

Observing Stochastic Processes by Timed Automata

Joost-Pieter Katoen

RWTH Aachen University
Software Modeling and Verification Group

<http://moves.rwth-aachen.de>

Workshop on Reachability Problems, Genova, 2011

joint work with Benoît Barbot, Taolue Chen,
Tingting Han and Alexandru Mereacre

September 25, 2011

Let's start easy

Discrete-time Markov chain

A **DTMC** \mathcal{D} is a tuple $(S, \mathbf{P}, \iota_{\text{init}})$ with:

- ▶ S is a countable nonempty set of **states**

Let's start easy

Discrete-time Markov chain

A **DTMC** \mathcal{D} is a tuple $(S, \mathbf{P}, \iota_{\text{init}})$ with:

- ▶ S is a countable nonempty set of **states**
- ▶ $\mathbf{P} : S \times S \rightarrow [0, 1]$, **transition probability function** s.t. $\sum_{s'} \mathbf{P}(s, s') = 1$

Let's start easy

Discrete-time Markov chain

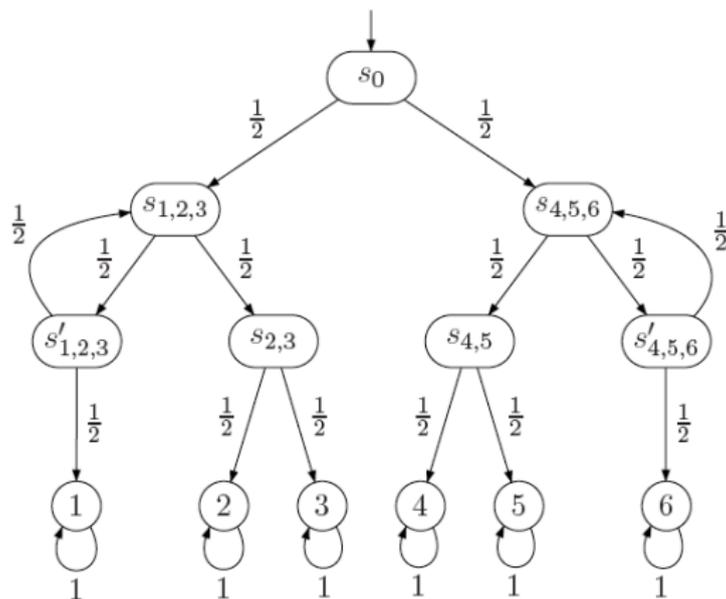
A **DTMC** \mathcal{D} is a tuple $(S, \mathbf{P}, \iota_{\text{init}})$ with:

- ▶ S is a countable nonempty set of **states**
- ▶ $\mathbf{P} : S \times S \rightarrow [0, 1]$, **transition probability function** s.t. $\sum_{s'} \mathbf{P}(s, s') = 1$
- ▶ $\iota_{\text{init}} : S \rightarrow [0, 1]$, the **initial distribution** with $\sum_{s \in S} \iota_{\text{init}}(s) = 1$

Initial states

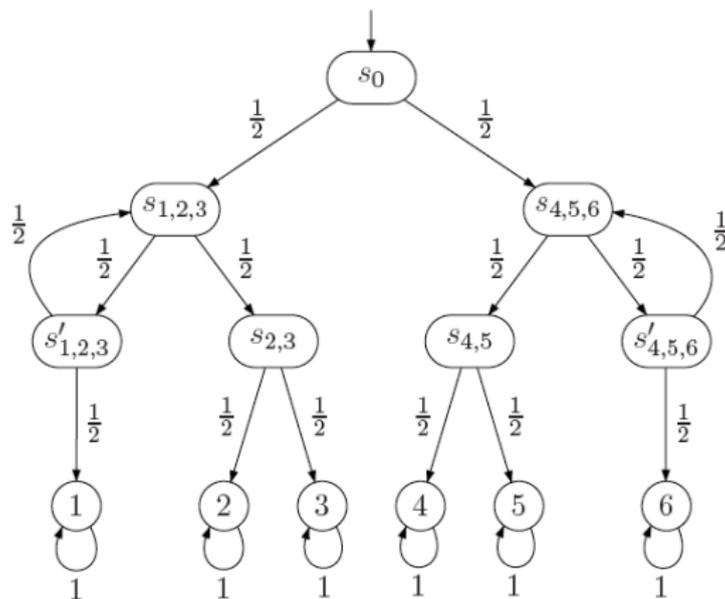
- ▶ $\iota_{\text{init}}(s)$ is the probability that DTMC \mathcal{D} starts in state s
- ▶ the set $\{s \in S \mid \iota_{\text{init}}(s) > 0\}$ are the possible **initial states**.

Simulating a die by a fair coin [Knuth & Yao]



Heads = "go left"; tails = "go right".

Simulating a die by a fair coin [Knuth & Yao]



Heads = “go left”; tails = “go right”. Does this DTMC adequately model a fair six-sided die?

Some events of interest

(Simple) reachability

Eventually reach a state in $G \subseteq S$.

Some events of interest

(Simple) reachability

Eventually reach a state in $G \subseteq S$. Formally:

$$\diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \}$$

Some events of interest

(Simple) reachability

Eventually reach a state in $G \subseteq S$. Formally:

$$\diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \}$$

Invariance, i.e., always stay in state in G :

$$\square G = \{ \pi \in Paths(\mathcal{D}) \mid \forall i \in \mathbb{N}. \pi[i] \in G \} = \overline{\overline{\diamond G}}.$$

Some events of interest

(Simple) reachability

Eventually reach a state in $G \subseteq S$. Formally:

$$\diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \}$$

Invariance, i.e., always stay in state in G :

$$\square G = \{ \pi \in Paths(\mathcal{D}) \mid \forall i \in \mathbb{N}. \pi[i] \in G \} = \overline{\overline{\diamond G}}.$$

Constrained reachability

Some events of interest

(Simple) reachability

Eventually reach a state in $G \subseteq S$. Formally:

$$\diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \}$$

Invariance, i.e., always stay in state in G :

$$\square G = \{ \pi \in Paths(\mathcal{D}) \mid \forall i \in \mathbb{N}. \pi[i] \in G \} = \overline{\overline{\diamond \bar{G}}}$$

Constrained reachability

Or “reach-avoid” properties where states in $F \subseteq S$ are forbidden:

Some events of interest

(Simple) reachability

Eventually reach a state in $G \subseteq S$. Formally:

$$\diamond G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \}$$

Invariance, i.e., always stay in state in G :

$$\square G = \{ \pi \in Paths(\mathcal{D}) \mid \forall i \in \mathbb{N}. \pi[i] \in G \} = \overline{\overline{\diamond G}}.$$

Constrained reachability

Or “reach-avoid” properties where states in $F \subseteq S$ are forbidden:

$$\overline{F} U G = \{ \pi \in Paths(\mathcal{D}) \mid \exists i \in \mathbb{N}. \pi[i] \in G \wedge \forall j < i. \pi[j] \notin F \}$$

In a similar way, $\square \diamond G$ and $\diamond \square G$ are defined.

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Aim: determine $Pr(s \models \diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \diamond G\}$.

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Aim: determine $Pr(s \models \diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \diamond G\}$.

Characterisation of reachability probabilities

- ▶ Let variable $x_s = Pr(s \models \diamond G)$ for any state s

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Aim: determine $Pr(s \models \diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \diamond G\}$.

Characterisation of reachability probabilities

- ▶ Let variable $x_s = Pr(s \models \diamond G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s = 0$

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Aim: determine $Pr(s \models \diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \diamond G\}$.

Characterisation of reachability probabilities

- ▶ Let variable $x_s = Pr(s \models \diamond G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s = 0$
 - ▶ if $s \in G$ then $x_s = 1$

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Aim: determine $Pr(s \models \diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \diamond G\}$.

Characterisation of reachability probabilities

- ▶ Let variable $x_s = Pr(s \models \diamond G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s = 0$
 - ▶ if $s \in G$ then $x_s = 1$
- ▶ For any state $s \in Pre^*(G) \setminus G$:

Reachability probabilities in finite DTMCs

Problem statement

Let \mathcal{D} be a DTMC with finite state space S , $s \in S$ and $G \subseteq S$.

Aim: determine $Pr(s \models \diamond G) = Pr_s\{\pi \in Paths(s) \mid \pi \in \diamond G\}$.

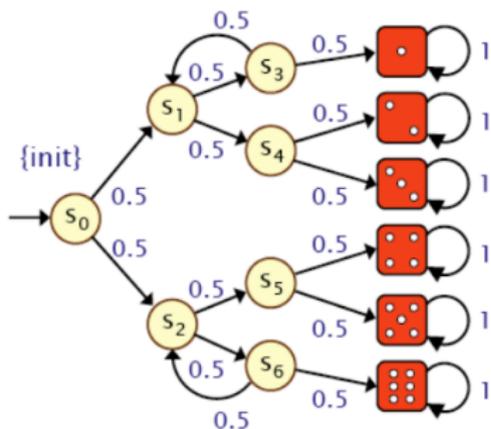
Characterisation of reachability probabilities

- ▶ Let variable $x_s = Pr(s \models \diamond G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s = 0$
 - ▶ if $s \in G$ then $x_s = 1$
- ▶ For any state $s \in Pre^*(G) \setminus G$:

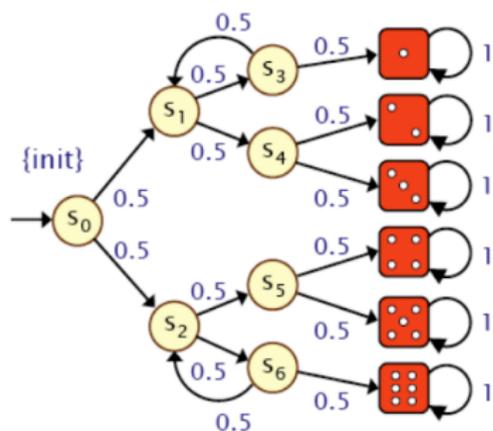
$$x_s = \underbrace{\sum_{t \in S \setminus G} \mathbf{P}(s, t) \cdot x_t}_{\text{reach } G \text{ via } t \in S \setminus G} + \underbrace{\sum_{u \in G} \mathbf{P}(s, u)}_{\text{reach } G \text{ in one step}}$$

Reachability probabilities: Knuth's die

- Consider the event $\diamond 4$



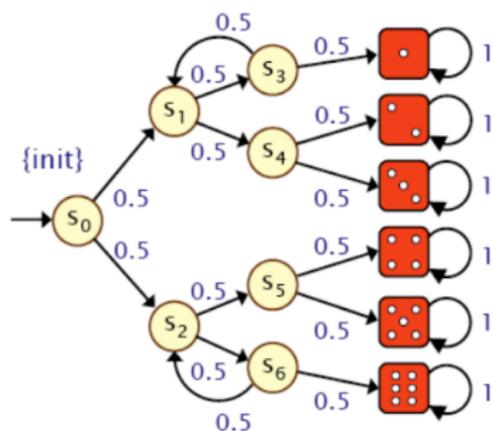
Reachability probabilities: Knuth's die



- ▶ Consider the event $\diamond 4$
- ▶ Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

Reachability probabilities: Knuth's die



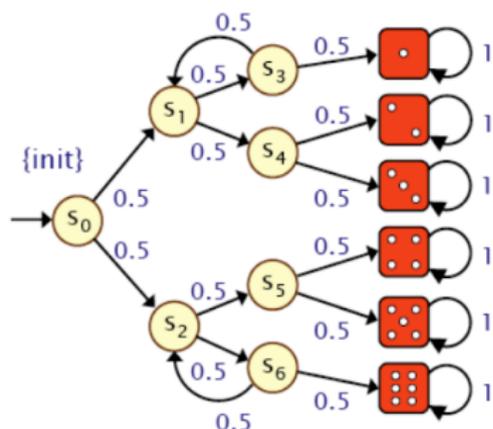
► Consider the event $\diamond 4$

► Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

Reachability probabilities: Knuth's die



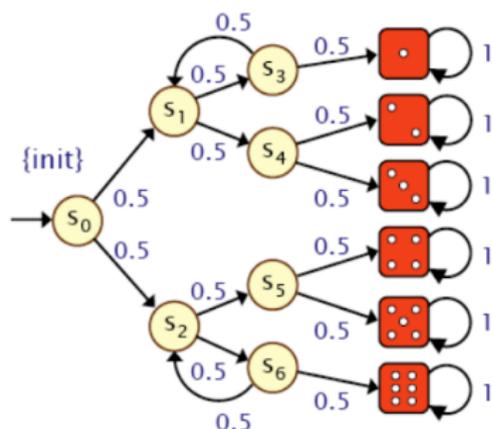
- ▶ Consider the event $\diamond 4$
- ▶ Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

$$x_{s_0} = \frac{1}{2}x_{s_1} + \frac{1}{2}x_{s_2}$$

Reachability probabilities: Knuth's die



- ▶ Consider the event $\diamond 4$
- ▶ Using the previous characterisation we obtain:

