

Making concurrent systems probabilistic

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Workshop *Pomsets and Related Structures (RaPS)*

April 24th, 2024

Outline

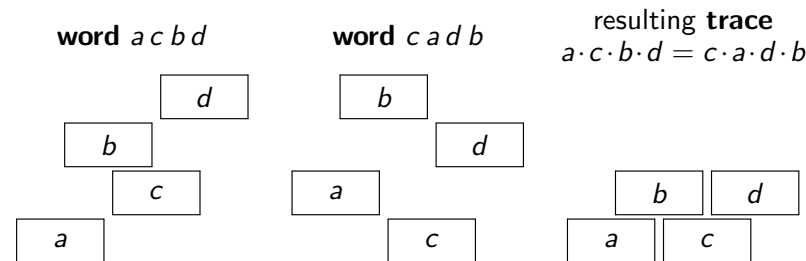
Framework

Looking for the probabilistic parameters

Several examples

Semantics and theory

Framework (1): an example of trace monoid



- ▶ Alphabet $\Sigma = \{a, b, c, d\}$ of **generators**
- ▶ **Traces** \iff heaps obtained in a “Tetris”-like way by letting generators fall
- ▶ Commutation relations between generators:

$$a \cdot c = c \cdot a$$

$$a \cdot d = d \cdot a$$

$$b \cdot d = d \cdot b$$

Framework (1): an example of trace monoid

Structure

- ▶ Heaps are in bijection with the elements of the **trace monoid**:

$$\mathcal{M} = \left\langle \underbrace{a, b, c, d}_{\text{generators}} \mid \underbrace{ac = ca, ad = da, bd = db}_{\text{commutation relations}} \right\rangle$$

Elements of \mathcal{M} are **words on** $\{a, b, c, d\}$ up to the congruence generated by the 3 **commutation relations**

Framework (1): an example of trace monoid

Structure

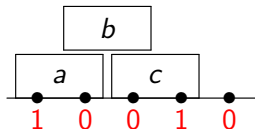
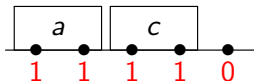
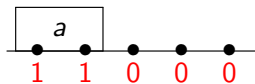
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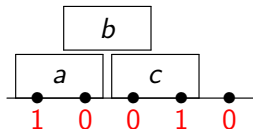
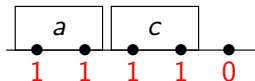
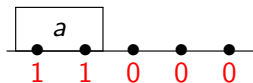
Elements of \mathcal{M} are **words on** $\{a, b, c, d\}$ up to the congruence generated by the 3 **commutation relations**

- ▶ \mathcal{M} comes with:
 - ▶ a *monoid structure*, i.e., an associative composition $(x, y) \in \mathcal{M} \times \mathcal{M} \mapsto x \cdot y$ and a neutral element ε (the empty trace)
 - ▶ a *length function* $x \in \mathcal{M} \mapsto |x| \in \mathbb{N}$, number of letters in x
 - ▶ a partial order relation: $x \leq y \iff \exists z \in \mathcal{M} \quad y = x \cdot z$

Framework (2): an example of action of trace monoid



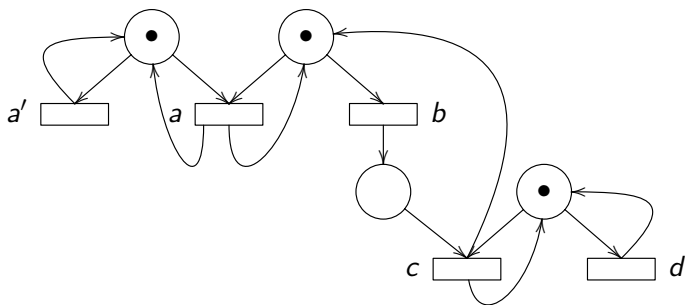
Framework (2): an example of action of trace monoid



