# Probabilistic Aspects of Computer Science: Probabilistic Automata

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#### MPRI M1

- Presentation
- Properties of Stochastic Languages
- 3 Decidability Results

### **Plan**

Presentation

**Properties of Stochastic Languages** 

**Decidability Results** 

### An introductive example

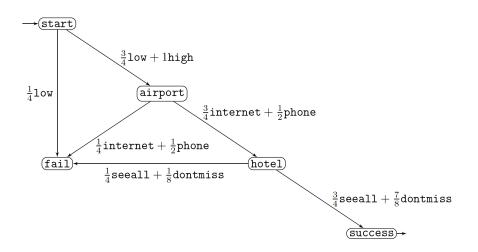
#### Planning holidays in a foreign country

- 1. Choosing which train or plane to use;
- 2. Renting an house or a room in an hotel;
- 3. Buying tickets for some exhibitions, etc.

Usually these actions must be planned before the holidays.

Thus one looks for an *a priori* optimal policy that maximizes the probability to *reach* a goal.

### **Formalisation**



The probability of success of lowcost  $\cdot$  internet  $\cdot$  seeall is  $\frac{27}{64}$ .

### Probabilistic automata

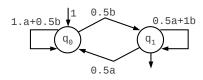
Probabilistic Automata (PA) are a variation of MDP where:

- ▶ The set of possible actions is the same in every state.
- There are no rewards.
- There is a subset of final states.

#### More formally, a PA $\mathcal{A} = (Q, A, \{\mathbf{P}_a\}_{a \in A}, \pi_0, F)$ is defined by:

- Q, the finite set of states;
- A, the finite alphabet;
- ▶ For all  $a \in A$ ,  $P_a$ , a probability transition matrix over S;
- $\bullet$   $\pi_0$ , the initial distribution over states and  $F \subseteq Q$  the final states.

### Illustration



- ▶  $A = \{a, b\};$
- $Q = \{q_0, q_1\}, F = \{q_1\};$
- $\bullet$   $\pi_0[q_0] = 1.$

An edge from a state to another one is labelled by a vector of transition probabilities indexed by A. The vector is denoted by a formal sum.

For instance, the transition from  $q_0$  to itself is labelled by 1a + 0.5b means that:

- when a is chosen in state  $q_0$ , the probability that the next state is  $q_0$ ,  $\mathbf{P}_a[q_0, q_0]$ , is equal to 1.
- when b is chosen in state  $q_0$ , the probability that the next state is  $q_0$ ,  $\mathbf{P}_b[q_0, q_0]$ , is equal to 0.5.

#### **Policies** in PA

Words are policies. When some finite word  $w \stackrel{\text{def}}{=} a_1 \dots a_n$  is selected, we are interested in the probability to be in a final state using w as a policy.

Given  $\mathcal{A}$  a PA and  $w \stackrel{\text{def}}{=} a_1 \dots a_n \in A^*$  a word, the *acceptance probability* of w by  $\mathcal{A}$  is defined by:

$$\mathbf{Pr}_{\mathcal{A}}(w) \stackrel{\mathsf{def}}{=} \sum_{q \in Q} \pi_{\mathbf{0}}[q] \sum_{q' \in F} \left( \prod_{i=1}^{n} \mathbf{P}_{a_i} \right) [q, q']$$

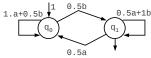
**Notation.** Given a word  $w \stackrel{\text{def}}{=} a_1 \dots a_n$ , the probability matrix  $\mathbf{P}_w$  is defined by  $\mathbf{P}_w \stackrel{\text{def}}{=} \prod_{i=1}^n \mathbf{P}_{a_i}$ . In particular  $\mathbf{P}_{\varepsilon} = \mathbf{Id}$ .

With these notations:

$$\mathbf{Pr}_{\mathcal{A}}(w) = \pi_0 \mathbf{P}_w \mathbf{1}_F^T$$

where  $\mathbf{1}_F$  is the indicator vector of subset F.

### Illustration



Observe that for all w,  $\mathbf{Pr}_{\mathcal{A}}(w) = \mathbf{Pr}(\mathsf{to} \mathsf{ be in } q_1 \mathsf{ after following policy of } w)$  and  $1 - \mathbf{Pr}_{\mathcal{A}}(w) = \mathbf{Pr}(\mathsf{to} \mathsf{ be in } q_0 \mathsf{ after following policy of } w)$ 

• 
$$\mathbf{Pr}_{\mathcal{A}}(\varepsilon) = 0$$
,  $\mathbf{Pr}_{\mathcal{A}}(a) = \frac{1}{2}\mathbf{Pr}_{\mathcal{A}}(\varepsilon) = 0$ 

$$\mathbf{Pr}_{\mathcal{A}}(ab) = \mathbf{Pr}_{\mathcal{A}}(a) + \frac{1}{2}(1 - \mathbf{Pr}_{\mathcal{A}}(a)) = \frac{1}{2}$$

$$\mathbf{Pr}_{\mathcal{A}}(abb) = \mathbf{Pr}_{\mathcal{A}}(ab) + \frac{1}{2}(1 - \mathbf{Pr}_{\mathcal{A}}(ab)) = \frac{3}{4}$$

$$\mathbf{Pr}_{\mathcal{A}}(abba) = \frac{1}{2} \mathbf{Pr}_{\mathcal{A}}(abb) = \frac{3}{8}$$