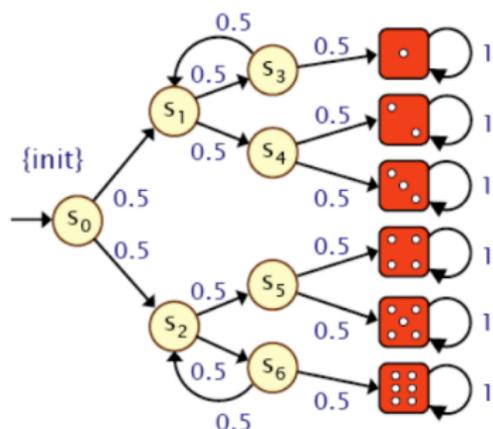
$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

$$x_{s_0} = \frac{1}{2}x_{s_1} + \frac{1}{2}x_{s_2}$$

$$x_{s_2} = \frac{1}{2}x_{s_5} + \frac{1}{2}x_{s_6}$$

Reachability probabilities: Knuth's die



- ▶ Consider the event $\diamond 4$
- ▶ Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

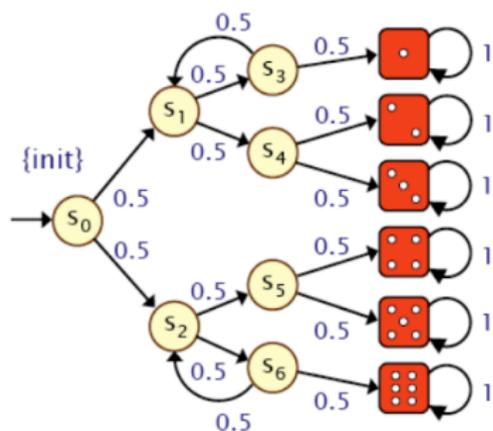
$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

$$x_{s_0} = \frac{1}{2}x_{s_1} + \frac{1}{2}x_{s_2}$$

$$x_{s_2} = \frac{1}{2}x_{s_5} + \frac{1}{2}x_{s_6}$$

$$x_{s_5} = \frac{1}{2}x_5 + \frac{1}{2}x_4$$

Reachability probabilities: Knuth's die



- ▶ Consider the event $\diamond 4$
- ▶ Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

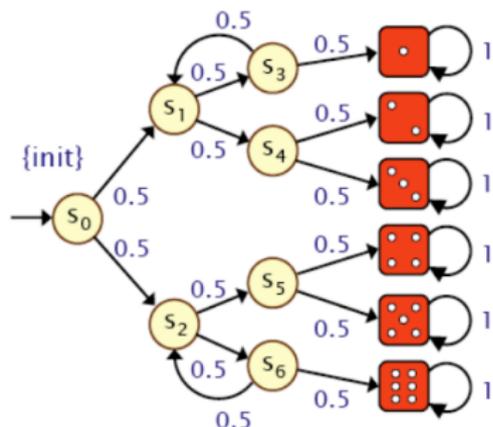
$$x_{s_0} = \frac{1}{2}x_{s_1} + \frac{1}{2}x_{s_2}$$

$$x_{s_2} = \frac{1}{2}x_{s_5} + \frac{1}{2}x_{s_6}$$

$$x_{s_5} = \frac{1}{2}x_5 + \frac{1}{2}x_4$$

$$x_{s_6} = \frac{1}{2}x_{s_2} + \frac{1}{2}x_6$$

Reachability probabilities: Knuth's die



- ▶ Consider the event $\diamond 4$
- ▶ Using the previous characterisation we obtain:

$$x_1 = x_2 = x_3 = x_5 = x_6 = 0 \text{ and } x_4 = 1$$

$$x_{s_1} = x_{s_3} = x_{s_4} = 0$$

$$x_{s_0} = \frac{1}{2}x_{s_1} + \frac{1}{2}x_{s_2}$$

$$x_{s_2} = \frac{1}{2}x_{s_5} + \frac{1}{2}x_{s_6}$$

$$x_{s_5} = \frac{1}{2}x_5 + \frac{1}{2}x_4$$

$$x_{s_6} = \frac{1}{2}x_{s_2} + \frac{1}{2}x_6$$

- ▶ Gaussian elimination yields:

$$x_{s_5} = \frac{1}{2}, x_{s_2} = \frac{1}{3}, x_{s_6} = \frac{1}{6}, \text{ and } \boxed{x_{s_0} = \frac{1}{6}}$$

Linear equation system

Reachability probabilities as linear equation system

Linear equation system

Reachability probabilities as linear equation system

- ▶ Let $S_{\neq} = Pre^*(G) \setminus G$, the states that can reach G by > 0 steps

Linear equation system

Reachability probabilities as linear equation system

- ▶ Let $S_? = Pre^*(G) \setminus G$, the states that can reach G by > 0 steps
- ▶ $\mathbf{A} = (\mathbf{P}(s, t))_{s, t \in S_?}$, the transition probabilities in $S_?$

Linear equation system

Reachability probabilities as linear equation system

- ▶ Let $S_\tau = Pre^*(G) \setminus G$, the states that can reach G by > 0 steps
- ▶ $\mathbf{A} = (\mathbf{P}(s, t))_{s, t \in S_\tau}$, the transition probabilities in S_τ
- ▶ $\mathbf{b} = (b_s)_{s \in S_\tau}$, the probs to reach G in 1 step, i.e., $b_s = \sum_{u \in G} \mathbf{P}(s, u)$

Linear equation system

Reachability probabilities as linear equation system

- ▶ Let $S_\gamma = Pre^*(G) \setminus G$, the states that can reach G by > 0 steps
- ▶ $\mathbf{A} = (\mathbf{P}(s, t))_{s, t \in S_\gamma}$, the transition probabilities in S_γ
- ▶ $\mathbf{b} = (b_s)_{s \in S_\gamma}$, the probs to reach G in 1 step, i.e., $b_s = \sum_{u \in G} \mathbf{P}(s, u)$

Then: $\mathbf{x} = (x_s)_{s \in S_\gamma}$ with $x_s = Pr(s \models \diamond G)$ is the **unique** solution of:

Linear equation system

Reachability probabilities as linear equation system

- ▶ Let $S_? = Pre^*(G) \setminus G$, the states that can reach G by > 0 steps
- ▶ $\mathbf{A} = (\mathbf{P}(s, t))_{s, t \in S_?}$, the transition probabilities in $S_?$
- ▶ $\mathbf{b} = (b_s)_{s \in S_?}$, the probs to reach G in 1 step, i.e., $b_s = \sum_{u \in G} \mathbf{P}(s, u)$

Then: $\mathbf{x} = (x_s)_{s \in S_?}$ with $x_s = Pr(s \models \diamond G)$ is the **unique** solution of:

$$\mathbf{x} = \mathbf{A} \cdot \mathbf{x} + \mathbf{b} \quad \text{or} \quad (\mathbf{I} - \mathbf{A}) \cdot \mathbf{x} = \mathbf{b}$$

where \mathbf{I} is the identity matrix of cardinality $|S_?| \times |S_?|$.

Repeated reachability and persistence

Long-run theorem

Repeated reachability and persistence

Long-run theorem

Almost surely any finite DTMC eventually reaches a BSCC and visits all its states infinitely often.

Repeated reachability and persistence

Long-run theorem

Almost surely any finite DTMC eventually reaches a BSCC and visits all its states infinitely often.

Repeated reachability = Reachability

For finite DTMC with state space S , $G \subseteq S$, and $s \in S$:

$$Pr(s \models \Box \Diamond G) = Pr(s \models \Diamond U)$$

where U is the union of all BSCCs T with $T \cap G \neq \emptyset$.

Repeated reachability and persistence

Long-run theorem

Almost surely any finite DTMC eventually reaches a BSCC and visits all its states infinitely often.

Repeated reachability = Reachability

For finite DTMC with state space S , $G \subseteq S$, and $s \in S$:

$$Pr(s \models \Box \Diamond G) = Pr(s \models \Diamond U)$$

where U is the union of all BSCCs T with $T \cap G \neq \emptyset$.

Persistency = Reachability

For finite DTMC with state space S , $G \subseteq S$, and $s \in S$:

$$Pr(s \models \Diamond \Box G) = Pr(s \models \Diamond U)$$

where U is the union of all BSCCs T with $T \subseteq G$.

Verifying ω -regular objectives = Reachability

Verifying ω -regular objectives = Reachability

Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$.

Verifying ω -regular objectives = Reachability

Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$. Then:

$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \diamond U) \quad \text{where } q_s = \delta(q_0, L(s)).$$

Verifying ω -regular objectives = Reachability

Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$. Then:

$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \diamond U) \quad \text{where } q_s = \delta(q_0, L(s)).$$

where U is the union of all accepting BSCCs in $\mathcal{D} \otimes \mathcal{A}$.

Verifying ω -regular objectives = Reachability

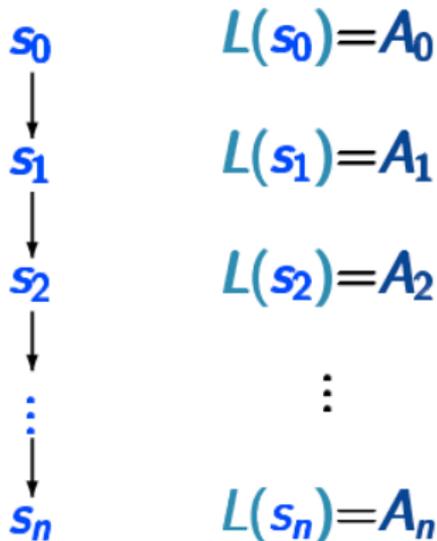
Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$. Then:

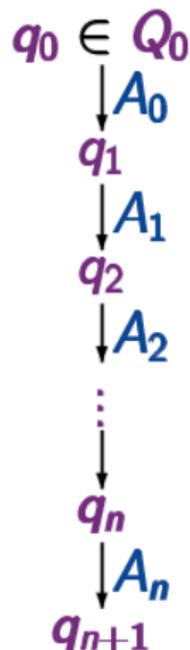
$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \diamond U) \quad \text{where } q_s = \delta(q_0, L(s)).$$

where U is the union of all accepting BSCCs in $\mathcal{D} \otimes \mathcal{A}$. BSCC $T \subseteq S \times Q$ is accepting if $T \cap (S \times L_i) = \emptyset$ and $T \cap (S \times K_i) \neq \emptyset$ for some i .

