Aspects of Enumeration and Generation with a String Automata Representation *

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Abstract

In general, the representation of combinatorial objects is decisive for the feasibility of several enumerative tasks. In this work, we show how a (unique) string representation for (complete) initially-connected deterministic automata (ICDFA's) with nstates over an alphabet of k symbols can be used for counting, exact enumeration, sampling and optimal coding, not only the set of ICDFA's but, to some extent, the set of regular languages. An exact generation algorithm can be used to partition the set of ICDFA's in order to parallelize the counting of minimal automata (and thus of regular languages). We present also a uniform random generator for ICDFA's that uses a table of pre-calculated values. Based on the same table it is also possible to obtain an optimal coding for ICDFA's.

Keyword: regular languages, initially-connected deterministic finite automata, enumeration, random generation

1 Introduction

In general, the representation of combinatorial objects is decisive for the feasibility of several enumerative tasks. In this work, we show how a (unique) string representation for (complete) initially-connected deterministic automata (ICDFA's) with n states over an alphabet of k symbols can be used for counting, exact enumeration, sampling and optimal coding, not only the set of ICDFA's but, to some extent, the set of regular languages. The key fact is that string representations are characterized by a set of rules that allow an exact and ordered generation of all its elements. An exact generation algorithm can be used to partition the set of ICDFA's in order to parallelize the counting of minimal automata, and thus of regular languages. With the same set of rules it is possible to design a uniform random generator for ICDFA's that uses a table of pre-calculated values (as usual in combinatorial decomposition approaches). Based on the same table it is also possible to obtain an optimal coding for ICDFA's (with or without final states).

^{*}Work partially funded by Fundação para a Ciência e Tecnologia (FCT) and Program POSI.

In the next section, some definitions and notation are introduced. In Section 3 we review the string representation of non-isomorphic ICDFA₀'s (i.e., ICDFA's without final states), and how it can be used to generate and enumerate all ICDFA's. We also relate those methods to the ones presented by Champarnaud and Paranthöen in [CP05], by giving a new enumerative result. In Section 4, we briefly describe the implementation of a generator algorithm for ICDFA₀'s. Section 5 presents the methods for parallelizing the counting of languages by slicing the universe of ICDFA₀'s is described in Section 6 along with some experimental results and statistical tests. Using the recurrence formulae defined in Section 6, we show in Section 7 how we can associate an integer with an ICDFA₀'s and vice-versa. Section 8 concludes with final remarks.

2 Preliminaries

Given two integers m < n we represent the set $\{i \in \mathbb{N} \mid m \leq i \leq n\}$ by [m,n]. A deterministic finite automaton (DFA) \mathcal{A} is a quintuple $(Q, \Sigma, \delta, q_0, F)$ where Q is a finite set of states, Σ the alphabet, i.e., a non-empty finite set of symbols, $\delta: Q \times \Sigma \to Q$ is the transition function, q_0 the initial state and $F \subseteq Q$ the set of final states. The size of the automaton is given by |Q|. We assume that the transition function is total, so we consider only *complete* DFA's. As we are not interested in the labels of the states, we can represent them by an integer $i \in [0, |Q| - 1]$. The transition function δ extends naturally to Σ^* . A DFA is *initially-connected*¹ (ICDFA) if for each state $q \in Q$ there exists a string $x \in \Sigma^*$ such that $\delta(q_0, x) = q$. The structure of an automaton (Q, Σ, δ, q_0) denotes a DFA without its final state information and is referred to as a DFA_{\emptyset} . For each structure, there will be 2^n DFA's, if |Q| = n. We denote by ICDFA_{\emptyset} the structure of an ICDFA. Two DFA's $\mathcal{A} = (Q, \Sigma, \delta, q_0, F)$ and $\mathcal{A}' = (Q', \Sigma, \delta', q'_0, F')$ are called *isomorphic* (by states) if there exists a bijection $f: Q \to Q'$ such that $f(q_0) = q'_0$ and for all $\sigma \in \Sigma$ and $q \in Q$, $f(\delta(q, \sigma)) = \delta'(f(q), \sigma)$. Furthermore, for all $q \in Q$, $q \in F$ if and only if $f(q) \in F'$. The language accepted by a DFA \mathcal{A} is $L(\mathcal{A}) = \{x \in \Sigma^* \mid \delta(q_0, x) \in F\}$. Two DFA are *equivalent* if they accept the same language. Obviously, two isomorphic automata are equivalent, but two non-isomorphic automata may also be equivalent. A DFA \mathcal{A} is minimal if there is no DFA \mathcal{A}' with fewer states equivalent to \mathcal{A} . Trivially a minimal DFA is an ICDFA. Minimal DFA's are unique up to isomorphism. Domaratzki et al. [DKS02] gave some asymptotic estimates and explicit computations of the number of distinct languages accepted by finite automata with n states over an alphabet of ksymbols. Given n and k, they denoted by $f_k(n)$ the number of pairwise non-isomorphic minimal DFA's and by $g_k(n)$ the number of distinct languages accepted by DFA's, where $g_k(n) = \sum_{i=1}^n f_k(i).$

