Génération aléatoire uniforme pour les réseaux d'automates

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Motivations (1/2)



Automata are omni-present in computer science.

Given a regular language, it is natural to ask

- what does a typical word of a fixed length n look like ?
- what does an infinite typical word look like ?

The literature provides answer based on

- Uniform sampling (from combinatorics);
- Maximal entropy measure (from information & ergodic theory)

when a deterministic finite state automaton (DFA) recognising the language is provided.

These methods are polynomial in the size of the given DFA.

Motivations (2/2)

Automata in verification of concurrent systems

- Computational systems (software or hardware) are often composed of several components that interact together;
- Networks of automata are an elegant and useful framework to model concurrent systems;
- The associated product automaton $\mathcal{A} = \mathcal{A}_1 \times \cdots \times \mathcal{A}_K$ is of exponential size $|\mathcal{A}| = |\mathcal{A}_1| \times \cdots \times |\mathcal{A}_K|$.

In this talk we will see how to do

- uniform sampling of words of a given length;
- sampling according to the maximal entropy measure;

for a network of DFAs in a compositional fashion.

A previous work on the subject by [Denise et al., STTT 2012] gives applications to model based testing.

Monolithic methods of sampling for a single DFA (a recap)

Compositional methods of sampling for Network of DFAs

Conclusion and perspective

Uniform sampling of words of an automaton (1/3).

Fixed length. Recursive Method.



Languages $\mathcal{L}_{p,k} \to \text{Cardinalities } |\mathcal{L}_{p,k}| \to \text{Probabilities } p_k(p \xrightarrow{b} q) = \frac{|\mathcal{L}_{q,n-k}|}{|\mathcal{L}_{p,n-k+1}|}$

$$\mathcal{L}_{p,k} = a\mathcal{L}_{p,k-1} \cup b\mathcal{L}_{q,k-1}; \quad \mathcal{L}_{q,k} = c\mathcal{L}_{p,k-1}.$$
$$|\mathcal{L}_{p,k}| = |\mathcal{L}_{p,k-1}| + |\mathcal{L}_{q,k-1}|; \quad |\mathcal{L}_{q,k}| = |\mathcal{L}_{p,k-1}|.$$
$$\begin{pmatrix} |\mathcal{L}_{p,k}| \\ |\mathcal{L}_{q,k}| \end{pmatrix} = M \begin{pmatrix} |\mathcal{L}_{p,k-1}| \\ |\mathcal{L}_{q,k-1}| \end{pmatrix} = M^k \begin{pmatrix} 1 \\ 1 \end{pmatrix} \text{ with } M = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}.$$
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Uniform sampling of words of an automaton (2/3).

Random length. Boltzmann Sampling [Duchon, Flajolet, Louchard, Schaeffer, ICALP'02].



Uniform sampling of words of an automaton (3/3). Infinite length. Parry sampling.



- For a strongly connected automaton.
- Defined by Shannon, known as Parry measure in ergodic theory. Here, we call it Boltzmann critic.

 ω -regular Languages $\mathcal{L}_{\rho,\omega} \to \mathsf{Perron}$ eigenvector $\mathbf{v} \to \mathsf{Probabilities} \ p_{\frac{1}{\rho}}(b) = rac{v_q}{\rho v_{\rho}}$

$$\mathcal{L}_{m{p},\omega} = a\mathcal{L}_{m{p},\omega} \cup b\mathcal{L}_{m{q},\omega}; \quad \mathcal{L}_{m{q},\omega} = c\mathcal{L}_{m{p},\omega}.$$

 $\rho v_p = v_p + v_q; \quad \rho v_q = v_p \text{ avec } \rho \text{ v.p. maximale.}$

$$\rho \mathbf{v} = M \mathbf{v}.$$
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Monolithic methods of sampling for a single DFA (a recap)

Compositional methods of sampling for Network of DFAs

Conclusion and perspective

A network of three DFAs with shared actions $\{\alpha, \beta, \gamma\}$



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A network of three DFAs with shared actions $\{\alpha, \beta, \gamma\}$



Example of words recognised: $\alpha ba\gamma d$ The product DFA:



The easy case: no shared action

Language of the product = shuffle of languages.