More generally, the following recursive equations hold:

$$\mathbf{Pr}_{\mathcal{A}}(wa) = \frac{1}{2}\mathbf{Pr}_{\mathcal{A}}(w) \text{ and } \mathbf{Pr}_{\mathcal{A}}(wb) = \frac{1}{2}(1 + \mathbf{Pr}_{\mathcal{A}}(w))$$

from which one can derive an explicit expression of the acceptance probability:

$$\mathbf{Pr}_{\mathcal{A}}(a_1 \dots a_n) = \sum_{i=1}^{n} 2^{i-n-1} \cdot \mathbf{1}_{a_i=b}$$

Which word maximizes the acceptance probability?



# **Stochastic languages**

We are interested in "useful" policies.

This directly leads to the introduction of stochastic languages. Let:

- $ightharpoonup \mathcal{A}$  be a probabilistic automaton;
- $\theta \in [0,1]$  be a threshold;
- $\blacktriangleright\bowtie\in\{<,\leq,>,\geq,=,\neq\}$  be a comparison operator.

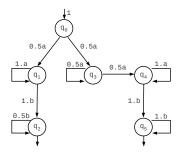
Then  $L_{\bowtie \theta}(\mathcal{A})$  is defined by:

$$L_{\bowtie \theta}(\mathcal{A}) = \{ w \in A^* \mid \mathbf{Pr}_{\mathcal{A}}(w) \bowtie \theta \}$$

For expressiveness and decidability issues, one also needs the following definitions.

- A rational PA is a PA with probability distributions over  $\mathbb{Q}^Q$ .
- ► A rational stochastic language is a stochastic language specified by a rational PA and a rational threshold.

### **Counting with PA**



(a succinct representation with an omitted absorbing rejecting state)

Any word z different from  $a^mb^n$  with m>0, n>0 cannot be accepted.

Let  $w \stackrel{\text{def}}{=} a^m b^n$  with m > 0, n > 0. w can be accepted by:

- ▶ a path  $q_0, q_1^m, q_2^n$  with probability  $\frac{1}{2^n}$ ;
- ▶ or by a family of paths  $q_0, q_3^r, q_4^s, q_5^n$  with 0 < r, s and r + s = m with cumulated probability  $\frac{1}{2} \frac{1}{2^m}$ .

Summing, one obtains:  $\frac{1}{2} + \frac{1}{2^n} - \frac{1}{2^m}$ .

Thus: 
$$\mathcal{L}_{=0.5}(\mathcal{A}) = \{a^n b^n \mid n > 0\}$$

### **Plan**

**Presentation** 

Properties of Stochastic Languages

**Decidability Results** 

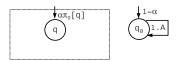
# **Expressiveness problems**

Provide a minimal set of comparison operators and thresholds.

Position the stochastic languages w.r.t. the Chomsky hierarchy.

Study the closure properties of the stochastic languages.

# A single threshold is enough



The value  $\alpha$  depends on  $\theta \neq \frac{1}{2}$  in the following way:

▶ If  $\theta > \frac{1}{2}$  then  $q_0 \notin F$  and  $\alpha \stackrel{\mathsf{def}}{=} \frac{1}{2\theta}$  so that for all  $w \in A^*$ ,

$$\mathbf{Pr}_{\mathcal{A}'}(w) = \frac{1}{2\theta} \mathbf{Pr}_{\mathcal{A}}(w)$$

Thus  $w \in L_{\bowtie \frac{1}{2}}(\mathcal{A}')$  iff  $w \in L_{\bowtie \theta}(\mathcal{A})$ .

▶ If  $\theta < \frac{1}{2}$  then  $q_0 \in F$  and  $\alpha \stackrel{\mathrm{def}}{=} \frac{1}{2(1-\theta)}$  so that for all  $w \in A^*$ ,

$$\mathbf{Pr}_{\mathcal{A}'}(w) = \frac{1-2\theta + \mathbf{Pr}_{\mathcal{A}}(w)}{2(1-\theta)}$$

Thus  $w \in L_{\bowtie \frac{1}{2}}(\mathcal{A}')$  iff  $w \in L_{\bowtie \theta}(\mathcal{A})$ .

# **Getting rid of (dis)equality**

Given a PA A, we build A' as follows.