3 Strings for ICDFA's

Reis et al. [RMA05] presented a unique string representation for non-isomorphic ICDFA $_{\emptyset}$'s. In this section, we briefly review this representation and how it can be used to generate

¹Also called *accessible*.

and enumerate all ICDFAs. We also give a new enumerative result and relate this representation to the one presented by Champarnaud and Paranthöen in [CP05].

Given a complete DFA_{\emptyset} (Q, Σ, δ, q_0) with |Q| = n and $|\Sigma| = k$, consider a total order < over Σ . We can define a canonical order over the set of the states by exploring the automaton in a breadth-first way choosing at each node the outgoing edges in the order considered for Σ . If we restrict this representation to $ICDFA_{\emptyset}$'s, then this representation is unique and defines an order over the set of its states. For instance, consider the following $ICDFA_{\emptyset}$ and consider the alphabetic order in $\{a, b, c\}$.



The states ordering is A,C,B,D and [1, 2, 0, 2, 3, 0, 3, 0, 2, 1, 3, 2] is its string representation. Formally, let $\Sigma = \{\sigma_i \mid i \in [0, k-1]\}$, with $\sigma_0 < \sigma_1 < \cdots < \sigma_{k-1}$. Given an ICDFA $_{\emptyset}$ (Q, Σ, δ, q_0) with |Q| = n, the representing string is of the form $(s_i)_{i \in [0, kn-1]}$ with $s_i \in [0, n-1]$ and $s_i = \delta(\lfloor i/k \rfloor, \sigma_{i \mod k})$.

Let $(s_i)_{i \in [0,kn-1]}$ with $s_i \in [0, n-1]$ be a string satisfying the following conditions:

$$(\forall m \in [2, n-1])(\forall i \in [0, kn-1])(s_i = m \Rightarrow (\exists j \in [0, i-1]) s_j = m-1).$$
 (R1)

$$(\forall m \in [1, n-1]) (\exists j \in [0, km-1]) s_j = m.$$
 (**R2**)

In [RMA05] the following theorem was proved.

Theorem 1 There is a one-to-one mapping between $(s_i)_{i \in [0,kn-1]}$ with $s_i \in [0, n-1]$ satisfying rules **R1** and **R2**, and the non-isomorphic ICDFA_{\emptyset}'s with n states, over an alphabet Σ of size k.

We note that this string representation can be extended to non-complete ICDFA $_{\emptyset}$'s, by representing all missing transitions with the value -1. In this case, rules **R1** and **R2** remain valid, and we can assume that the transitions from this state are into itself. However for enumeration and generation purposes we do not consider non-complete ICDFA $_{\emptyset}$'s.