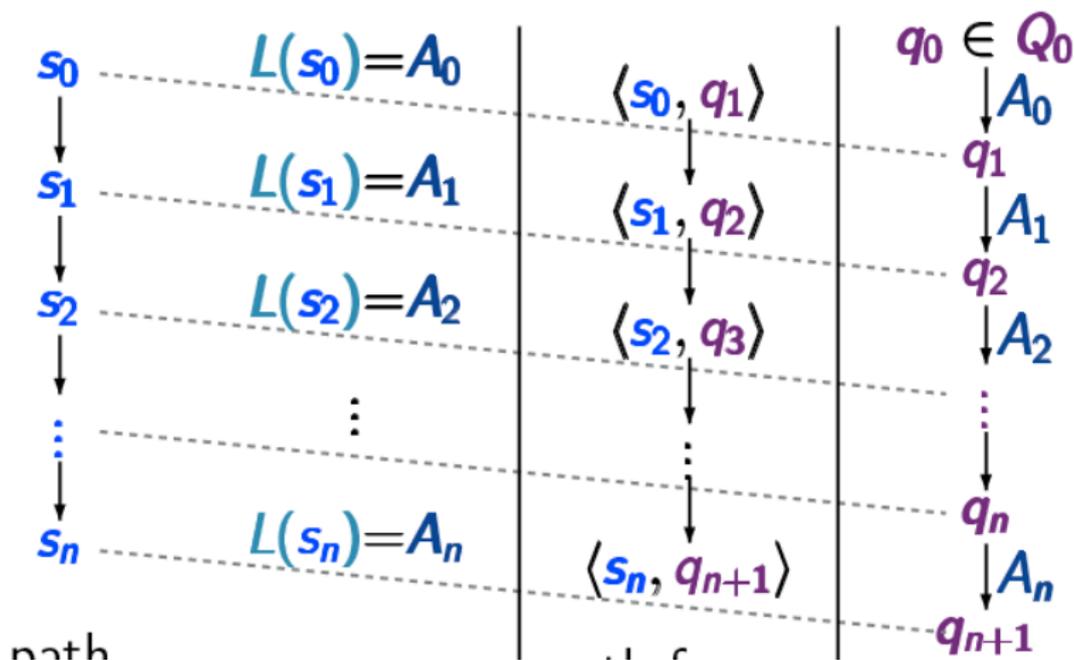
Synchronous product construction

DTMC \mathcal{D} with state space S 

path

DRA \mathcal{A} with state space Q 

Synchronous product construction \otimes

DTMC \mathcal{D} with state space S DRA \mathcal{A} with state space Q 

path

product $\mathcal{D} \otimes \mathcal{A}$

Verifying ω -regular objectives = Reachability

Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$. Then:

$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \diamond U) \quad \text{where } q_s = \delta(q_0, L(s)).$$

where U is the union of all accepting BSCCs in $\mathcal{D} \otimes \mathcal{A}$.

Verifying ω -regular objectives = Reachability

Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$. Then:

$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \diamond U) \quad \text{where } q_s = \delta(q_0, L(s)).$$

where U is the union of all accepting BSCCs in $\mathcal{D} \otimes \mathcal{A}$.

Thus the computation of probabilities for satisfying ω -regular properties boils down to computing the **reachability probabilities** for certain BSCCs in $\mathcal{D} \otimes \mathcal{A}$.

Verifying ω -regular objectives = Reachability

Verifying DRA objectives theorem

Let \mathcal{D} be a finite DTMC, s a state in \mathcal{D} , \mathcal{A} a DRA (**deterministic Rabin automaton**) with acceptance set $\{(L_1, K_1), \dots, (L_n, K_n)\}$. Then:

$$Pr^{\mathcal{D}}(s \models \mathcal{A}) = Pr^{\mathcal{D} \otimes \mathcal{A}}(\langle s, q_s \rangle \models \diamond U) \quad \text{where } q_s = \delta(q_0, L(s)).$$

where U is the union of all accepting BSCCs in $\mathcal{D} \otimes \mathcal{A}$.

Thus the computation of probabilities for satisfying ω -regular properties boils down to computing the **reachability probabilities** for certain BSCCs in $\mathcal{D} \otimes \mathcal{A}$. A graph analysis and solving systems of linear equations suffice.

Random timing



Negative exponential distribution

Negative exponential distribution

Density of exponential distribution

The density of an *exponentially distributed* r.v. Y with *rate* $\lambda \in \mathbb{R}_{>0}$ is:

$$f_Y(x) = \lambda \cdot e^{-\lambda \cdot x} \quad \text{for } x > 0 \quad \text{and } f_Y(x) = 0 \text{ otherwise}$$

Negative exponential distribution

Density of exponential distribution

The density of an *exponentially distributed* r.v. Y with *rate* $\lambda \in \mathbb{R}_{>0}$ is:

$$f_Y(x) = \lambda \cdot e^{-\lambda \cdot x} \quad \text{for } x > 0 \quad \text{and } f_Y(x) = 0 \text{ otherwise}$$

The cumulative distribution of r.v. Y with rate $\lambda \in \mathbb{R}_{>0}$ is:

$$F_Y(d) = \int_0^d \lambda \cdot e^{-\lambda \cdot x} dx = [-e^{-\lambda \cdot x}]_0^d = 1 - e^{-\lambda \cdot d}.$$

Negative exponential distribution

Density of exponential distribution

The density of an *exponentially distributed* r.v. Y with *rate* $\lambda \in \mathbb{R}_{>0}$ is:

$$f_Y(x) = \lambda \cdot e^{-\lambda \cdot x} \quad \text{for } x > 0 \quad \text{and } f_Y(x) = 0 \text{ otherwise}$$

The cumulative distribution of r.v. Y with rate $\lambda \in \mathbb{R}_{>0}$ is:

$$F_Y(d) = \int_0^d \lambda \cdot e^{-\lambda \cdot x} dx = [-e^{-\lambda \cdot x}]_0^d = 1 - e^{-\lambda \cdot d}.$$

The rate $\lambda \in \mathbb{R}_{>0}$ uniquely determines an exponential distribution.

Negative exponential distribution

Density of exponential distribution

The density of an *exponentially distributed* r.v. Y with *rate* $\lambda \in \mathbb{R}_{>0}$ is:

$$f_Y(x) = \lambda \cdot e^{-\lambda \cdot x} \quad \text{for } x > 0 \quad \text{and } f_Y(x) = 0 \text{ otherwise}$$

The cumulative distribution of r.v. Y with rate $\lambda \in \mathbb{R}_{>0}$ is:

$$F_Y(d) = \int_0^d \lambda \cdot e^{-\lambda \cdot x} dx = [-e^{-\lambda \cdot x}]_0^d = 1 - e^{-\lambda \cdot d}.$$

The rate $\lambda \in \mathbb{R}_{>0}$ uniquely determines an exponential distribution.

Variance and expectation

Let r.v. Y be exponentially distributed with rate $\lambda \in \mathbb{R}_{>0}$. Then:

Negative exponential distribution

Density of exponential distribution

The density of an *exponentially distributed* r.v. Y with *rate* $\lambda \in \mathbb{R}_{>0}$ is:

$$f_Y(x) = \lambda \cdot e^{-\lambda \cdot x} \quad \text{for } x > 0 \quad \text{and } f_Y(x) = 0 \text{ otherwise}$$

The cumulative distribution of r.v. Y with rate $\lambda \in \mathbb{R}_{>0}$ is:

$$F_Y(d) = \int_0^d \lambda \cdot e^{-\lambda \cdot x} dx = [-e^{-\lambda \cdot x}]_0^d = 1 - e^{-\lambda \cdot d}.$$

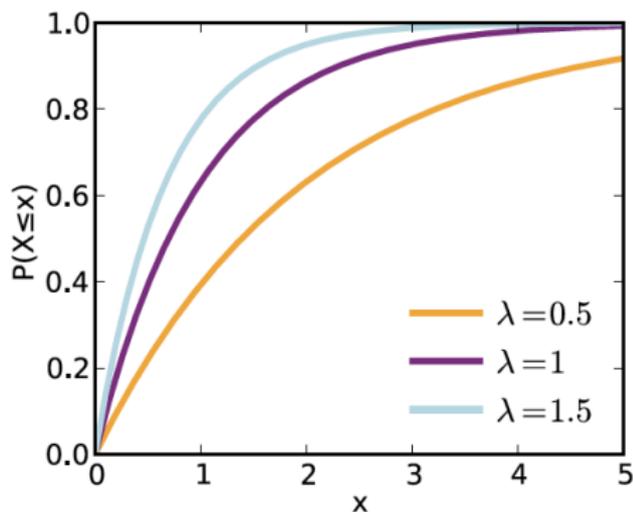
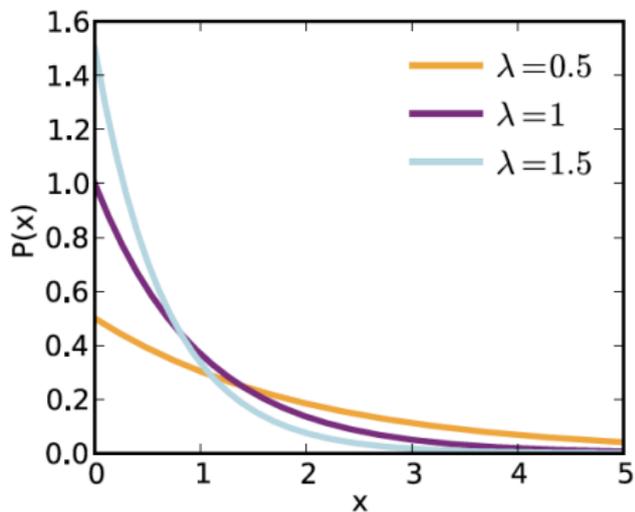
The rate $\lambda \in \mathbb{R}_{>0}$ uniquely determines an exponential distribution.

Variance and expectation

Let r.v. Y be exponentially distributed with rate $\lambda \in \mathbb{R}_{>0}$. Then:

$$\text{Expectation } E[Y] = \frac{1}{\lambda} \text{ and variance } \text{Var}[Y] = \frac{1}{\lambda^2}$$

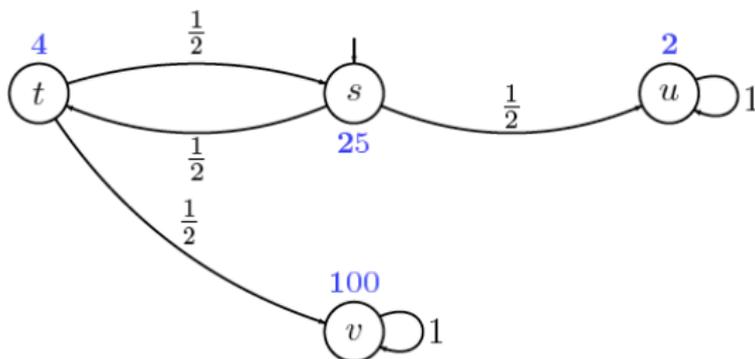
Exponential pdf and cdf



The higher λ , the faster the cdf approaches 1.

Continuous-time Markov chains

A CTMC is a DTMC with an *exit rate* function $r : S \rightarrow \mathbb{R}_{>0}$ where $r(s)$ is the rate of an exponential distribution.



$$r(s) = 25, r(t) = 4, r(u) = 2 \text{ and } r(v) = 100$$

Example: a classical perspective

A CTMC is a DTMC with an *exit rate* function $r : S \rightarrow \mathbb{R}_{>0}$ where $r(s)$ is the rate of an exponential distribution.

Example: a classical perspective

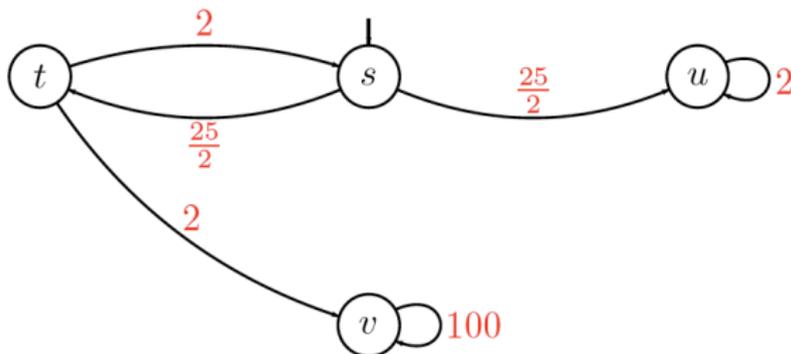
A CTMC is a DTMC with an *exit rate* function $r : S \rightarrow \mathbb{R}_{>0}$ where $r(s)$ is the rate of an exponential distribution.

A CTMC is a DTMC where transition probability function \mathbf{P} is replaced by a *transition rate* function \mathbf{R} .

Example: a classical perspective

A CTMC is a DTMC with an *exit rate* function $r : S \rightarrow \mathbb{R}_{>0}$ where $r(s)$ is the rate of an exponential distribution.

A CTMC is a DTMC where transition probability function \mathbf{P} is replaced by a *transition rate* function \mathbf{R} . We have $\mathbf{R}(s, s') = \mathbf{P}(s, s') \cdot r(s)$.



$$r(s) = 25, r(t) = 4, r(u) = 2 \text{ and } r(v) = 100$$

CTMC semantics

CTMC semantics

State-to-state timed transition probability

The probability to *move* from non-absorbing s to s' in $[0, t]$ is:

$$\frac{R(s, s')}{r(s)} \cdot \left(1 - e^{-r(s) \cdot t}\right).$$

CTMC semantics

State-to-state timed transition probability

The probability to *move* from non-absorbing s to s' in $[0, t]$ is:

$$\frac{R(s, s')}{r(s)} \cdot (1 - e^{-r(s) \cdot t}).$$

Residence time distribution

The probability to *take some* outgoing transition from s in $[0, t]$ is:

$$\int_0^t r(s) \cdot e^{-r(s) \cdot x} dx = 1 - e^{-r(s) \cdot t}$$

CTMCs are omnipresent!

