Unranking algorithms applied to MUPAD

X. Molinero and J. Vives

Abstract—We present an improvement of the implementation of some unlabeled unranking algorithms of the open-source algebraic combinatorics package MUPAD-COMBINAT for the computer algebra system MUPAD. We compare our implementation with the current one. Moreover, we have also developed unranking algorithms applied to some unlabeled admissible operators that are not still implemented in the package MUPAD-COMBINAT. These algorithms are also able to develop some structures useful to generate molecules applied to chemistry and influence graphs applied to game theory and social networks, among other topics.

Index Terms—Unranking Algorithms, MuPAD, Generating Molecules, Generating Influence Games.

I. INTRODUCTION

The problem of *unranking* asks for the generation of the *i*th combinatorial object of size n in some combinatorial class \mathcal{A} , according to some well defined order among the objects of size n of the class. Efficient unranking algorithms have been devised for many different combinatorial classes, like binary and Cayley trees, Dyck paths, permutations, strings or integer partitions, but most of the work in this area concentrates in efficient algorithms for particular classes, whereas we aim at generic algorithms that apply to a broad family of combinatorial classes. The problem of unranking is intimately related with its converse, the ranking problem, as well as with the problems of random generation and exhaustive generation of all combinatorial objects of a given size. The interest of this whole subject is witnessed by the vast number of research papers and books that has appeared in over five decades (see, for instance, [24], [12], [9], [8], [11], [25], [10], [20], [19], [21], [3]).

[14], [13] designed *generic* unranking algorithms for a large family of combinatorial classes, namely, those which can be inductively built from the basic ϵ -class (a class which contains only one object of size 0), atomic classes (classes that contain only one object of size 1 or *atom*) and a collection of admissible combinatorial operators: disjoint unions, labeled and unlabeled products, sequence, set, etc. Now we use such techniques to implement those algorithms in MUPAD [2], [18]. In the open-source algebraic combinatorics package MUPAD-COMBINAT [1] for the computer algebra system MUPAD there are implemented the unranking for some admissible combinatorial operators, but now we improve such implementation for unlabeled unions and products (and sequences). Moreover, we

Unlabeled class	Specification
Binary trees	$\mathcal{B} = Z + \mathcal{B} imes \mathcal{B}$
Unary-binary trees or	
Motzkin trees	$\mathcal{M} = Z + Z imes \mathcal{M} + Z imes \mathcal{M} imes \mathcal{M}$
Integer partitions	$\mathcal{P} = Set(Seq(Z, card \ge 1))$
Integer compositions	$C = $ Seq (Set (Z , card ≥ 1))
Non-ordered rooted trees or	
Rooted unlabeled trees	$\mathcal{T} = Z imes Set(\mathcal{T})$
Binary sequences	$\mathcal{A} = \mathbf{Seq}(Z + Z)$
Non plane ternary trees	$\mathcal{D} = Z + Set(\mathcal{D}, card=3)$
Integer partitions with	
distinct parts	$\mathcal{E} = PowerSet(Seq(Z,card{\geq}1))$

Fig. 1. Examples of unlabeled classes and their specifications

have also implemented other operators as unlabeled sets and powersets (with and without restrictions).

The paper just considers unlabeled combinatorial classes and it is organized as follows. In Section II we briefly review basic definitions and concepts, the unranking algorithms and the theoretical analysis of their performance. Afterwards, from the computer algebra system MUPAD, we compare the required CPU time of our implementation with the required CPU time of the current implementation in the package MUPAD-COMBINAT. Moreover, we also explain our current and future work in this subject.

II. PRELIMINARIES

As it will become apparent, all the unranking algorithms in this paper require an efficient algorithm for counting, that is, given a specification of a class and a size, they need to compute the number of objects with the given size. Hence, we will only deal with (some of) the so-called *admissible combinatorial classes* [6], [7]. Those are constructed from *admissible operators*, operations over classes that yield new classes, and such that the number of objects of a given size in the new class can be computed from the number of objects of that size or smaller sizes in the constituent classes. In this paper we just consider unlabeled objects (those whose atoms are indistinguishable¹) built from these admissible combinatorial operators.