In order to have an algorithm for the enumeration and generation of ICDFA_{\emptyset} 's, instead of rules **R1** and **R2** an alternative set of rules were used. For n = 1 there is only one (non-isomorphic) ICDFA_{\emptyset} for each $k \geq 1$, so we assume in the following that n > 1. In a string representing an ICDFA_{\emptyset} , let $(f_j)_{j \in [1,n-1]}$ be the sequence of indexes of the first occurrence of each state label j. For explanation purposes, we call those indexes *flags*. It is easy to see that (**R1**) and (**R2**) correspond respectively to (**G1**) and (**G2**):

$$(\forall j \in [2, n-1])(f_j > f_{j-1});$$
 (G1)

$$(\forall m \in [1, n-1]) \ (f_m < km). \tag{G2}$$

This means that $f_1 \in [0, k-1]$, and $f_{j-1} < f_j < kj$ for $j \in [2, n-1]$. We begin by counting the number of sequences of flags allowed.

Theorem 2 Given k and n, the number of sequences $(f_j)_{j \in [1,n-1]}$, $F_{k,n}$, is given by

$$F_{k,n} = \sum_{f_1=0}^{k-1} \sum_{f_2=f_1+1}^{2k-1} \cdots \sum_{f_{n-1}=f_{n-2}+1}^{k(n-1)-1} 1 = \binom{kn}{n} \frac{1}{(k-1)n+1} = C_n^{(k)};$$

where $C_n^{(k)}$ are the (generalised) Fuss-Catalan numbers.

Proof 1 The first equality follows directly from the definition of the $(f_j)_{j\in[1,n-1]}$. For the second, note that $C_n^{(k)}$ enumerates k-ary trees with n internal nodes, \mathcal{T}_n^k (see for instance [SF96]). In particular, for k = 2, C_n^2 are exactly the Catalan numbers that count binary trees with n internal nodes. This sequence appears in Sloane [Slo03] as **A00108** and for k = 3 and k = 4 as **A001764** and **A002293** sequences, respectively. So it suffices to give a bijection between these trees and the sequences of flags. Recall that a k-ary tree is an external node or an internal node attached to an ordered sequence of k, k-ary sub-trees.



Figure 1: Two 3-ary trees with 4 internal nodes and the correspondent sequence of flags.

Let \mathcal{T}_n^k be a k-ary tree and let < be a total order over Σ . For each internal node i of \mathcal{T}_n^k its outgoing edges can be ordered left-to-right and attached a unique symbol of Σ according to <. Considering a breadth-first, left-to-right, traversal of the tree and ignoring the root node (that is considered the 0-th internal node), we can represent \mathcal{T}_n^k , uniquely, by a bitmap where a 0 represents an external node and a 1 represents an internal node. As the number of external nodes are (k-1)n+1, the length of the bitmap is kn. Moreover the j+1-th block of k bits corresponds to the children of the j-th internal node visited, for $j \in [0, n - 1]$ 1]. For example, the bitmaps of the trees in Figure 1 are [0,0,1,0,0,1,0,0,1,0,0,0] and [0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0], respectively. The positions of the 1's in the bitmaps correspond to a sequence of flags, $(f_i)_{i \in [1,n-1]}$, i.e., f_i corresponds to the number of nodes visited before the *i*-th internal node (excluding the root node). It is obvious that $(f_i)_{i \in [1,n-1]}$ verify G1. For G2, note that for the each internal node the outdegree of the previous internal nodes is k. Conversely, given a sequence of flags $(f_j)_{j \in [1,n-1]}$, we construct the bitmap such that $b_{f_i}=1$ for $i \in [1, n-1]$ and $b_j=0$ for the remaining values, for $j \in [0, kn-1]$. As above, for the representation of the j + 1-th internal node, $|f_j/k|$ gives the parent and $f_j \mod k$ gives its position between its siblings (in breadth-first, left-to-right traversal).

