

Boltzmann Sampling

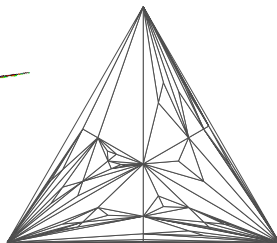
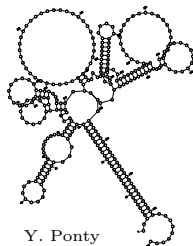
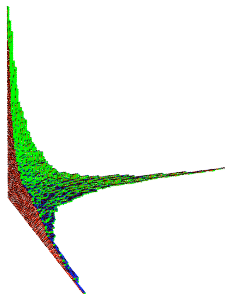
Carine Pivoteau

Université PARIS EST

december 2010

Motivations

- observation, experimentation, statistical study
 - conjectures,
 - parameters distribution,
 - limit properties,
 - validation : correctness/model adequacy.
- simulations, tests
 - programs validity,
 - software robustness.
- applications
 - combinatorics,
 - bio-informatics,
 - software testing,
 - statistical physics,
 - ...



Need of **efficient** and **automatic** samplers !

Generate series-parallel circuits

$$\mathcal{S} = \mathcal{Z} + \text{SEQ}_{\geq 2}(\mathcal{P})$$

$$\mathcal{P} = \mathcal{Z} + \text{MSET}_{\geq 2}(\mathcal{S})$$

$$\sum_{n \geq 0} c_n z^n \quad \downarrow \text{generating series}$$

$$S(z) = z + \frac{1}{1 - P(z)} - 1 - P(z)$$

$$P(z) = z + \exp\left(\sum_{k \geq 0} \frac{S(z^k)}{k}\right) - 1 - S(z)$$

coefficients

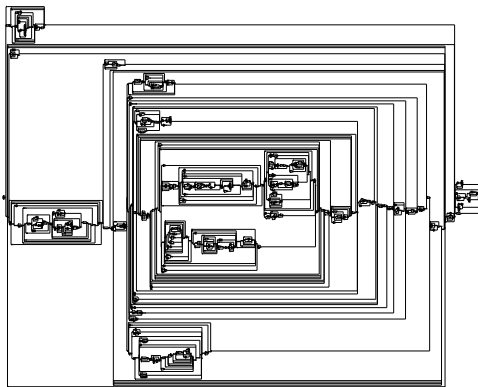
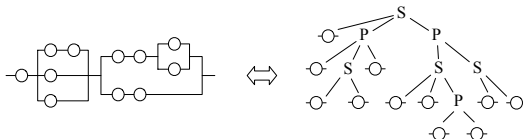
oracle

Recursive
Sampling

[FlZiVC94]

Boltzmann
Method

[DuFLoSc04]



Decomposable classes

[Flajolet, Sedgewick]

Symbolic method [FlSe08] :

- Combinatorial specification (recursive grammar) :

$$\mathcal{Y} = \mathcal{H}(\mathcal{Z}, \mathcal{Y}) \equiv \begin{cases} \mathcal{Y}_1 = \mathcal{H}_1(\mathcal{Z}, \mathcal{Y}_1, \mathcal{Y}_2, \dots, \mathcal{Y}_m), \\ \mathcal{Y}_2 = \mathcal{H}_2(\mathcal{Z}, \mathcal{Y}_1, \mathcal{Y}_2, \dots, \mathcal{Y}_m), \\ \vdots \\ \mathcal{Y}_m = \mathcal{H}_m(\mathcal{Z}, \mathcal{Y}_1, \mathcal{Y}_2, \dots, \mathcal{Y}_m), \end{cases}$$

- usual combinatorial constructions (labelled or not) :

$\mathcal{E}$, $\mathcal{Z}$, $+$, $\times$, sequence, cycle, set

- automatic generating series for any combinatorial class $\mathcal{C}$:

$$\text{ordinary : } C(z) = \sum_{n \geq 0} c_n z^n \qquad \text{exponential : } \hat{C}(z) = \sum_{n \geq 0} c_n \frac{z^n}{n!}$$

where c_n is the number of structures of $\mathcal{C}$ that are of size n .

Generating series dictionary

	construction	ordinary g.s. (unlabelled)	exponential g.s. (labelled)
$\mathcal{E}$ / atom	1 / $\mathcal{Z}$	1 / z	1 / z
Union	$\mathcal{A} + \mathcal{B}$	$A(z) + B(z)$	$A(z) + B(z)$
Product	$\mathcal{A} \times \mathcal{B}$	$A(z) \times B(z)$	$A(z) \times B(z)$
Sequence	$\text{SEQ}(\mathcal{A})$	$\frac{1}{1 - A(z)}$	$\frac{1}{1 - A(z)}$
Set	$\text{PSET}(\mathcal{A})$	$\exp\left(\sum_{k=1}^{\infty} \frac{(-1)^{k-1}}{k} A(z^k)\right)$	$\exp(A(z))$
Multi-Set	$\text{MSET}(\mathcal{A})$	$\exp\left(\sum_{k=1}^{\infty} \frac{1}{k} A(z^k)\right)$	—
Cycle	$\text{CYC}(\mathcal{A})$	$\sum_{k=1}^{\infty} \frac{\varphi(k)}{k} \log \frac{1}{1 - A(z^k)}$	$\log \frac{1}{1 - A(z)}$