For unlabeled classes, the finite specifications are generated from the ϵ -class, atomic classes, and combinatorial operators including disjoint union ('+'), Cartesian product ('×'), sequence ('Seq'), powerset ('PowerSet'), set ('Set')², and sequence, powerset and set (or multiset) with restricted cardinality. Figure 1 gives a few examples of unlabeled admissible classes.

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¹On the contrary, each of the n atoms of a *labeled* object of size n bears a distinct label drawn from the numbers 1 to n.

²Also denoted by 'multisets' (MultiSet) to emphasize that repetition is allowed.

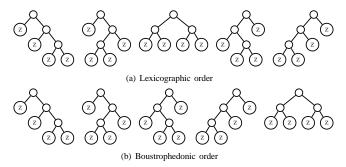


Fig. 2. Binary trees of size 4.

For the rest of this paper, we will use calligraphic uppercase letters to denote classes: $\mathcal{A}, \mathcal{B}, \mathcal{C}, \ldots$ Given a class \mathcal{A} and a size n, \mathcal{A}_n will denote the subset of objects of size n in \mathcal{A} .

The order \prec_{C_n} among the objects of size *n* for a class C = A + B is naturally defined by $\gamma \prec_{C_n} \gamma'$ if both γ and γ' belong to the same class (either A_n or B_n) and $\gamma \prec \gamma'$ within their class, or if $\gamma \in A_n$ and $\gamma' \in B_n$. It is then clear that although A + B and B + A are isomorphic ("the same class"), these two specifications induce quite different orders. The unranking algorithm for disjoint unions compares the given rank with the cardinality of A_n to decide if the sought object belongs to A or to B and then solves the problem by recursively calling the unranking on whatever class (A or B) is appropriate.

For Cartesian products the order in $C_n = (\mathcal{A} \times \mathcal{B})_n$ depends on whether $\gamma = (\alpha, \beta)$ and $\gamma' = (\alpha', \beta')$ have first components of the same size. If $|\alpha| = |\alpha'| = j$ then we have $\gamma \prec_{C_n} \gamma'$ if $\alpha \prec_{\mathcal{A}_j} \alpha'$ or $\alpha = \alpha'$ and $\beta \prec_{\mathcal{B}_{n-j}} \beta'$. But when $|\alpha| \neq |\alpha'|$, we must provide a criterion to order γ and γ' . The *lexicographic* order stems from the specification

$$C_n = A_0 \times B_n + A_1 \times B_{n-1} + \ldots + A_n \times B_0,$$

in other words, the smaller object is that with smaller first component. On the other hand, the *boustrophedonic* order is induced by the specification

$$C_n = \mathcal{A}_0 \times \mathcal{B}_n + \mathcal{A}_n \times \mathcal{B}_0 + \mathcal{A}_1 \times \mathcal{B}_{n-1} + \mathcal{A}_{n-1} \times \mathcal{B}_1 + \mathcal{A}_2 \times \mathcal{B}_{n-2} + \mathcal{A}_{n-2} \times \mathcal{B}_2 + \dots,$$

in other words, we consider that the smaller pairs of total size n are those whose A-component has size 0, then those with A-component of size n, then those with A-component of size 1, and so on. Figure II shows the lists of unlabeled binary trees of size 4 in lexicographic (a) and boustrophedonic order (b).

Of course, other orders are also possible, but they either do not help improving the performance of unranking or they are too complex to be useful or of general applicability. For powersets, among some natural orders (see [14], [16]) we can choose

$$\begin{array}{l} \mathsf{PowerSet}(\mathcal{A}) = \epsilon + \\ \bigcup_{n>0} \bigcup_{j=1}^{n} \bigcup_{k=n \div j}^{1} \quad \left(\mathsf{PowerSet}(\mathcal{A}_{j}, \mathsf{card} = k) \times \right. \\ \\ \mathsf{PowerSet}_{n-kj}(\mathcal{A}_{>j}) \right) \end{array}$$

where

PowerSet(
$$\mathcal{A}_j$$
, card= k) = $\bigcup_{\alpha \in \mathcal{A}_j} \left(\alpha \times \operatorname{PowerSet}(\mathcal{A}_j^{(\succ \alpha)}, \operatorname{card} = k - 1) \right)$

being $\mathcal{A}^{(\succ\alpha)} = \{ \alpha' \in \mathcal{A} : \alpha' \succ \alpha \}$, and PowerSet $(\mathcal{A}_{>j})$ is a powerset with \mathcal{A} -components of size at least equal to j + 1. Other orders described in [14], [16] do not change the complexity and they could also be easily adapted to our implementation.