$$\mathcal{M} = \langle a, b, c, d \mid ac = ca, ad = da, bd = db \rangle$$

$$X \subseteq \{0, 1\}^5$$

Framework (2): an example of 1-safe Petri net



$$\left. \begin{array}{l}
 X = \{0, 1\} \quad \text{the two markings of the net} \\
 \mathcal{M} = \left\langle \underbrace{a, a', b, c, d}_{\text{transitions of the net}} \mid \underbrace{a'b = ba', \dots, bd = db}_{\bullet t \cap \bullet t' = \emptyset} \right\rangle
 \end{array} \right\} \begin{array}{l}
 \text{True-concurrent} \\
 \text{semantics: action} \\
 \text{of } \mathcal{M} \text{ on } X
 \end{array}$$

Framework (2): action of \mathcal{M} on X

A trace monoid \mathcal{M} **acts** on a set of **states** X with an additional **sink** state \perp

$$\begin{cases} (X \cup \{\perp\}) \times \mathcal{M} & \rightarrow & X \cup \{\perp\} \\ (\alpha, x) & \mapsto & \alpha \cdot x \end{cases}$$

$$\alpha \cdot (x \cdot y) = (\alpha \cdot x) \cdot y, \quad \alpha \cdot \varepsilon = \varepsilon, \quad \perp \cdot x = \perp$$

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Semantics

- ▶ $\alpha \cdot x \neq \perp \iff$
 $\begin{cases} x \text{ is an } \mathbf{execution} \text{ of the system from the } \mathbf{initial state} \alpha \\ \alpha \cdot x \text{ is the } \mathbf{state} \text{ reached after playing } x \in \mathcal{M} \text{ from } \alpha \end{cases}$
- ▶ $\alpha \cdot x = \perp \iff$ playing x from α is **forbidden**

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Randomizing the executions

- ▶ **Data:** concurrent system (\mathcal{M}, X, \perp)
- ▶ **Goal:** determine the probabilistic parameters $\lambda_\alpha(a)$ as the “probability of playing the letter $a \in \Sigma$ from the state $\alpha \in X$ ”

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With the intuitive properties:

$$\begin{aligned}\alpha \cdot a = \perp &\iff \lambda_\alpha(a) = 0 \\ a \cdot b = b \cdot a &\implies \underbrace{\lambda_\alpha(a)\lambda_{\alpha \cdot a}(b)}_{\text{probability of playing } a \cdot b \text{ from } \alpha} = \underbrace{\lambda_\alpha(b)\lambda_{\alpha \cdot b}(a)}_{\text{probability of playing } b \cdot a \text{ from } \alpha}\end{aligned}$$

Randomizing the executions

- ▶ **Data:** concurrent system (\mathcal{M}, X, \perp)
- ▶ **Goal:** determine the probabilistic parameters $\lambda_\alpha(a)$ as the “probability of playing the letter $a \in \Sigma$ from the state $\alpha \in X$ ”

Without concurrency:

$$\forall \alpha \in X \quad \sum_{a \in \Sigma} \lambda_\alpha(a) = 1$$

What is the correct *normalization condition* with concurrency?

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Semantics and theory

Basic case: the trace monoid with one state

Trivial action $\{*\} \times \mathcal{M} \rightarrow \{*\}$: as if there was no state

$\mathcal{M} = \langle a, b, c, d \mid ac = ca, ad = da, bd = db \rangle$

Denote by the same letters a, b, c, d the sought probabilistic parameters. Then:

$$1 - a - b - c - d + \underbrace{ac + ad + bd}_{\substack{\text{commutation cliques} \\ \text{of order 2}}} = 0$$