- ▶ Markovian queueing networks (Kleinrock 1975)
- ▶ Stochastic Petri nets (Molloy 1977)
- ▶ Stochastic activity networks (Meyer & Sanders 1985)
- ▶ Stochastic process algebra (Herzog *et al.*, Hillston 1993)
- ▶ Probabilistic input/output automata (Smolka *et al.* 1994)
- ▶ Calculi for biological systems (Priami *et al.*, Cardelli 2002)

CTMCs are omnipresent!

- ▶ Markovian queueing networks (Kleinrock 1975)
- ▶ Stochastic Petri nets (Molloy 1977)
- ▶ Stochastic activity networks (Meyer & Sanders 1985)
- ▶ Stochastic process algebra (Herzog *et al.*, Hillston 1993)
- ▶ Probabilistic input/output automata (Smolka *et al.* 1994)
- ▶ Calculi for biological systems (Priami *et al.*, Cardelli 2002)

CTMCs are one of the most prominent models in performance analysis

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

$$\pi = s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \dots$$

such that $s_j \in S$ and $t_j \in \mathbb{R}_{>0}$.

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

$$\pi = s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \dots$$

such that $s_j \in S$ and $t_j \in \mathbb{R}_{>0}$.

Time instant t_j is the amount of time spent in state s_j .

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

$$\pi = s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \dots$$

such that $s_j \in S$ and $t_j \in \mathbb{R}_{>0}$.

Time instant t_j is the amount of time spent in state s_j .

Notations

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

$$\pi = s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \dots$$

such that $s_j \in S$ and $t_j \in \mathbb{R}_{>0}$.

Time instant t_j is the amount of time spent in state s_j .

Notations

- ▶ Let $\pi[i] := s_i$ denote the $(i+1)$ -st state along the timed path π .

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

$$\pi = s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \cdots$$

such that $s_j \in S$ and $t_j \in \mathbb{R}_{>0}$.

Time instant t_j is the amount of time spent in state s_j .

Notations

- ▶ Let $\pi[i] := s_i$ denote the $(i+1)$ -st state along the timed path π .
- ▶ Let $\pi @ t$ be the state occupied in π at time $t \in \mathbb{R}_{\geq 0}$,

Paths in a CTMC

Timed paths

Paths in CTMC \mathcal{C} are maximal (i.e., infinite) paths of alternating states and time instants:

$$\pi = s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \cdots$$

such that $s_j \in S$ and $t_j \in \mathbb{R}_{>0}$.

Time instant t_j is the amount of time spent in state s_j .

Notations

- ▶ Let $\pi[i] := s_i$ denote the $(i+1)$ -st state along the timed path π .
- ▶ Let $\pi @ t$ be the state occupied in π at time $t \in \mathbb{R}_{\geq 0}$, i.e. $\pi @ t := \pi[i]$ where i is the smallest index such that $\sum_{j=0}^i t_j > t$.

Zeno theorem

¹Zeno of Elea (490–430 BC), philosopher, famed for his paradoxes.

Zeno theorem

Zeno path

Path $s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \xrightarrow{t_2} s_3 \dots$ is called **Zeno**¹ if $\sum_i t_i$ converges.

¹Zeno of Elea (490–430 BC), philosopher, famed for his paradoxes.

Zeno theorem

Zeno path

Path $s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \xrightarrow{t_2} s_3 \dots$ is called **Zeno**¹ if $\sum_i t_i$ converges.

Example

$$s_0 \xrightarrow{1} s_1 \xrightarrow{\frac{1}{2}} s_2 \xrightarrow{\frac{1}{4}} s_3 \dots s_i \xrightarrow{\frac{1}{2^i}} s_{i+1} \dots$$

¹Zeno of Elea (490–430 BC), philosopher, famed for his paradoxes.

Zeno theorem

Zeno path

Path $s_0 \xrightarrow{t_0} s_1 \xrightarrow{t_1} s_2 \xrightarrow{t_2} s_3 \dots$ is called **Zeno**¹ if $\sum_i t_i$ converges.

Example

$$s_0 \xrightarrow{1} s_1 \xrightarrow{\frac{1}{2}} s_2 \xrightarrow{\frac{1}{4}} s_3 \dots s_i \xrightarrow{\frac{1}{2^i}} s_{i+1} \dots$$

In timed automata, such executions are typically excluded from the analysis.

Zeno theorem

For all states s in any CTMC, $Pr\{\pi \in Paths(s) \mid \pi \text{ is Zeno}\} = 0$.

¹Zeno of Elea (490–430 BC), philosopher, famed for his paradoxes.

Timed reachability events

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I .

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I . Formally:

$$\diamond^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \}$$

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I . Formally:

$$\diamond^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \}$$

Invariance, i.e., always stay in state in G in the interval I :

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I . Formally:

$$\diamond^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \}$$

Invariance, i.e., always stay in state in G in the interval I :

$$\square^I G = \{ \pi \in Paths(\mathcal{C}) \mid \forall t \in I. \pi @ t \in G \} = \overline{\overline{\diamond^I \overline{G}}}$$

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I . Formally:

$$\diamond^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \}$$

Invariance, i.e., always stay in state in G in the interval I :

$$\square^I G = \{ \pi \in Paths(\mathcal{C}) \mid \forall t \in I. \pi @ t \in G \} = \overline{\diamond^I \overline{G}}.$$

Constrained timed reachability

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I . Formally:

$$\diamond^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \}$$

Invariance, i.e., always stay in state in G in the interval I :

$$\square^I G = \{ \pi \in Paths(\mathcal{C}) \mid \forall t \in I. \pi @ t \in G \} = \overline{\diamond^I \overline{G}}.$$

Constrained timed reachability

Or “reach-avoid” properties where states in $F \subseteq S$ are forbidden:

Timed reachability events

Let CTMC \mathcal{C} with (possibly infinite) state space S .

(Simple) timed reachability

Eventually reach a state in $G \subseteq S$ in the interval I . Formally:

$$\diamond^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \}$$

Invariance, i.e., always stay in state in G in the interval I :

$$\square^I G = \{ \pi \in Paths(\mathcal{C}) \mid \forall t \in I. \pi @ t \in G \} = \overline{\diamond^I \overline{G}}$$

Constrained timed reachability

Or “reach-avoid” properties where states in $F \subseteq S$ are forbidden:

$$\overline{F} U^I G = \{ \pi \in Paths(\mathcal{C}) \mid \exists t \in I. \pi @ t \in G \wedge \forall d < t. \pi @ d \notin F \}$$

Measurability

Measurability

Measurability theorem

Events $\diamond^! G$, $\square^! G$, and $\overline{F} U^! G$ are measurable on any CTMC.

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Aim: $Pr(s \models \diamond^{\leq t} G) = Pr_s\{\pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G\}$

where Pr_s is the probability measure in CTMC \mathcal{C} with single initial state s .

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Aim: $Pr(s \models \diamond^{\leq t} G) = Pr_s\{\pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G\}$

where Pr_s is the probability measure in CTMC \mathcal{C} with single initial state s .

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr(s \models \diamond^{\leq t} G)$ for any state s

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Aim: $Pr(s \models \diamond^{\leq t} G) = Pr_s\{\pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G\}$

where Pr_s is the probability measure in CTMC \mathcal{C} with single initial state s .

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Aim: $Pr(s \models \diamond^{\leq t} G) = Pr_s\{\pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G\}$

where Pr_s is the probability measure in CTMC \mathcal{C} with single initial state s .

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t
 - ▶ if $s \in G$ then $x_s(t) = 1$ for all t

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Aim: $Pr(s \models \diamond^{\leq t} G) = Pr_s\{\pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G\}$

where Pr_s is the probability measure in CTMC \mathcal{C} with single initial state s .

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t
 - ▶ if $s \in G$ then $x_s(t) = 1$ for all t
- ▶ For any state $s \in Pre^*(G) \setminus G$:

Timed reachability probabilities in finite CTMCs

Problem statement

Let \mathcal{C} be a CTMC with finite state space S , $s \in S$, $t \in \mathbb{R}_{\geq 0}$ and $G \subseteq S$.

Aim: $Pr(s \models \diamond^{\leq t} G) = Pr_s\{\pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G\}$

where Pr_s is the probability measure in CTMC \mathcal{C} with single initial state s .

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t
 - ▶ if $s \in G$ then $x_s(t) = 1$ for all t
- ▶ For any state $s \in Pre^*(G) \setminus G$:

$$x_s(t) = \int_0^t \sum_{s' \in S} \underbrace{R(s, s') \cdot e^{-r(s) \cdot x}}_{\text{probability to move to state } s' \text{ at time } x} \cdot \underbrace{x_{s'}(t-x)}_{\text{prob. to fulfill } \diamond^{\leq t-x} G \text{ from } s'} dx$$

Reachability

Reachability

Reachability probabilities in finite DTMCs and CTMCs

Solve a system of **linear** equations (using some efficient techniques).

Reachability

Reachability probabilities in finite DTMCs and CTMCs

Solve a system of **linear** equations (using some efficient techniques).

Timed reachability probabilities in finite CTMCs

Solve a system of **Volterra integral** equations.

Reachability

Reachability probabilities in finite DTMCs and CTMCs

Solve a system of **linear** equations (using some efficient techniques).

Timed reachability probabilities in finite CTMCs

Solve a system of **Volterra integral** equations. This is in general non-trivial, inefficient, and has several pitfalls such as numerical stability.

Reachability

Reachability probabilities in finite DTMCs and CTMCs

Solve a system of **linear** equations (using some efficient techniques).

Timed reachability probabilities in finite CTMCs

Solve a system of **Volterra integral** equations. This is in general non-trivial, inefficient, and has several pitfalls such as numerical stability.

Solution

Reduce the problem of computing $Pr(s \models \diamond^{\leq t} G)$ to an alternative problem for which well-known efficient techniques exist:

Reachability

Reachability probabilities in finite DTMCs and CTMCs

Solve a system of **linear** equations (using some efficient techniques).

Timed reachability probabilities in finite CTMCs

Solve a system of **Volterra integral** equations. This is in general non-trivial, inefficient, and has several pitfalls such as numerical stability.

Solution

Reduce the problem of computing $Pr(s \models \diamond^{\leq t} G)$ to an alternative problem for which well-known efficient techniques exist: computing **transient** probabilities.

Timed reachability probabilities = transient probabilities

Aim

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} .

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important.

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Let CTMC $\mathcal{C} = (S, \mathbf{P}, r, \iota_{\text{init}})$ and $G \subseteq S$.

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Let CTMC $\mathcal{C} = (S, \mathbf{P}, r, \iota_{\text{init}})$ and $G \subseteq S$. The CTMC $\mathcal{C}[G] = (S, \mathbf{P}_G, r, \iota_{\text{init}})$ with $\mathbf{P}_G(s, t) = \mathbf{P}(s, t)$ if $s \notin G$ and $\mathbf{P}_G(s, s) = 1$ if $s \in G$.

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Let CTMC $\mathcal{C} = (S, \mathbf{P}, r, \iota_{\text{init}})$ and $G \subseteq S$. The CTMC $\mathcal{C}[G] = (S, \mathbf{P}_G, r, \iota_{\text{init}})$ with $\mathbf{P}_G(s, t) = \mathbf{P}(s, t)$ if $s \notin G$ and $\mathbf{P}_G(s, s) = 1$ if $s \in G$.

All outgoing transitions of $s \in G$ are replaced by a single self-loop at s .

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Let CTMC $\mathcal{C} = (S, \mathbf{P}, r, \iota_{\text{init}})$ and $G \subseteq S$. The CTMC $\mathcal{C}[G] = (S, \mathbf{P}_G, r, \iota_{\text{init}})$ with $\mathbf{P}_G(s, t) = \mathbf{P}(s, t)$ if $s \notin G$ and $\mathbf{P}_G(s, s) = 1$ if $s \in G$.

All outgoing transitions of $s \in G$ are replaced by a single self-loop at s .