For sets we have analogous isomorphisms but allowing repetitions.

The theoretical performance of these unranking algorithms is summarized in [16], [13].

Theorem 1: The worst-case time complexity of unranking for objects of size n in any admissible labeled class \mathcal{A} using lexicographic ordering is of $\mathcal{O}(n^2)$ arithmetic operations.

Theorem 2: The worst-case time complexity of unranking for objects of size n in any admissible labeled class \mathcal{A} using boustrophedonic ordering is of $\mathcal{O}(n \log n)$ arithmetic operations.

III. OUR IMPLEMENTATION V.S. MUPAD-COMBINAT IMPLEMENTATION

In this section we compare our implementation³ for unranking in MUPAD with the current implementation of the package MUPAD-COMBINAT(using MUPAD Pro 4.0). All our experiments run under Linux in a AMD64X2 4400 at 2.2 GHz with 4 Gb of RAM, and they use the basic facilities for counting already provided by the package MUPAD-COMBINAT.

For instance, the interface for binary trees has the following inputs:

where spec is the specification⁴, Lexi or Bous forces the lexicographic or boustrophedonic order, respectively, and rank and size are the considered rank and size, respectively. Thus, the following commands provide all binary trees of size 8 in lexicographic order:

```
spec := {B = Union(Z, BB), BB = Prod(B, B)};
p1 := combinat::
        decomposableObjects(spec, Lexi);
for i from 0 to p1::count(8) - 1 do
    p1::unrank(i, 8);
end_for
```

³It is available on request from the first author; send an E-mail to xavier.molinero@upc.edu.

 $^{4}\mbox{The first class defined in the specification is the considered class (B in this case).}$

Lexicographic order					Boustrophedonic order				
	Size	$ au_T$	τ_T'	ρ	Size	$ au_T$	τ_T'	ρ	
	25	1.24	4.55	0.27	25	0.90	2.95	0.30	
	50	3.63	11.95	0.30	50	2.50	6.52	0.38	
	75	6.08	20.17	0.30	75	4.27	10.24	0.41	
	100	9.27	29.65	0.31	100	6.20	13.97	0.44	
	125	12.46	39.98	0.31	125	6.99	17.91	0.39	
	150	16.54	51.73	0.31	150	8.39	20.80	0.40	
	175	21.37	66.16	0.32	175	9.67	24.90	0.38	
	200	26.85	79.93	0.33	200	11.63	28.87	0.40	

TABLE I

Average CPU time (in milliseconds) for Binary Trees, $\mathcal{B}=Z+\mathcal{B}\times\mathcal{B}$. The CPU time required to calculate the pre-computed tables in our unranking is 380 milliseconds.

Lexicographic order					Boustrophedonic order						
S	Size	$ au_T$	τ_T'	ρ			Size	τ_T	τ_T'	ρ	
	25	1.12	3.24	0.34			25	1.06	2.57	0.41	-
	50	2.60	9.33	0.27			50	2.71	5.75	0.47	
	75	4.81	16.61	0.28			75	4.19	9.74	0.43	
	100	7.18	25.72	0.27			100	5.95	13.76	0.43	
	125	9.65	34.27	0.28			125	8.01	17.66	0.45	
	150	12.28	42.57	0.28			150	9.80	22.43	0.43	
	175	15.42	54.44	0.28			175	11.49	27.76	0.41	
	200	18.83	67.08	0.28			200	13.56	32.67	0.41	

TABLE II

Average CPU time (in milliseconds) for Motzkin Trees, \mathcal{M} =Z+Z \times \mathcal{M} +Z \times \mathcal{M} \times \mathcal{M} . The CPU time required to calculate the pre-computed tables in our unranking is 690 milliseconds.

Notice that, in general, pl::count(size) returns the number of objects of pl with size size.