1 Introduction

2 Boltzmann Samplers

- Labelled Boltzmann Samplers
- Unlabelled samplers

3 Computing the oracle

- Transfer of convergence
- Newton iterations

4 Conclusion

The recursive method on an example

Rooted binary trees

$$1 + z + 2z^2 + 5z^3 + 14z^4 + \dots$$

$$\mathcal{T} = \mathcal{E} + \mathcal{Z} \times \mathcal{T}^2$$

The recursive method on an example

Rooted binary trees

$$1 + z + 2z^2 + 5z^3 + 14z^4 + \dots$$



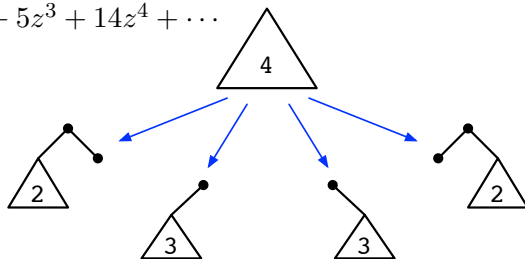
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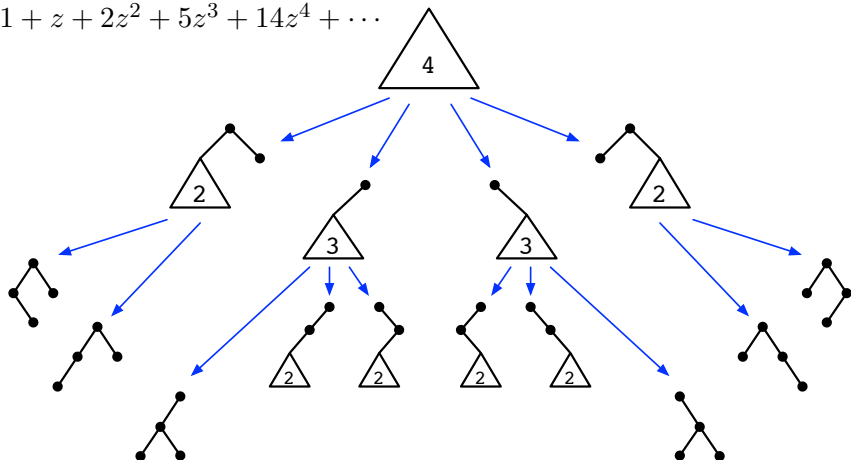


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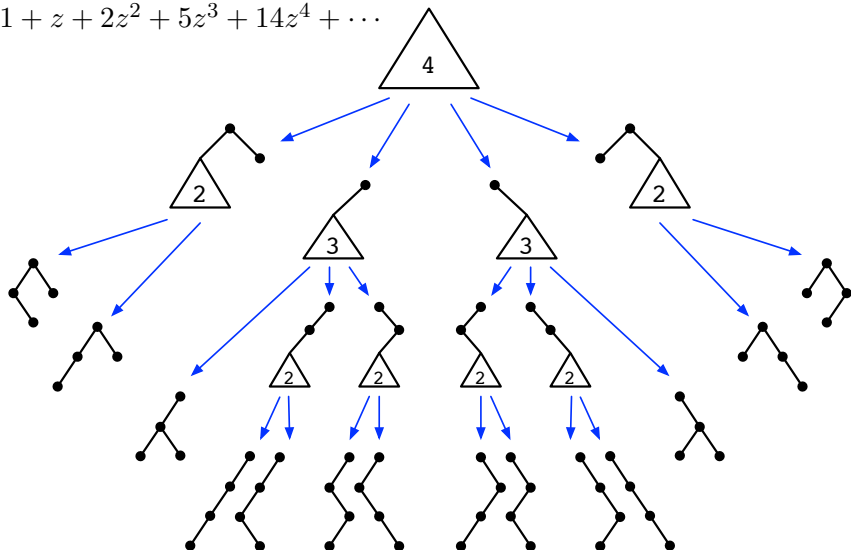


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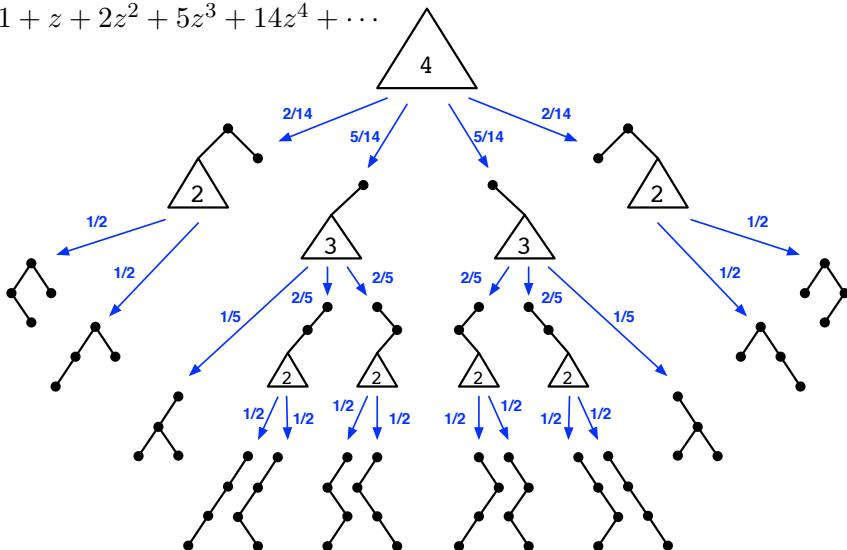


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Boltzmann model

Size distribution spread over the whole combinatorial class, but **uniform** for a sub-class of objects of **the same size**.

- **approximate size** sampling (control parameter).
- **very large structures** can be generated.

Definition (Duchon, Flajolet, Louchard, Schaeffer – 2004)

For a given $x < \rho_C$, the *unlabelled Boltzmann model* associates to any object c in the class $\mathcal{C}$ the probability :

$$\mathbb{P}_x(c) = \frac{x^{|c|}}{C(x)}$$

A Boltzmann sampler $\Gamma_C(x)$ for the class $\mathcal{C}$ is an algorithm that produces objects of $\mathcal{C}$ according to this model.

2 structures of the same size $\rightarrow$ same probability.