$$\mathcal{L}(\mathcal{A}^{(1)} imes \cdots imes \mathcal{A}^{(K)}) = \mathcal{L}(\mathcal{A}^{(1)}) \sqcup \cdots \sqcup \mathcal{L}(\mathcal{A}^{(K)})$$

Shuffle of languages

- Shuffle of words *ab* $\sqcup cd = \{abcd, acbd, acdb, cabd, cdab\}$
- Shuffle of two languages:

$$\mathcal{L}^{(1)} \sqcup\!\!\sqcup \mathcal{L}^{(2)} = igcup_{(w^{(1)},w^{(2)}) \in \mathcal{L}^{(1)} imes \mathcal{L}^{(2)}} w^{(1)} \sqcup\!\!\sqcup w^{(2)}$$

• Naturally extends to K languages.

Computing the cardinalities of shuffle of languages

For the shuffle of two languages

$$|(\mathcal{L}\sqcup \mathcal{L}')_n| = \sum_{k=0}^n \binom{n}{k} |\mathcal{L}_k| \cdot |\mathcal{L}'_{n-k}|. \tag{1}$$

For the shuffle of K languages $\mathcal{L} = \mathcal{L}^{(1)} \sqcup \cdots \sqcup \mathcal{L}^{(K)}$

Do not use

$$|\mathcal{L}_{n}| = \sum_{n^{(1)} + \dots + n^{(K)} = n} {\binom{n}{n^{(1)}, \dots, n^{(K)}}} |\mathcal{L}_{n^{(1)}}^{(1)}| \cdots |\mathcal{L}_{n^{(K)}}^{(K)}|$$

There are exponentially many coefficients!

• Instead apply equation (1) K - 1 times $\mathcal{L} = (\dots ((\mathcal{L}^{(1)} \sqcup \mathcal{L}^{(2)}) \sqcup \mathcal{L}^{(3)}) \sqcup \dots) \sqcup \mathcal{L}^{(K)}.$

This can be transformed into a recursive method of sampling for $\mathcal{L} = \mathcal{L}^{(1)} \sqcup \cdots \sqcup \mathcal{L}^{(K)}$.

Generating functions for shuffle of languages Exponential generating functions $\hat{L}(z) = \sum_{n \in \mathbb{N}} |\mathcal{L}_n| z^n / n!$ Exponential Boltzmann measure $\hat{\mu}_z(w) = \frac{z^{|w|}}{|w|!\hat{L}(z)}$

• Given
$$\mathcal{L} = \mathcal{L}^{(1)} \sqcup \cdots \sqcup \mathcal{L}^{(K)}$$

$$\hat{L}(z) = \hat{L}^{(1)}(z) \times \cdots \times \hat{L}^{(K)}(z)$$

•
$$L(z) = \int_0^{+\infty} e^{-u} \hat{L}(zu) du$$

Boltzmann sampler of parameter z for \mathcal{L}

- Choose *u* according to weight function: $u \mapsto e^{-u} \hat{L}(zu) = e^{-u} \prod_{i=1}^{K} \hat{L}^{(i)}(zu);$
- For i = 1 to K, let $w^{(i)}$ be chosen using an exponential Boltzmann sampler of parameter zu for $\mathcal{L}^{(i)}$.
- Return a word uniformly at random in $w^{(1)} \sqcup \cdots \sqcup w^{(K)}$

Shannon Parry-Markov chain for the shuffle of languages

Recap of the definition

$$P(p \xrightarrow{a} q) = v_q / (\rho v_p)$$
 with $Mv = \rho v$

Lemma

Let $\mathcal{A} = \mathcal{A}^{(1)} \times \cdots \times \mathcal{A}^{(K)}$ be the product of K strongly connected DFAs without synchronisation. Then $\rho = \sum_{i=1}^{n} \rho^{(i)}$, $v_s = \prod_{i=1}^{K} v_{s^{(i)}}^{(i)}$.