The selected collection for our experiments are two classical classes: binary trees ($\mathcal{B} = Z + \mathcal{B} \times \mathcal{B}$) and, unary-binary trees or Motzkin trees ($\mathcal{M} = Z + Z \times \mathcal{M} + Z \times \mathcal{M} \times \mathcal{M}$).

Essentially, we have used two techniques in our implementation. First, we have appropriately used the command option remember. Second, we have also used some precomputed tables to store the counting of each considered class and size. The access to the indices of such tables is notably faster than the access to the command count. Tables I and II show the improvement of the average CPU time (in milliseconds). We have pre-computed the counting tables and, afterwards, we have generated 10000 random objects of the considered class and size. τ_T is our average time required to unrank a random rank of the considered class and size, τ_T' is MUPAD-COMBINAT average time required to unrank a random rank of the considered class and size, and ρ is the ratio τ_T/τ'_T . For any case, it looks as the improvements tend to be stable when the size n increase. For lexicographic binary trees it approaches to $\rho = 0.33$, for Boustrophedonic binary trees it approaches to $\rho = 0.40$, for lexicographic Motzkin trees it approaches to $\rho = 0.28$, and for Boustrophedonic Motzkin trees it approaches to $\rho = 0.41$. Thus, all results are satisfactorily better. Note that even the pre-computed tables require some CPU time, the average CPU time (when the number of generated objects increase) of our implementation substantially improves the previous one. We have meaningfully improved the average CPU time required to generate a random unranking: In lexicographic order, our implementation spends about 30% of the CPU time of the previous version; and, in boustrophedonic order, it spends about 40% of the CPU time of the previous version.

Sequences are done from unions and products, thus the timing improvements have similar advantages.

On the other hand, we have also done some experiments with classes that involve sets or cycles, for instance, we have considered the so-called *functional graphs* defined by \mathcal{F} = Set(Cycle(\mathcal{T})) with \mathcal{T} =×(Z, Set(\mathcal{T})). In such cases, our implementation spends between 60% and 80% of the CPU time of the previous version.

A. Future implementation

The current implementation in MUPAD-COMBINAT does not consider all admissible combinatorial operators as well as restricted cardinalities in sets or powersets. We have added some of these operators in our implementation. In particular, we have considered admissible operators like

$$\varphi(\mathcal{B}, \operatorname{card} \tau k)$$

where $\varphi \in \{\text{Seq}, \text{Set}, \text{PowerSet}\}, \tau \in \{\leq | =\geq\} \text{ and } k \in \mathbb{N}.$

By the way, the required average CPU time for the implemented operators is clearly competitive. Now one of the following open problems is to develop the corresponding precomputed tables of counting for powersets and sets (see the described isomorphisms for powersets and sets in Section II).

IV. CONCLUSIONS AND FUTURE WORK

We have implemented in MUPAD the unranking applied to some basic unlabeled admissible combinatorial operators: disjoint unions, Cartesian products, and sequences. We are now working on the implementation for (unlabeled) powersets and sets (with and without restricted cardinalities).

Our implementation is making two main improvements for the unranking of unlabeled admissible classes in front of the implementation in the package MUPAD-COMBINAT. First, we have significantly reduced the average CPU time required to generate a random unranking. Second, we are programming more unlabeled admissible combinatorial operators (powersets and sets with and without restricted cardinalities).

Future work is to implement even more unlabeled admissible combinatorial operators (substitution, the open problem for unlabeled cycles, the union among non-disjoint classes, the intersection among classes, etc.).

Another line of research is to study similar operators but from the labeled point of view, that is, to consider that the nodes of the combinatorial structures of size n can be distinguished by labels from 1 to n.

The ranking, exhaustive and random generation should also be implemented [24], [12], [16].

On the other hand, these algorithms are also able to develop some structures useful to generate molecules [4], [5] applied to chemistry and influence graphs [17] applied to game theory and social networks, among other topics [22].

Finally, to what we know, it is still open to study the unranking, ranking and exhaustive generation of combinatorial structures from the viewpoint of genetic algorithms [15], [23]. Thus, it should be very interesting to establish some genetic algorithms to solve these problems.

ACKNOWLEDGMENTS

We thank anonymous referees for their useful comments and suggestions that helped us to improve the contents of the paper.

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