

# Deriving a Fast Inverse of the Generalized Cantor N-tupling Bijection

Paul Tarau

Dept. of Computer Science and Engineering  
University of North Texas, Denton, Texas, USA  
tarau@cs.unt.edu

---

## Abstract

We attack an interesting open problem (an efficient algorithm to invert the generalized Cantor N-tupling bijection) and solve it through a sequence of equivalence preserving transformations of logic programs, that take advantage of unique strengths of this programming paradigm. An extension to set and multiset tuple encodings, as well as a simple application to a “fair-search” mechanism illustrate practical uses of our algorithms.

The code in the paper (a literate Prolog program, tested with SWI-Prolog and Lean Prolog) is available at <http://logic.cse.unt.edu/tarau/research/2012/pcantor.pl>.

**1998 ACM Subject Classification** F.4.1 Mathematical Logic, Logic Programming

**Keywords and phrases** generalized Cantor  $n$ -tupling bijection, bijective data type transformations, combinatorial number system, solving combinatorial problems in Prolog, optimization through program transformation, logic programming and software engineering

**Digital Object Identifier** 10.4230/LIPIcs.ICLP.2012.312

## 1 Introduction

It is by no means a secret that logic programming is an ideal paradigm for solving combinatorial problems. Built-in backtracking, unification and availability of constraint solvers facilitates quick prototyping for problems involving search or generation of combinatorial objects. It also provides an easy path from executable specification to optimal implementation through a well-understood set of program transformations. From a software engineering perspective, problem solving with help of logic programming tools is a natural fit to *agile development* practices as it encourages a fast moving iterative process consisting of incremental refinements.

This paper reports on tackling a somewhat atypical problem solving instance: finding a fast inverse of a generalization of Cantor’s pairing bijection to  $n$ -tuples. This generalization is mentioned in two relatively recent papers [2, 7] with a possible attribution in [2] to Skolem as a first reference.

The formula, given in [2] p.4, looks as follows:

$$K_n(x_1, \dots, x_n) = \binom{n-1+x_1+\dots+x_n}{n} + \dots + \binom{k-1+x_1+\dots+x_k}{k} + \dots + \binom{1+x_1+x_2}{2} + \binom{x_1}{1}$$

where  $\binom{n}{k}$  represents the number of subsets of  $k$  elements of a set of  $n$  elements and  $K_n(x_1, \dots, x_n)$  denotes the natural number associated to the tuple  $x_1, \dots, x_n$ . So the problem of inverting it means finding a solution of the *Diophantine equation*

$$\binom{x_1}{1} + \binom{1+x_1+x_2}{2} + \dots + \binom{n-1+x_1+\dots+x_n}{n} = z \quad (1)$$



© Paul Tarau;

licensed under Creative Commons License ND

Technical Communications of the 28th International Conference on Logic Programming (ICLP’12).

Editors: A. Dovier and V. Santos Costa; pp. 312–322

Leibniz International Proceedings in Informatics



LIPIcs Schloss Dagstuhl – Leibniz-Zentrum für Informatik, Dagstuhl Publishing, Germany

and proving that it is unique. Unfortunately, despite extensive literature search, we have not found any attempt to devise an algorithm that computes the inverse of the function  $K_n$ , so we had to accept the fact that we were looking at an *open problem* with possibly interesting implications, given that for  $n = 2$  it reduces to Cantor's pairing function that has been used in hundreds of papers on foundations of mathematics, logic, recursion theory as well as in some practical applications (dynamic  $n$ -dimensional arrays) like [11].

As an inductive proof that  $K_n$  is a bijection is given in [7] (Theorem 2.1), we know that a solution exists and is unique, so the problem reduces to computing the first solution of the Diophantine equation (1).

Unfortunately, solving an arbitrary Diophantine equation is Turing-equivalent. This is a consequence of the negative answer to Hilbert's 10-th problem, proven by Matiyasevich [8], based on earlier work by Robinson, Davis and Putnam [4, 10], and some fairly simple instances of it, like Fermat's  $\exists x, y, z > 0, \exists n \geq 3, x^n + y^n = z^n$  have waited for centuries before being solved.