The sampling according to the Shannon-Parry Markov chain Repeat forever the following: With probability $\rho^{(i)}/\rho$ make one step $(s^{(i)}, a, t^{(i)})$ of the Shannon-Parry Markov chain number *i*, write *a* on the output tape;

Difficulties come from synchronisation

Recap no shared actions=shuffle of languages=everything is easy;

All letters shared

• Language of the product = intersection of languages :

$$\mathcal{L}(\mathcal{A}^{(1)} imes \cdots imes \mathcal{A}^{(K)}) = \mathcal{L}(\mathcal{A}^{(1)}) \cap \cdots \cap \mathcal{L}(\mathcal{A}^{(K)})$$

•
$$\mathcal{L}(\mathcal{A}^{(1)}) \cap \cdots \cap \mathcal{L}(\mathcal{A}^{(K)}) \stackrel{?}{=} \emptyset$$
 is a PSPACE-complete problem.

In our framework

We introduce the **reduced automaton**:

- It keeps only the synchronised part of the product automaton (the true difficulty that needs sequential reasoning).
- The non-synchronised part is projected out (easy to treat by combining independent local works).

The reduced automaton

The reduced automaton of a DFA $\mathcal{A} = (Q, \Sigma, \iota, F, \delta)$ is a finite automaton $\mathcal{A}_{red} = (Q_{red}, \Sigma_{red}, \iota_{red}, F_{red}, \Delta_{red})$ such that

- Q_{red} ⊆ Q are states occurring just after a shared action + initial state ι;
- Σ_{red} set of shared action;
- $\iota_{red} = \iota$ (same initial state);
- Final states F_{red} irrelevant
- $\Delta_{\text{red}} = \{(s, \alpha, t) \mid s \xrightarrow{u\alpha} t \text{ for some } u \in (\Sigma \setminus \Sigma_{\text{red}})^*\}$



Do not compute \mathcal{A}_{red} from the product DFA $\mathcal{A} = \mathcal{A}^1 \times \cdots \times \mathcal{A}^K$ but use $\mathcal{A}_{red} = \mathcal{A}^1_{red} \times \cdots \times \mathcal{A}^K_{red}$.

Languages associated to the reduced automaton

Given a DFA ${\mathcal A}$ and its reduced automaton ${\mathcal A}_{\text{red}}.$

• $\tilde{\mathcal{L}}_s$: language from state *s* without shared action.

•
$$\mathcal{L}_{\delta} = \{ u \in (\Sigma \setminus \Sigma_{\texttt{red}})^* \mid s \xrightarrow{u\alpha} t \}$$
, for $\delta = (s, \alpha, t) \in \Delta_{\texttt{red}}$

These language are obtained by modifying slightly the automaton. Example $\tilde{\mathcal{L}}_{111}$ and $\mathcal{L}_{(112,\gamma,323)}$



In fact, compute everything locally and use shuffle of languages: $\mathcal{L}_{(112,\gamma,323)} = \mathcal{L}_{(1,\gamma,3)}^{(1)} \sqcup \mathcal{L}_{(1,\gamma,2)}^{(2)} \sqcup \mathcal{L}_{(3,\gamma,3)}^{(3)} = a \sqcup (bc)^* b \sqcup \varepsilon.$

Equations on languages related to the reduced automaton

Theorem: Equations on languages

$$\mathcal{L}_{s} = \tilde{\mathcal{L}}_{s} \cup \bigcup_{\delta = (s, \alpha, t) \in \Delta_{\text{red}}} \mathcal{L}_{\delta} \cdot \alpha \cdot \mathcal{L}_{t}$$
$$\tilde{\mathcal{L}}_{s} = \bigsqcup_{i=1}^{K} \tilde{\mathcal{L}}_{s^{(i)}}^{(i)}; \quad \mathcal{L}_{\delta} = \bigsqcup_{i=1}^{K} \mathcal{L}_{\delta^{(i)}}^{(i)}$$

Our generic recipe to randomly generate a word $w \in \mathcal{L}_s$

- Choose whether a synchronisation will occur or not; if not choose w ∈ L̃_s = □^K_{i=1}L̃⁽ⁱ⁾_{c(i)}; otherwise
- choose $\delta = (s, \alpha, t) \in \Delta_{\texttt{red}}$;

• choose
$$u \in \mathcal{L}_{\delta} = \bigsqcup_{i=1}^{K} \mathcal{L}_{\delta^{(i)}}^{(i)}$$
;

• write $u\alpha$ and repeat from t to generate the rest of the word.