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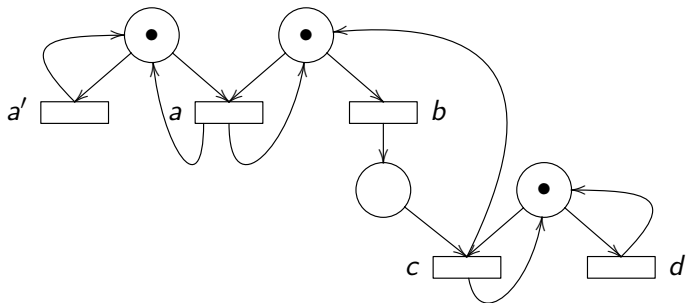
$$1 - a - b - c - d + \underbrace{ac + ad + bd}_{\substack{\text{commutation cliques} \\ \text{of order 2}}} = 0$$

Particular case: all parameters are equal for the *uniform measure*

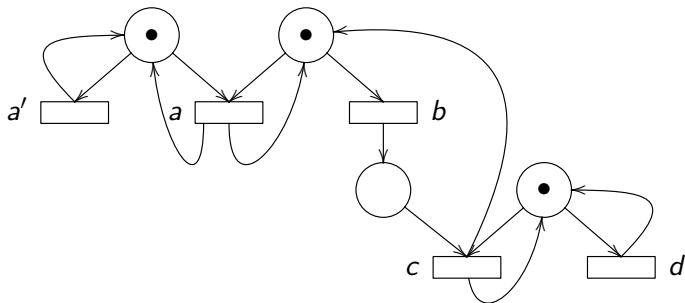
root of the Möbius polynomial of \mathcal{M} : $1 - 4t + 3t^2 = 0 \longrightarrow t = \frac{1}{3}$ (!!)

Interpretation: All traces of a given length k have the same probability $(\frac{1}{3})^k$ to appear at the beginning of a random infinite execution

The Petri net example (1)



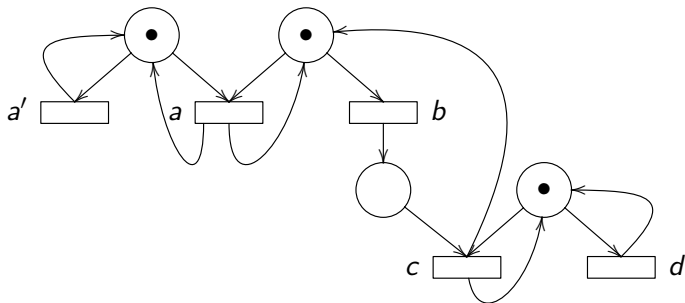
The Petri net example (1)



At marking 0: parameters a'_0, a_0, b_0, c_0, d_0

At marking 1: parameters a'_1, a_1, b_1, c_1, d_1

The Petri net example (1)

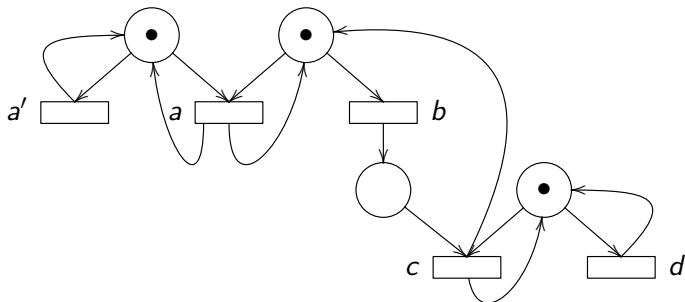


$$0 \cdot c = \perp \implies c_0 = 0$$

$$1 \cdot a = \perp \implies a_1 = 0$$

$$1 \cdot b = \perp \implies b_1 = 0$$

The Petri net example (1)