To generate all the ICDFA_{\emptyset}'s, for each allowed sequence of flags $(f_j)_{j \in [1,n-1]}$, all the

remaining symbols s_i can be generated according to the following rules:

$$i < f_1 \Rightarrow s_i = 0;$$
 (G3)

$$(\forall j \in [1, n-2])(f_j < i < f_{j+1} \Rightarrow s_i \in [0, j]);$$
 (G4)

 $i > f_{n-1} \Rightarrow s_i \in [0, n-1].$ (G5)

In [RMA05] a simple combinatorial argument was given to show that

Theorem 3 The number of strings $(s_i)_{i \in [0,kn-1]}$ representing ICDFA_{\emptyset}'s with n states over an alphabet of k symbols is given by

$$B_{k,n} = \sum_{f_1=0}^{k-1} \sum_{f_2=f_1+1}^{2k-1} \sum_{f_3=f_2+1}^{3k-1} \cdots \sum_{f_{n-1}=f_{n-2}+1}^{k(n-1)-1} \prod_{i=2}^n i^{f_i-f_{i-1}-1};$$
(1)

where $f_n = kn$.

In Section 6 we give other recursive definition that is more adequate for tabulation.

3.1 Analysis of the Champarnaud et al. Method

Champarnaud and Paranthoën in [CP05, Par04], generalizing work of Nicaud [Nic00] presented a method to generate and enumerate ICDFA₀'s, although not giving an explicit and compact representation for them, as the string representation used here. An order < over Σ^* is a *prefix order* if $(\forall x \in \Sigma^*)(\forall \sigma \in \Sigma)x < x\sigma$. Let \mathcal{A} be an ICDFA₀ over Σ with k symbols and n states. Given a prefix order in Σ^* , each automaton state is ordered according to the first word $x \in \Sigma^*$ that reaches it in a simple path from the initial state. The sets of this words $\{\mathcal{P}\}$ are in bijection with k-ary trees with n internal nodes, and therefore to the set of sequences of flags, in our representation². Then it is possible to obtain a valid ICDFA₀ by adding other transitions in a way that preserves the previous state labelling. For the generation of the sets \mathcal{P} it is used another set of objects that are in bijection with k-ary trees with n internal nodes and are called generalised tuples. The number of ICDFA₀'s is computed using recursive formulae associated with generalized tuples, akin the ones we present in Section 6.

4 Generating ICDFA $_{\emptyset}$'s

In this section, we present a method to generate all ICDFA_{\emptyset}'s, given k and n. We start with an initial string, and then consecutively iterate over all allowed strings until the last one is reached. The main procedure is the one that given a string returns the *next* legal one. For each k and n, the first ICDFA_{\emptyset} is represented by the string $0^{k-1}10^{k-1} \dots (n-1)0^k$ and the last is represented by $12 \dots (n-1)(n-1)^{(k-1)n+1}$. According to the rules **G1- G5**, we first generate a sequence of flags, and then, for each one, the set of strings representing the ICDFA_{\emptyset}'s in lexicographic order. The algorithm to generate the next sequence of flags is the following, where the initial sequence of flags is $(ki - 1)_{i \in [1,n-1]}$:

²Indeed our order on the states induces a prefix order in Σ^{\star} .

To generate a new sequence, we must call nextflags(n-1). Given the rules G1 and G2 the correctness of the algorithm is easily proved. When a new sequence of flags is generated, the first ICDFA₀ is represented by a string with 0s in all other positions (i.e., the lower bounds in rules G3–G5). The following strings, with the same sequence of flags, are computed lexicographically using the procedure **nexticdfa**, called with a = n - 1 and b = k - 1:

Note that the last string for each sequence of flags has the value $s_l = j$ for $l \in [f_j + 1, f_{j+1} - 1]$, with $j \in [1, n - 1]$. The time complexity of the generator is linear in the number of automata. As an example, for k = 2 and n = 9 it took about 12 hours to generate all the 705068085303 ICDFA $_{\emptyset}$'s, using a AMD Athlon at 2.5GHz. Finally, for the generation of ICDFA's we only need to add to the string representation of an ICDFA $_{\emptyset}$, a string of n 0's and 1's, correspondent to one of the 2^n possible choices of final states.