Our generic recipe to randomly generate a word $w \in \mathcal{L}_{s,n}$ (1/3) Fixed length uniform sampling

1. Choose whether a synchronisation will occur or not;

• No synchronisation with probability $|\tilde{\mathcal{L}}_{s,n}|/|\mathcal{L}_{s,n}|$.

if not choose $w \in \tilde{\mathcal{L}}_s = oxplus_{i=1}^K \tilde{\mathcal{L}}_{s^{(i)}}^{(i)}$; otherwise

2. choose
$$\delta = (s, lpha, t) \in \Delta_{ t red}$$
;

• choose the length *m* with weight

$$\frac{\sum_{\delta = (s, \alpha, t) \in \Delta_{\text{red}}} |\mathcal{L}_{\delta, m-1}|}{\sum_{m=1}^{n} \sum_{\delta = (s, \alpha, t) \in \Delta_{\text{red}}} |\mathcal{L}_{\delta, m-1}|}$$

• choose $\delta = (s, \alpha, t) \in \Delta_{red}$ with weight $\frac{|\mathcal{L}_{\delta, m-1}|}{\sum_{\delta' = (s, \alpha', t')} |\mathcal{L}_{\delta', m-1}|}$;

- 3. choose $u \in \mathcal{L}_{\delta,m-1} = \bigsqcup_{i=1}^{K} \mathcal{L}_{\delta^{(i)},m-1}^{(i)}$;
- 4. write $u\alpha$ and repeat from t to generate the rest of the word of length n m.

Our generic recipe to randomly generate a word $w \in \mathcal{L}_s$ (2/3) Boltzmann sampling

Recap:
$$L_s(z) = \tilde{L}_s(z) + z \sum_{\delta = (s, \alpha, t) \in \Delta_{red}} L_{\delta}(z) L_t(z).$$
 (2)

1. Choose whether a synchronisation will occur or not;

• No synchronisation with probability $\tilde{L}_s(z)/L_s(z)$.

if not choose $w \in \tilde{\mathcal{L}}_s = \bigsqcup_{i=1}^K \tilde{\mathcal{L}}_{s^{(i)}}^{(i)}$ using Boltzmann sampling with parameter z; otherwise

2. choose $\delta = (s, \alpha, t) \in \Delta_{\texttt{red}}$ with probability

$$\frac{L_{\delta}(z)L_{t}(z)}{\sum_{\delta'=(s,\alpha',t')\in\Delta_{\mathrm{red}}}L_{\delta'}(z)L_{t'}(z)}$$

- 3. choose $u \in \mathcal{L}_{\delta} = \bigsqcup_{i=1}^{K} \mathcal{L}_{\delta^{(i)}}^{(i)}$ with probability $z^{|u|}/L_{\delta}(z)$ using Boltzmann sampling with parameter z;
- 4. write $u\alpha$ and repeat from t to generate the rest of the word.

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Our generic recipe to randomly generate a word $w \in \mathcal{L}_{s,\omega}$ (3/3) Parry sampling

Assume the product automaton is strongly connected and let $v \ge 0$ and ρ such that $Mv = \rho v$.

- 1. A synchronisation occurs in the future with probability 1;
- 2. choose $\delta = (s, \alpha, t) \in \Delta_{\texttt{red}}$ with probability

 $L_{\delta}(1/\rho)\frac{v_t}{\rho v_s}$

3. choose $u \in \mathcal{L}_{\delta} = \bigsqcup_{i=1}^{K} \mathcal{L}_{\delta^{(i)}}^{(i)}$ with probability

$$rac{1}{
ho^{|u|}L_{\delta}(1/
ho)}$$

using Boltzmann sampling with parameter $1/\rho$;

4. write $u\alpha$ and repeat from t to generate the rest of the word.