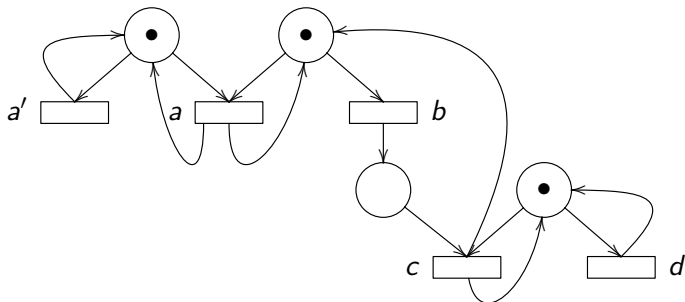


equality in \mathcal{M} : $a' \cdot b = b \cdot a' \implies a'_0 b_0 = b_0 a'_1 \implies a'_0 = a'_1$

equality in \mathcal{M} : $b \cdot d = d \cdot b \implies b_0 d_1 = d_0 b_0 \implies d_0 = d_1$

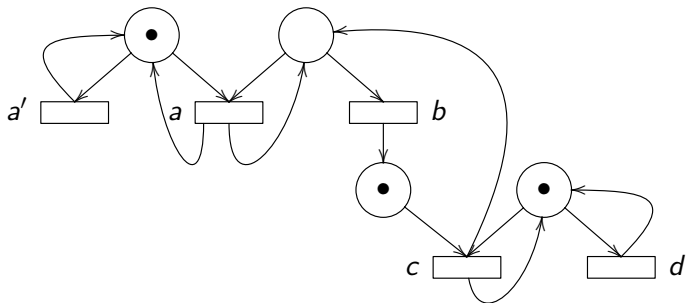
5 parameters only: $a' = a'_0$, $a = a_0$, $b = b_0$, $c = c_1$, $d = d_1$

The Petri net example (1)



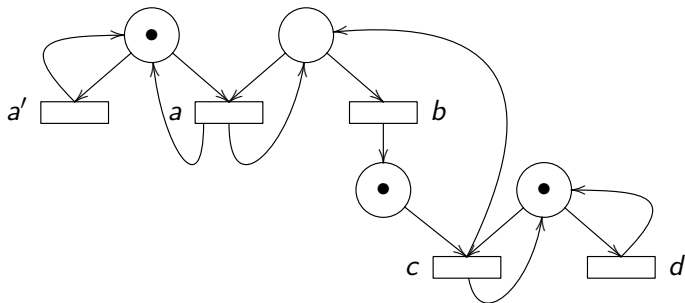
$$\text{At } 0: \quad 1 - \underbrace{a' - a - b - d}_{\substack{\text{commutation cliques} \\ \text{of order 1 that can} \\ \text{play at marking 0}}} + \underbrace{a'b + a'd + ad + bd}_{\text{idem order 2}} - \underbrace{a'bd}_{\text{idem order 3}} = 0$$

The Petri net example (1)



$$\text{At 1: } 1 - a' - c - d + a'c + a'd = 0$$

The Petri net example (1)

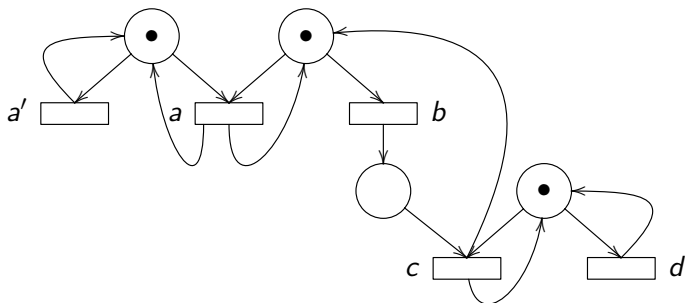


Question: is there a solution to satisfy both normalization conditions?

$$(1 - a' - a - b + a'b)(1 - d) = 0$$

$$(1 - c - d)(1 - a') = 0$$

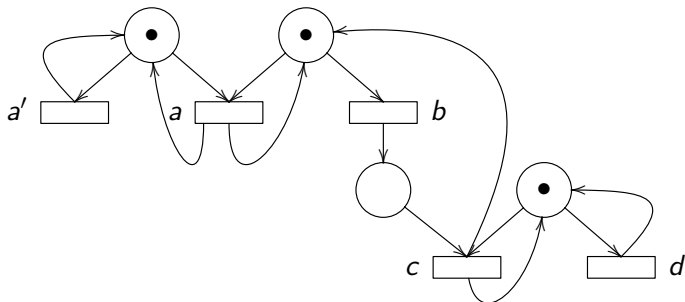
The Petri net example (2): the uniform measure