- ▶ The set of states  $Q' \stackrel{\text{def}}{=} Q \times Q$ ;
- $\mathbf{P}'_a[(q_1, q_2), (q'_1, q'_2)] \stackrel{\text{def}}{=} \mathbf{P}_a[q_1, q'_1] \mathbf{P}_a[q_2, q'_2];$
- $\bullet$   $\pi'_0[q_1, q_2] \stackrel{\text{def}}{=} \pi_0[q_1]\pi_0[q_2]$  and  $F' \stackrel{\text{def}}{=} F \times (Q \setminus F)$ .

Once a word w is selected,

the two components of the DES behave independently and so:

$$\mathbf{Pr}_{\mathcal{A}'}(w) = \mathbf{Pr}_{\mathcal{A}}(w)(1 - \mathbf{Pr}_{\mathcal{A}}(w))$$

Consequently  $\mathbf{Pr}_{\mathcal{A}'}(w) \leq \frac{1}{4}$  with equality iff  $\mathbf{Pr}_{\mathcal{A}}(w) = \frac{1}{2}$ . Thus:

$$L_{>\frac{1}{4}}(\mathcal{A}') = L_{=\frac{1}{2}}(\mathcal{A})$$

# Getting rid of "lower (or equal) than"

Given a PA A, we build A' by complementing the final states. Then:

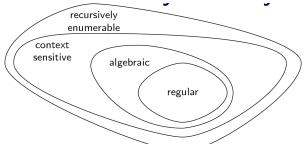
$$\mathbf{Pr}_{\mathcal{A}'}(w) = 1 - \mathbf{Pr}_{\mathcal{A}}(w)$$

And so:

$$L_{>\theta}(\mathcal{A}') = L_{<\theta}(\mathcal{A})$$

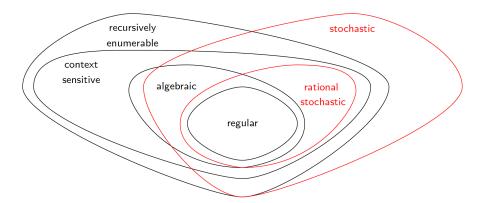
$$L_{>\theta}(\mathcal{A}') = L_{<\theta}(\mathcal{A})$$

# The Chomsky hierarchy

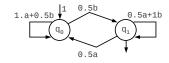


Class	Grammar	Device	
Regular language	$L \to aR a \varepsilon$	Finite automaton	
	with $L,R\in\Delta$ , $a\in\Sigma$		
Algebraic language	$L  o R_1 \dots R_n$ with	Stack automaton	
	$L \in \Delta$ and $R_i \in \Delta \cup \Sigma$		
Context-sensitive	$L_1 \dots L_m \to R_1 \dots R_n$	Non determ. Turing	
language	$m \leq n$ , $(S \rightarrow \varepsilon)$	machine performing in	
	with $L_i, R_j \in \Delta \cup \Sigma$	linear space	
Recursively enumerable	$L_1 \dots L_m \to R_1 \dots R_n$	Turing machine	
language	avec $L_i, R_j \in \Delta \cup \Sigma$		

# Revisiting the Chomsky hierarchy



# Non recursively enumerable languages



Define  $v_a \stackrel{\text{def}}{=} 0$  and  $v_b \stackrel{\text{def}}{=} 1$ .

The acceptance probability of  $w_1 \dots w_n$  is the binary number  $0.v_{w_n} \dots v_{w_1}$ . So  $\mathcal{L}_{>\theta}(\mathcal{A})$  is the set of representations of numbers (with finite binary development) greater than  $\theta$ .

Thus given  $0 \le \theta < \theta' \le 1$ ,

$$\mathcal{L}_{>\theta'}(\mathcal{A}) \subsetneq \mathcal{L}_{>\theta}(\mathcal{A})$$

So there is an uncountable number of stochastic languages implying that "most" of them are non recursively enumerable.

This result does not hold for rational stochastic languages.



# Regular versus stochastic languages

A deterministic automaton is a stochastic automaton with probabilities in  $\{0,1\}$ .

Thus regular languages are stochastic languages.

The language  $\{a^nb^n \mid n>0\}$  is a rational stochastic non regular language.

# Non stochastic context-free languages (1)

$$L \stackrel{\mathsf{def}}{=} \{a^{n_1}ba^{n_2}b\dots a^{n_k}ba^* \mid \exists i > 1 \ n_i = n_1\}$$
 is a non stochastic context-free language.

#### Proof.

L is context-free. Use a non deterministic automaton with a counter.

- $\blacktriangleright$  With a counter one counts  $n_1$  the number of a's until the first occurrence of b.
- ▶ Then one guesses an occurrence of *b* and decrements the counter by the occurrences of *a* until the next occurrence of *b*.
- ▶ If the counter is zero the word is accepted.

Assume that (1) 
$$L = L_{>\theta}(A)$$
 or (2)  $L = L_{>\theta}(A)$ .

Let  $\sum_{i=0}^{n} c_i X^i$  be the minimal polynomial of  $\mathbf{P}_a$ .

Since 1 is an eigenvalue of  $P_a$ , one gets  $\sum_{i=0}^n c_i = 0$  and there are positive and negative coefficients.

