Reachability for Multi-Priced Timed Automata and Diophantine Approximation

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- Multi-Priced Timed Automata (MPTA) further extend timed automata with variables, called observers, that have a non-negative slope that can change from one location to another.
- Such variables can model the accumulation of costs or the use of resources along a computation, such as energy and memory consumption in embedded systems, or bandwidth in communication networks

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- Address a more expressive variant of MPTA than previously considered: namely those in which observers can have both positive and negative rates
- Alternatively, and equivalently, one can consider MPTA with nonnegative rates, but in which one allows reachability specifications to contain constraints on the difference between two observers rather than just threshold constraints that compare observers to constants
- This extension is motivated by the desire to measure net resource use along computations

Definition

2.1 Multi-Priced Timed Automata

Let $\mathbb{R}_{\geq 0}$ denote the set of non-negative real numbers. Given a set $\mathcal{X} = \{x_1, \dots, x_n\}$ of clocks, the set $\Phi(\mathcal{X})$ of clock constraints is generated by the grammar

$$\varphi ::= \mathsf{true} \mid x \leq k \mid x \geq k \mid \varphi \wedge \varphi$$
,

where $k \in \mathbb{N}$ is a natural number and $x \in \mathcal{X}$. A clock valuation is a mapping $\nu : \mathcal{X} \to \mathbb{R}_{\geq 0}$ that assigns to each clock a non-negative real number. We denote by $\mathbf{0}$ the valuation such that $\mathbf{0}(x) = 0$ for all clocks $x \in \mathcal{X}$. We write $\nu \models \varphi$ to denote that ν satisfies the constraint φ . Given $t \in \mathbb{R}_{\geq 0}$, we let $\nu + t$ be the clock valuation such that $(\nu + t)(x) = \nu(x) + t$ for all clocks $x \in \mathcal{X}$. Given $\lambda \subseteq \mathcal{X}$, let $\nu[\lambda \leftarrow 0]$ be the clock valuation such that $\nu[\lambda \leftarrow 0](x) = 0$ if $x \in \lambda$, and $\nu[\lambda \leftarrow 0](x) = \nu(x)$ otherwise.

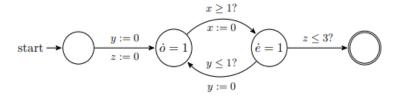
A multi-priced timed automaton (MPTA) $A = \langle L, \ell_0, L_f, X, \mathcal{Y}, E, R \rangle$ comprises a finite set L of locations, an initial location $\ell_0 \in L$, a set $L_f \subseteq L$ of accepting locations, a finite set X of clock variables, a finite set Y of observers, a set $E \subseteq L \times \Phi(X) \times 2^X \times L$ of edges, and a rate function $R : L \to \mathbb{Z}^Y$. Here $R(\ell)(y)$ is the derivative of the observer $y \in Y$ in location ℓ . Denote by ||A|| the length of the description of A, where all integers are writen in binary.

A state of A is a triple (ℓ, ν, t) where ℓ is a location, ν a clock valuation, and $t \in \mathbb{R}_{\geq 0}$ is a time stamp. A run of A is an alternating sequence of states and edges

$$\rho = (\ell_0, \nu_0, t_0) \xrightarrow{e_1} (\ell_1, \nu_1, t_1) \xrightarrow{e_2} \dots \xrightarrow{e_m} (\ell_m, \nu_m, t_m)$$

where $t_0=0,\ \nu_0=\mathbf{0},\ t_{i-1}\leq t_i$ for all $i\in\{1,\dots,m\}$, and $e_i=\langle\ell_{i-1},\varphi,\lambda,\ell_i\rangle\in E$ is such that $\nu_{i-1}+(t_i-t_{i-1})\models\varphi$ and $\nu_i=(\nu_{i-1}+(t_i-t_{i-1}))[\lambda\leftarrow 0]$ for $i=1,\dots,m$. The run is accepting if $\ell_m\in L_f$. The value of such a run is a vector $\mathrm{val}(\rho)\in\mathbb{R}^{\mathcal{Y}}$, defined by $\mathrm{val}(\rho)=\sum_{i=0}^{m-1}(t_{i+1}-t_i)R(\ell_i)$. We refer to Figure 1 for an example of an MPTA and its operational semantics.

Example



Definition

Given L, the set of finite locations, let $R:L\to\mathbb{Z}^{\mathcal{Y}}$ be the rate function, where \mathcal{Y} is a finite set of observers.

