## Boltzmann sampling

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based on work by P. Duchon, P. Flajolet, E. Fusy, G. Louchard, C. Pivoteau and G. Schaeffer


## Random generation: different approaches

Fixed size random uniform generation:

- Ad hoc methods
- bijections, surjections, ...

$$
\begin{aligned}
& \mathcal{A}=\phi(\mathcal{B}) \text { and } \Gamma \mathcal{B}(n) \Rightarrow \text { random sampler } \Gamma \mathcal{A}(n) \\
& a_{n}=f\left(a_{n-1}\right) \quad \Rightarrow \quad \text { incremental algorithm } \Gamma \mathcal{A}(n)
\end{aligned}
$$

- rejection $\mathcal{A} \subset \mathcal{B}$ and $\Gamma \mathcal{B}(n) \quad \Rightarrow \quad$ random sampler $\Gamma \mathcal{A}(n)$
- Recursive method : counting + recursive process
- Nijenhuis, Wilf, 1978
- Flajolet, Zimmermann, Van Cutsem, 1994
preprocessing time (to compute g.f. coefficients): $O\left(n^{2}\right)$ random generation time : $O(n \log n)$

Approximate size random uniform generation:

- Boltzmann sampling...


## Boltzmann method

Random sampling under Boltzmann model

- approximate size sampling,
- size distribution spread over the whole combinatorial class, but uniform for a sub-class of objects of the same size,
- control parameter,
- automatized sampling: the sampler is compiled from specification automatically,
- very large objects can be sampled.
$\rightarrow$ large scale simulations
$\rightarrow$ observation of random structures limit properties...
Boltzmann samplers for the random generation of combinatorial structures.
P. Duchon, P. Flajolet, G. Louchard, G. Schaeffer. Combinatorics, Probability and Computing, 13(4-5):577-625, 2004. Special issue on Analysis of Algorithms. Boltzmann sampling of unlabelled structures. Ph. Flajolet, E. Fusy, C. Pivoteau. Proceedings of ANALCO07, january 2007.


## Model definition

## Definition

In the unlabelled case, Boltzmann model assigns to any object $c \in \mathcal{C}$ the following probability:

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

In the labelled case, this probability becomes:

$$
\mathbb{P}_{x}(c)=\frac{1}{\hat{C}(x)} \frac{x^{|c|}}{|c|!}
$$

A free Boltzmann sampler $\Gamma C(x)$ for the class $\mathcal{C}$ is a process that produces objects from $\mathcal{C}$ according to this model.
$\rightarrow 2$ objects of the same size will be drawn with the same probability.

## Unlabelled unions, products, sequences

Suppose $\Gamma A(x)$ and $\Gamma B(x)$ are given:

## Disjoint unions

Boltzmann sampler $\Gamma C$ for $\mathcal{C}=\mathcal{A} \cup \mathcal{B}$ :
With probability $\frac{A(x)}{C(x)}$ do $\Gamma A(x)$ else do $\Gamma B(x) \quad \rightarrow$ Bernoulli.

## Products

Boltzmann sampler $\Gamma C$ for $\mathcal{C}=\mathcal{A} \times \mathcal{B}$ :
Generate a pair $\langle\Gamma A(x), \Gamma B(x)\rangle \quad \rightarrow$ independent calls.

## Sequences

Boltzmann sampler $\Gamma C$ for $\mathcal{C}=\operatorname{SEQ}(\mathcal{A})$ :
Generate $k$ according to a geometric law of parameter $A(x)$
Generate a $k$-tuple $\langle\Gamma A(x), \ldots, \Gamma A(x)\rangle \rightarrow$ independent calls.
Remark: $A(x), B(x)$ and $C(x)$ are given by an oracle.

## Binary trees

$$
\begin{aligned}
\mathcal{B} & =\mathcal{Z}+\mathcal{B} \times \mathcal{B} \\
B(z) & =z+B(z)^{2}=\frac{1-\sqrt{1-4 z}}{2}
\end{aligned}
$$



## Algorithm: $\Gamma B(x)$

$b \leftarrow \operatorname{Bern}(x / B(x))$;
if $b=1$ then
Return $\square$
else
Return $\langle\Gamma B(x), \Gamma B(x)\rangle ;$ end if

## Examples of specifications with $\{U, \times$, Seq $\}$

Regular specifications (non recursive).

- integer compositions, permutations,...
- polyominos that have rational g.f.: column-convex,

- regular languages,

Context-free specifications.

- any algebraic language,
- tree-like structures
- $k$-ary, 2-3-4 trees, ...,
- triangulations,
- noncrossing graphs,
- general planar rooted trees,
- ...



## Labelled classes

Same algorithms, with exponential generating functions

| construction | sampler |
| :--- | :--- |
| $\mathcal{C}=\emptyset$ or $\mathcal{Z}$ | $\Gamma C(x):=\varepsilon$ or atom |
| $\mathcal{C}=\mathcal{A}+\mathcal{B}$ | $\Gamma C(x): \left.=\operatorname{Bern} \frac{\hat{A}(x)}{\hat{C}(x)} \longrightarrow \Gamma A(x) \right\rvert\, \Gamma B(x)$ |
| $\mathcal{C}=\mathcal{A} \times \mathcal{B}$ | $\Gamma C(x):=\langle\Gamma A(x) ; \Gamma B(x)\rangle$ |
| $\mathcal{C}=\operatorname{SeQ}(\mathcal{A})$ | $\Gamma C(x):=\operatorname{Geom} \hat{A}(x) \Longrightarrow \Gamma A(x)$ |

Put the labels at the end !

## Size control - parameter tuning

- Free samplers: produce objects with randomly varying sizes!
- Approximate and exact size samplers: use rejection.
- Tuned samplers: choose $x$ so that expected size is $n$.

$$
\mathbb{E}_{x}(N)=x \frac{C^{\prime}(x)}{C(x)} \quad \text { or } \quad x \frac{\hat{C}^{\prime}(x)}{\hat{C}(x)}
$$

- Size distribution determines the cost of rejection.