By definition,  $\sum_{i=0}^{n} c_i \mathbf{P}_{a^i} = 0$  and so for any word w,  $\sum_{i=0}^{n} c_i \mathbf{P}_{a^i w} = 0$ .

# Non stochastic context-free languages (2)

#### Proof (continued).

Let  $Pos = \{i \mid 0 \le i \le n \land c_i > 0\}$  and  $NonPos = \{i \mid 0 \le i \le n \land c_i \le 0\}$ . Write Pos as  $\{i_1, \ldots, i_k\}$ .

Choose  $w \stackrel{\mathsf{def}}{=} ba^{i_1}b \dots ba^{i_k}b$ .

Case 
$$L=L_{>\theta}(\mathcal{A})$$
. Let  $0\leq i\leq n$ , by definition of  $L$ , 
$$\pi_0\mathbf{P}_{a^iw}\mathbf{1}_F^T>\theta \text{ iff } i\in\{i_1,\ldots,i_k\}$$

$$0 = \sum_{i=0}^{n} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T = \sum_{i \in Pos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T$$

$$> (\sum_{i \in Pos} c_i) \theta + (\sum_{i \in NonPos} c_i) \theta = (\sum_{i=0}^{n} c_i) \theta = 0$$

leading to a contradiction.

Case 
$$L=L_{\geq \theta}(\mathcal{A})$$
. Let  $0\leq i\leq n$ , by definition of  $L$ , 
$$\pi_0\mathbf{P}_{a^iw}\mathbf{1}_F^T\geq \theta \text{ iff } i\in\{i_1,\ldots,i_k\}$$

So: 
$$0 = \sum_{i=0}^{n} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T = \sum_{i \in Pos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T + \sum_{i \in NonPos} c_i \mathbf{1}_F + \sum_{i \in NonPos} c_i$$

leading to a contradiction.

# Non context-free stochastic languages (1)

$$L \stackrel{\mathsf{def}}{=} \{ a^n b^n c^n \mid n > 0 \}$$

is a non context-free rational stochastic language.

#### Proof.

Using Ogden's lemma it can be easily proved that L is not context-free.

Observe that 
$$L=L_1\cap L_2$$
 with  $L_1\stackrel{\mathrm{def}}{=}\{a^nb^nc^+\mid n>0\}$  and  $L_2\stackrel{\mathrm{def}}{=}\{a^+b^nc^n\mid n>0\}.$ 

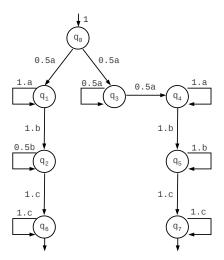
So we prove that:

- for  $i \in \{1,2\}$ ,  $L_i = L_{=\frac{1}{2}}(\mathcal{A}_i)$  for some  $\mathcal{A}_i$
- the family of languages  $\{L=L_{=\frac{1}{2}}(\mathcal{A})\}_{\mathcal{A}}$  is closed under intersection.

# Non context-free stochastic languages (2)

Proof (continued).

$$L_{=\frac{1}{2}}(\mathcal{A}) = \{a^n b^n c^+ \mid n > 0\}$$



# Non context-free stochastic languages (3)

#### Proof (ended).

Let  $L_{=\frac{1}{2}}(\mathcal{A}_1)$  and  $L_{=\frac{1}{2}}(\mathcal{A}_2)$  be two arbitrary languages.

Using the previous construction, let  $\mathcal{A}_1'$  and  $\mathcal{A}_2'$  be automata such that:

- ▶ For any word w,  $\mathbf{Pr}_{\mathcal{A}'_i}(w) \leq \frac{1}{4}$ ;
- $L_{=\frac{1}{2}}(\mathcal{A}_i) = L_{=\frac{1}{4}}(\mathcal{A}'_i).$

#### One builds A as follows:

- ▶ The set of states  $Q \stackrel{\text{def}}{=} Q'_1 \times Q'_2$ ;
- $\mathbf{P}_a[(q_1, q_2), (q_1', q_2')] \stackrel{\mathsf{def}}{=} (\mathbf{P}_1')_a[q_1, q_1'] (\mathbf{P}_2')_a[q_2, q_2'];$
- $\qquad \qquad \pi_0'[q_1,q_2] \stackrel{\mathsf{def}}{=} \pi_{1,0}[q_1]\pi_{2,0}[q_2] \text{ and } F \stackrel{\mathsf{def}}{=} F_1' \times F_2'.$

By construction,  $\mathbf{Pr}_{\mathcal{A}}(w) = \mathbf{Pr}_{\mathcal{A}'_1}(w)\mathbf{Pr}_{\mathcal{A}'_2}(w)$ .

So for all word w,  $\mathbf{Pr}_{\mathcal{A}}(w) \leq \frac{1}{16}$  and  $\mathbf{Pr}_{\mathcal{A}}(w) = \frac{1}{16}$  iff  $\mathbf{Pr}_{\mathcal{A}_1'}(w) = \mathbf{Pr}_{\mathcal{A}_2'}(w) = \frac{1}{4}$ .