the Möbius matrix:
$$M(t) = \begin{pmatrix} 1 - 3t + 2t^2 & -t + 2t^2 - t^3 \\ -t + t^2 & 1 - 2t + t^2 \end{pmatrix} \begin{matrix} 0 \\ 1 \end{matrix}$$

$$\theta(t) = \det M(t) = (1 - t)^3(1 - 2t - t^2) \longrightarrow \boxed{r = \sqrt{2} - 1}$$

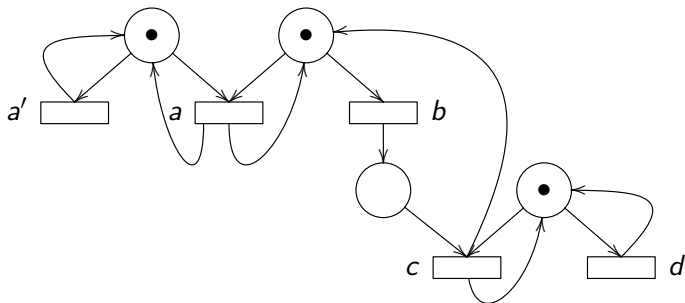
The Petri net example (2): the uniform measure



$$M(r) = \begin{pmatrix} 10 - 7\sqrt{2} & 2(7 - 5\sqrt{2}) \\ 4 - 3\sqrt{2} & 2(3 - 2\sqrt{2}) \end{pmatrix}$$

$$\ker M(r) = \text{Vect}(u) \quad \text{with } u = \begin{pmatrix} 2 \\ \sqrt{2} \end{pmatrix} = \begin{pmatrix} u_0 \\ u_1 \end{pmatrix} \quad \text{a positive vector}$$

The Petri net example (2): the uniform measure

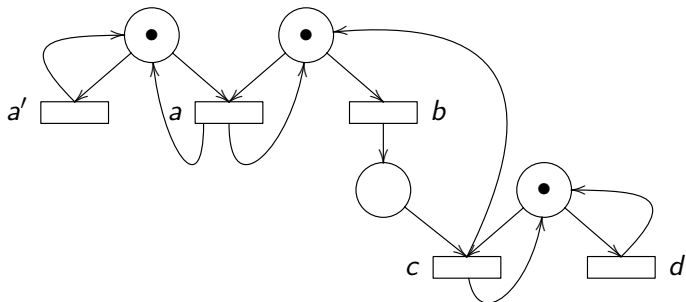


Use the following normalized parameters:

$$\text{for } \alpha = 0, 1 \text{ and } \sigma = a', a, b, c, d \quad \lambda_\alpha(\sigma) = r \frac{u_{\alpha \cdot \sigma}}{u_\alpha}$$

$$\text{and then for any } x = \sigma_1 \cdot \dots \cdot \sigma_k \in \mathcal{M}: \quad f_\alpha(x) = r^{|x|} \frac{u_{\alpha \cdot x}}{u_\alpha}$$

The Petri net example (2): the uniform measure



$\sigma \in \Sigma$	a'	a	b	c	d
$\lambda(\sigma)$	$-1 + \sqrt{2}$	$-1 + \sqrt{2}$	$1 - \frac{1}{2}\sqrt{2}$	$2 - \sqrt{2}$	$-1 + \sqrt{2}$
$\approx \lambda(\sigma)$.4142	.4142	.2929	.5858	.4142

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Semantics

For a trace monoid \mathcal{M} :

- ▶ $\partial\mathcal{M}$ is the space of *infinite traces*
- ▶ for $x \in \mathcal{M}$ the *visual cylinder* of base x is
$$\uparrow x = \{\omega \in \partial\mathcal{M} : x \leq \omega\}$$