5 Counting Regular Languages (in Slices)

To obtain the number of languages accepted by DFA's with n states over an alphabet of k symbols, we can generate all ICDFA's, determine which of them are minimal $(f_k(n))$ and calculate the value of $g_k(n)$. Obviously, this is in general an intractable procedure. But for small values of n and k some experiments can take place. We must have an efficient implementation of a minimization algorithm, not because of the size of each automaton but because the number of automata we need to cope with. For that we implemented Hopcroft's minimization algorithm [Hop71], using efficient set representations. For very small values of n and k (n + k < 16) we represented sets as bitmaps and for larger values, AVL trees [Av] were used.

The problem can be parallelized providing that the space search can be safely par-

	n	ICDFA∅	ICDFA	Minimal $(f_k(n))$	Minimal %	Time (s)
k = 2	2	12	48	24	50%	0
	3	216	1728	1028	59%	0.018
	4	5248	83968	56014	66%	0.99
	5	160675	5141600	3705306	72%	79.12
	6	5931540	379618560	286717796	75%	8700
	7	256182290	32791333120	25493886852	77%	1237313
k = 3	2	56	224	112	50%	0.002
	3	7965	63720	41928	65%	0.7
	4	2128064	34049024	26617614	78%	494.72
	5	914929500	29277744000	25184560134	86%	652703
k = 4	2	240	960	480	50%	0.01
	3	243000	1944000	1352732	69%	23.5
	4	642959360	10287349760	7756763336	75%	184808
k = 5	2	992	3968	1984	50%	0.041
	3	6903873	55230984	36818904	66%	756.2

Table 1: Performance and number of minimal automata.

titioned. Using the method presented in Section 4, we can easily generate *slices* of $ICDFA_{\emptyset}$'s and feed them to the minimization algorithm. A *slice* is a sequence of $ICDFA_{\emptyset}$'s and is defined by a pair (start, last), where start is the first automaton in the sequence and *last* is the last one. If we have a set of CPUs available, each one can receive a slice, generate all ICDFA_{\emptyset}'s (in that slice), generate all the necessary ICDFA's and feed them to the minimization algorithm. For the generation of ICDFA's, we used the observation by Domaratzki *et al.* [DKS02], that is enough to test 2^{n-1} sets of final states, using the fact that a DFA is minimal *iff* its complementary automaton is minimal too. In this way, we can safely divide the search space and distribute each slice to a different CPU. Note that this approach relies in the assumption that we have a much more efficient way to partition the search space than to actually perform the search (in this case a minimization algorithm). The task of creating the slices can be taken by a central process that successively generates the next slice and at the end assembles all the results. The server can run interactively with its *slaves*, or it can generate all the slices at once to be used later. The server generates a *slice* using the generator algorithm presented in Section 4. For this experiment we used two approaches. We developed a simple slave management system called Hydra — based on Python threads, that was composed by a server and a variable set of *slaves*. In this case, the slaves can be any computer³. For each slice a process was executed via ssh, and the result was returned to the server. Another approach was to use a computer grid, in particular 24 AMD Opteron 250 2.4GHz (dual core).

5.1 Experimental results

In Table 1, we summarise some experimental results. Most of the values for k = 2 and k = 3, were already given by Domaratzki *et al.* in [DKS02] and the new results are in bold in the table. For k = 2, n = 8 we have divided the universe of ICDFA₀'s in 254 slices and the estimated CPU time for each one to be processed is 11 days.

Moreover, the slicing process can give new insights about the distribution of minimal automata. Figure 2 presents two examples of the values obtained for the rate of minimal DFA's. For n = 7 and k = 2 we give the percentage of minimal automata for each of the

³We used all the normal desktop computers of our colleagues in the CS Department.

