

6. Probabilities: Markov chains and statistical model checking

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<https://fm-dcc.github.io/sv2425>



CISTER - Research Centre in
Real-Time & Embedded
Computing Systems

Where we are

- Introduction to model-checking
- CCS: a simple language for concurrency
 - Syntax
 - Semantics
 - Equivalence
 - mCRL2: modelling
- Dynamic logic
 - Syntax
 - Semantics
 - Relation with equivalence
 - mCRL2: verification
- Timed Automata
 - Syntax
 - Semantics (composition, Zeno)
 - Equivalence
 - UPPAAL: modelling
- Temporal logics (LTL/CTL)
 - Syntax
 - Semantics
 - UPPAAL: verification
- Probabilistic and stochastic systems
 - Going probabilistic
 - UPPAAL: monte-carlo

Going probabilistic

Systems can get very complex

- E.g., 5 components, 3 possible traces each
- No communication (pure interleaving)
- Many permutations

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- E.g., 5 components, 3 possible traces each
- No communication (pure interleaving)
- Many permutations
- More components, more traces – untreatable
- Verifying deadlock freedom (and others) requires traversing all states
- **Approximation:**
 - traverse only part of the states
 - give more **priority** to some actions
 - return (statistically) likelihood of a given property

- $\alpha : S \rightarrow N \times S$ Moore machine
- $\alpha : S \rightarrow \text{Bool} \times S^N$ deterministic automata
- $\alpha : S \rightarrow \text{Bool} \times P(S)^N$ non-deterministic automata (reactive)
- $\alpha : S \rightarrow P(N \times S)$ non deterministic LTS (generative)
- $\alpha : S \rightarrow (S + 1)^N$ partial deterministic LTS
- $\alpha : S \rightarrow P(S)$ unlabelled TS
- $\alpha : S \rightarrow D(S)$ Markov chain

Markov chains

$$\alpha : S \rightarrow D(S)$$

where $D(S)$ is the set of all **discrete probability distributions** on set S

A Markov chain goes from a state s to a state s' with probability p if

$$\alpha(s) = \mu \quad \text{with} \quad \mu(s') = p > 0$$

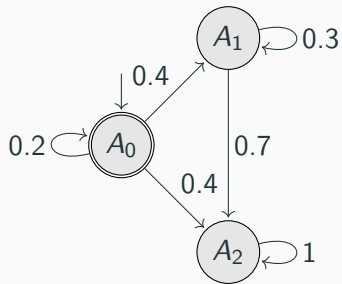
Recall

$\mu : S \rightarrow [0, 1]$ is a **discrete probability distribution** if

- $\{s \in S \mid \mu(s) > 0\}$, is finite (called the **support** of μ), and
- $\sum_{s \in S} \mu(s) = 1$

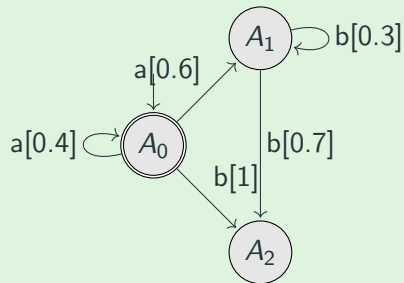
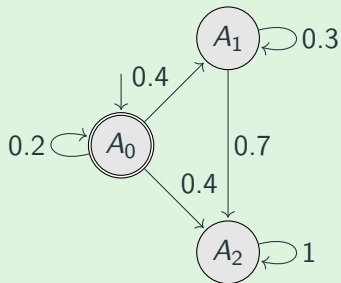
Examples

- **Dirac distribution:** $\mu_s^1 = \{s \rightarrow 1\}$
- **Product distribution:** $(\mu_1 \times \mu_2)\langle s, t \rangle = \mu_1(s) \times \mu_2(t)$



$$\alpha : S \rightarrow (D(S) + 1)^N$$

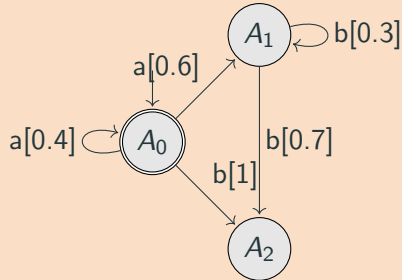
Ex. 6.1: Formalise the systems below as functions