Lemma

$$\underbrace{Pr(s \models \diamond^{\leq t} G)}_{\text{timed reachability in } \mathcal{C}} =$$

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Let CTMC $\mathcal{C} = (S, \mathbf{P}, r, \iota_{\text{init}})$ and $G \subseteq S$. The CTMC $\mathcal{C}[G] = (S, \mathbf{P}_G, r, \iota_{\text{init}})$ with $\mathbf{P}_G(s, t) = \mathbf{P}(s, t)$ if $s \notin G$ and $\mathbf{P}_G(s, s) = 1$ if $s \in G$.

All outgoing transitions of $s \in G$ are replaced by a single self-loop at s .

Lemma

$$\underbrace{Pr(s \models \diamond^{\leq t} G)}_{\text{timed reachability in } \mathcal{C}} = \underbrace{Pr(s \models \diamond^{=t} G)}_{\text{timed reachability in } \mathcal{C}[G]} =$$

Timed reachability probabilities = transient probabilities

Aim

Compute $Pr(s \models \diamond^{\leq t} G)$ in CTMC \mathcal{C} . Observe that once a path π reaches G within t time, then the remaining behaviour along π is not important. This suggests to make all states in G absorbing.

Let CTMC $\mathcal{C} = (S, \mathbf{P}, r, \iota_{\text{init}})$ and $G \subseteq S$. The CTMC $\mathcal{C}[G] = (S, \mathbf{P}_G, r, \iota_{\text{init}})$ with $\mathbf{P}_G(s, t) = \mathbf{P}(s, t)$ if $s \notin G$ and $\mathbf{P}_G(s, s) = 1$ if $s \in G$.

All outgoing transitions of $s \in G$ are replaced by a single self-loop at s .

Lemma

$$\underbrace{Pr(s \models \diamond^{\leq t} G)}_{\text{timed reachability in } \mathcal{C}} = \underbrace{Pr(s \models \diamond^{=t} G)}_{\text{timed reachability in } \mathcal{C}[G]} = \underbrace{\underline{p}(t)}_{\text{transient prob. in } \mathcal{C}[G]} \text{ with } \underline{p}(0) = \mathbf{1}_s.$$

Transient distribution theorem

Theorem: transient distribution as ordinary differential equation

The **transient** probability vector $\underline{p}(t) = (p_{s_1}(t), \dots, p_{s_k}(t))$ satisfies:

$$\underline{p}'(t) = \underline{p}(t) \cdot (\mathbf{R} - \mathbf{r}) \quad \text{given} \quad \underline{p}(0)$$

where \mathbf{r} is the diagonal matrix of vector \underline{r} .

Transient distribution theorem

Theorem: transient distribution as ordinary differential equation

The **transient** probability vector $\underline{p}(t) = (p_{s_1}(t), \dots, p_{s_k}(t))$ satisfies:

$$\underline{p}'(t) = \underline{p}(t) \cdot (\mathbf{R} - \mathbf{r}) \quad \text{given} \quad \underline{p}(0)$$

where \mathbf{r} is the diagonal matrix of vector \underline{r} .

Solution technique:

Transform the CTMC (again), and then truncate a Taylor-MacLaurin expansion.

Transient distribution theorem

Theorem: transient distribution as ordinary differential equation

The **transient** probability vector $\underline{p}(t) = (p_{s_1}(t), \dots, p_{s_k}(t))$ satisfies:

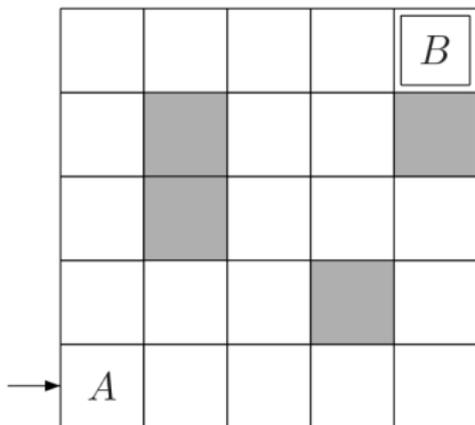
$$\underline{p}'(t) = \underline{p}(t) \cdot (\mathbf{R} - \mathbf{r}) \quad \text{given} \quad \underline{p}(0)$$

where \mathbf{r} is the diagonal matrix of vector \underline{r} .

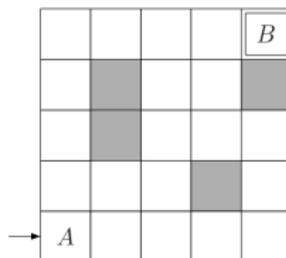
Solution technique:

Transform the CTMC (again), and then truncate a Taylor-MacLaurin expansion. This yields a **polynomial-time approximation** algorithm.

Robot navigation

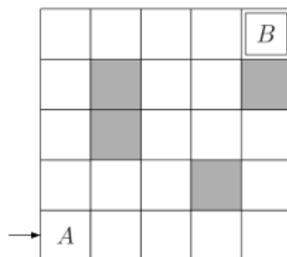


Robot navigation



- ▶ The robot randomly moves through the cells, and resides in a cell for an **exponentially** distributed amount of time.
- ▶ Gray cells are **dangerous**; the robot should leave them quickly.

Robot navigation



- ▶ The robot randomly moves through the cells, and resides in a cell for an **exponentially** distributed amount of time.
- ▶ Gray cells are **dangerous**; the robot should leave them quickly.

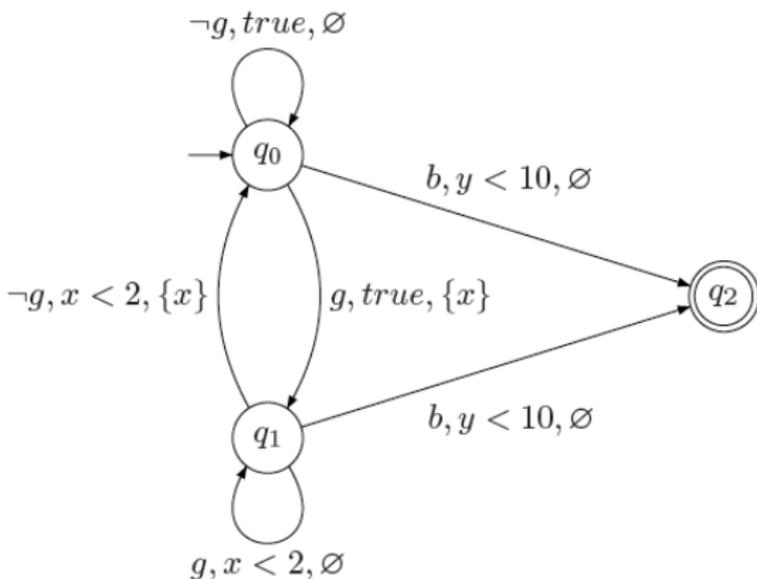
Property:

What is the probability to reach *B* from *A* within 10 time units while residing in any **dangerous** zone for at most 2 time units?

Robot navigation: property

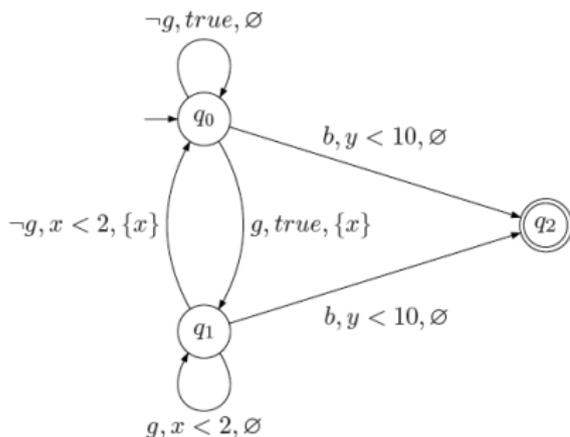
Property:

What is the probability to reach B from A within 10 time units while residing in any **dangerous** zone for at most 2 time units?



Deterministic timed automata

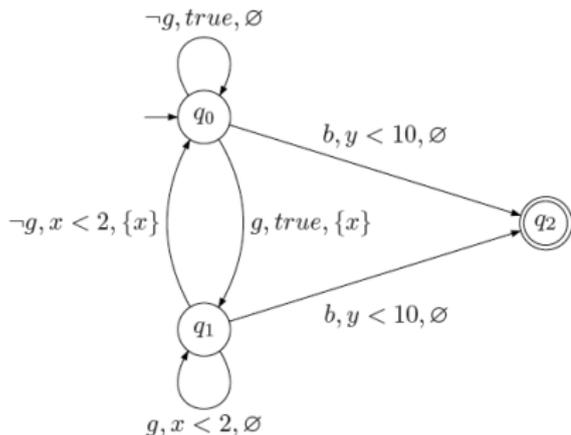
A **D**eterministic **T**imed **A**utomaton (DTA) A is a tuple $(\Sigma, X, Q, q_0, F, \rightarrow)$:



- ▶ Σ - *alphabet*
- ▶ X - finite set of *clocks*
- ▶ Q - finite set of *locations*
- ▶ $q_0 \in Q$ - *initial* location
- ▶ $F \subseteq Q$ - *accept* locations
- ▶ $\rightarrow \in Q \times \Sigma \times \mathcal{C}(X) \times 2^X \times Q$
- *transition relation*;

Deterministic timed automata

A **D**eterministic **T**imed **A**utomaton (DTA) A is a tuple $(\Sigma, X, Q, q_0, F, \rightarrow)$:



- ▶ Σ - *alphabet*
- ▶ X - finite set of *clocks*
- ▶ Q - finite set of *locations*
- ▶ $q_0 \in Q$ - *initial* location
- ▶ $F \subseteq Q$ - *accept* locations
- ▶ $\rightarrow \in Q \times \Sigma \times \mathcal{C}(X) \times 2^X \times Q$
- *transition relation*;

Determinism: $q \xrightarrow{a, g, X} q'$ and $q \xrightarrow{a, g', X'} q''$ implies $g \cap g' = \emptyset$

What are we interested in?

Problem statement:

Given model CTMC \mathcal{C} and specification DTA \mathcal{A} , determine the fraction of runs in \mathcal{C} that satisfy \mathcal{A} :

$$Pr(\mathcal{C} \models \mathcal{A}) := Pr^{\mathcal{C}}\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$$

Theoretical facts

Well-definedness

For any CTMC \mathcal{C} and DTA \mathcal{A} , the set $\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$ is measurable.

Theoretical facts

Well-definedness

For any CTMC \mathcal{C} and DTA \mathcal{A} , the set $\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$ is measurable.

Characterizing the probability of $\mathcal{C} \models \mathcal{A}$

$Pr(\mathcal{C} \models \mathcal{A})$ equals the probability of accepting paths in $\mathcal{C} \otimes \mathcal{A}$.

Theoretical facts

Well-definedness

For any CTMC \mathcal{C} and DTA \mathcal{A} , the set $\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$ is measurable.

Characterizing the probability of $\mathcal{C} \models \mathcal{A}$

$\Pr(\mathcal{C} \models \mathcal{A})$ equals the probability of accepting paths in $\mathcal{C} \otimes \mathcal{A}$.

Zone graph construction

1. Reachability probabilities in $\mathcal{C} \otimes \mathcal{A}$ and $ZG(\mathcal{C} \otimes \mathcal{A})$ coincide
2. $ZG(\mathcal{C} \otimes \mathcal{A})$ and $\mathcal{C} \otimes ZG(\mathcal{A})$ are isomorphic
3. $\mathcal{C} \otimes ZG(\mathcal{A})$ is a piecewise-deterministic Markov process [Davis, 1993]

Theoretical facts

Well-definedness

For any CTMC \mathcal{C} and DTA \mathcal{A} , the set $\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$ is measurable.