257 slices we had used to divide the search space (32791333120 ICDFA_{\emptyset}'s). Each slice had about 100000 ICDFA_{\emptyset}'s, and so 128000000 ICDFA's, and it took about 78 minutes to conclude the process. The whole set of automata was processed in 12 hours of real time of a CPU grid, that corresponds to 344 hours of CPU time.



Figure 2: Rate of minimal DFA's with (k = 3, n = 5) for 915 slices and with (k = 2, n = 7) for 257 slices.

6 A Uniform Random Generator

The ICDFA $_{\emptyset}$ representation presented (Section 3) permits an easy random generation for ICDFAs, and thus for DFAs. To randomly generate a DFA for a given n and k, it is only necessary to: (i) randomly generate a valid sequence of flags $(f_i)_{i \in [1,n-1]}$ according to G1 and G2; (ii) followed by the random generation of the rest of the nk elements of the string following G3–G5 rules; (iii) and finally the random generation of the set of final states. The uniformity issue for steps (ii) and (iii) is quite straightforward. For step (iii) it is just necessary to use a uniform random integer generator for a value $i \in [0, 2^n]$. It is enough, for step (ii) the repeated use of the same number generator for values in the range [0, i] for $0 \le i < n$ according to rules **G3–G5**. Step (i) is the only step that needs special care. Consider the case n = 5 and k = 2. Because of rule **R1** flag f_1 can only be on positions 0 or 1. But there are 140450 ICDFA_{\emptyset}'s with f_1 in the first case and only 20225 in the second. Thus the random generation of flags, to be uniform, must take this into account by making the first case more probable than the second. We can generate a random ICDFA_{\emptyset} generating its representing string from left to right. Supposing that flag f_{m-1} is already placed at position i and all the symbols to its left are generated, i.e., the prefix $s_0 s_1 \cdots s_i$ is already defined, then the process can be described by:

$$r = \operatorname{random}(1, \sum_{j=i+1}^{mk-1} N_{m,j})$$

for $j = i+1$ to $mk-1$:
if $r \in \left[\sum_{l=i}^{j-1} N_{m,l}, \sum_{l=i}^{j} N_{m,l}\right]$ then return i

where random(a,b) is an uniform random generated integer between a and b, and $N_{m,j}$ is the number of ICDFA₀s with prefix $s_0s_1\cdots s_i$ with the first occurrence of symbol m in position j, making $N_{m,i} = 0$ to simplify the expressions. The values for $N_{m,j}$ could be obtained from expressions similar to Equation (1), and used in a program. But the program would have a exponential time complexity. By expressing $N_{m,j}$ in a recursive form, we have, given k and n

$$N_{n-1,j} = n^{nk-1-j} \qquad \text{with } j \in [n-2, (n-1)k-1];$$

$$N_{m,j} = \sum_{i=0}^{(m+1)k-j-2} (m+1)^i N_{m+1,j+i+1} \qquad \text{with } m \in [1, n-2], \qquad (2)$$

$$j \in [m-1, mk-1].$$

This evidences the fact that we keep repeating the same computations with very small variations, and thus, if we use some kind of tabulation of this values $(N_{m,j})$, with the obvious price of memory space, we can create a version of a uniform random generator, that apart of a constant overhead used for tabulation of the function refered, has a complexity of $\mathcal{O}(n^3k)\mathcal{O}(\text{random})$. The algorithm is described by the following:

for
$$i = (n-1)k-1$$
 downto $n-2$:
 $N_{n-1,i} = n^{nk-1-i}$
for $m = n-2$ downto 1:
 $N_{m,mk+1} = \sum_{i=0}^{k-1} (m+1)^i N_{m+1,mk+i}$
for $i = mk-2$ downto $m-1$:
 $N_{m,i} = (m+1)N_{m,i+1} + N_{m+1,i+1}$
for $i = 1$ to $n-1$:
 $f = \text{generateflag}(i, g+1)$
for $j = g+1$ to $f-1$:
print random $(0, i-1)$
print i
 $g = f$