On the other hand, things do not look that bad in this case, as it is easy to show that in the equation (1),  $\forall i, x_i \leq z$  holds. Therefore, an enumeration of all tuples  $x_1, \dots, x_n$  for  $0 \leq x_i \leq z$  provides an obvious but dramatically inefficient solution.

So our open problem reduces to *finding an efficient, linear or low polynomial algorithm for computing the inverse*. This paper provides a surprisingly simple solution to it in section 7, after telling the story of our incremental refinements (as well as backtracking steps) leading to it. Section 2 overviews the well-known solution for  $n = 2$ . Section 3 provides the Prolog implementation of the mapping from  $n$ -tuples to natural numbers. Section 4 describes the successive refinements of the inverse function, from its specification to a moderately useful implementation. Section 5 introduces a *list-to-set bijection* that will turn out to be helpful in "connecting the dots" to a well-known combinatorial problem that leads to our solution described in section 7 (after a small "backtracking step" shown in section 6). Section 8 extends the bijection to sets and multisets. Section 10 discusses related work and section 11 concludes the paper.

## 2 The Classic Result: Cantor's Pairing Function and its Inverse

Cantor's pairing function is a polynomial of degree 2, obtained from the generalized one for  $n = 2$ , given by the formula  $f(x_1, x_2) = x_1 + \frac{(x_1+x_2+1)(x_1+x_2)}{2}$ .

The following Prolog code implements it:

```
cantor_pair(X1,X2,P) :- P is X1 + (((X1+X2+1) * (X1+X2)) // 2).
```

Note that by composing it  $n$  times, one can obtain an  $n$ -tupling function, but unfortunately the resulting polynomial is of degree  $2^n$ , in contrast to the generalized  $n$ -tupling bijection which is a polynomial of degree  $n$ . On the other, hand, as the following Prolog code shows, the problem of finding its inverse efficiently is relatively easy. Basically, the inverse of Cantor's pairing function is obtained by solving a second degree equation while keeping in mind that solutions should be natural numbers [17].

```
cantor_unpair(P,K1,K2) :- E is 8*P+1, intSqrt(E,R), I is (R-1)//2,
    K1 is P-((I*(I+1))//2), K2 is ((I*(3+I))//2)-P.
```

We face a small bump here – Prolog's ordinary square root returning a fixed size float or double does not make sense when working with arbitrary size integers, so we need to

implement an “integer square root” of  $N$  returning the natural number that provides the largest perfect square  $\leq N$ . Fortunately, we can ensure fast convergence using Newton’s method:

```
intSqrt(0,0).
intSqrt(N,R) :- N>0, iterate(N,N,K), K2 is K*K, (K2>N -> R is K-1 ; R=K).

iterate(N,X,NewR) :- R is (X+(N//X))//2, A is abs(R-X),
(A<2 -> NewR=R ; iterate(N,R,NewR)).
```

As the following example shows, computations with larger than 64-bit operands are handled, provided that the underlying Prolog system supports arbitrary size integers.

```
?- cantor_pair(1234567890,9876543210,P),cantor_unpair(P,A,B).
P = 61728394953703703760, A = 1234567890, B = 9876543210.
```

### 3 Implementing the Generalized Cantor $n$ -tupling Bijection

Tupling/untupling functions are a natural generalization of pairing/unpairing operations. They are called *ranking/unranking* functions by combinatorialists as they map bijectively various combinatorial objects to  $\mathbb{N}$  (ranking) and back (unranking).

The natural generalization of Cantor’s pairing bijection described in [2] is introduced using geometric considerations that make it obvious that it defines a bijection  $K_n : \mathbb{N}^n \rightarrow \mathbb{N}$ . More precisely, they observe that the enumeration in  $\mathbb{N}^2$  of integer coordinate pairs laying on the anti-diagonals  $x_1 + x_2 = c$  can be lifted to points with integer coordinates laying on hyperplanes of the form  $x_1 + x_2 + \dots + x_k = c$ . The same result, using a slightly different formula is proven algebraically, by induction in [7]. We remind that the bijection  $K_n$  is defined by the formula

$$K_n(x_1, \dots, x_n) = \sum_{k=1}^n \binom{k-1+x_1+\dots+x_k}{k} \quad (2)$$

where  $\binom{n}{k}$ , also called “binomial coefficient” denotes the number of subsets of  $n$  with  $k$  elements as well as the coefficient of  $x^k$  in the expansion of the binomial  $(x+y)^n$ .