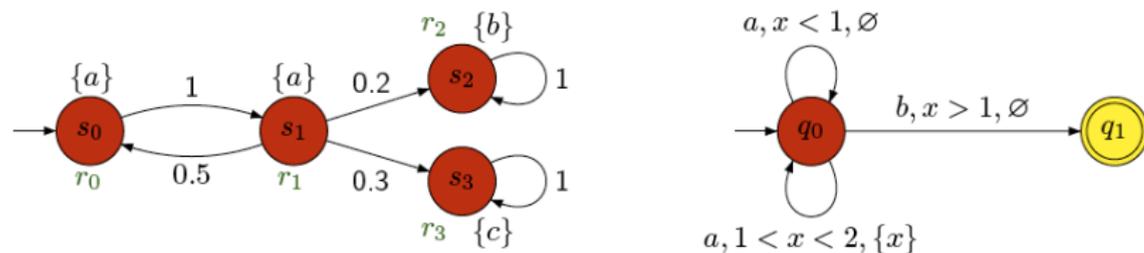
Characterizing the probability of $\mathcal{C} \models \mathcal{A}$ under finite acceptance

$Pr(\mathcal{C} \models \mathcal{A})$ equals the probability of accepting paths in $\mathcal{C} \otimes ZG(\mathcal{A})$.

Characterizing the probability of $\mathcal{C} \models \mathcal{A}$ under Muller acceptance

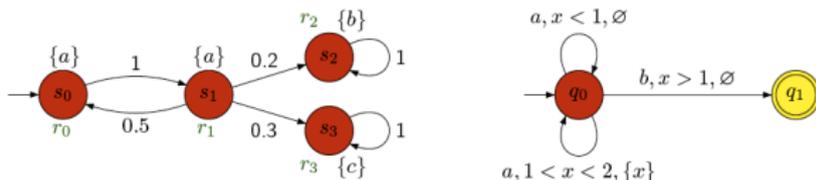
$Pr(\mathcal{C} \models \mathcal{A})$ equals the probability of accepting BSCCs in $\mathcal{C} \otimes ZG(\mathcal{A})$.

Product construction: example

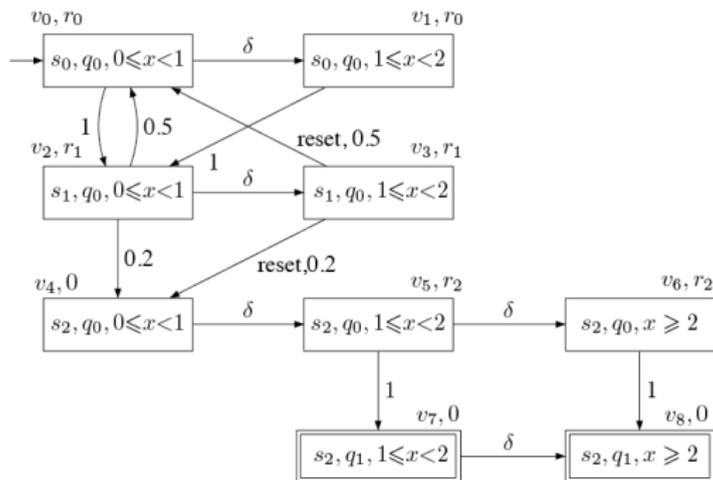


An example CTMC \mathcal{C} (left) and DTA \mathcal{A} (right)

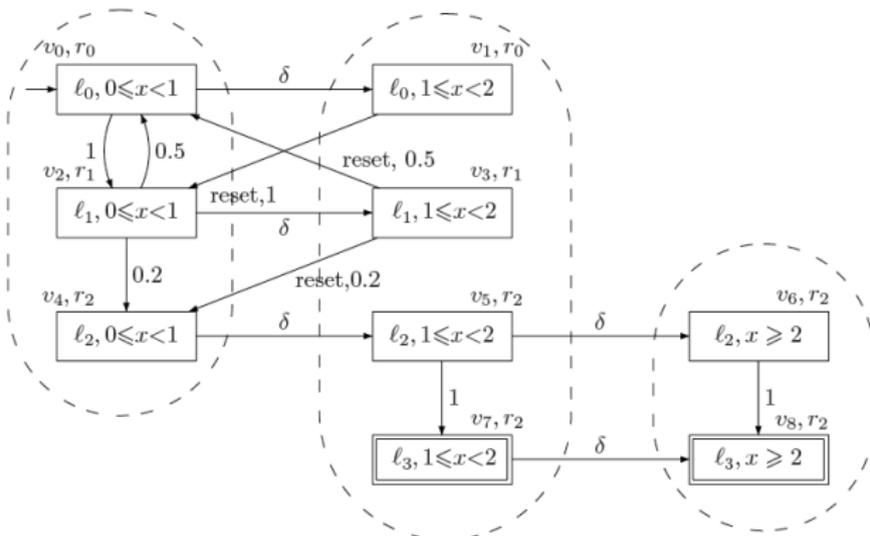
Product construction: example



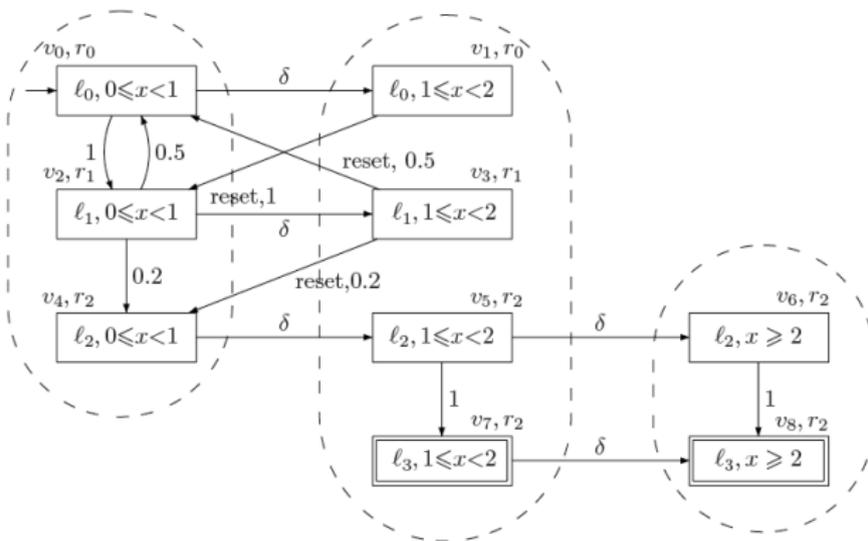
An example CTMC \mathcal{C} (left up) and DTA \mathcal{A} (right up) and $\mathcal{C} \otimes \text{ZG}(\mathcal{A})$ (below)



One-clock DTA: partitioning $\mathcal{C} \otimes ZG(\mathcal{A})$

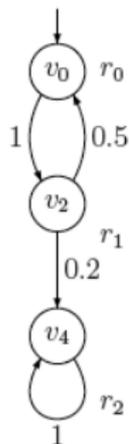
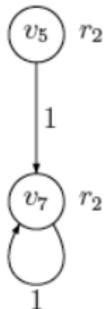
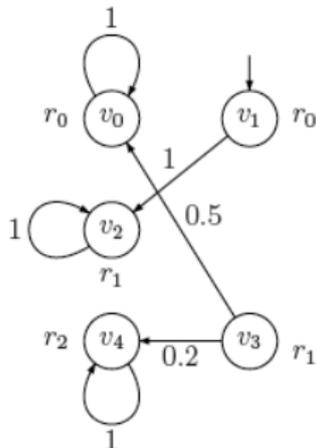
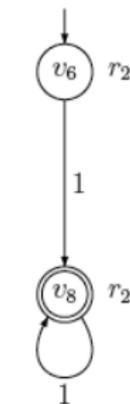
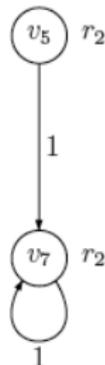


One-clock DTA: partitioning $\mathcal{C} \otimes ZG(\mathcal{A})$



- ▶ constants $c_0 < \dots < c_m$ in \mathcal{A} yields $m+1$ subgraphs.
- ▶ subgraph i captures behaviour of \mathcal{C} and \mathcal{A} in $[c_i, c_{i+1})$.
- ▶ any subgraph is a CTMC, resets lead to subgraph 0, delays to $i+1$.
- ▶ a subgraph with its resets yields an “augmented” CTMC.

One-clock DTA: partitioning $\mathcal{C} \otimes ZG(\mathcal{A})$

(a) \mathcal{C}_0 (b) \mathcal{C}_1 (c) \mathcal{C}_1^a (d) \mathcal{C}_2

One-clock DTA: characterizing $Pr(\mathcal{C} \models \mathcal{A})$

Theorem

For CTMC \mathcal{C} with initial distribution ι_{init} and 1-clock DTA \mathcal{A} we have:

$$Pr(\mathcal{C} \models \mathcal{A}) = \iota_{\text{init}} \cdot \mathbf{u}$$

where \mathbf{u} is the solution of the linear equation system $\mathbf{x} \cdot \mathbf{M} = \mathbf{f}$, with

$$\mathbf{M} = \left(\begin{array}{c|c} \mathbf{I}_{n_0} - \mathbf{B}_{m-1} & \mathbf{A}_{m-1} \\ \hline \hat{\mathbf{P}}_m^a & \mathbf{I}_{n_m} - \mathbf{P}_m \end{array} \right)$$

and \mathbf{f} is the characterizing vector of the final states in subgraph m , and \mathbf{A} and \mathbf{B} are obtained from transient probabilities in all subgraphs.

One-clock DTA: characterizing $Pr(\mathcal{C} \models \mathcal{A})$

Theorem

For CTMC \mathcal{C} with initial distribution ι_{init} and 1-clock DTA \mathcal{A} we have:

$$Pr(\mathcal{C} \models \mathcal{A}) = \iota_{\text{init}} \cdot \mathbf{u}$$

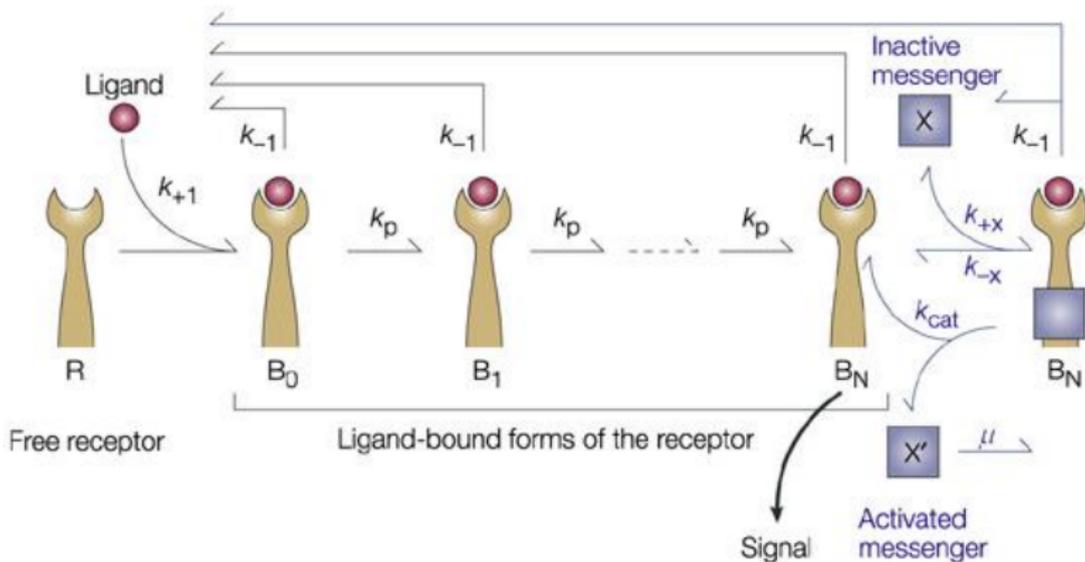
where \mathbf{u} is the solution of the linear equation system $\mathbf{x} \cdot \mathbf{M} = \mathbf{f}$, with

$$\mathbf{M} = \left(\begin{array}{c|c} \mathbf{I}_{n_0} - \mathbf{B}_{m-1} & \mathbf{A}_{m-1} \\ \hline \hat{\mathbf{P}}_m^a & \mathbf{I}_{n_m} - \mathbf{P}_m \end{array} \right)$$

and \mathbf{f} is the characterizing vector of the final states in subgraph m , and \mathbf{A} and \mathbf{B} are obtained from transient probabilities in all subgraphs.

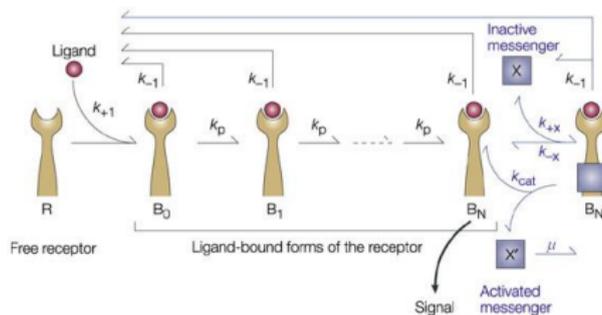
For **single-clock** DTA, reachability probabilities in (our) PDPs are characterized by the least solution of a **linear equation system**, whose coefficients are solutions of ODEs (= transient probabilities in CTMCs).