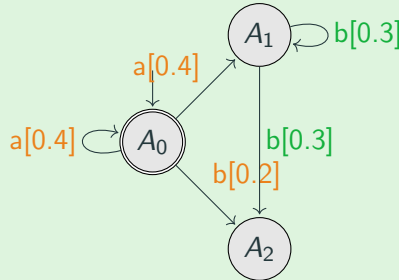
Notions of bisimulation arise naturally.

$$\alpha : S \rightarrow D((S \times N) + 1)$$

Before (reactive)

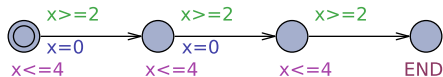


Ex. 6.2: Now (generative) – formalise it

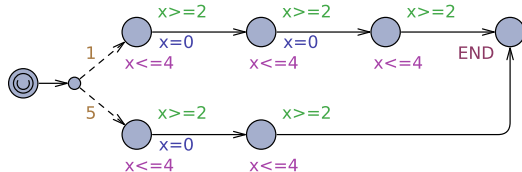


Probabilities in Uppaal

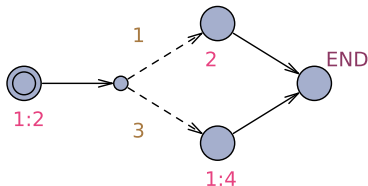
A1



A2



A3



$$\langle L, L_0, Act, C, Tr, Inv \rangle$$

where

- L is a set of **locations**, and $L_0 \subseteq L$ the set of **initial** locations
- Act is a set of **actions** and C a set of **clocks**
- $Tr \subseteq L \times \mathcal{C}(C) \times Act \times \mathcal{P}(C) \times \mathbb{N} \times L$ is the **transition relation**

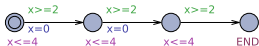
$$\ell_1 \xrightarrow{g, a, U, w} \ell_2$$

denotes a transition from location ℓ_1 to ℓ_2 , **labelled** by a , enabled if **guard** g is valid, which, when performed, **resets** the set U of **clocks**, **with a probability given by the weight** w

- $Inv : L \longrightarrow \mathcal{C}(C) + \mathbb{Q}$ is the assignment of **invariants or rates** (of an **exponential distribution**) to locations

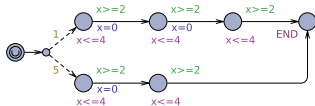
where $\mathcal{C}(C)$ denotes the set of clock constraints over a set C of clock variables

A1

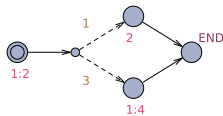


- Probability of $\langle A1_0, \bar{0} \rangle \xrightarrow{0.5} \langle A1_0, \bar{0.5} \rangle$?
- Probability of $\langle A2_0, \bar{0} \rangle \xrightarrow{0.5} \langle A2_0, \bar{0.5} \rangle$?

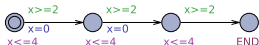
A2



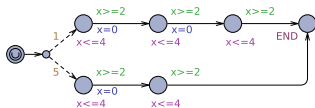
A3



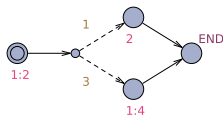
A1



A2

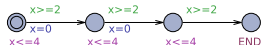


A3

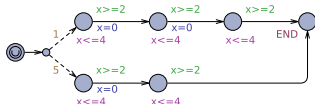


- Probability of $\langle A1_0, \bar{0} \rangle \xrightarrow{0.5} \langle A1_0, \bar{0.5} \rangle$?
- Probability of $\langle A2_0, \bar{0} \rangle \xrightarrow{0.5} \langle A2_0, \bar{0.5} \rangle$?
- Probability of $\langle A3_0, \bar{0} \rangle \xrightarrow{0.5} \langle A3_0, \bar{0.5} \rangle$?

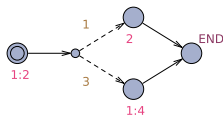
A1



A2

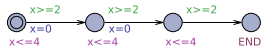


A3

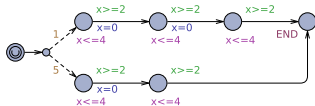


- Probability of $\langle A1_0, \bar{0} \rangle \xrightarrow{0.5} \langle A1_0, \bar{0.5} \rangle$?
- Probability of $\langle A2_0, \bar{0} \rangle \xrightarrow{0.5} \langle A2_0, \bar{0.5} \rangle$?
- Probability of $\langle A3_0, \bar{0} \rangle \xrightarrow{0.5} \langle A3_0, \bar{0.5} \rangle$?
- Probability of reaching $A1_1$?
- Probability of reaching $A2_1$?