It is easy to see that the generalized Cantor  $n$ -tupling function defined by equation (2) is a polynomial of degree  $n$  in its arguments, and a conjecture, attributed in [2] to Rudolf Fueter (1923), states that it is the only one, up to a permutation of the arguments. As mentioned in section 1, as we have found out through extensive literature search, while hoping for the contrary, it was also an *open problem* to find an efficient inverse for it.

Our first step is an efficient implementation of the function  $K_n : \mathbb{N}^k \rightarrow \mathbb{N}$ . By all means, this is the easy part, just summing up a set of binomial coefficients.

#### 3.1 Binomial Coefficients, efficiently

Computing binomial coefficients efficiently is well-known

$$\binom{k}{n} = \frac{n!}{k!(n-k)!} = \frac{n(n-1)\dots(n-(k-1))}{k!} \quad (3)$$

However, we will need to make sure that we avoid unnecessary computations and reduce memory requirements by using a tail-recursive loop. After simplifying the slow formula in the first part of the equation (3) with the faster one based on falling factorial  $n(n-1)\dots(n-(k-1))$ , and performing divisions as early as possible to avoid generating excessively large intermediate results, one can derive the `binomial_loop` tail-recursive predicate:

```
binomial_loop(_,K,I,P,R) :- I>=K, !, R=P.
binomial_loop(N,K,I,P,R) :- I1 is I+1, P1 is ((N-I)*P) // I1,
    binomial_loop(N,K,I1,P1,R).
```

Note that, as a simple optimization, when  $N - K \leq K$ , the faster computation of  $\binom{N}{N-K}$  is used to reduce the number of steps in `binomial_loop`.

The resulting predicate `binomial(N,K,R)` computes  $\binom{N}{K}$  and unifies the result with `R`.

```
binomial(N,K,R) :- N<K, !, R=0.
binomial(N,K,R) :- K1 is N-K, K>K1, !, binomial_loop(N,K1,0,1,R).
binomial(N,K,R) :- binomial_loop(N,K,0,1,R).
```

### 3.2 The $\mathbb{N}^k \rightarrow \mathbb{N}$ bijection

We are ready to implement a first version of the  $\mathbb{N}^k \rightarrow \mathbb{N}$  ranking function as a tail-recursive computation using the accumulator pairs  $L1 \rightarrow L2$ , that hold the states of the length of the list processed so far, and  $S1 \rightarrow S2$ , that hold the state of the prefix sum of  $X_1, X_2, \dots, X_k$  computed so far.

```
from_cantor_tuple1(Xs,R) :- from_cantor_tuple1(Xs,0,0,0,R).

from_cantor_tuple1([],_L,_S,B,B).
from_cantor_tuple1([X|Xs],L1,S1,B1,Bn) :- L2 is L1+1, S2 is S1+X, N is S2+L1,
    binomial(N,L2,B), B2 is B1+B,
    from_cantor_tuple1(Xs,L2,S2,B2,Bn).
```

The following examples illustrate the fact that the values of the result are relatively small, independently of the length or the size of the values on the input list.

```
?- from_cantor_tuple([],N).
N = 0.
?- from_cantor_tuple([0,2012,999,0,10],N).
N = 2107259417045595.
?- from_cantor_tuple([9,8,7,6,5,4,3,2,1,0,0,1,2,3,4,5,6,7,8,9],N).
N = 3706225144988231392404.
```

## 4 Refining the Specification of the Inverse

We start with an executable specification of the inverse, seen as defining, for a given  $K$ , a bijection  $g_K : \mathbb{N} \rightarrow \mathbb{N}^K$ .

### 4.1 Enumerating, naively

The predicate `to_cantor_tuple1(K,N,Ns)` computes, for each  $K$ , the function  $g_K$  associating to the natural number  $N$  a tuple represented as a list `Ns` of length  $K$ .

```
to_cantor_tuple1(K,N,Ns) :- numlist(0,N,Is), cartesian_power(K,Is,Ns),
    from_cantor_tuple1(Ns,N),
    !. % just an optimization - no other solutions exist
```

Note that the built-in `numlist(From, To, Is)` is used to generate a list of integers in the interval `[From..To]`.