Générateurs

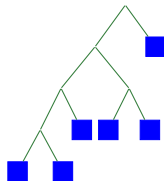
[DuFlLoSc04]

	ordinary g.s. (unlabelled)	exponential g.s. (labelled)	
$1 / \mathcal{Z}$	$1 / z$	$1 / z$	
$\mathcal{A} + \mathcal{B}$	$A(z) + B(z)$	$A(z) + B(z)$	→ Bernoulli law
$\mathcal{A} \times \mathcal{B}$	$A(z) \times B(z)$	$A(z) \times B(z)$	→ no choice
$\text{SEQ}(\mathcal{A})$	$\frac{1}{1 - A(z)}$	$\frac{1}{1 - A(z)}$	→ geometric law ▷ $\mathbb{P}(X = k) = (1 - \lambda)\lambda^k$
$\text{PSET}(\mathcal{A})$	...	$\exp(A(z))$	→ Poisson law ▷ $\mathbb{P}(X = k) = e^{-\lambda} \frac{\lambda^k}{k!}$
$\text{MSET}(\mathcal{A})$	...	—	
$\text{CYC}(\mathcal{A})$	...	$\log \frac{1}{1 - A(z)}$	→ logarithmic law ▷ $\mathbb{P}(X = k) = \frac{1}{\log(1 - \lambda)^{-1}} \frac{\lambda^k}{k}$

Example : binary trees

$$\mathcal{B} = \mathcal{Z} + \mathcal{B} \times \mathcal{B}$$

$$B(z) = z + B(z)^2 = \frac{1 - \sqrt{1 - 4z}}{2}$$



Algorithm : $\Gamma B(x)$

$$b \leftarrow \text{Bern}(x/B(x));$$

if $b = 1$ then

Return 

else

Return $\langle \Gamma B(x) , \Gamma B(x) \rangle$;

end if

Boltzmann method

[DuFlLoSc04]

Theorem (Labelled samplers – DuFlLoSc04)

For any **labelled** class $\mathcal{C}$ specified (possibly recursively) using the following constructions :

$$\varepsilon, \mathcal{Z}, +, \times, \text{SEQ}, \text{SET}, \text{CYC},$$

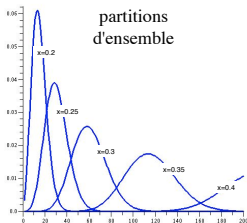
the **free Boltzmann sampler** $\Gamma_{\mathcal{C}}(x)$ works in **linear time** with respect to the size of the output.

- **Free sampler** : structures of **any sizes** !
 - ▷ **target** size n : choose x such that **the expected** size is n .

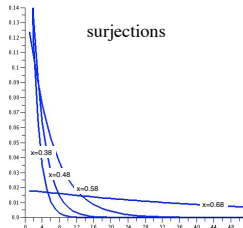
$$\mathbb{E}_x(N) = x \frac{C'(x)}{C(x)}$$

- **Approximate** or **exact** size sampler : use **rejection**.
 - ▷ the size distribution determine the **cost of rejection**.

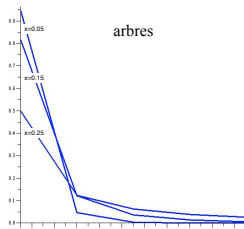
Approximate/exact size sampling



$$C(x) = e^{(e^x - 1)}$$



$$C(x) = \frac{1}{2 - e^x}$$



$$C(x) = \frac{1 - \sqrt{1 - 4x}}{2}$$

Theorem (Complexities – DuFiloSc04)

distribution type	bumpy	flat	picked
approximate size	$\mathcal{O}(n)$	$\mathcal{O}(n)$	$o(n^2)$
exact size	$o(n^2)$	$o(n^2)$	—

▷ pointing
singular sampler

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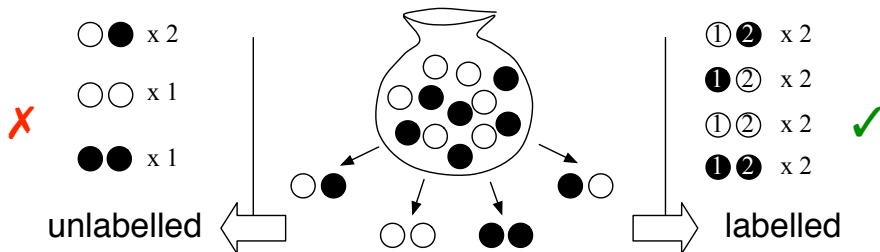
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3 Computing the oracle

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4 Conclusion

The symmetries issue



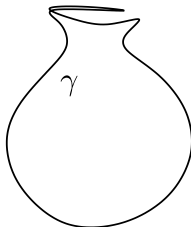
Pólya operators

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The multi-set example

 $\text{MSet}(\mathcal{A})$

$$\prod_{k=1}^{\infty} \exp\left(\frac{1}{k} A(x^k)\right)$$

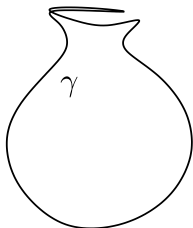


- Draw k such that : $\mathbb{P}(K \leq k) = \prod_{j \leq k} \exp(A(x^j)/j)$
- For any index i up to k
 - Draw p – the number of elements to generate – according to a Poisson law of parameter $A(x^i)/i$.
 - Perform p independent call to the sampler of the class $\mathcal{A}$, with parameter x^i , then make i copies of the result.

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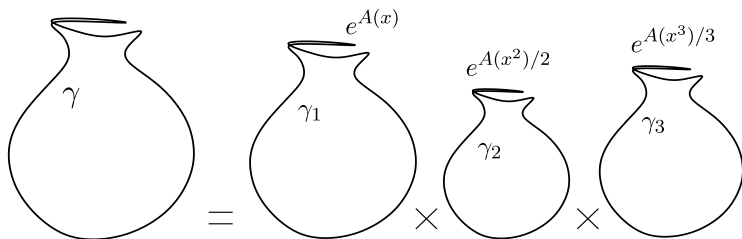


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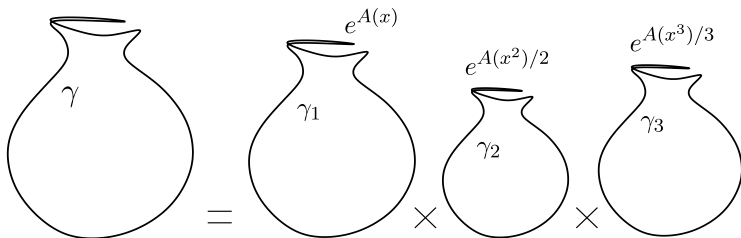


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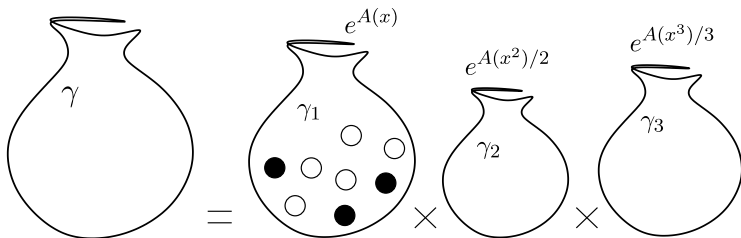