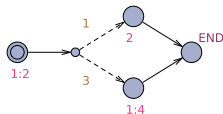
A1



A2



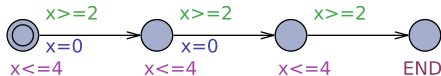
A3



- Probability of $\langle A1_0, \bar{0} \rangle \xrightarrow{0.5} \langle A1_0, \bar{0.5} \rangle$?
- Probability of $\langle A2_0, \bar{0} \rangle \xrightarrow{0.5} \langle A2_0, \bar{0.5} \rangle$?
- Probability of $\langle A3_0, \bar{0} \rangle \xrightarrow{0.5} \langle A3_0, \bar{0.5} \rangle$?
- Probability of reaching $A1_1$?
- Probability of reaching $A2_1$?
- Probability of reaching $A3_{END}$ in less than 4.3?
= ...

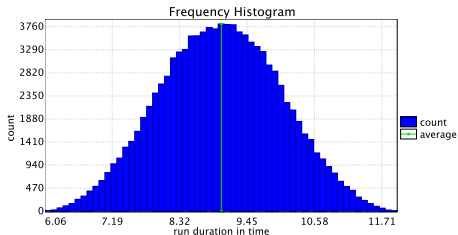
A1: When does it end?

A1



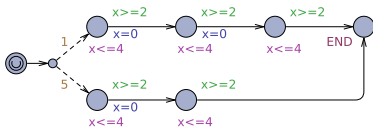
- Run 102000 times
- Histogram: how many times it took [9..9.1] seconds?
- ...

A1's histogram



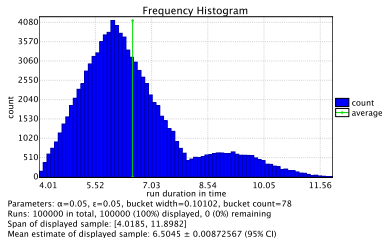
Parameters: $\alpha=0.05$, $\epsilon=0.05$, bucket width=0.10002, bucket count=59
Runs: 102000 in total, 102000 (100%) displayed, 0 (0%) remaining
Span of displayed sample: [6.06618, 11.9675]
Mean estimate of displayed sample: 9.00732 ± 0.00614869 (95% CI)

A2

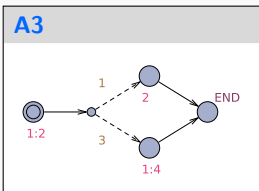


- Run 100000 times
- Histogram: how many times it took [9..9.1] seconds?
- ...

A2's histogram

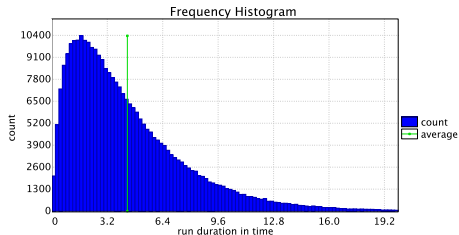


A3: When does it end?



- Run 300000 times
- Histogram: how many times it took [9..9.1] seconds?
- ...

A3's histogram

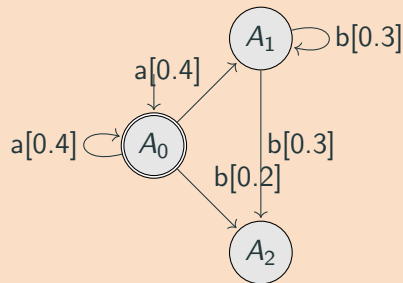


Parameters: $\alpha=0.05$, $\epsilon=0.05$, bucket width=0.19991, bucket count=100
Runs: 300000 in total, 299181 (99.727%) displayed, 819 (0.273%) remaining
Span of displayed sample: [0.0045298, 19.9958]
Mean estimate of displayed sample: 4.31514 ± 0.0118734 (95% CI)