Consequently,

$$L_{=\frac{1}{16}}(\mathcal{A}) = L_{=\frac{1}{2}}(\mathcal{A}_1) \cap L_{=\frac{1}{2}}(\mathcal{A}_2)$$

# **Inclusion in context-sensitive languages**

The class of rational stochastic languages is strictly included in the class of context-sensitive languages.

#### Proof.

Context-sensitive languages are the languages for which membership checking can be performed by a non deterministic procedure in linear space.

A deterministic procedure in linear space (far from being optimal) Pre-computation in constant space.

- ► Compute the l.c.m., say b, of denominators of  $\theta$ , items of matrices  $\{\mathbf{P}_a\}_{a\in A}$ , and items of vector  $\pi_0$ .
- ▶ Build the integer matrices  $\mathbf{P}'_a \stackrel{\text{def}}{=} b\mathbf{P}_a$  and vector  $\pi'_0 \stackrel{\text{def}}{=} b\pi_0$ .

For word  $w \stackrel{\text{def}}{=} a_1 \dots a_n$ , the problem becomes  $\pi'_0(\prod_{i=1}^n \mathbf{P}'_{a_i})\mathbf{1}_F^T \bowtie \theta b^{n+1}$ ?

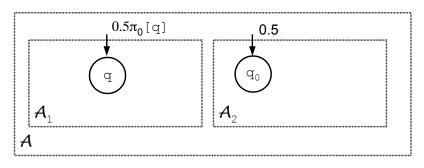
- ▶ Compute  $\theta b^{n+1}$  in space O(n).
- ► Compute  $\mathbf{v} \stackrel{\text{def}}{=} \pi_0'(\prod_{i=1}^n \mathbf{P}_{a_i}')$  by initializing  $\mathbf{v}$  to  $\pi_0'$  and then iteratively multiply it by  $\mathbf{P}_{a_i}'$ . The vectors are bounded by  $b^{n+1}$ . So this is performed in space O(n).
- ▶ The sum and comparison are also done in space O(n).

# Operations with regular languages

The family of (rational) stochastic languages is closed under intersection and union with regular languages.

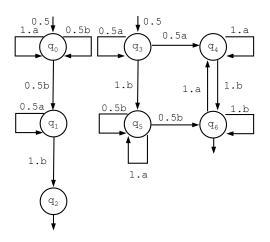
#### Proof.

Let  $L_{\bowtie \theta}(\mathcal{A}_1)$  be a (rational) stochastic language (with  $\bowtie \in \{>, \geq\}$ ) and  $L_{=1}(\mathcal{A}_2)$  be a regular language.



$$L_{\bowtie \frac{\theta}{2}}(\mathcal{A}) = L_{\bowtie \theta}(\mathcal{A}_1) \cup L_{=1}(\mathcal{A}_2) \text{ and } L_{\bowtie \frac{1+\theta}{2}}(\mathcal{A}) = L_{\bowtie \theta}(\mathcal{A}_1) \cap L_{=1}(\mathcal{A}_2)$$

### A stochastic language



$$L_{=\frac{1}{2}}(\mathcal{A}) = \{a^{m_1}b \dots ba^{m_k}b \mid 1 < k \land m_1 = m_k\}$$

since 
$$\mathbf{Pr}_{\mathcal{A}}(a^{m_1}b\dots ba^{m_k}b) = \frac{1}{2}\left(\left(\frac{1}{2}\right)^{k+m_k-1} + 1 - \left(\frac{1}{2}\right)^{k+m_1-1}\right)$$

### **Concatenation**

The family of (rational) stochastic languages is not closed under concatenation with a regular language.

#### Proof.

Let 
$$L \stackrel{\text{def}}{=} \{a^{m_1}b \dots ba^{m_k}b \mid 1 < k \wedge m_1 = m_k\}$$
 be the previous stochastic language.

Then  $LA^* = \{a^{m_1}ba^{m_2}b\dots a^{m_k}ba^* \mid \exists i > 1 \ m_i = m_1\}$  which is not a stochastic language.

### **Iteration**

The family of (rational) stochastic languages is not closed under Kleene star.

#### Proof.

Let  $L \stackrel{\mathrm{def}}{=} \{a^{m_1}b \dots ba^{m_k}b \mid 1 < k \wedge m_1 = m_k\}$  be the previous stochastic language. Assume that  $L^* = L_{\bowtie \theta}(\mathcal{A})$  with  $\bowtie \in \{>, \geq\}$ .

Let  $\sum_{i=0}^n c_i X^i$  be the minimal polynomial of  $\mathbf{P}_a$ . Since 1 is an eigenvalue of  $\mathbf{P}_a$ , one gets  $\sum_{i=0}^n c_i = 0$  and there are positive and negative coefficients.