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- ▶ a *probability measure* ν on $\partial\mathcal{M}$ is entirely determined by the countable collection of values $\nu(\uparrow x)$, for $x \in \mathcal{M}$
- ▶ $\nu(\uparrow x)$ is the probability of seeing x at the “beginning” of a random infinite execution

Semantics

For a concurrent system (\mathcal{M}, X, \perp) :

- ▶ a *Markov measure* is a collection $(\nu_\alpha)_{\alpha \in X}$ of probability measures on $\partial\mathcal{M}$ such that:

$$\forall \alpha \in X \quad \forall x \in \mathcal{M} \quad \nu_\alpha(\uparrow x) > 0 \iff \alpha \cdot x \neq \perp$$

and satisfying the chain rule: $\forall \alpha \in X \quad \forall x, y \in \mathcal{M}$

$$\nu_\alpha(\uparrow(x \cdot y)) = \nu_\alpha(\uparrow x) \nu_{\alpha \cdot x}(\uparrow y)$$

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- ▶ Question 1: do Markov measures always exist?
- ▶ Question 2: determine all of them
- ▶ Question 3: how to simulate a given Markov measure?

Semantics

Reformulation

The *valuation* associated to the Markov measure ν is the family $f = (f_\alpha)_{\alpha \in X}$ of functions $f_\alpha : \mathcal{M} \rightarrow \mathbb{R}_{\geq 0}$ defined by

$$f_\alpha(x) = \nu_\alpha(\uparrow x)$$

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The functions f_α satisfy:

$$1) f_\alpha(x) > 0 \iff \alpha \cdot x \neq \perp$$

$$2) f_\alpha(x \cdot y) = f_\alpha(x)f_{\alpha \cdot x}(y)$$

$$\implies f_\alpha(a \cdot b \cdot c) = f_\alpha(b)f_{\alpha \cdot a}(b)f_{\alpha \cdot a \cdot b}(c)$$

f is determined by the *finite* collection of parameters

$$\lambda_\alpha(a) = f_\alpha(a), \quad (\alpha, a) \in X \times \Sigma$$

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- ▶ **Conversely:** What are the normalization conditions on $(\lambda_\alpha(a))_{(\alpha,a) \in X \times \Sigma}$ for the associated valuation f to be *probabilistic* (i.e., induced by some probability measure)?

Theory: probabilistic trace monoids

In case of a trace monoid with the trivial action
 $\{*\} \times \mathcal{M} \rightarrow \{*\}$ **(no state, no constraint)**

The valuation f is a single function $f : \mathcal{M} \rightarrow \mathbb{R}_{>0}$ satisfying:

$$\forall x, y \in \mathcal{M} \quad f(x \cdot y) = f(x)f(y)$$

and determined by the collection $(f(a))_{a \in \Sigma}$

$$f(a_1 \cdot \dots \cdot a_k) = f(a_1) \dots f(a_k)$$

The associated probability measure is said to be *memoryless*
(rather than *Markov*)

Theory: probabilistic trace monoids

Given a function $f : \mathcal{M} \rightarrow \mathbb{R}_{>0}$, its *Möbius* transform is the function $h : \mathcal{C} \rightarrow \mathbb{R}$

$$\forall \gamma \in \mathcal{C} \quad h(\gamma) = \sum_{\gamma' \in \mathcal{C} : \gamma \leq \gamma'} (-1)^{|\gamma'| - |\gamma|} f(\gamma')$$

where \mathcal{C} is the (finite) set of *commutation cliques* of \mathcal{M} and then the *Möbius inversion formula* holds:

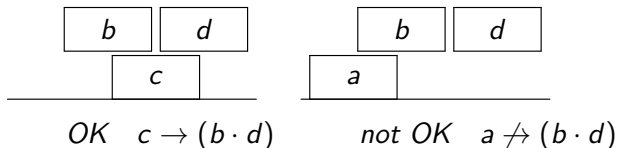
$$\forall \gamma \in \mathcal{C} \quad f(\gamma) = \sum_{\gamma' \in \mathcal{C} : \gamma \leq \gamma'} h(\gamma')$$