The predicate `to_cantor_tuple1` uses `cartesian_power(K,Is,Ns)` to enumerate candidates of length  $K$ , drawn from the initial segment `[0..N]` of  $\mathbb{N}$ .

```

cartesian_power(0,_, []).
cartesian_power(K,Is,[X|Xs]) :- K>0, K1 is K-1, member(X,Is),
    cartesian_power(K1,Is,Xs).

```

As `cartesian_power` backtracks over this finite set of potential solutions, the predicate `from_cantor_tuple1(Ns,N)` is called until the first (and known to be unique) solution is found. Given the unicity of the solution, the CUT in the predicate `to_cantor_tuple1` is simply an optimization without an effect on the meaning of the program.

The following example illustrates the correctness of this executable specification.

```

?- to_cantor_tuple1(3,42,R), from_cantor_tuple(R,S).
R = [1, 2, 2], S = 42.

```

Unfortunately, performance deteriorates quickly around  $K$  larger than 5 and  $N$  larger than 100 as the time complexity of this program is at least  $O(N^K)$ . However, given our reliance on Prolog's backtracking, the search uses at most  $O(K \log(N))$  space when filtering through lists of length  $K$  containing numbers of at most the bitsize of  $N$ .

## 4.2 A better algorithm, using a tighter upper limit

The next step in deriving an efficient untupling function is a bit trickier. First we observe that, as `from_cantor_tuple(K,Ns,N)` runs through successive hyperplanes  $X_1 + \dots + X_k = M$ , for each of them the sum maxes out when  $X_1 = M$  and  $X_k = 0$  for  $1 \leq k \leq N$ . We can compute directly this maximum value with the predicate `largest_binomial_sum` as follows:

```

largest_binomial_sum(K,M,R) :- largest_binomial_sum(K,M,0,R).

largest_binomial_sum(0,_,R,R).
largest_binomial_sum(K,M,R1,Rn) :- K>0, K1 is K-1, M1 is M+K1,
    binomial(M1,K,B), R2 is R1+B,
    largest_binomial_sum(K1,M,R2,Rn).

```

The predicate `largest_binomial_sum(K,M,R)` computes the same  $R$  as `cantor_tuple([M, 0, ..., 0], R)`, with  $K-1$  0s following  $M$ .

Next we compute the upper limit for possible values of the sum  $M$  of  $[X_1, \dots, X_k]$  such that the relation `to_cantor_tuple([X1, ..., Xk], N)` holds, i.e. we find the hyperplane  $X_1 + \dots + X_k = M$  defining the Cantor  $K$ -tuple. This computation, is implemented by the predicate `find_hyper_plane(K,N,M)` which, when given the inputs  $K$  and  $M$ , finds the value of the sum  $M$  that defines the hyperplane containing our tuple.

```

find_hyper_plane(0,_,0).
find_hyper_plane(K,N,M) :- K>0, between(0,N,M), largest_binomial_sum(K,M,R), R>=N,!.

```

Note the use of the built-in `between(From,To,I)` that backtracks over integers in the interval `[From..To]`.

We are now ready to define a more efficient inverse of the `from_cantor_tuple1` bijection, called `to_cantor_tuple2`, as a search through the set of lists such that the relation `from_cantor_tuple1(Xs,N)` holds.

```

to_cantor_tuple2(K,N,Ns) :- find_hyper_plane(K,N,M),
    sum_bounded_cartesian_power(K,M,Xs),
    from_cantor_tuple1(Xs,N),
    !,
    Ns=Xs.

```

The search, restricted this time to integers in the interval  $[0..M]$  is implemented by the predicate `sum_bounded_cartesian_power`.

```
sum_bounded_cartesian_power(0,0, []).
sum_bounded_cartesian_power(K,M,[X|Xs]) :- K>0, M>=0, K1 is K-1,
    between(0,M,X), M1 is M-X,
    sum_bounded_cartesian_power(K1,M1,Xs).
```

Note that, after applying the upper limit  $M$  computed by `find_hyper_plane`, to ensure that only tuples summing up to  $M$  are explored, we are using a customized cartesian product computation, in the predicate `sum_bounded_cartesian_power` backtracking over lists  $[X_1..X_k]$  that sum-up to  $M$ . However, as the query

```
?- findall(M,(between(0,31,N),P is 2^N,find_hyper_plane(2,P,M)),Ms).
Ms = [1,1,2,3,5,7,10,15,22,31,44,63,90,127,180,255,361,511,723,1023,
    1447,2047,2895,4095,5792,8191,11584,16383,23169,32767,46340,65535]
```

indicates, while  $M$  grows significantly slower than  $P$  it can reach intractable ranges quite quickly.