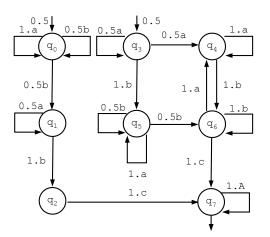
By definition,  $\sum_{i=0}^{n} c_i \mathbf{P}_{a^i} = 0$  and so for any word w,  $\sum_{i=0}^{n} c_i \mathbf{P}_{a^i w} = 0$ . Let  $c_{i_1}, \ldots, c_{i_k}$  be the positive coefficients of this polynomial.

Let  $w \stackrel{\text{def}}{=} ba^{i_1}b(a^{i_2}b)^2 \dots (a^{i_k}b)^2$ .  $a^iw \in L^*$  iff  $i \in \{i_1, \dots, i_k\}$ .

Case  $L^* = L_{>\theta}(\mathcal{A})$ . Let  $0 \le i \le n$ ,  $\pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T > \theta$  iff  $i \in \{i_1, \dots, i_k\}$ . So:  $0 = \sum_{i=0}^n c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T > (\sum_{i=0}^n c_i)\theta = 0$  leading to a contradiction.

Case  $L^* = L_{\geq \theta}(\mathcal{A})$ . Let  $0 \leq i \leq n$ ,  $\pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T \geq \theta$  iff  $i \in \{i_1, \dots, i_k\}$ . So:  $0 = \sum_{i=0}^n c_i \pi_0 \mathbf{P}_{a^i w} \mathbf{1}_F^T > (\sum_{i=0}^n c_i)\theta = 0$  leading to a contradiction.

### A stochastic language



$$L_{=\frac{1}{2}}(\mathcal{A}) = \{a^{m_1}b \dots ba^{m_k}bcA^* \mid 1 < k \land m_1 = m_k\}$$

### Homomorphism

The family of (rational) stochastic languages is not closed under homomorphism.

#### Proof.

Let  $L \stackrel{\text{def}}{=} \{a^{m_1}b \dots ba^{m_k}bcA^* \mid 1 < k \land m_1 = m_k\}$  be the previous stochastic language.

Define the homomorphism h from A to  $A' \stackrel{\mathsf{def}}{=} \{a, b\}$  by:

$$h(a) \stackrel{\mathsf{def}}{=} a \qquad h(b) \stackrel{\mathsf{def}}{=} b \qquad h(c) \stackrel{\mathsf{def}}{=} \varepsilon$$

Then  $h(L) = \{a^{m_1}ba^{m_2}b\dots a^{m_k}ba^* \mid \exists i > 1 \ m_i = m_1\}$  which is not a stochastic language.

### **Plan**

**Presentation** 

**Properties of Stochastic Languages** 

Oecidability Results

### Two decision problems

Let A and A' be probabilistic automata.

#### First problem

Are A and A' equivalent?

$$\forall w \in A^* \mathbf{Pr}_{\mathcal{A}}(w) = \mathbf{Pr}_{\mathcal{A}'}(w)$$

#### Second problem

Is 
$$L_{\bowtie \theta}(\mathcal{A})$$
 equal to  $L_{\bowtie' \theta'}(\mathcal{A}')$ ?

For deterministic automata this is the same problem. It can be solved in polynomial time by a product construction which provides a witness of non equivalence of size less than |Q||Q'|.

### Linear algebra recalls

Let  $\mathbf{v_0} \in \mathbb{R}^n$  and  $\mathbf{v_1}, \dots, \mathbf{v_k}$  be linearly independent vectors of  $\mathbb{R}^n$ .

How to check whether  $v_0$  is a linear combination of  $v_1, \dots, v_k$ ?

• Solve in  $O(k^3 + n^2)$ 

$$\begin{pmatrix} \mathbf{v}_1[1] & \dots & \mathbf{v}_k[1] \\ \dots & \dots & \dots \\ \mathbf{v}_1[n] & \dots & \mathbf{v}_k[n] \end{pmatrix} \begin{pmatrix} x_1 \\ \vdots \\ x_k \end{pmatrix} = \begin{pmatrix} \mathbf{v}_0[1] \\ \vdots \\ \mathbf{v}_0[n] \end{pmatrix}$$

ullet When  $\mathbf{v_1}, \dots, \mathbf{v_k}$  are orthogonal (i.e. for all  $a \neq b$ ,  $\mathbf{v}_a \cdot \mathbf{v}_b \stackrel{\text{def}}{=} \sum_{i=1}^n \mathbf{v}_a[i]\mathbf{v}_b[i] = 0$ )

$$\stackrel{\text{ef}}{=} \sum_{i=1}^{n} \mathbf{v}_a[i] \mathbf{v}_b[i] = 0)$$

Compute in O(kn) the orthogonal projection

$$\mathbf{w}_0 = \sum_{i=1}^k \frac{\mathbf{v}_0 \cdot \mathbf{v}_i}{\mathbf{v}_i \cdot \mathbf{v}_i} \mathbf{v}_i$$

Check in O(n) whether  $\mathbf{v}_0 = \mathbf{w}_0$ .