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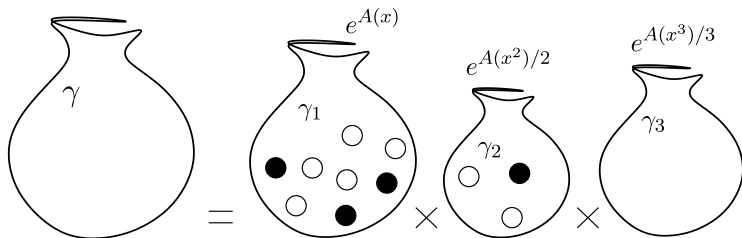


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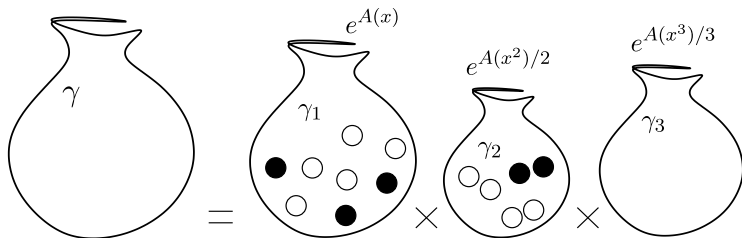


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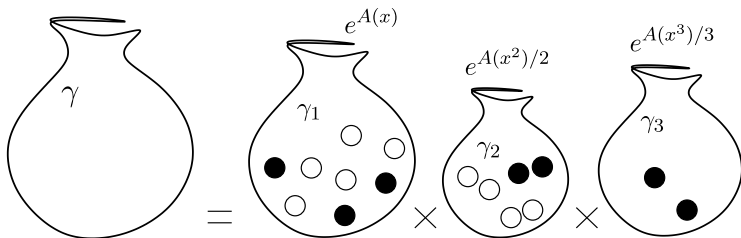


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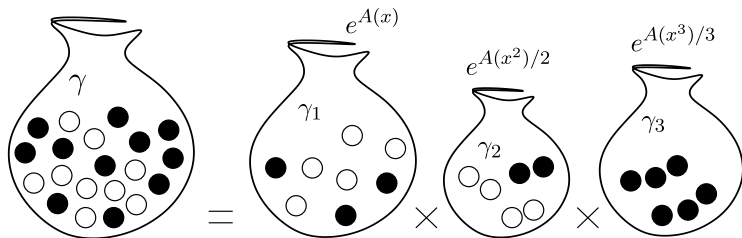


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Unlabelled Boltzmann samplers

Theorem (Unlabelled samplers – FlFuPi07)

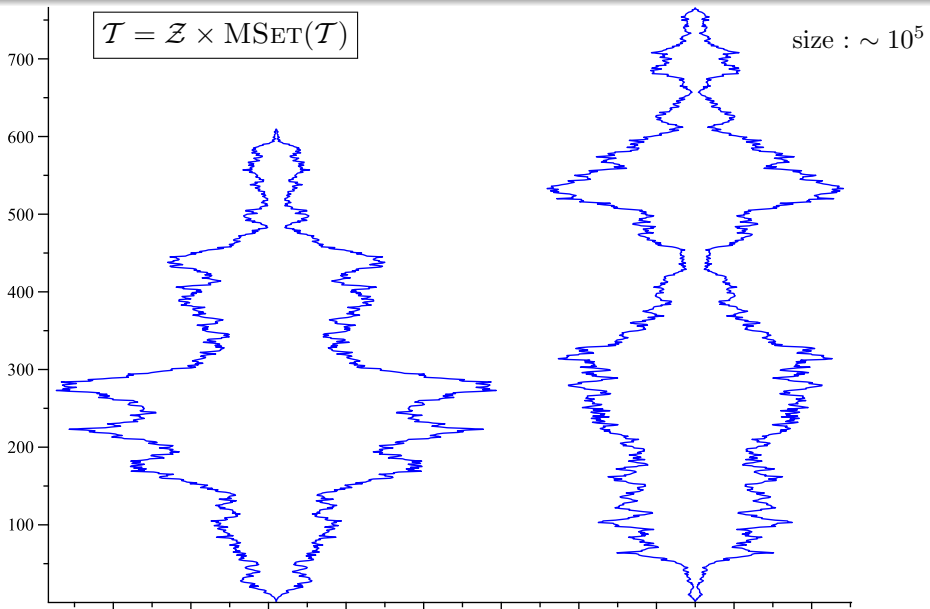
For any **unlabelled** class $\mathcal{C}$ specified (possibly recursively) using the following constructions :

$$\varepsilon, \mathcal{Z}, +, \times, \text{SEQ}, \text{SEQ}_k, \text{MSET}, \text{MSET}_k, \text{CYC}, \text{CYC}_k,$$

the **free Boltzmann sampler** $\Gamma_{\mathcal{C}}(x)$ works in **linear time** with respect to the size of the output.

[FlFuPi07] *Boltzmann sampling of unlabelled structures*, P. Flajolet, É. Fusy and C. Pivoteau

General non planar rooted trees



Experimental results

Approximate size sampling $\pm 10\%$ (times in seconds)

	bumpy distribution	flat distribution		picked distribution		
	Integer partitions	Cyclic compositions	Functional graphs	Otter trees	General trees	Circuits
size	$O(\sqrt{n})$	$O(n)$	$O(n)$	$O(n)$	$O(n)$	$O(n)$
10^2	0.0020	0.0047	0.0084	0.024	0.018	0.047
10^3	0.039	0.051	0.080	0.20	0.17	0.43
10^4	0.078	0.57	0.82	2.2	1.6	4.7
10^5	0.17	6.2	8.4	22	18	43
10^6	0.47	60	86	240	170	530
10^7	1.5	570	—	—	—	—
10^8	5.2	—	—	—	—	—

up to 10^{10}

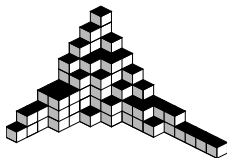
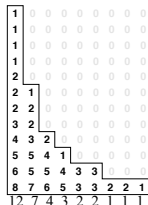
with Maple

with OCaml, on an Intel processor, 3.2 GHz with a 2 GB RAM

Application to plane partitions

$$\mathcal{P} = \text{MSET}(\mathcal{Z} \times \text{SEQ}(\mathcal{Z})^2)$$

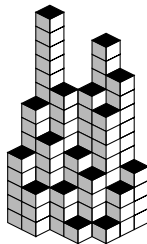
$$P(z) = \prod_{i,j \geq 0} \frac{1}{1 - z^{i+j+1}}$$



plane partition

$$\mathcal{S}_D \prod_{(i,j) \in D} \text{SEQ}(\mathcal{Z}^{h(i,j)})$$

$$S_D(z) \prod_{(i,j) \in D} \frac{1}{1 - z^{h(i,j)}}$$



skew plane partition

Theorem (Average complexity)