The predicate `to_cantor_tuple2` is a good improvement over `to_cantor_tuple1`, but it is by no means the efficient algorithm we are seeking.

Clearly, a “paradigm shift” is needed at this point, as obvious optimizations only promise diminishing returns. The highest hope would be to find a deterministic predicate similar to the integer square root based inverse for the case  $N = 2$ , but this time the arbitrary degree  $N$  of our polynomial looks like an insurmountable obstacle.

## 5 The Missing Link: from Lists to Sets and Back

After rewriting the formula for the  $\mathbb{N}^k \rightarrow \mathbb{N}$  bijection as:

$$K_n(x_1, \dots, x_n) = \sum_{k=1}^n \binom{k-1+s_k}{k} \quad (4)$$

where  $s_k = \sum_{i=1}^k x_i$ , we recognize the *prefix sums*  $s_k$  incremented with values of  $k$  starting at 0.

At this point, as our key “Eureka step”, we instantly recognize here the “set side” of the bijection between sequences of  $n$  natural numbers and sets of  $n$  natural numbers described in [13]<sup>1</sup>. We can compute the bijection `list2set` together with its inverse `set2list` as

```
list2set(Ns,Xs) :- list2set(Ns,-1,Xs).

list2set([],_, []).
list2set([N|Ns],Y,[X|Xs]) :- X is (N+Y)+1, list2set(Ns,X,Xs).

set2list(Xs,Ns) :- set2list(Xs,-1,Ns).

set2list([],_, []).
set2list([X|Xs],Y,[N|Ns]) :- N is (X-Y)-1, set2list(Xs,X,Ns).
```

<sup>1</sup> In [13] a general framework for bijective data transformations provides such conversion algorithms between a large number of fundamental data types.

The following examples illustrate how it works:

```
?- list2set([2,0,1,2],Set).
Set = [2, 3, 5, 8].
```

```
?- set2list([2, 3, 5, 8],List).
List = [2, 0, 1, 2].
```

As a side note, this bijection is mentioned in [5] and implicitly in [2], with indications that it might even go back to the early days of the theory of recursive functions.

## 6 Backtracking one step: revisiting the $\mathbb{N}^k \rightarrow \mathbb{N}$ bijection

It is time to step back at this point, and factor out `list2set` from our tail-recursive “untupling” loop `from_cantor_tuple1`.

The predicate `from_cantor_tuple` implements the the  $\mathbb{N}^k \rightarrow \mathbb{N}$  bijection in Prolog, using the iterative computation of the binomial  $\binom{n}{k}$  as well as the sequence to set transformer `list2set`. In contrast to `from_cantor_tuple1`, `untupling_loop` does not need to add the increments  $1, 2, \dots, L - 1$  as this task has been factored out and processed by `list2set`.

```
from_cantor_tuple(Ns,N) :-
    list2set(Ns,Xs),
    untupling_loop(Xs,0,0,N).

untupling_loop([],_L,B,B).
untupling_loop([X|Xs],L1,B1,Bn) :- L2 is L1+1, binomial(X,L2,B), B2 is B1+B,
    untupling_loop(Xs,L2,B2,Bn).
```

This shifts the problem of computing its inverse from lists to sets, an apparently minor use of a *bijective data type transformation*, that will turn out to be the single most critical step toward our solution.

## 7 The Efficient Inverse

We have now split our problem in two simpler ones: inverting `untupling_loop` and then applying `set2list` to get back from sets to lists.

Our first attempt was to try out constraint solving as it can sometime reverse arithmetic operations. Moreover, global constraints like `all_different` can take advantage of the fact that we are dealing with sets. However, the code (included as a comment in the companion Prolog file), turned out to be orders of magnitude slower than `to_cantor_tuple2`. This happened despite of the fact that we have tried also to take advantage of the optimizations implemented by the predicate `to_cantor_tuple2`, most likely because delaying computations brought unnecessary overhead without changing the essentially nondeterministic nature of the search.

