [4] is a binary relation $\text{sim} \subseteq \Omega^1 \times \Omega^2$ where Ω^1 and Ω^2 are sets of states of \mathcal{A}_1 and \mathcal{A}_2 . We say \mathcal{A}_2 time simulates \mathcal{A}_1 ($\mathcal{A}_1 \preceq_t \mathcal{A}_2$) if the following conditions holds:

1. initial states correspondence: $(\langle l_0^1, \mathbf{0} \rangle, \langle l_0^2, \mathbf{0} \rangle) \in \text{sim}$
2. timed transition: For every $(\langle l_1, v_1 \rangle, \langle l_2, v_2 \rangle) \in \text{sim}$
 if $\langle l_1, v_1 \rangle \xrightarrow{d} \langle l_1, v_1 + d \rangle$, there exists $\langle l_2, v_2 + d \rangle$
 such that $\langle l_2, v_2 \rangle \xrightarrow{d} \langle l_2, v_2 + d \rangle$ and $(\langle l_1, v_1 + d \rangle, \langle l_2, v_2 + d \rangle) \in \text{sim}$
3. discrete transition: for every $(\langle l_1, v_1 \rangle, \langle l_2, v_2 \rangle) \in \text{sim}$
 1. if a is an observable action,

if $\langle l_1, v_1 \rangle \xrightarrow{a} \langle l'_1, v'_1 \rangle$, there exists $\langle l'_2, v'_2 \rangle$ such that
 $\langle l_2, v_2 \rangle \xrightarrow{a} \langle l'_2, v'_2 \rangle$ and $(\langle l'_1, v'_1 \rangle, \langle l'_2, v'_2 \rangle) \in \text{sim}$
 2. if a is not an observable action,
 if $\langle l_1, v_1 \rangle \xrightarrow{a} \langle l'_1, v'_1 \rangle$, there exists $\langle l'_2, v'_2 \rangle$ such that
 $\langle l_2, v_2 \rangle \xrightarrow{a|e} \langle l'_2, v'_2 \rangle$ and $(\langle l'_1, v'_1 \rangle, \langle l'_2, v'_2 \rangle) \in \text{sim}$

Definition: Transition Enabled Interval

Transition enabled interval is defined on an edge's guard and its output location's invariant, which indicates the time interval that an edge can be enabled. For an edge $e = (l, a, r, l')$, the enabled interval is $I(l) \wedge G(e)$.

For example, an edge e has the guard $t \geq 3$ and l has the invariant $t \leq 6$, the enabled interval is $[3, 6]$.

If the enabled interval of an edge e is changed from $[a, b]$ to $[a - \epsilon_g, b + \epsilon_i]$, with $\epsilon_g \in \mathbb{N}$, $\epsilon_i \in \mathbb{N}$ and $\epsilon_g + \epsilon_i > 0$, we say that the enabled interval of the edge e is extended.

Note that as we assuming that there is no deadlock, therefore $I(l) \wedge G(e) \wedge I(l') = I(l) \wedge G(e)$.

Followings are formal proofs of the abstraction rules.

Theorem 4. $\mathcal{A}_2 = \mathcal{R}_1(\mathcal{A}_1, \Delta_G, \Delta_I) \Rightarrow \mathcal{A}_2$ timed simulates \mathcal{A}_1 .

Proof. After applying the \mathcal{R}_1 on \mathcal{A}_1 , we get $\mathcal{A}_2 = GI(\mathcal{A}_1, \Delta_G, \Delta_I)$ whose transition enabled intervals are extended. Because the only differences of \mathcal{A}_2 from \mathcal{A}_1 is the guards G and invariants I .

Therefore, the proof idea is that for any timed transition or discrete transition, there exist transitions that have the same location l and clock assignment v and differ from the transition enabled interval.

Let $\text{sim} \subseteq \Omega^1 \times \Omega^2$ where Ω^1 and Ω^2 are sets of states of \mathcal{A}_1 and \mathcal{A}_2 . It can be seen that \mathcal{A}_2 time simulates \mathcal{A}_1 ($\mathcal{A}_1 \preceq_t \mathcal{A}_2$) because the following conditions holds:

1. initial states correspondence: $(\langle l_0, \mathbf{0} \rangle, \langle l_0, \mathbf{0} \rangle) \in \text{sim}$
2. timed transition: $\forall (\langle l_1, v_1 \rangle, \langle l_2, v_2 \rangle) \in \text{sim}, \text{ where } \langle l_1, v_1 \rangle \in \mathcal{A}_1 \text{ and } \langle l_2, v_2 \rangle \in \mathcal{A}_2,$

we want to prove $\langle l_1, v_1 \rangle \xrightarrow{d} \langle l_1, v_1 + d \rangle \Rightarrow \langle l_2, v_2 \rangle \xrightarrow{d} \langle l_2, v_2 + d \rangle$.

If $\langle l_1, v_1 \rangle \xrightarrow{d} \langle l_1, v_1 + d \rangle$, then we know that $v_1 \models \text{inv}_1(l_1, X_1, N_1)$.

– Because $\mathcal{A}_2 = GI(\mathcal{A}_1, \Delta_G, \Delta_I)$, then there exists $\langle l_2, v_2 \rangle = \langle l_1, v_1 \rangle$ such that $\text{inv}_2(l_2, X_2, N_2) = \text{inv}_1(l_1, X_1, N_1 + \Delta_I[1])$.

– Because $v_1 \models \text{inv}_1(l_1, X_1, N_1)$, $v_2 = v_1$, then $v_2 \models \text{inv}_1(l_1, X_1, N_1 + \Delta_I[1]) = \text{inv}_2(l_2, X_2, N_2)$.