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 - \blacktriangleright This motivates the Gap Domination Problem, where we include a slack paramter ϵ
 - ▶ If there is some run ρ such that $\operatorname{val}(\rho) \leq \gamma \epsilon$, the output should be "dominated", and if there is no run such that $\operatorname{val}(\rho) \leq \gamma$ the output should be "not dominated"

Reduction to a bilinear problem

Definition

A semilinear set is the finite union of linear sets

Lemma

Let \mathcal{A} be an MPTA with set of observers \mathcal{Y} of cardinality d+1. Then there is a semilinear set $\mathcal{S}_{\mathcal{A}} \subset (\mathbb{Z}^{\mathcal{Y}})^{d+1}$, such that for every accepting run ρ of \mathcal{A} there exists $(\gamma_1, \ldots, \gamma_{d+1}) \in \mathcal{S}_{\mathcal{A}}$, for which the val (ρ) lies in the convex hull of these vectors.

Moreover, $\mathcal{S}_{\mathcal{A}}$ can be written as a collection of linear sets that can be computed in time exponential in $||\mathcal{A}||$, each of which has a description length polynomial in $||\mathcal{A}||$

Reduction to a bilinear problem

Lemma

Given $\gamma \in \mathbb{R}^x$, where x depends on \mathcal{A} , there exists a run ρ on \mathcal{A} with $val(\rho) \leq \gamma$ if and only if the following mixed integer-real system of non-linear inequalities has a solution:

$$\lambda_{1}\gamma_{1} + \dots + \lambda_{d+1}\gamma_{d+1} \leq \gamma, \ 1 = \sum_{i=1}^{d+1} \lambda_{i}$$
$$(\gamma_{1}, \dots, \gamma_{d+1}) \in \mathcal{S}_{\mathcal{A}}, \ 0 \leq \lambda_{i}$$
$$\gamma_{i} \in \mathbb{Z}^{\times}, \ \lambda_{i} \in \mathbb{R}$$

What we actually think about

Definition

A mixed-integer bilinear (MIB) system is a collection of constraints in integer variables ${\bf x}$ and real variables ${\bf y}$ of the form

$$\mathbf{x}^T A_i \mathbf{y} < b_i, i = 1, ..., l,$$
 $C \mathbf{x} \leq \mathbf{d},$
 $E \mathbf{y} \leq \mathbf{f},$
 $\mathbf{x} \in \mathbb{Z}^m, \mathbf{y} \in \mathbb{R}^n.$

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We say the system is bounded if the polyhedron

$$\{\mathbf{y} \in \mathbb{R}^n : E\mathbf{y} \leq \mathbf{f}\}$$

is bounded, i.e. a polytope.

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Gap satisfiability problem: given $\epsilon > 0$ and MIB system \mathcal{S} , find procedure that returns SAT if \mathcal{S} has a satisfying assignment with slack ϵ , and UNSAT if \mathcal{S} is unsatisfiable.

No requirement on output if $\mathcal S$ is satisfiable, but with no satisfying assignment with slack ϵ .

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The Gap Domination Problem for MPTA is decidable in non-deterministic exponential time

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 - the use the flatness theorem from Diophantine Approximation to round these solutions to integers
- Loosely, the flatness theorem is like Minkowski's theorem on lattices
- Essentially, it shows if a convex polyhedron is "wide" enough it contains an integer point

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Theorem (Multidimensional Dirichlet)

Let (i_1, \ldots, i_m) be an m-tuple of real numbers satisfying

$$0 < i_1, \ldots, i_m < 1 \text{ and } \sum_{t=1}^m i_t = 1.$$

Then, for any $x=(x_1,\ldots,x_m)\in\mathbb{R}^m$ and $N\in\mathbb{N}$, there exists a $q\in\mathbb{Z}$ such that $1\leq q\leq N$ and

$$\max \left\{ \| \mathit{qx}_1 \|^{1/i_1}, \ldots, \, \| \mathit{qx}_m \|^{1/i_m} \right\} < \mathit{N}^{-1},$$

where by ||x|| we mean the distance to the integer closest to x.

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 - ▶ WARNING! When does $\frac{\mathbf{y}^*}{a}$ still satisfy

$$Ey < f$$
?

Another Idea?

Theorem (Minkowski's Theorem for Systems of Linear Forms)

Let $\beta_{i,j} \in \mathbb{R}$, where $1 \leq i, j \leq q$ and let $C_1, \ldots, C_k > 0$. If

$$|\det(\beta_{i,j})| \leq \prod_{i=1}^k C_i,$$

then there exists a non-zero integer point $x = (x_1, ..., x_k)$ such that

$$|x_1\beta_{i,1} + \dots + x_k\beta_{i,k}| < C_i, \quad 1 \le i \le k-1$$

 $|x_1\beta_{k,1} + \dots + x_k\beta_{k,k}| \le C_k.$