The key “Eureka step” at this point is to observe that `untupling_loop` implements the sum of the combinations  $\binom{X_1}{1} + \binom{X_2}{2} + \dots + \binom{X_K}{K} = N$ , which is nothing but the representation of  $N$  in the *combinatorial number system of degree  $K$* , [16], due to [6]. Fortunately, efficient conversion algorithms between the conventional and the combinatorial number system are well known, [1, 5].

For instance, theorem **L** in [5] describes the precise position of a given sequence in the lexicographic order enumeration of all sequences of length  $k$ .





The two transformations can be seen as defining a bijection between strictly increasing and nondecreasing sequences of natural numbers:

```
?- set2mset([2,5,6,8,9],Mset), mset2set(Mset,Set).
Mset = [2, 4, 4, 5, 5], Set = [2, 5, 6, 8, 9].
```

We can combine this bijection with the Cantor  $n$ -tupling bijection and obtain

```
from_cantor_multiset_tuple(Ms,N) :- mset2set(Ms,Xs), from_cantor_set_tuple(Xs,N).
```

```
to_cantor_multiset_tuple(K,N,Ms) :- to_cantor_set_tuple(K,N,Xs), set2mset(Xs,Ms).
```

For instance, when dealing with *commutative and associative operations*, such multiset encodings turn out to be a natural match.

## 9 A Simple Application: Fair Search

One might ask, legitimately, why would one bother with pairing and  $n$ -tupling bijections. While the case has been made (see for instance [11]) for various applications besides theoretical computer science, that range from indexing multi-dimensional data and geographic information systems to cryptography and coding theory, we will focus here on a simple application with immediate relevance to logic programming: fair search through a multi-parameter search space.

A theorem conjectured by Bachet and proven by Lagrange, states that “*every natural number is the sum of at most four squares*”. Let’s assume that one wants to find, a “simple” solution to the equation (5), knowing that, as a consequence of this theorem, a solution always exists.

$$N = X^2 + Y^2 + Z^2 + U^2 \quad (5)$$

Let us define “simple solution” as a solution bounded by  $O(X + Y + Z + U)$ . We want to enumerate “simpler” candidates first, efficiently. To this end, we can use the fast inverse of the Cantor  $n$ -tupling function (specialized to multisets, given that both the “\*” and “+” operations, involved in the equation 5, are associative and commutative). We can write a generic `fair_multiset_tuple_generator` as:

```
fair_multiset_tuple_generator(From,To,Length, Tuple) :- between(From,To,N),
to_cantor_multiset_tuple(Length,N,Tuple).
```

We can specialize `fair_multiset_tuple_generator` for our specific problem as:

```
to_lagrange_squares(N,Ms) :- M is N^2, % conservative upper limit
fair_multiset_tuple_generator(0,M,4,Ms),
maplist(square,Ms,MMs), sumlist(MMs,N),
!. % keep the first solution only

square(X,S) :- S is X*X.
```

The algorithm is quite efficient, for instance, it takes only a few seconds to find a decomposition for 2012:

```
?- time(to_lagrange_squares(2012,Xs)), maplist(square,Xs,Ns), sumlist(Ns,N).
% 9,685,955 inferences, 4.085 CPU in 4.085 seconds (100% CPU, 2371347 Lips)
Xs = [15, 23, 23, 27], Ns = [225, 529, 529, 729], N = 2012.
```

The algorithm is also simple enough to be used as an executable specification and it ensures optimality of the solution, in the sense that our search scans hyperplanes of the form  $X_1 + X_2 + X_3 + X_4 = K$  for progressively larger and larger values of  $K$ . Also, given the multiset representation, the associativity and commutativity of “\*” and “+” are factored in, reducing the search space significantly. However, our simple algorithm is no match to the  $O(\log^2(N))$  randomized algorithm of [9]. As a side note, deriving a faster algorithm for this decomposition is a fascinating task on its own, starting with the observation that it needs only to be computed for the prime factors of a number and involving some elegant identities holding for Hurwitz quaternions [18]. More importantly, the mechanism sketched here can also be used in iterative deepening algorithms as a fair a goal selector (for both conjunctions and disjunctions). This can be done initially in a meta-interpreter and possibly partially evaluated or moved to the underlying Prolog abstract machine.