Plane partitions :

- approx. size : $O(n \ln(n)^3)$
- exact size : $O(n^{\frac{4}{3}})$

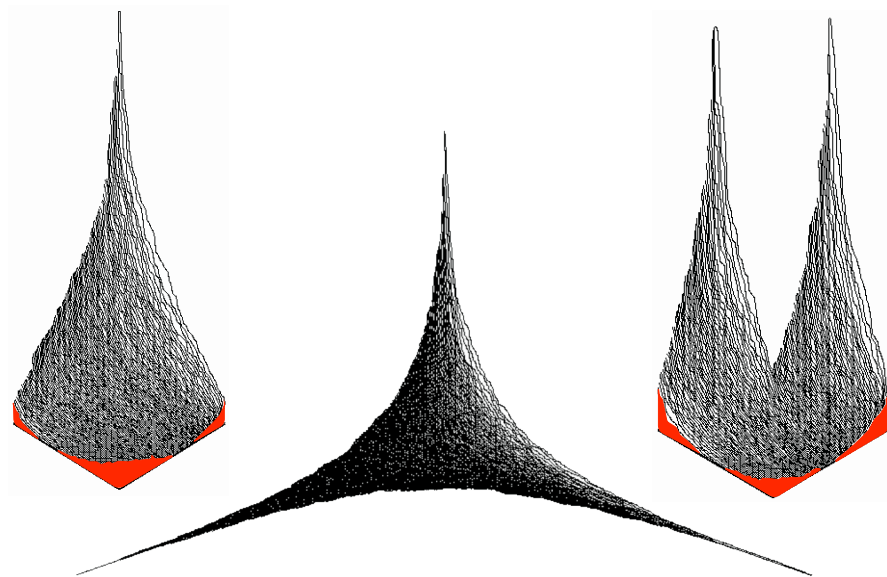
Boxed or skew partitions (on a domain D) :

- approximate size :
 $O(1)$ when $n \rightarrow \infty$
- exact size :
bounded by $C|D|.n$,
 C being a constant.

[BoFuPi06]

Random sampling of plane partitions,
O. Bodini, E. Fusy and C. Pivoteau

Giant partitions



Related work (non exhaustive list)

Applications

- Accessible det. automata : *F. Bassino, J. David, C. Nicaud.*
- Planar graphs, *E. Fusy.*
- RNA structures, *Y. Ponty.*
- Apollonian network, *A. Darrasse, M. Soria.*
- Software testing, *J. Oudinet.*

Extensions

- Unbiased pointing operator for unlabeled structures, *M. Bodirsky, E. Fusy, M. Kang, S. Vigerske.*
- Colored structures, *O. Bodini, A. Jacquot.*
- Ordered product, Shuffle, Hadamard product, *A. Darrasse, K. Panagiotou, O. Roussel, M. Soria.*
- Multivariate Boltzmann samplers, *O. Bodini, Y. Ponty.*

Miscellaneous

- Properties of Random Graphs via Boltzmann Samplers, *K. Panagiotou, A. Weißl.*

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Example

Combinatorial specification for an algebraic language :

$$\mathcal{C}_0 = \mathcal{Z}\mathcal{C}_1\mathcal{C}_2\mathcal{C}_3(\mathcal{C}_1 + \mathcal{C}_2)$$

$$\mathcal{C}_1 = \mathcal{Z} + \mathcal{Z}\text{SEQ}(\mathcal{C}_1^2\mathcal{C}_3^2)$$

$$\mathcal{C}_2 = \mathcal{Z} + \mathcal{Z}^2\text{SEQ}(\mathcal{Z}\mathcal{C}_2^2\text{SEQ}(\mathcal{Z}))\text{SEQ}(\mathcal{C}_2)$$

$$\mathcal{C}_3 = \mathcal{Z} + \mathcal{Z}(3\mathcal{Z} + \mathcal{Z}^2 + \mathcal{Z}^2\mathcal{C}_1\mathcal{C}_3)\text{SEQ}(\mathcal{C}_1^2)$$

The associated system on generating series :

$$C_0(z) = zC_1(z)C_2(z)C_3(z)(C_1(z) + C_2(z))$$

$$C_1(z) = z + z/(1 - C_1(z)^2C_3(z)^2)$$

$$C_2(z) = z + z^2/((1 - zC_2(z)^2/(1 - z))(1 - C_2(z)))$$

$$C_3(z) = z + z(3z + z^2 + z^2C_1(z)C_3(z))/(1 - C_1^2(z))$$

Example

Combinatorial specification for an algebraic language :

$$\mathcal{C}_0 = \mathcal{Z}\mathcal{C}_1\mathcal{C}_2\mathcal{C}_3(\mathcal{C}_1 + \mathcal{C}_2)$$

$$\mathcal{C}_1 = \mathcal{Z} + \mathcal{Z}\text{SEQ}(\mathcal{C}_1^2\mathcal{C}_3^2)$$

$$\mathcal{C}_2 = \mathcal{Z} + \mathcal{Z}^2\text{SEQ}(\mathcal{Z}\mathcal{C}_2^2\text{SEQ}(\mathcal{Z}))\text{SEQ}(\mathcal{C}_2)$$

$$\mathcal{C}_3 = \mathcal{Z} + \mathcal{Z}(3\mathcal{Z} + \mathcal{Z}^2 + \mathcal{Z}^2\mathcal{C}_1\mathcal{C}_3)\text{SEQ}(\mathcal{C}_1^2)$$