This means that with the same AMD Athlon 64 at 2.5GHz, using a C implementation with libgmp [GMP] the times reported in Table 2 were observed. It is possible, without

	k = 2	k = 3	k = 5	k = 10	k = 15
n = 10	0.10s	0.16s	0.29s	0.61s	1.30s
n = 20	0.31s	0.49s	1.26s	4.90s	12.24s
n = 30	0.54s	1.37s	3.19s	19.91s	62.12
n = 50	1.61s	3.86s	17.58s	2.22m	947.71s
n = 75	3.96s	12.98s	76.69s	700.20s	2459.34s
n = 100	7.92s	36.33s	215.32s	$2219.04 \mathrm{s}$	8091.30s

Table 2: Times for the random generation of 10000 automata.

unreasonable amounts of RAM to generate random automata for unusually large values of n and k. For example, with n = 1000 and k = 2 the memory necessary is less than 450MB. The amount of memory used is so large not only because of the amount of tabulated values, but because the size of the values is enormous. To understand that, it is enough to note that the total number of ICDFA₀'s for these values of n and k is greater than 10^{3350} , and the values tabulated are only bounded by this number.

6.1 Statistical test of the random generator

Although the method used to generate random automata is, by its own construction, uniform, we used χ^2 test to evaluate the random generation quality. The universe of ICDFA₀'s with 6 states and 2 symbols has a total size of 5931540. This size is large enough for a test with some significance and it is still reasonable, both in time and space, to perform the test. We generated three different sets of 3000000 ICDFA₀'s and perform the test in each one. Because of the size of the data, we could not find any tabulated values for acceptance, and thus the following formula was used with v = 30000000 - 1 and x_p being the significance level (1% in this case):

$$v + 2\sqrt{vx_p} + \frac{3}{4}x_p^2 - \frac{2}{3}.$$

The size of the data sets and the repetition of the test for three times, is the recommended procedure by Knuth ([Knu81], pages 35–39). For the three experiments the values obtained were, respectively, 5933268.92456, 5925676.75108 and 5935733.28172, that are all smaller than the acceptance limit, that for this case was 5938980.75468.

7 Enumeration of $ICDFA_{\emptyset}$'s

In this section, we show how, given a string representation of an ICDFA_{\emptyset}'s of size *n* over an alphabet of *k* symbols, we can compute its number in the generation order (described in Section 4) and vice-versa, i.e., given a number less than $B_{k,n}$, we obtain the corresponding ICDFA_{\emptyset}. This provides an optimal encoding for ICDFA_{\emptyset}'s, as defined by M. Lothaire in [Lot05], Chapter 9. This bijection is accomplished using the tables defined in Section 6 that correspond to partial sums of Equation (1).

Theorem 4
$$B_{k,n} = \sum_{l=0}^{k-1} N_{1,l}.$$

Proof 2 The result follows easily by expanding $N_{m,j}$ using Equations (2) and Equation (1).

7.1 From ICDFA $_{\emptyset}$'s to Integers

Let $(s_i)_{i \in [0,kn-1]}$ be an ICDFA $_{\emptyset}$'s string representation, and let $(f_j)_{j \in [1,n-1]}$ be the corresponding sequence of flags. From the sequence of flags we obtain the following number, n_f ,

$$n_f = \sum_{i=1}^{n-1} \sum_{j=f_i+1}^{ik-1} (i^{j-f_i} N_{i,j} (\prod_{m=1}^{i-1} (m^{f_{m+1}-f_m-1}))$$
(3)

which is the number of the first ICDFA $_{\emptyset}$ with flags $(f_j)_{j \in [1,n-1]}$. Now we must add the information provided by the rest of the elements of the string $(s_i)_{i \in [0,kn-1]}$:

$$n_r = \sum_{j=1}^{n-1} \left(\sum_{l=f_j+1}^{f_{j+1}-1} s_l(j+1)^{f_{j+1}-1-l} \left(\prod_{m=j+1}^{n-1} (m+1)^{f_{m+1}-f_m-1} \right) \right)$$
(4)

And the corresponding number is $n_s = n_f + n_r$.