Systems biology: immune-receptor signaling



[Goldstein et. al., Nat. Reviews Immunology, 2004]

Systems biology: immune-receptor signaling



- ▶ M ligands can react with a receptor R with rate k_{+1} yielding a ligand-receptor LR
- ▶ LR undergoes a sequence of N modifications with a constant rate k_p yielding B_1, \dots, B_N
- ▶ LR B_N can link with an inactive messenger with rate k_{+x} yielding a ligand-receptor-messenger (LRM).
- ▶ The LRM decomposes into an **active** messenger with rate k_{cat}

Verification results

M	#CTMC states	No lumping		With lumping			
		# \otimes states	time(s)	#blocks	time(s)	%transient	%lumping
1	18	31	0	13	0	0%	0%
2	150	203	0.06	56	0.05	58%	39%
3	774	837	1.36	187	0.84	64%	30%
4	3024	2731	17.29	512	9.19	73%	24%
5	9756	7579	152.54	1213	73.4	76%	21%
6	27312	18643	1547.45	2579	457.35	78%	20%
7	68496	41743	11426.46	5038	3185.6	85%	14%
8	157299	86656	23356.5	9200	11950.8	81%	18%
9	336049	169024	71079.15	15906	38637.28	76%	22%
10	675817	312882	205552.36	26256	116314.41	71%	26%

In the case of no lumping, 99% of time is spent on transient analysis

Multi-multi-core model checking

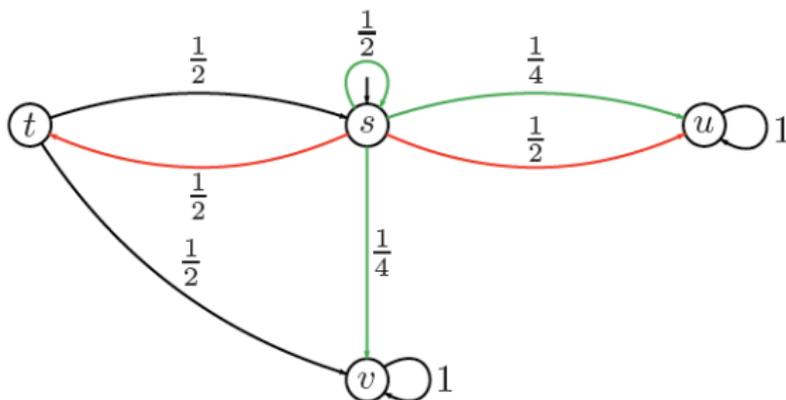
N	4 Cores		20 Cores	
	time(s)	speedup	time(s)	speedup
3	0.45	3.03	0.42	3.22
4	5.3	3.26	3.44	5.02
5	44.73	3.41	15.87	9.61
6	620.16	2.50	160.58	9.64
7	4142.19	2.76	949.32	12.04
8	8168.62	2.86	1722.63	13.56
9	23865.17	2.98	5457.01	13.03
10	70623.46	2.91	16699.22	12.31

Parallelization of the transient analysis only; not the lumping.

Non-determinism: MDP

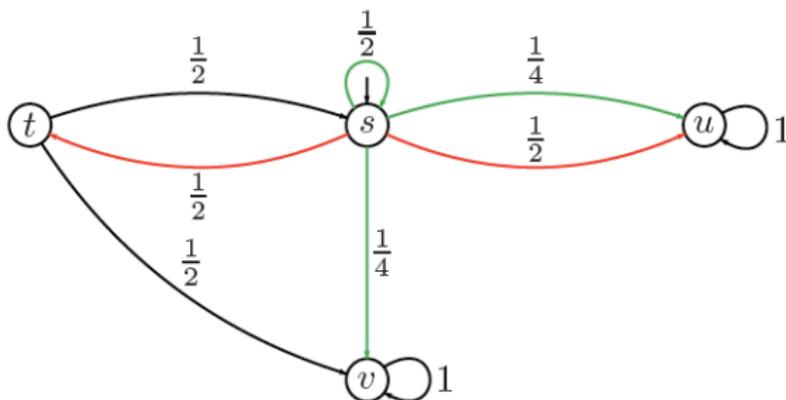
Non-determinism: MDP

An MDP is a DTMC in which in any state a non-deterministic choice between probability distributions exists.



Non-determinism: MDP

An MDP is a DTMC in which in any state a non-deterministic choice between probability distributions exists.



Set of enabled distributions (= colors) in state s is $Act(s) = \{\alpha, \beta\}$ where

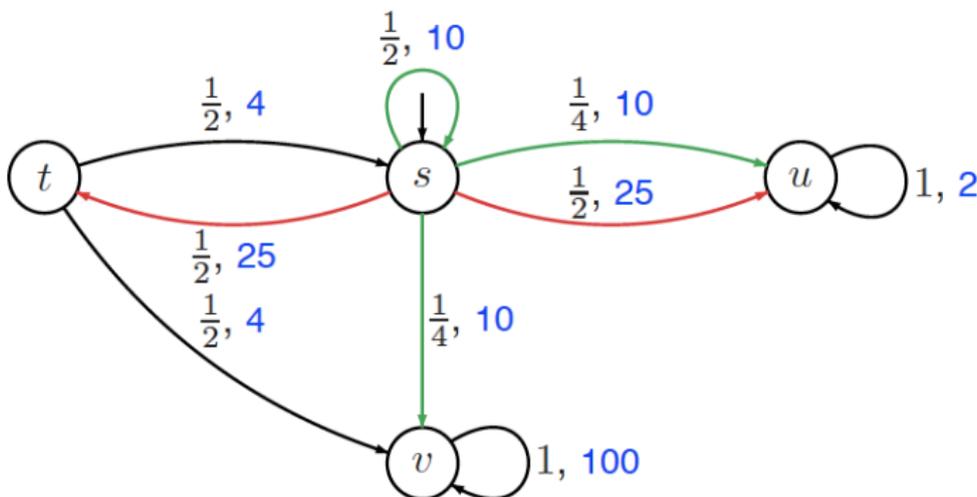
- ▶ $\mathbf{P}(s, \alpha, s) = \frac{1}{2}$, $\mathbf{P}(s, \alpha, t) = 0$ and $\mathbf{P}(s, \alpha, u) = \mathbf{P}(s, \alpha, v) = \frac{1}{4}$
- ▶ $\mathbf{P}(s, \beta, s) = \mathbf{P}(s, \beta, v) = 0$, and $\mathbf{P}(s, \beta, t) = \mathbf{P}(s, \beta, u) = \frac{1}{2}$

Continuous-time Markov decision processes

A CTMDP is an MDP with an *exit rate* function $r : S \times Act \rightarrow \mathbb{R}_{>0}$ where $r(s, \alpha)$ is the rate of an exponential distribution.

Continuous-time Markov decision processes

A CTMDP is an MDP with an *exit rate* function $r : S \times Act \rightarrow \mathbb{R}_{>0}$ where $r(s, \alpha)$ is the rate of an exponential distribution. State residence times thus depend on the selected distribution.



$$r(s, \alpha) = 10 \text{ and } r(s, \beta) = 25$$

Timed reachability objectives

Policy

Non-determinism is reduced by a [policy](#).

Timed reachability objectives

Policy

Non-determinism is reduced by a **policy**. A policy \mathfrak{G} is a (measurable) function that takes a state and the elapsed time so far, and maps this onto a distribution (= color).

Timed reachability objectives

Policy

Non-determinism is reduced by a **policy**. A policy \mathfrak{G} is a (measurable) function that takes a state and the elapsed time so far, and maps this onto a distribution (= color).

Timed reachability

Let $G \subseteq S$ be a finite set of goal states and $t \in \mathbb{R}_{\geq 0}$ a deadline.

Time-bounded reachability probability from state s under policy \mathfrak{G} :

$$Pr^{\mathfrak{G}}(s \models \diamond^{\leq t} G) = Pr_s^{\mathcal{L}^{\mathfrak{G}}} \{ \pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G \}$$

Timed reachability objectives

Policy

Non-determinism is reduced by a **policy**. A policy \mathfrak{G} is a (measurable) function that takes a state and the elapsed time so far, and maps this onto a distribution (= color).

Timed reachability

Let $G \subseteq S$ be a finite set of goal states and $t \in \mathbb{R}_{\geq 0}$ a deadline.

Time-bounded reachability probability from state s under policy \mathfrak{G} :

$$Pr^{\mathfrak{G}}(s \models \diamond^{\leq t} G) = Pr_s^{\mathcal{L}^{\mathfrak{G}}} \{ \pi \in Paths(s) \mid \pi \models \diamond^{\leq t} G \}$$

Analysis focuses on obtaining **lower-** and **upper**bounds, e.g.,

$$Pr^{\max}(s \models \diamond^{\leq t} G) = \sup_{\mathfrak{G}} Pr^{\mathfrak{G}}(s \models \diamond^{\leq t} G)$$

where \mathfrak{G} ranges over all possible policies.

Maximal timed reachability

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr^{\max}(s \models \diamond^{\leq t} G)$ for any state s

Maximal timed reachability

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr^{\max}(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t

Maximal timed reachability

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr^{\max}(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t
 - ▶ if $s \in G$ then $x_s(t) = 1$ for all t

Maximal timed reachability

Characterisation of timed reachability probabilities

- ▶ Let function $x_s(t) = Pr^{\max}(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t
 - ▶ if $s \in G$ then $x_s(t) = 1$ for all t
- ▶ For any state $s \in Pre^*(G) \setminus G$:

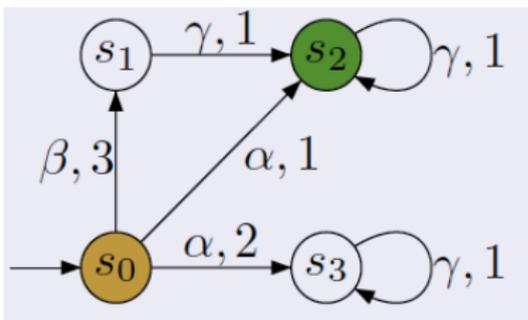
Maximal timed reachability

Characterisation of timed reachability probabilities

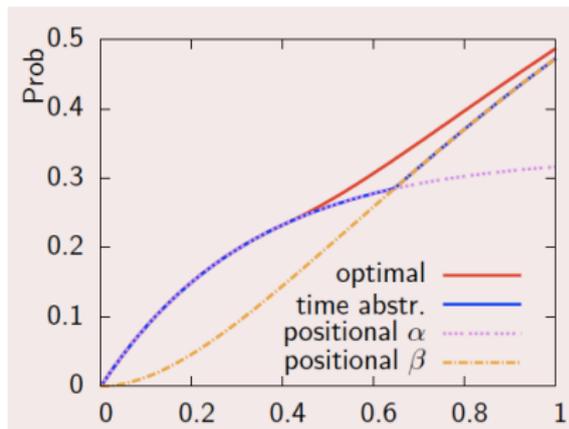
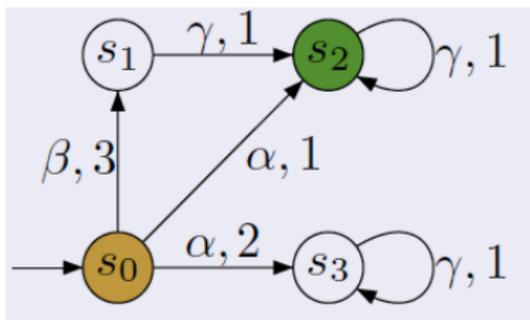
- ▶ Let function $x_s(t) = Pr^{\max}(s \models \diamond^{\leq t} G)$ for any state s
 - ▶ if G is not reachable from s , then $x_s(t) = 0$ for all t
 - ▶ if $s \in G$ then $x_s(t) = 1$ for all t
- ▶ For any state $s \in Pre^*(G) \setminus G$:

$$x_s(t) = \max_{\alpha \in Act(s)} \int_0^t \sum_{s' \in S} \underbrace{\mathbf{R}(s, \alpha, s') \cdot e^{-r(s, \alpha) \cdot x}}_{\substack{\text{probability to move to} \\ \text{state } s' \text{ at time } x \\ \text{under action } \alpha}} \cdot \underbrace{x_{s'}(t-x)}_{\substack{\text{max. prob.} \\ \text{to fulfill } \diamond^{\leq t-x} G \\ \text{from } s'}} dx$$