The associated system on generating series : with $z = 0.27$

$$C_0(z) = zC_1(z)C_2(z)C_3(z)(C_1(z) + C_2(z))$$

$$C_1(z) = z + z/(1 - C_1(z)^2C_3(z)^2)$$

$$C_2(z) = z + z^2/((1 - zC_2(z)^2/(1 - z))(1 - C_2(z)))$$

$$C_3(z) = z + z(3z + z^2 + z^2C_1(z)C_3(z))/(1 - C_1^2(z))$$

Example

Combinatorial specification for an algebraic language :

$$C_0 = ZC_1C_2C_3(C_1 + C_2)$$

$$C_1 = Z + Z\text{SEQ}(C_1^2C_3^2)$$

$$C_2 = Z + Z^2\text{SEQ}(ZC_2^2\text{SEQ}(Z))\text{SEQ}(C_2)$$

$$C_3 = Z + Z(3Z + Z^2 + Z^2C_1C_3)\text{SEQ}(C_1^2)$$

The associated system on generating series : with $z = 0.27$

$$C_0 = 0.27C_1C_2C_3(C_1 + C_2)$$

$$C_1 = 0.27 + 0.27/(1 - C_1^2C_3^2)$$

$$C_2 = 0.27 + 0.0729/((1 - 0.2754211138C_2^2)(1 - C_2))$$

$$C_3 = 0.27 + 0.27(0.8829 + 0.0729C_1C_3)/(1 - C_1^2)$$

with Maple

$$\text{sys} := \left[\begin{array}{l} C0 = x \, C3 \, C1 \, C2 \, (C1 + C2), \, C3 = x + \frac{x \, (3 \, x + x^2 + C1 \, C3 \, x^2)}{1 - C1^2}, \\ \\ C2 = x + \frac{x^2}{\left(1 - \frac{x \, C2^2}{1 - x}\right) (1 - C2)}, \, C1 = x + \frac{x}{1 - C1^2 \, C3^2} \end{array} \right]$$

solve(subs(x=0.27,sys));

{ C2 = 0.3988484105, C3 = 0.643292746, C0 = 0.03981177934, C1 = 0.5844488219 },
 { C2 = 0.3988484105, C0 = 0.08483583330, C1 = 0.7669225413, C3 = 0.881136046 },
 { C2 = 0.3988484105, C1 = 3.827601486, C3 = 0.25115016, C0 = 0.4375296896 },
 { C3 = 0.643292746, C1 = 0.5844488219, C2 = 0.8243443421, C0 = 0.1178894124 },
 { C1 = 0.7669225413, C3 = 0.881136046, C2 = 0.8243443421, C0 = 0.2393370429 },
 { C1 = 3.827601486, C3 = 0.25115016, C2 = 0.8243443421, C0 = 0.9953303530 },
 { C3 = 0.643292746, C1 = 0.5844488219, C2 = 1.702776766, C0 = 0.3953535314 },
 { C1 = 0.7669225413, C3 = 0.881136046, C2 = 1.702776766, C0 = 0.7672908697 },
 { C1 = 3.827601486, C3 = 0.25115016, C2 = 1.702776766, C0 = 2.444198592 }

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Idea

By iteration

Boltzmann Oracle :
numerical iteration that converges
to the unique **relevant** solution

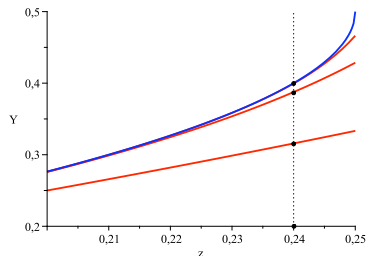
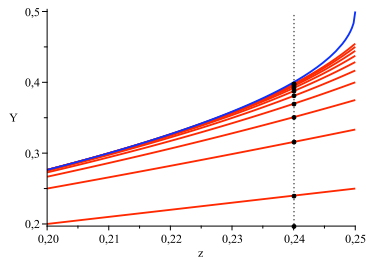


convergence of the iteration
on counting series



convergence of the iteration on
combinatorial functional systems

Fast oracle : Newton iteration



Binary trees : $B(z) = z + B(z)^2$

Theorem (Transfer of convergence – PiSaSo08)

Let $\mathcal{Y} = \mathcal{H}(\mathcal{Z}, \mathcal{Y})$ be a system such that $\mathcal{H}(0, 0) = 0$ and $\partial\mathcal{H}/\partial\mathcal{Y}(0, 0)$ is nilpotent. If the combinatorial iteration :

$$\mathcal{Y}^{[n+1]} = \mathcal{F}(\mathcal{Z}, \mathcal{Y}^{[n]}), \quad \text{avec } \mathcal{Y}^{[0]} = 0,$$

converges to the family of species $\mathcal{Y}$, solution of $\mathcal{Y} = \mathcal{H}(\mathcal{Z}, \mathcal{Y})$, then the iteration :

$$\mathbf{Y}^{[n+1]}(z) = \mathbf{F}(z, \mathbf{Y}^{[n]}(z)), \quad \text{avec } \mathbf{Y}^{[0]}(z) = 0,$$

converges to the vector of generating series $\mathbf{Y}(z)$ of $\mathcal{Y}$. Besides, if $\mathcal{F}$ is analytic, for any α such that $|\alpha| < \rho$, the iteration :

$$\mathbf{y}^{[n+1]} = \mathbf{F}(\alpha, \mathbf{y}^{[n]}), \quad \text{avec } \mathbf{y}^{[0]} = 0,$$

converges to the vector $\mathbf{Y}(\alpha)$ of values of the series $\mathbf{Y}(z)$ at point α .

[PiSaSo08] *Boltzmann oracle for combinatorial systems*, C. Pivoteau, B. Salvy and M. Soria.

Validity of this approach

Combinatorial convergence

- Implicit Species Theorem [Joyal 81] : **existence** and **unicity** of the solution for the system $\mathcal{Y} = \mathcal{H}(\mathcal{Z}, \mathcal{Y})$:
 - $\mathcal{H}(\mathbf{0}, \mathbf{0}) = \mathbf{0}$,
 - the jacobian matrix $\frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathbf{0}, \mathbf{0})$ is nilpotent.
- Measure of convergence : **contact** between combinatorial classes.