7.2 From Integers to $ICDFA_{\emptyset}$'s

Given an integer $0 \le m < B_{k,n}$ a string representing uniquely an ICDFA $_{\emptyset}$ can be obtained using a method inverse of the one in the last section. The flags $(f_j)_{j \in [1,n-1]}$ are generated from right-to-left, by successive subtractions. The rest of the string $(s_i)_{i \in [0,kn-1]}$ is generate considering the remainders of integer divisions. The algorithms are the following:

$$\begin{split} s &= 1 \\ \text{for } i &= 1 \text{ to } n-1: \\ j &= i * k - 1 \\ p &= i^{j-f_{i-1}-1} \\ \text{while } j &> = i-1 \text{ and } m \geq p * s * N_{i,j}: \\ m &= m - N_{i,j} * p * s \\ j &= j-1 \\ p &= p/i \\ s &= s * i^{j-f_{i-1}-1} \\ f_i &= j \end{split}$$
 $\begin{aligned} & i &= k * n-1 \\ j &= n-1 \\ \text{while } m > 0 \text{ and } j > 0: \\ \text{while } m > 0 \text{ and } i > f_j: \\ s_i &= m \mod (j+1) \\ m &= m \div (j+1) \\ i &= i-1 \\ j &= j-1 \end{aligned}$

8 Final Remarks

The methods here presented were implemented and tested to obtain both exact and approximate values for the density of minimal automata. Champarnaud *et al.* in [CP05], checked a conjecture of Nicaud that for k = 2 the number of minimal ICDFA's is about 80% of the total, by sampling automata with 100 states (for all possible number of final states). Our results also corroborate that conjecture, being the exact values for some small values of n and samples for greater values. In particular, for k = 2 and n = 100 we obtained the same results as Champarnaud *et al.*. It seems that for k > 2 almost all ICDFA's are minimal. For k = 3, 5 and n = 100 that was also checked by Champarnaud *et al.*. For a confidence interval of 99% and significance level of 1% the following table presents the percentages of minimal ICDFA's for several values of k and n, and each possible number of final states.

$k \setminus n$	5	6	7	8	9	10	20	40	80	160
3	85.8%	90.8%	93.3%	95.0%	96.1%	96.7%	98.7%	99.4%	99.7%	99.8%
5	93.0%	96.5%	98.2%	99.1%	99.5%	99.8%	100.0%	100.0%	100.0%	100.0%
7	93.7%	96.8%	98.4%	99.2%	99.6%	99.8%	100.0%	100.0%	100.0%	-
9	93.7%	96.9%	98.4%	99.2%	99.6%	99.8%	100.0%	100.0%	-	-
11	93.8%	96.9%	98.4%	99.2%	99.6%	99.8%	100.0%	100.0%	-	-
13	93.7%	96.9%	98.4%	99.2%	99.6%	99.8%	100.0%	100.0%	-	-

A web interface to the random generator can be found in the **FAdo** project web page [Fad]. Bassino and Nicaud in [BN] presented also a random generator of ICDFA's based on Boltzmann Samplers, recently introduced by Duchon *et al.* [DFLS04]. However the sampler is uniform for partitions of a set with kn elements into n nonempty subsets (not for the universe of automata). These partitions are related with string representations that verify only rule **R1**. Based on the work here presented, it would be interesting to study a better approximation, that would satisfy rule **R2**.

9 Acknowledgments

We thank the anonymous referees for their comments that helped to improve this paper.

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