# Principles of equivalence checking

#### **Enumeration of words**

Looking for a counter-example whose length is increasing starting with word  $\varepsilon$ .

#### A stack

Managing a stack of words w in order to find counter-examples aw for all  $a \in A$ . For efficiency purposes, the stack contains tuples  $(\mathbf{P}_w \mathbf{1}_F, \mathbf{P}_w' \mathbf{1}_{F'}, w)$ .

#### An orthogonal family for restricting the enumeration

Gen is a set of independent orthogonal vectors of  $\mathbb{R}^{Q \cup Q'}$ .

If w is not a counter-example, check if  $\mathbf{v} \stackrel{\text{def}}{=} (\mathbf{P}_w \mathbf{1}_F, \mathbf{P}'_w \mathbf{1}_{F'})$  is generated by Gen.

- ightharpoonup producing  $\mathbf{v}'$  the orthogonal projection of  $\mathbf{v}$  on subspace spanned by Gen;
- ightharpoonup comparing  $\mathbf{v}'$  to  $\mathbf{v}$ .

If  $v' \neq v$  then:

- ▶ w is added to the stack:
- $\mathbf{v} \mathbf{v}'$  is added to Gen.

# The algorithm

```
If \pi_0 \cdot \mathbf{1}_F \neq \pi_0' \cdot \mathbf{1}_{F'} then return(false, \varepsilon)
Gen \leftarrow \{(\mathbf{1}_F, \mathbf{1}_{F'})\}; \mathbf{Push}(Stack, (\mathbf{1}_F, \mathbf{1}_{F'}, \varepsilon))
Repeat
     (\mathbf{v}, \mathbf{v}', w) \leftarrow \mathbf{Pop}(Stack)
     For a \in A do
        \mathbf{z} \leftarrow \mathbf{P}_a \mathbf{v}; \ \mathbf{z}' \leftarrow \mathbf{P}'_a \mathbf{v}'
        If \pi_0 \cdot \mathbf{z} \neq \pi_0' \cdot \mathbf{z}' then return(false, aw)
        \mathbf{v} \leftarrow \mathbf{0} : \mathbf{v}' \leftarrow \mathbf{0}
        For (\mathbf{x}, \mathbf{x}') \in Gen do
           \mathbf{y} \leftarrow \mathbf{y} + \frac{\mathbf{z} \cdot \mathbf{x}}{\mathbf{x} \cdot \mathbf{x}} \mathbf{x}
           \mathbf{y}' \leftarrow \mathbf{y}' + \frac{\mathbf{z}' \cdot \mathbf{x}'}{\mathbf{z}'} \mathbf{x}'
        If (\mathbf{z}, \mathbf{z}') \neq (\mathbf{v}, \mathbf{v}') then
            \mathbf{Push}(Stack,(\mathbf{z},\mathbf{z}',aw))
            Gen \leftarrow Gen \cup \{(\mathbf{z} - \mathbf{y}, \mathbf{z}' - \mathbf{y}')\}
Until IsEmpty(Stack)
return(true)
```

# **Complexity**

#### Time complexity

An item is pushed on the stack iff an item is added to Gen.

There can be no more than |Q| + |Q'| items in Gen.

So there are at most |Q| + |Q'| iterations of the external loop.

The index of the first inner loop ranges over A while the index of the most inner loop ranges over Gen.

The operations inside the most inner loop are done in O(|Q| + |Q'|).

This leads to an overall time complexity of  $O((|Q| + |Q'|)^3 |A|)$ .

#### Length of witnesses

In addition, the length of the witness is at most |Q|+|Q'|. (also valid for deterministic automata)

### **Correctness**

Assume that the automata are not equivalent and that the algorithm returns **true**.

Let u be a non examined word such that  $\mathbf{Pr}_{\mathcal{A}}(u) \neq \mathbf{Pr}_{\mathcal{A}'}(u)$ .

Let  $u\stackrel{\mathrm{def}}{=} w'w$  with  $w(\neq u)$  the greatest suffix examined by the algorithm.

Among such words u, pick one word such that |w'| is minimal.

**Claim.** There exists w'' that has been inserted in the stack before w such that  $\mathbf{Pr}_{\mathcal{A}}(w'w'') \neq \mathbf{Pr}_{\mathcal{A}'}(w'w'')$ .