Generating series

- Counting of combinatorial structures.
- Contact $\Leftrightarrow$ **valuation** of series.

Numerical iteration

- Numerical values $\Leftrightarrow$ evaluation of series at point α .
- Analyticity of H in α and $y^{[n]}$.

Example of general plane trees

(Newton)

$$\mathcal{Y}^{[0]} = \emptyset$$

$$\mathcal{T} = \mathcal{Z} \times \text{SEQ}(\mathcal{T})$$

$$\mathcal{Y}^{[1]} = \boxed{\bullet} \boxed{\bullet} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \dots$$

$$\mathcal{Y}^{[2]} = \boxed{\bullet} \boxed{\bullet} \boxed{\begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array}} \boxed{\begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array}} \boxed{\begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array} \begin{array}{c} \bullet \\ \bullet \end{array}} \dots$$

$$Y^{[0]}(z) = \mathbf{0}$$

$$Y^{[1]}(z) = \mathbf{z} + \mathbf{z}^2 + z^3 + z^4 + z^5 + z^6 + z^7 + z^8 + z^9 + \dots$$

$$Y^{[2]}(z) = \mathbf{z} + \mathbf{z}^2 + \mathbf{2z}^3 + \mathbf{5z}^4 + \mathbf{14z}^5 + \mathbf{42z}^6 + 131z^7 + 417z^8 + 1341z^9 \dots$$

$$Y^{[3]}(z) = \mathbf{z} + \mathbf{z}^2 + \mathbf{2z}^3 + \mathbf{5z}^4 + \mathbf{14z}^5 + \mathbf{42z}^6 + \mathbf{132z}^7 + \dots + \mathbf{742900z}^{14} + \dots$$

$$Y^{[4]}(z) = \mathbf{z} + \mathbf{z}^2 + \mathbf{2z}^3 + \mathbf{5z}^4 + \mathbf{14z}^5 + \mathbf{42z}^6 + \dots + \mathbf{1002242216651368z}^{30} + \dots$$

$$y^{[0]} = \mathbf{0}$$

with $z = 0.2$

$$y^{[1]} = \mathbf{0.2}500000000000000000000000000000 \dots$$

$$y^{[2]} = \mathbf{0.27}586206896551724137931034482758 \dots$$

$$y^{[3]} = \mathbf{0.27639}296187683284457478005865104 \dots$$

$$y^{[4]} = \mathbf{0.2763932022}4997168102435963404908 \dots$$

$$y^{[5]} = \mathbf{0.27639320225002103035908263}104683 \dots$$

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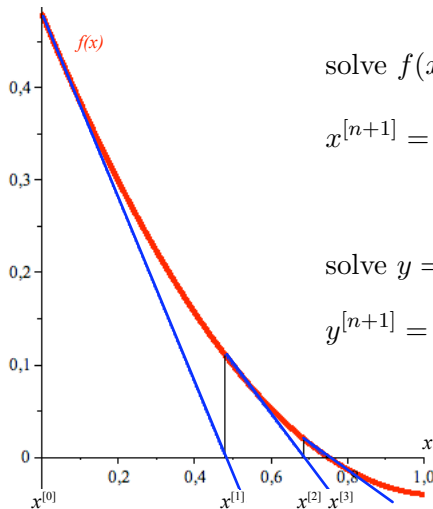
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Classical Newton numerical iteration



solve $f(x) = 0$:

$$x^{[n+1]} = x^{[n]} - \frac{f(x^{[n]})}{f'(x^{[n]})}$$

solve $y = f(y)$:

$$y^{[n+1]} = y^{[n]} + \frac{1}{1 - f'(y^{[n]})} (f(y^{[n]}) - y^{[n]})$$

Combinatorial Newton iteration

For a unique equation [Décoste, Labelle, Leroux 82] :

$$\mathcal{Y}^{[n+1]} = \mathcal{Y}^{[n]} + \text{SEQ}\left(\frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathcal{Z}, \mathcal{Y}^{[n]})\right) \times \left(\mathcal{H}(\mathcal{Z}, \mathcal{Y}^{[n]}) - \mathcal{Y}^{[n]}\right), \quad \mathcal{Y}^{[0]} = 0$$

▷ combinatorial derivative

For a system :

$$\mathcal{Y}^{[n+1]} = \mathcal{Y}^{[n]} + \left(\text{Id} - \frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathcal{Z}, \mathcal{Y}^{[n]})\right)^{-1} \left(\mathcal{H}(\mathcal{Z}, \mathcal{Y}^{[n]}) - \mathcal{Y}^{[n]}\right), \quad \mathcal{Y}^{[0]} = 0$$

▷ jacobian matrix

▷ “combinatorial bloomings” [Labelle 85]

Optimization

Compute the matrix $\left(\text{Id} - \frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathcal{Z}, \mathcal{Y}^{[n]})\right)^{-1}$ by Newton iteration

▷ iterations in parallel.

Combinatorial derivative

$\partial\mathcal{H}/\partial\mathcal{T}$: *derivative* of $\mathcal{H}(\mathcal{Z}, \mathcal{Y}_1, \dots, \mathcal{Y}_m)$ with respect to $\mathcal{T}$.

	notation		derivative	
atom	$\mathcal{Z}$ ou $\mathcal{E}$		$\emptyset$	
$\mathcal{Y}_i$	$\mathcal{Y}_i = \mathcal{T}$	$\mathcal{Y}_i \neq \mathcal{T}$	$\mathcal{E}$	$\mathcal{Y}'_i$
union	$\mathcal{Y}_i + \mathcal{Y}_j$		$\mathcal{Y}_i + \mathcal{Y}'_i + \mathcal{Y}'_j$	
product	$\mathcal{Y}_i \times \mathcal{Y}_j$		$\mathcal{Y}'_i \times \mathcal{Y}_j + \mathcal{Y}_i \times \mathcal{Y}'_j$	
sequence	$\text{SEQ}(\mathcal{Y}_i)$		$\text{SEQ}(\mathcal{Y}_i) \times \mathcal{Y}'_i \times \text{SEQ}(\mathcal{Y}_i)$	
cycle	$\text{CYC}(\mathcal{Y}_i)$		$\text{SEQ}(\mathcal{Y}_i) \times \mathcal{Y}'_i$	
set	$\text{SET}(\mathcal{Y}_i)$		$\text{SET}(\mathcal{Y}_i) \times \mathcal{Y}'_i$	

Example : $\mathcal{H}(\mathcal{Z}, \mathcal{T}) = \mathcal{T}^2$, $\partial\mathcal{H}/\partial\mathcal{T}(\mathcal{Z}, \mathcal{T}) = 2\mathcal{T}$.

