On Zone-Based Reachability Computation for Duration-Probabilistic Automata

How the Timed Automaton Lost its Tail

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Summary

- Processes that take time
- Worst-case versus average case reasoning about time

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- Duration probabilistic automata
- Forward reachability and density transformers
- Concluding remarks

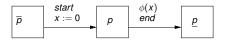
Processes that Take Time

- We are interested in processes that take some time to conclude after having started
- Can model almost anything:
 - Transmission delays in a network
 - Propagation delays in digital gates
 - Execution time of programs
 - Duration of a production step in a factory
 - Time to produce proteins in a cell
 - Cooking recipes
 - Project planning

Mathematically they are simple timed automata:

$$\boxed{\overline{p}} \xrightarrow{start} x := 0 \qquad p \qquad end \qquad \underline{p}$$

Processes that Take Time



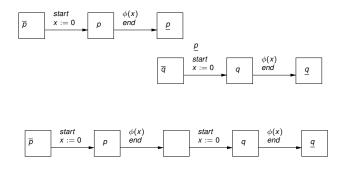
- A waiting state p;
- A start transition which resets a clock x to measure time elapsed in active state p
- An **end** transition guarded by a temporal condition $\phi(x)$

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- Condition \u03c6 can be
 - true (no constraint)
 - x = d (deterministic)
 - $x \in [a, b]$ (non-deterministic)
 - Probabilistically distributed

Composition

- Such processes can be combined:
- Sequentially, to represent precedence relations between tasks, for example p precedes q:

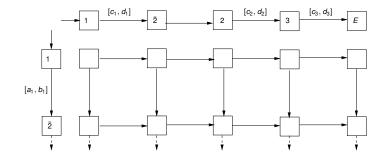


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Composition

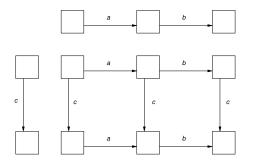
- Such processes can be combined:
- In parallel, to express partially-independent processes, sometimes competing with each other



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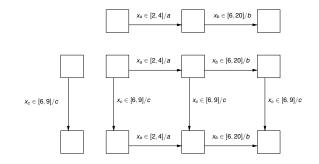
Levels of Abstraction: Untimed

- Consider two parallel processes, one doing a · b and the other doing c
- Untimed (asynchronous) modeling assumes nothing concerning duration
- Each process may take between zero and infinity time
- Consequently any interleaving in $(a \cdot b) ||c|$ is possible



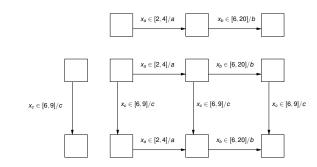
Levels of Abstraction: Timed

- Timed automata and similar formalisms add more detail
- Assume a (positive) lower bound and (finite) upper bound for the duration of each processing step



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Levels of Abstraction: Timed



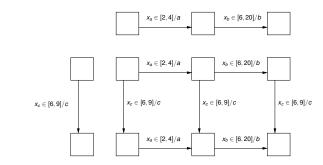
- The arithmetics of time eliminates some paths:
- Since 4 < 6, a must precede c and the set of possible paths is reduced to a ⋅ (b||c) = abc + acb</p>

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But how likely is *abc* to occur?

Possible but Unlikely

How likely is abc to occur?



Each run corresponds to a point in the duration space

$$(y_a, y_b, y_c) \in Y = [2, 4] \times [6, 20] \times [6, 9]$$

- Event *b* precedes *c* only when $y_a + y_b < y_c$
- Since y_a + y_b ranges in [8, 24] and y_c ∈ [6, 9], this is less likely than c preceding b

Levels of Abstraction: Probabilistic Timed

- Interpreting temporal guards probabilistically
- This gives precise quantitative meaning to this intuition
- It allows us to:
 - Compute probabilities of different paths (equivalence classes of qualitative behaviors)
 - Compute and compare the expected performance of schedulers, for example for job-shop problems with probabilistic step durations
 - Discard low-probability paths in verification and maybe reduce some of the state and clock explosion

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Levels of Abstraction: Probabilistic Timed

- Of course, continuous-time stochastic processes are not our invention
- But, surprisingly(?), computing these probabilities for such composed processes has rarely been attempted
- Some work in the probabilistic verification community deals with the very special case of exponential (memoryless) distribution

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- With this distribution, the time that has already elapsed since start does not influence the probability over the remaining time to termination
- Notable exceptions:
 - Alur and Bernadsky (GSMP)
 - Vicario et al (stochastic PN)

Probabilistic Interpretation of Timing Uncertainty

- We interpret a duration interval [a, b] as a uniform distribution: all values in the interval are equally likely
- This is expressed via density function

$$\phi(\mathbf{y}) = \left\{ egin{array}{cc} 1/(b-a) & ext{if } a \leq \mathbf{y} \leq b \ 0 & ext{otherwise} \end{array}
ight.$$

- Interval [a, b] is the **support** of ϕ
- The probability that the actual duration is in some interval [c, d] is

$$\mathsf{P}([\mathsf{c},\mathsf{d}]) = \int_{\mathsf{c}}^{\mathsf{d}} \phi(au) \mathsf{d} au$$

Minkowski Sum vs. Convolution

- Consider two processes with durations in [a, b] and [a', b'] that execute sequentially
- ► Their total duration is inside the Minkowski sum
 [a, b] ⊕ [a', b'] = [a + a', b + b']
- This is what timed automata will compute for you
- ► With the intervals interpreted as uniform distributions φ, φ' the total duration is distributed as their convolution

$$\phi * \phi'(\mathbf{y}) = \int \phi(\mathbf{y} - \tau) \phi'(\tau) d\tau$$

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Duration Probabilistic Automata