Let  $Gen = \{w_1, \dots, w_k\}$  when examining w, there exist  $\lambda_1, \dots, \lambda_k$  such that:

So: 
$$\mathbf{P}_w \mathbf{1}_F = \sum_{i=1}^k \lambda_i \mathbf{P}_{w_i} \mathbf{1}_F$$
 and  $\mathbf{P}_w' \mathbf{1}_{F'} = \sum_{i=1}^k \lambda_i \mathbf{P}_{w_i}' \mathbf{1}_{F'}$ 

$$\mathbf{Pr}_{\mathcal{A}}(w'w) \stackrel{\text{def}}{=} \pi_0 \mathbf{P}_{w'} \mathbf{P}_w \mathbf{1}_F = \sum_{i=1}^k \lambda_i \pi_0 \mathbf{P}_{w'} \mathbf{P}_{w_i} \mathbf{1}_F = \sum_{i=1}^k \lambda_i \mathbf{Pr}_{\mathcal{A}}(w'w_i)$$
Similarly:  $\mathbf{Pr}_{\mathcal{A}'}(w'w) = \sum_{i=1}^k \lambda_i \mathbf{Pr}_{\mathcal{A}'}(w'w_i)$ 

So there exists i, with  $\mathbf{Pr}_{\mathcal{A}}(w'w_i) \neq \mathbf{Pr}_{\mathcal{A}'}(w'w_i)$ .

Let  $w' \stackrel{\text{def}}{=} w'''a$ .  $aw_i$  is examined by the algorithm.

So the word  $u' \stackrel{\text{def}}{=} w'w_i$  has a decomposition  $u' \stackrel{\text{def}}{=} z'z$  where z the greatest suffix examined by the algorithm has for suffix  $aw_i$ . So |z'| < |w'|: a contradiction.

# Undecidability of the equality problem

Given  $\mathcal A$  a rational stochastic automaton, the question  $L_{=\frac{1}{2}}(\mathcal A)=\{\varepsilon\}$ ? is undecidable.

#### Proof.

By reduction of the undecidable Post correspondence problem (PCP): Given an alphabet A and two morphisms  $\varphi_1, \varphi_2$  from A to  $\{0,1\}^+$ , does there exist a word  $w \in A^+$  such that  $\varphi_1(w) = \varphi_2(w)$ ?

Already undecidable for a restriction where the images of letters lie in  $(10 + 11)^+$ . Inserting a 1 before each letter of images reduces the former problem to the latter.

A word  $w \stackrel{\text{def}}{=} a_1 \dots a_n \in (10+11)^+$  defines a value  $val(w) \in [0,1]$  by:

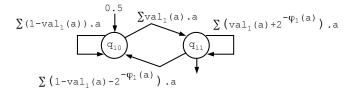
$$val(w) \stackrel{\mathsf{def}}{=} \sum_{i=1}^{n} \frac{a_i}{2^{n+1-i}}$$

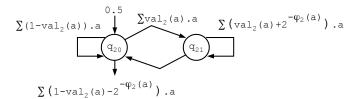
Since every word starts with a 1, val(w) = val(w') implies w = w'.



### Reduction of PCP

For  $w \in A^+$  and  $i \in \{1, 2\}$ , define  $val_i(w) \stackrel{\text{def}}{=} val(\varphi_i(w))$ .

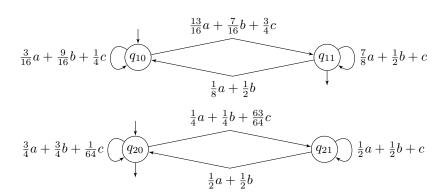




### Illustration of the reduction

A	a	b	c
$\varphi_1$	(1)0(1)1	(1)0(1)0	(1)1
$\varphi_2$	(1)0	(1)0	(1)1(1)1(1)1

A	a	b	c
$val_1$	$\frac{13}{16}$	$\frac{7}{16}$	$\frac{3}{4}$
$val_2$	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{63}{64}$



### Correctness of the reduction

The recurrence equation:

$$\mathbf{1}_{q_{i0}} \mathbf{P}_{wa} \mathbf{1}_{q_{i1}}^T = \mathbf{1}_{q_{i0}} \mathbf{P}_w \mathbf{1}_{q_{i1}}^T (val_i(a) + 2^{-|\varphi_i(a)|}) + (1 - \mathbf{1}_{q_{i0}} \mathbf{P}_w \mathbf{1}_{q_{i1}}^T) val_i(a)$$
$$= val_i(a) + 2^{-|\varphi_i(a)|} \mathbf{1}_{q_{i0}} \mathbf{P}_w \mathbf{1}_{q_{i1}}^T$$

By induction we obtain that for all  $w \stackrel{\text{def}}{=} a_1 \dots a_n$ :

$$\mathbf{1}_{q_{i0}} \mathbf{P}_{w} \mathbf{1}_{q_{i1}}^{T} = \sum_{i=1}^{n} val_{i}(a_{j}) 2^{-\sum_{j < k \le n} |\varphi_{i}(a_{k})|} = val_{i}(w)$$

So for  $w \in A^+$ :  $\mathbf{Pr}_{\mathcal{A}}(w) = \frac{1}{2}(val_1(w) + 1 - val_2(w))$ .

Thus  $w \in L_{=\frac{1}{5}}(\mathcal{A})$  iff  $val(\varphi_1(w)) = val(\varphi_2(w))$  implying that  $\varphi_1(w) = \varphi_2(w)$ .