▷ Combinatorial interpretation of the product by a derivative

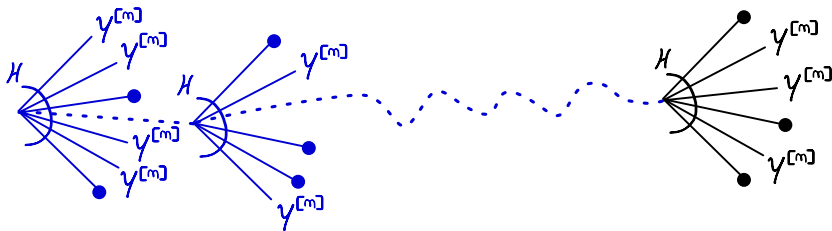
$\partial\mathcal{H}/\partial\mathcal{Y}$: la *jacobian matrix* de $\mathcal{H}(\mathcal{Z}, \mathcal{Y})$ with respect to $\mathcal{Y}$,
its entries are the partial derivatives $\partial\mathcal{H}_i/\partial\mathcal{Y}_j(\mathcal{Z}, \mathcal{Y})$.

Convergence of the combinatorial Newton iteration

Combinatorial interpretation of Newton iteration

[Décoste, Labelle, Leroux 82]

$$\mathcal{Y}^{[n+1]} = \mathcal{Y}^{[n]} + \text{SEQ}\left(\frac{\partial \mathcal{H}}{\partial \mathcal{Y}}(\mathcal{Z}, \mathcal{Y}^{[n]})\right) \times \left(\mathcal{H}(\mathcal{Z}, \mathcal{Y}^{[n]}) - \mathcal{Y}^{[n]}\right), \quad \mathcal{Y}^{[0]} = 0$$



▷ quadratic convergence!

Newton iteration on series

$$\mathbf{Y}^{[k+1]}(z) = \mathbf{Y}^{[k]}(z) + \left(I - \frac{\partial \mathbf{H}}{\partial \mathbf{Y}}(z, \mathbf{Y}^{[k]}(z)) \right)^{-1} \left(\mathbf{H}(z, \mathbf{Y}^{[k]}(z)) - \mathbf{Y}^{[k]}(z) \right)$$

Fast algorithm to compute the coefficients !

Arithmetic and binary complexity : quasi-optimal

Theorem (Arithmetic complexity)

Cost of the calculation of the first N terms of $\mathbf{Y}(z)$: $O(M(N))$

- ▷ $M(N)$: cost of multiplying series of order N .
- ▷ with FFT : $O(N \log N)$.
- ▷ better complexity for the recursive method.

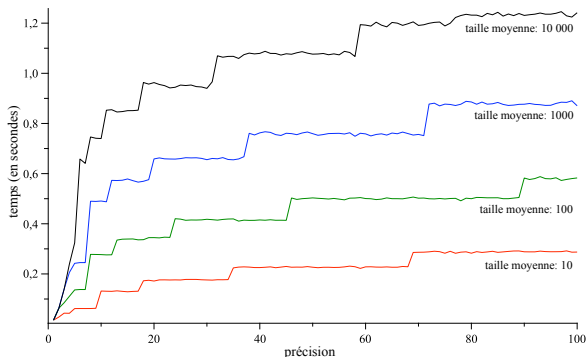
Numerical oracle

Theorem (Boltzmann oracle– PiSaSo08)

Let $\mathcal{Y} = \mathcal{H}(\mathcal{Z}, \mathcal{Y})$ be a combinatorial specification (without Pólya operator). The following numerical iteration **converges to $\mathbf{Y}(\alpha)$** :

$$\mathbf{y}^{[k+1]} = \mathbf{y}^{[k]} + \left(\mathbf{I} - \frac{\partial \mathbf{H}}{\partial \mathbf{Y}}(\alpha, \mathbf{y}^{[k]}) \right)^{-1} \left(\mathbf{H}(\alpha, \mathbf{y}^{[k]}) - \mathbf{y}^{[k]} \right), \quad \mathbf{y}^{[0]} = \mathbf{0}$$

Evolution of the computation time of the oracle for the **SP circuits** with respect to the precision, for different values of α .



Implementation

Maple prototype

(specifications without Pólya operators) :

- tests on random grammars,
- XML grammars, $\sim 10^3$ equations [Darrasse 08],
- Uniform random walks in very large models (software testing) [Oudinet 07],
- ...

# equations	4	10	50		100		500
# constructions/eqn	10	10	10	50	10	50	50
$\sim$ size max scc	2.47	3.42	7.95	18.62	10.93	67.18	339.1
time (0.99 ρ)	0.05	0.11	0.17	0.47	0.23	7.29	61.73
time (0.999999 ρ)	0.08	0.16	0.19	0.56	0.25	8.11	61.86
$\sim$ expected size	$4.1 \cdot 10^{14}$	$1.4 \cdot 10^7$	$2.2 \cdot 10^5$	$1.0 \cdot 10^5$	$1.2 \cdot 10^6$	$5.0 \cdot 10^4$	$3.3 \cdot 10^4$

in seconds, with Maple 11, on an Intel processor, 3.2 GHz, with 2 GB RAM.

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Conclusion

- Efficient sampling method, with large scale applications
- Toward a generic package of Boltzmann samplers :
Maple? Sage? C?

Work in progress :

- Pólya operators
- specifications involving $\mathcal{E}$
- binary complexity

Perspectives :

- Convergence acceleration
- Singularities computation
 - ▷ automatic asymptotic analysis
- Boltzmann parameter tuning