 Duration probabilistic automata (DPA) consist of a composition of simple DPA (SDPA) and a scheduler

$$\mathcal{A} = \mathcal{A}^1 \circ \mathcal{A}^2 \circ \cdots \circ \mathcal{A}^n \circ S$$

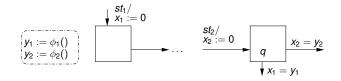
SDPA: (acyclic) alternations of waiting and active states

$$\underbrace{ \begin{array}{c} y^1 := \phi^1() \\ y^k := \phi^k() \end{array} }_{ \begin{array}{c} y^1 := 0 \end{array} } \underbrace{ \begin{array}{c} s^1 \\ x^1 := 0 \end{array} }_{ \begin{array}{c} q^1 \end{array} } \underbrace{ \begin{array}{c} x^1 = y^1 \\ e^1 \end{array} }_{ \begin{array}{c} q^1 \end{array} } \underbrace{ \begin{array}{c} s^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^{k+1} \end{array} } \underbrace{ \begin{array}{c} x^{k-1} \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^{k+1} \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} } \underbrace{ \begin{array}{c} x^k = y^k \\ e^k \end{array} }_{ \begin{array}{c} q^k \end{array} }_{ \begin{array}{c} x^k = y^k }_{ \begin{array}{c} x^k = y^k \end{array} }_{ \begin{array}{c} x^k = y^k }_{ \begin{array}{c} x^k = y^k \end{array} }_{ \begin{array}{c} x^k = y^k }_{ \begin{array}{c} x^k = y^k \end{array} }_{ \begin{array}{c} x^k = y^k }_{ \begin{array}{c} x^k }_{ \begin{array}{c} x^k = y^k }_{ \begin{array}{c} x^k }_{ \begin{array}{c} x^k = y^k }_{ \begin{array}{c} x^k }_{ \end{array} }_{ \end{array} }_{ \begin{array}{c} x^k }_{ \end{array} }_{ \end{array} }_{ \end{array} }_{ \end{array} }_{ \begin{array}{c} x^k }_{ \end{array} }_{ \end{array}$$
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- The y variables are "static" random variable drawn uniformly from the duration space
- The x variables are clocks reset to zero upon start transitions and compared to y upon end transitions
- The scheduler issues the start transitions

Clocks in Timed Automata and DPA

- A global state of a DPA is a tuple consisting of local states, some active and some inactive (waiting)
- For each active component, its corresponding clock measures the time since its start transition
- The clock values determine which end transition can be taken from this state (and in which probability)
- In other words, which active processes can "win the race"



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Zones, Symbolic States and Forward Reachability

- In timed automata the possible paths are computed using forward reachability over symbolic states
- A symbolic state is (q, Z) where q is a global state and Z is a set of clock valuations with which it is possible to reach q
- The reachability tree/graph is constructed iteratively from the initial state using a successor operator
- For every transition δ from q to q' the successor operator Post_δ is defined as:
- $(q', Z') = Post_{\delta}(q, Z')$ if Z' is the set of clock valuations possible at q' given that Z is the set of possible clock valuations at q

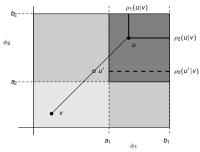
Forward Reachability for DPA

- We adapt this idea to DPA using extended symbolic states of the form (q, Z, ψ) where ψ is a (partial) density over the clock values
- Symbolic state (q', Z', ψ') is a successor of (q, Z, ψ) if:
- Given density ψ on the clock values upon entering q, the density upon taking the transition to q' is ψ'
- The successor operator is a density transformer which for start transitions is rather straightforward
- The crucial point is how to compute it in a state admitting several competing end transitions

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Intuition

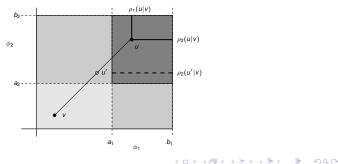
- ► Consider state q with two competing processes with durations distributed uniformly φ₁ = [a₁, b₁], φ₂ = [a₂, b₂]
- What is the probability *ρ_i(u|v)* that transition *i* wins at some clock valuation *u* = (*u*₁, *u*₂), i.e., *u_i* = *y_i*, **given** the state was entered at some *v*?
- First, this probability is non-zero only if v is a time-predecessor of u, v ∈ π(u)



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Intuition

- For transition 1 to win, process 1 should choose duration u₁ while process 2 chooses some y₂ > u₂
- Thus \(\rho_1(u|v)\) is obtained by summing up the duration probabilities above u
- If the state was entered with density ψ over clocks, we can sum up ρ_i(u|v) over π(u) according to ψ to obtain the expected ρ_i(u) as well as new densities ψ_i on clock values upon taking transition i



Definition

- For every end transition e_i outgoing from a state q with m active processes we define a density transformer T_{ri}
- ► The transformer T_{ei} computes the clock density at the time when process *i* wins the race, given the density was ψ upon entering the state

• It is defined as
$$\psi_i = \mathcal{T}_{e_i}(\psi)$$
 if

$$\psi_i(x_1,\ldots,x_m,y_1,\ldots,y_m) =$$

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Concluding Remarks

- All those densities are piecewise polynomial and can be computed. The degrees of the polynomials and their piecewiseness grow with the number of steps
- So far we consider only acyclic DPA. For cyclic ones, we need some progress in fixpoint techniques for linear operators in the spirit of Asarin and Degorre (2009)
- A prototype implementation by M. Bozga, using a slightly different technique for computing volumes, can handle, for example a product of 2 SPDA with 10 steps each
- ► As in timed automata the larger is the ratio (b a)/a the more paths have to be considered

There is still much to be done

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- There is still much to be done
- Thank you