Characterisation of the generating functions in the reduced automaton Recap equations on languages:

$$\mathcal{L}_{s} = \tilde{\mathcal{L}}_{s} \cup \bigcup_{\delta = (s, \alpha, t) \in \Delta_{\text{red}}} \mathcal{L}_{\delta} \cdot \alpha \cdot \mathcal{L}_{t}$$
(3)

Theorem: Equations on generating functions

$$L_s(z) = ilde{L}_s(z) + z \sum_{\delta = (s, lpha, t) \in \Delta_{ ext{red}}} L_\delta(z) L_t(z)$$

In matrix form

Let $\mathfrak{M}(z)$ be the $Q_{red} imes Q_{red}$ matrix defined by

$$\mathfrak{M}_{s,t}(z) = \sum_{\delta = (s,\alpha,t) \in \Delta_{\mathrm{red}}} L_{\delta}(z)$$
 (4)

 $\mathbf{L}(z) = \tilde{\mathbf{L}}(z) + z\mathfrak{M}(z)\mathbf{L}(z); \text{ then } \mathbf{L}(z) = (I - z\mathfrak{M}(z))^{-1}\tilde{\mathbf{L}}(z) \quad (5)_{z_1/z_2}$

Computing cardinalities for all languages

Let n be the length of words to sample.

Languages without synchronisation $(|\tilde{\mathcal{L}}_{s,m}|)_{m \leq n, s \in Q_{red}}$ and $(|\mathcal{L}_{\delta,m}|)_{m \leq n, \delta \in \Delta_{red}}$

See before, shuffle of languages. Polynomial in n and K.

Languages with synchronisations $(|\mathcal{L}_{s,m}|)_{m \leq n, s \in Q_{red}}$

- Write $\tilde{\mathbf{L}}_{s}(z) \mod z^{n+1} = \sum_{m=0}^{n} |\tilde{\mathcal{L}}_{s,m}| z^{m}$ and $\mathfrak{M}_{s,t}(z) \mod z^{n+1} = \sum_{m=0}^{n} \sum_{\delta = (s,\alpha,t) \in \Delta_{\mathrm{red}}} |\mathcal{L}_{\delta,m}| z^{m}$
- Find $L(z) \mod z^{n+1}$ by taking all operations modulo z^{n+1} in

$$\mathbf{L}(z) = (I - z\mathfrak{M}(z))^{-1}\mathbf{\tilde{L}}(z).$$

Polynomial in n and $|\mathcal{A}_{red}|$.

A Perron Frobenius Theorem for the reduced automaton

Let ${\cal A}$ be a product automaton that is strongly connected and ${\cal A}_{\rm red}$ its reduced automaton.

Spectral attributes of the matrix $\mathfrak{M}(z)$

Given $\lambda \in \mathbb{C}$ and $\mathbf{v} \neq \mathbf{0}$. If $\mathfrak{M}(1/\lambda)\mathbf{v} = \lambda \mathbf{v}$ then λ is called a reduced eigenvalue and v a reduced eigenvector.

Theorem

- Existence of ρ and v_{red}:
 - There exists a reduced eigenvalue $\rho > 0$ such that $|\lambda| \le \rho$ for every reduced eigenvalue λ .
 - There exists a unique $\mathbf{v}_{red} \ge 0$ (up to a multiplicative constant) which is a reduced eigenvector. It satisfies $\mathfrak{M}(1/\rho)\mathbf{v}_{red} = \rho\mathbf{v}_{red}$.
- Link with ${\mathcal A}$ and its adjacency matrix M
 - ρ is the spectral radius of M
 - \mathbf{v}_{red} is the restriction to Q_{red} of the unique eigenvector $\mathbf{v} \ge 0$ (it satisfies $M\mathbf{v} = \rho\mathbf{v}$)

Monolithic methods of sampling for a single DFA (a recap)

Compositional methods of sampling for Network of DFAs

Conclusion and perspective

What we have seen

- A recap in the monolithic case of
 - Uniform sampling
 - Boltzmann sampling
 - Sampling according to Shannon-Parry Markov chain

and their link to entropy

• Compositional methods for these sampling for network of DFAs based on the notion of reduced automata.

Possible further works

- Precise study of numerical computations (e.g. for finding reduced spectral radius).
- Design of algorithms with better bit complexity.
- Implementations and applications to
 - statistical model checking;
 - model based testing.
- Extension of the theory to weighted automata.
- Extension of the theory to timed automata.