– Then we have $\langle l_2, v_2 \rangle \xrightarrow{d} \langle l_2, v_2 + d \rangle$. Therefore $(\langle l_1, v_1 + d \rangle, \langle l_2, v_2 + d \rangle) \in \text{sim}$.

3. discrete transition: $\forall (\langle l_1, v_1 \rangle, \langle l_2, v_2 \rangle) \in \text{sim}, \text{ where } \langle l_1, v_1 \rangle \in \mathcal{A}_1 \text{ and } \langle l_2, v_2 \rangle \in \mathcal{A}_2.$

We want to prove $\langle l_1, v_1 \rangle \xrightarrow{a} \langle l'_1, v'_1 \rangle \Rightarrow \exists \langle l'_2, v'_2 \rangle \in \mathcal{A}_2, \langle l_2, v_2 \rangle \xrightarrow{a} \langle l'_2, v'_2 \rangle$.

If $\langle l_1, v_1 \rangle \xrightarrow{a} \langle l'_1, v'_1 \rangle$, we know that $v_1 \models g_1(e_1, X_1, N_1)$, where $e_1 = (l_1, l'_1)$.

- Because $\mathcal{A}_2 = GI(\mathcal{A}_1, \Delta_G, \Delta_I)$, then there exists $\langle l'_2, v'_2 \rangle = \langle l'_1, v'_1 \rangle$ such that $g_2(e_2, X_2, N_2) = g_1(e_1, X_1, N_1 - \Delta_G[1])$.
- Because $v_1 \models g_1(e_1, X_1, N_1)$, $v_2 = v_1$, then $v_2 \models g_1(e_1, X_1, N_1 - \Delta_G[1]) = g_2(e_2, X_2, N_2)$, where $e_2 = \langle l_2, l'_2 \rangle$.
- Then we have $\langle l_2, v_2 \rangle \xrightarrow{a} \langle l'_2, v'_2 \rangle$.

Therefore, $(\langle l'_1, v'_1 \rangle, \langle l'_2, v'_2 \rangle) \in \text{sim}$.

Theorem 5. \mathcal{A}_2 timed simulates $\mathcal{A}_1 \Rightarrow \mathcal{L}(\mathcal{A}_1) \subseteq \mathcal{L}(\mathcal{A}_2)$.

Proof. The language of a model is defined on the observable sending actions. Let w be any timed word of \mathcal{A}_1 then there must exist a simulation procedure $p_1 = \langle l_0^1, \mathbf{0} \rangle \langle l_1^1, v_1^1 \rangle \langle l_2^1, v_2^1 \rangle \dots \langle l_n^1, v_n^1 \rangle$ of \mathcal{A}_1 that produces w .

Let Ω^1 be the sets of states of \mathcal{A}_1 and Ω^2 be the sets of states of \mathcal{A}_2 . Let $\text{sim} \subseteq \Omega^1 \times \Omega^2$ be the timed simulation relation. Because \mathcal{A}_2 timed simulates \mathcal{A}_1 , then we have

1. $\exists \langle l_0^2, \mathbf{0} \rangle \in \Omega_2$ such that $(\langle l_0^1, \mathbf{0} \rangle, \langle l_0^2, \mathbf{0} \rangle) \in \text{sim}$. We can set a global time $t_g^1 = 0$ for \mathcal{A}_1 and $t_g^2 = 0$ for \mathcal{A}_2 at the initial state.
2. $\forall \langle l_i, v_i \rangle, \langle l_{i+1}, v_{i+1} \rangle \in \Omega^1$,

(Timed Transition) If the two neighbor states have a time increase by d , i.e.

$l_{i+1}^1 = l_i^1$, $v_{i+1}^1 = v_i^1 + d$ and $\langle l_i^1, v_i^1 \rangle \xrightarrow{d} \langle l_{i+1}^1, v_{i+1}^1 \rangle$, then t_g^1 increase d between the two states.

From the timed simulation relation we know that $\exists \langle l_j^2, v_j^2 \rangle, \langle l_{j+1}^2, v_{j+1}^2 \rangle \in \Omega_2$ such that $l_{j+1}^2 = l_j^2$, $v_{j+1}^2 = v_j^2 + d$, $\langle l_j^2, v_j^2 \rangle \xrightarrow{d} \langle l_{j+1}^2, v_{j+1}^2 \rangle$, which means t_g^2 can increase the same time d .

(Discrete Transition with Observable Events) If the two neighbor states have a transition that does not have an observable action, in other words, $\langle l_i^1, v_i^1 \rangle \xrightarrow{a} \langle l_{i+1}^1, v_{i+1}^1 \rangle$, where a is an observable action.

From the timed simulation relation we know that $\exists \langle l_j^2, v_j^2 \rangle, \langle l_{j+1}^2, v_{j+1}^2 \rangle \in \Omega_2$ such that $\langle l_j^2, v_j^2 \rangle \xrightarrow{a} \langle l_{j+1}^2, v_{j+1}^2 \rangle$, which means \mathcal{A}_2 can send the same observable signal a as \mathcal{A}_1 .

Until here we know 1. \mathcal{A}_1 and \mathcal{A}_2 have the same initial global time as *zero*.

2. For any timed transition of \mathcal{A}_1 that increases t_g^1 by d , there is the same procedure in \mathcal{A}_2 such that t_g^2 increases the same time d .

3. For any transition with observable event in \mathcal{A}_1 , there is the same procedure that produces the same observable event in \mathcal{A}_2 .

For any timed word w of \mathcal{A}_1 that send some observable action at some time, \mathcal{A}_2 can simulate the same timed word, i.e. send the same observable action at the same time.

Therefore, $\mathcal{L}(\mathcal{A}_1) \subseteq \mathcal{L}(\mathcal{A}_2)$.