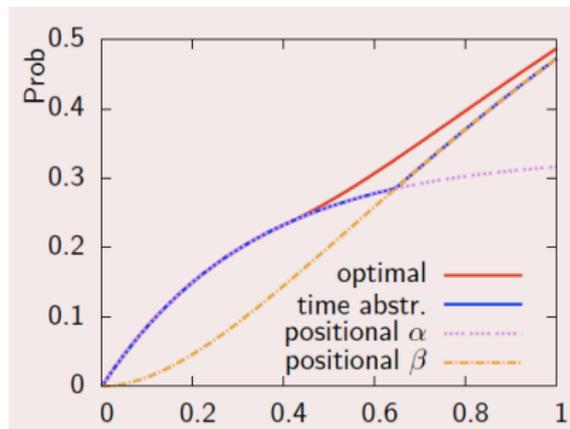
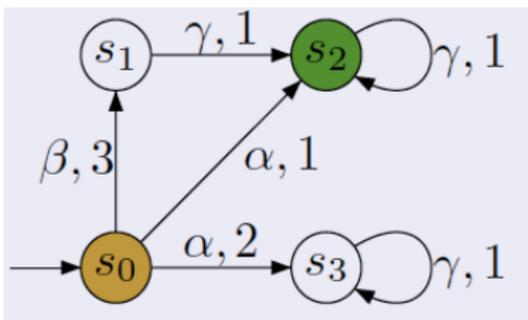
Timed policies are optimal



Timed policies are optimal

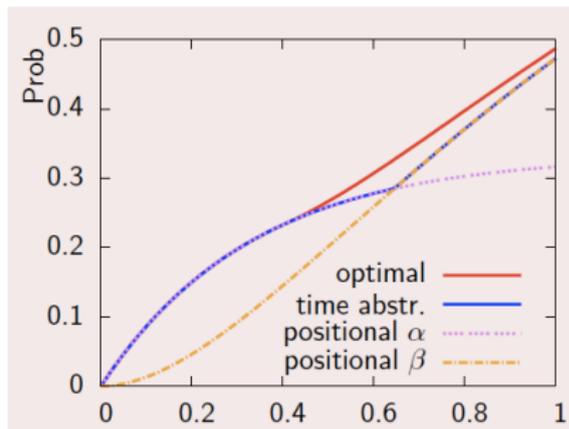
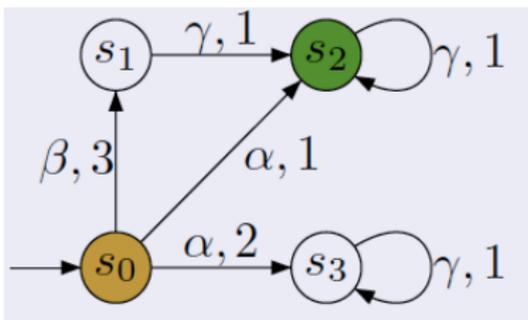


Timed policies are optimal



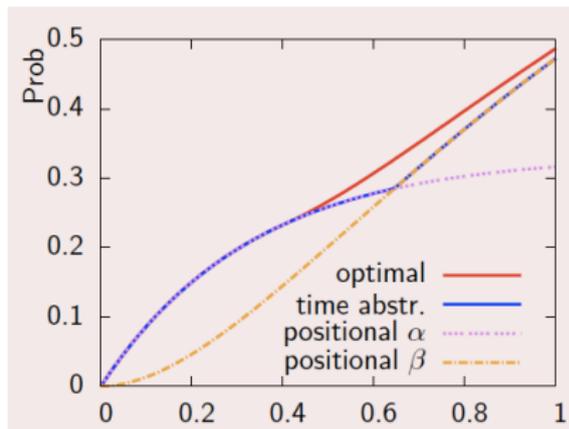
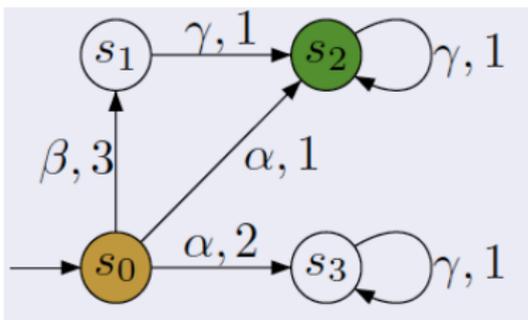
- ▶ Timed policies are optimal; any time-abstract policy is inferior.

Timed policies are optimal



- ▶ Timed policies are optimal; any time-abstract policy is inferior.
- ▶ If long time remains: choose β ; if short time remains: choose α .

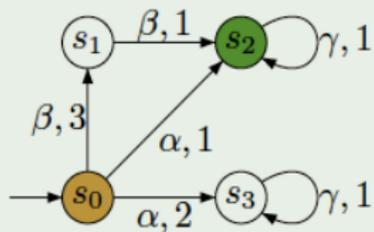
Timed policies are optimal



- ▶ Timed policies are optimal; any time-abstract policy is inferior.
- ▶ If long time remains: choose β ; if short time remains: choose α .
- ▶ Optimal policy for $t=1$: choose α if $1-t_0 \leq \ln 3 - \ln 2$, otherwise β

Discretisation

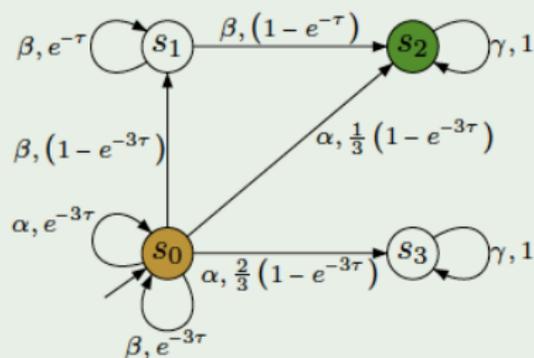
Continuous-time MDP \mathcal{C}



Exponential distributions

Reachability in d time \approx

Discrete-time MDP \mathcal{C}_τ



Discrete probability distributions

Reachability in $\frac{d}{\tau}$ steps

Checking CTMDPs against DTA objectives

Problem statement:

Given model CTMDP \mathcal{C} and specification DTA \mathcal{A} , determine the maximal fraction of runs in \mathcal{C} that satisfying \mathcal{A} :

$$Pr^{\max}(\mathcal{C} \models \mathcal{A}) := \sup_{\mathcal{G}} Pr^{\mathcal{G}}\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$$

Checking CTMDPs against DTA objectives

Problem statement:

Given model CTMDP \mathcal{C} and specification DTA \mathcal{A} , determine the maximal fraction of runs in \mathcal{C} that satisfying \mathcal{A} :

$$Pr^{\max}(\mathcal{C} \models \mathcal{A}) := \sup_{\mathcal{G}} Pr^{\mathcal{G}}\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$$

Characterizing the maximal probability of $\mathcal{C} \models \mathcal{A}$

1. $Pr^{\max}(\mathcal{C} \models \mathcal{A})$ equals the maximal probability of accepting paths in $\mathcal{C} \otimes \mathcal{A}$.

Checking CTMDPs against DTA objectives

Problem statement:

Given model CTMDP \mathcal{C} and specification DTA \mathcal{A} , determine the maximal fraction of runs in \mathcal{C} that satisfying \mathcal{A} :

$$Pr^{\max}(\mathcal{C} \models \mathcal{A}) := \sup_{\mathcal{G}} Pr^{\mathcal{G}}\{\text{Paths in } \mathcal{C} \text{ accepted by } \mathcal{A}\}$$

Characterizing the maximal probability of $\mathcal{C} \models \mathcal{A}$

1. $Pr^{\max}(\mathcal{C} \models \mathcal{A})$ equals the maximal probability of accepting paths in $\mathcal{C} \otimes \mathcal{A}$.
2. equals the maximal probability of accepting paths in $\mathcal{C} \otimes ZG(\mathcal{A})$.

One-clock DTA: characterizing $Pr^{\max}(C \models A)$

One-clock DTA: characterizing $Pr^{\max}(\mathcal{C} \models \mathcal{A})$

Verifying a CTMC against a 1-clock DTA

$Pr(\mathcal{C} \models \mathcal{A})$ can be characterised as the unique solution of a **linear equation system** whose coefficients are **transient probabilities** in CTMC \mathcal{C} .

One-clock DTA: characterizing $Pr^{\max}(\mathcal{C} \models \mathcal{A})$

Verifying a CTMC against a 1-clock DTA

$Pr(\mathcal{C} \models \mathcal{A})$ can be characterised as the unique solution of a **linear equation system** whose coefficients are **transient probabilities** in CTMC \mathcal{C} .

Verifying a CTMDP against a 1-clock DTA

$Pr^{\max}(\mathcal{C} \models \mathcal{A})$ can be characterised as the unique solution of a **linear inequation system** whose coefficients are **maximal timed reachability probabilities** in CTMDP \mathcal{C} .

For details, please consult the paper in the RP'11 proceedings.

Related work

- ▶ Observers for timed automata (Aceto *et al.* [JLAP 2003](#))
- ▶ Timed automata for GSMPs (Brazdil *et al.* [HSCC 2011](#))
- ▶ PTCTL model checking of PTA (Kwiatkowska *et al.* [TCS 2002](#))
- ▶ CSL with regular expressions (Baier *et al.* [IEEE TSE 2007](#))
- ▶ CSL with one-clock DTA as time constraints (Donatelli *et al.* [IEEE TSE 2009](#))
 - ▶ for single-clock DTA, our results coincide
 - ▶ ... but we obtain the results in a different manner
- ▶ Probabilistic semantics of TA (Baier *et al.* [LICS 2008](#))

Epilogue

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP
- ▶ Efficient numerical algorithm for 1-clock DTA:
 - ▶ using **standard** means: zone graph construction, graph analysis, transient analysis, linear equation systems.

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP
- ▶ Efficient numerical algorithm for 1-clock DTA:
 - ▶ using **standard** means: zone graph construction, graph analysis, transient analysis, linear equation systems.
 - ▶ **three orders** of magnitude faster than alternative approaches.

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP
- ▶ Efficient numerical algorithm for 1-clock DTA:
 - ▶ using **standard** means: zone graph construction, graph analysis, transient analysis, linear equation systems.
 - ▶ **three orders** of magnitude faster than alternative approaches.
 - ▶ natural support for **parallelisation** and **bisimulation minimisation**.

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP
- ▶ Efficient numerical algorithm for 1-clock DTA:
 - ▶ using **standard** means: zone graph construction, graph analysis, transient analysis, linear equation systems.
 - ▶ **three orders** of magnitude faster than alternative approaches.
 - ▶ natural support for **parallelisation** and **bisimulation minimisation**.
- ▶ Discretization approach for multiple-clock DTA with error bounds.

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP
- ▶ Efficient numerical algorithm for 1-clock DTA:
 - ▶ using **standard** means: zone graph construction, graph analysis, transient analysis, linear equation systems.
 - ▶ **three orders** of magnitude faster than alternative approaches.
 - ▶ natural support for **parallelisation** and **bisimulation minimisation**.
- ▶ Discretization approach for multiple-clock DTA with error bounds.
- ▶ For CTMDPs: similar approach using linear **inequations**.

Epilogue

Take-home messages

- ▶ Timed reachability in a CTMC \mathcal{C} = transient analysis of \mathcal{C}
- ▶ DTA acceptance of a CTMC \mathcal{C} = reachability probability in a PDP
- ▶ Efficient numerical algorithm for 1-clock DTA:
 - ▶ using **standard** means: zone graph construction, graph analysis, transient analysis, linear equation systems.
 - ▶ **three orders** of magnitude faster than alternative approaches.
 - ▶ natural support for **parallelisation** and **bisimulation minimisation**.
- ▶ Discretization approach for multiple-clock DTA with error bounds.
- ▶ For CTMDPs: similar approach using linear **inequations**.
- ▶ Prototypical tool-support for 1-clock DTA (to be in PRISM).