Theory: probabilistic trace monoids

Normal form of traces

- ▶ A pair (γ, γ') of cliques is *normal*, denoted $\gamma \rightarrow \gamma'$, if:

$$\forall z \in \gamma' \quad \exists y \in \gamma \quad \neg(y \parallel z)$$



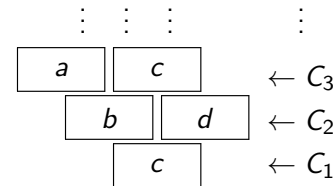
Theory: probabilistic trace monoids

Normal form of traces

- ▶ For every trace x there exists a unique integer $k \geq 0$ and a unique sequence (c_1, \dots, c_k) of cliques such that

$$1) c_i \rightarrow c_{i+1} \qquad 2) x = c_1 \cdot \dots \cdot c_k$$

- ▶ The same works for *infinite traces*
- ▶ If $\partial\mathcal{M}$ is equipped with a probability measure ν , then the sequence $(C_i)_{i \geq 1}$ is a *random process* (every C_i is a *finite random variable*)



Theory: probabilistic trace monoids

Proposition

If ν is a probability measure on $\partial\mathcal{M}$ and $f(x) = \nu(\uparrow x)$, then the law of C_1 is h , the Möbius transform of f :

$$\nu(C_1 = \gamma) = \sum_{\gamma' \in \mathcal{C} : \gamma \leq \gamma'} (-1)^{|\gamma'| - |\gamma|} f(\gamma')$$

Theory: probabilistic trace monoids

Consequence

If $f(a) = r$ for all $a \in \Sigma$, then $f(x) = r^{|x|}$

total probability law for C_1 :
$$\sum_{\gamma \in \mathcal{C}} \nu(C_1 = \gamma) = 1$$

yielding $\boxed{\mu(r) = 0}$ (Möbius polynomial of \mathcal{M})

Actually: r must be the root of **smallest modulus** of $\mu(t)$, which is unique and in $(0, 1)$

Example

$$\mathcal{M} = \langle a_0, \dots, a_4 \mid a_i a_j = a_j a_i \quad |i - j| \pmod{5} \geq 2 \rangle$$

$$\mu(t) = 1 - 5t + 5t^2 \quad \rightarrow r = \underbrace{\frac{5 \pm \sqrt{5}}{10}}_{\text{both in } (0, 1)} \quad \rightarrow r = \frac{5 - \sqrt{5}}{10}$$

Theory: probabilistic trace monoids

Law of the first clique for the uniform measure on

$$\mathcal{M} = \langle a, b, c, d \mid ac = ca, ad = da, bd = db \rangle$$

$$\mu(t) = 1 - 4t + 3t^2 \quad \rightarrow r = \frac{1}{3}$$

$$h(a) = f(a) - f(a \cdot c) - f(a \cdot d) = r - 2r^2 = \frac{1}{9}$$

$$h(b) = f(b) - f(b \cdot d) = r - r^2 = \frac{2}{9}$$

$$h(a \cdot c) = f(a \cdot c) = r^2 = \frac{1}{9}$$

$\gamma \in \mathcal{C}$	a	b	c	d	$a \cdot c$	$a \cdot d$	$b \cdot d$
$h(\gamma)$ (proba of $C_1 = \gamma$)	$\frac{1}{9}$	$\frac{2}{9}$	$\frac{2}{9}$	$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$
$f(\gamma)$ (proba of $C_1 \geq \gamma$)	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{9}$	$\frac{1}{9}$	$\frac{1}{9}$

Theory: probabilistic concurrent systems

For concurrent systems, an extension of the previous techniques works in connection with

- ▶ linear algebra
- ▶ analytic combinatorics