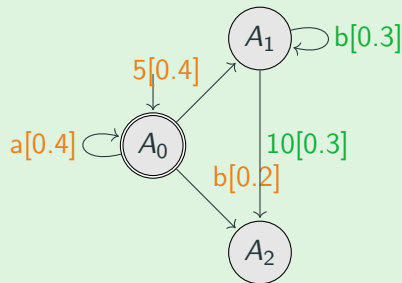
$$\alpha : S \rightarrow D_{disc}((S \times N) + 1)$$

$$\alpha : S \rightarrow D_{cont}((S \times (N + \mathcal{R}_0^+) + 1))$$

Before (PTS)



Ex. 6.3: Now (Timed PTS) – formalise it



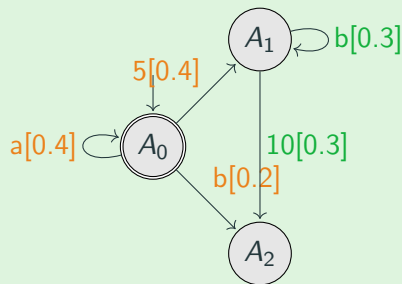
$$\alpha : S \rightarrow D_{disc}((S \times N) + 1)$$

$$\alpha : S \rightarrow D_{cont}((S \times (N + \mathcal{R}_0^+) + 1)$$

Notes

- Continuous time: continuous distribution
- Probabilities both at continuous **delays** and discrete **transitions**.

Ex. 6.3: Now (Timed PTS) – formalise it



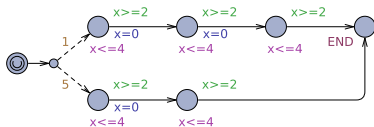
Probabilistic queries in Uppaal

- `Pr[c<=10; 100] ([] safe)` – runs 100 stochastic simulations and estimates the probability of safe remaining true within 10 cost units, based on 100 runs.
- `Pr[<=10] (<> good)` – runs a number of stochastic simulations and estimates the probability of good eventually becoming true within 10 time units. The number of runs is decided based on the probability interval precision ($\pm\epsilon$) and confidence level (level of significance α).
- `Pr[<=10] (<> good) >= 0.5` – checks if the probability of reaching good within 10 time units is greater than 50% (less runs than calculating the probability, using “Walds’s algorithm”)
- `E[<=10; 100] (max: cost)` runs 100 stochastic simulations and estimates the maximal value of cost expression over 10 time units of stochastic simulation.

More at https://docs.uppaal.org/language-reference/query-syntax/statistical_queries/

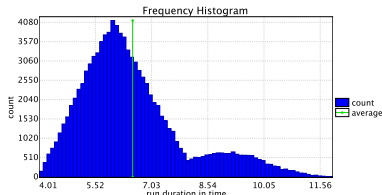
- `simulate[<=10] { x, y }` creates one stochastic simulation run of up to 10 time units in length and plot the values of `x` and `y` expressions over time (after checking, right-click the query and choose a plot).
- Variations: `[c<=10]` / `[#<=10]` – based on clock `c` or based on the number of transitions.

A2



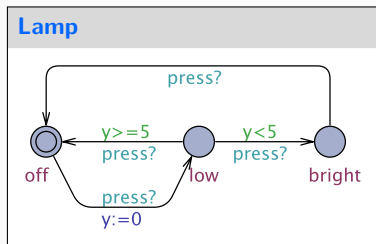
Ex. 6.4: Replicate the visualisation

A2's histogram



Parameters: $\alpha=0.05$, $\epsilon=0.05$, bucket width=0.10102, bucket count=78
 Runs: 100000 in total, 100000 (100%) displayed, 0 (0%) remaining
 Span of displayed sample: [4.0185, 11.8982]
 Mean estimate of displayed sample: 6.5045 \pm 0.00872567 (95% CI)

Ex. 6.5: Replicate the visualisation also for A1 and A3



Ex. 6.6: Adapt the model to make it stochastic

Ex. 6.7: Adapt requirements to make them probabilistic

1. The lamp can become bright;
2. The lamp will eventually become bright;
3. The lamp can never be on for more than 3600s;
4. It is possible to never turn on the lamp;
5. Whenever the light is bright, the clock y is non-zero;
6. Whenever the light is bright, it will eventually become off.