Note also that, depending on the natural representation of the candidate data tuple (i.e. set, multiset or sequence), one can customize the fair tuple generator accordingly.

## 10 Related Work

We have found the first reference to the generalization of Cantor’s pairing function to  $n$ -tuples in [2], and benefited from the extensive study of its properties in [7].

There are a large number of papers referring to the original “Cantor pairing function” among which we mention the surprising result that, together with the successor function it defines a decidable subset of arithmetic [3]. Combinatorial number systems can be traced back to [6] and one can find efficient conversion algorithms to conventional number systems in [5] and [1]. Finally, the “once you have seen it, obvious” `list2set` / `set2list` bijection is borrowed from [13], but not unlikely to be common knowledge of people working in combinatorics or recursion theory. This simple bijection between lists and sets of natural numbers shows the unexpected usefulness of the framework supporting bijective data type transformations [13, 15, 12], of which, a large Haskell-based<sup>2</sup> instance is described in [14].

## 11 Conclusion

We have derived through iterative refinements a fairly surprising solution to an open problem for which we had no a priori idea if it is solvable, or within which complexity bounds could be solved. The key “Eureka step” was to recognize a bijective data type transformation that suddenly brought us to a relatively well known equivalent problem for which efficient algorithms were available. Through the process, the ability to automate search algorithms relying directly on an executable declarative specification has been a major catalyst. The ability to derive equivalent logic programs using simple transformations has been also unusually helpful. From a software engineering perspective, this recommends logic programming as an ideal problem solving tool. Last but not least, proven sources of fundamental algorithms like [5] and the unusually high quality of Wikipedia articles on related topics have helped “connecting the dots” quickly and effectively.

## Acknowledgement

This research has been supported by NSF research grant 1018172.

---

<sup>2</sup> but designed in a guarded Horn-clause style, for virtually automatic transliteration to Prolog

---

**References**


---

- 1 B. P. Buckles and M. Lybanon. Generation of a Vector from the Lexicographical Index [G6]. *ACM Transactions on Mathematical Software*, 5(2):180–182, June 1977.
- 2 Patrick Cégielski and Denis Richard. On arithmetical first-order theories allowing encoding and decoding of lists. *Theoretical Computer Science*, 222(1–2):55 – 75, 1999.
- 3 Patrick Cégielski and Denis Richard. Decidability of the Theory of the Natural Integers with the Cantor Pairing Function and the Successor. *Theor. Comput. Sci.*, 257(1-2):51–77, 2001.
- 4 Martin Davis, Hilary Putnam, and Julia Robinson. The decision problem for exponential diophantine equations. *The Annals of Mathematics*, 74(4):425–436, nov 1961.
- 5 Donald E. Knuth. *The Art of Computer Programming, Volume 4, Fascicle 3: Generating All Combinations and Partitions*. Addison-Wesley Professional, 2005.
- 6 D. H. Lehmer. The machine tools of combinatorics. In *Applied combinatorial mathematics*, pages 5–30. Wiley, New York, 1964.
- 7 Meri Lisi. Some remarks on the Cantor pairing function. *Le Matematiche*, 62(1), 2007.
- 8 Yuri Matiyasevich. *Hilbert’s Tenth Problem*. MIT Press, Cambridge, London, 1993.
- 9 Michael O. Rabin and Jeffery O. Shallit. Randomized algorithms in number theory. *Communications on Pure and Applied Mathematics*, 39(S1):S239–S256, 1986.
- 10 Julia Robinson. Unsolvable diophantine problems. *Proceedings of the American Mathematical Society*, 22(2):534–538, aug 1969.
- 11 Arnold L. Rosenberg. Efficient pairing functions – and why you should care. *International Journal of Foundations of Computer Science*, 14(1):3–17, 2003.
- 12 Paul Tarau. A Groupoid of Isomorphic Data Transformations. In J. Carette, L. Dixon, C. S. Coen, and S. M. Watt, editors, *Intelligent Computer Mathematics, 16th Symposium, Calculemus 2009, 8th International Conference MKM 2009*, pages 170–185, Grand Bend, Canada, July 2009. Springer, LNAI 5625.
- 13 Paul Tarau. An Embedded Declarative Data Transformation Language. In *Proceedings of 11th International ACM SIGPLAN Symposium PPDP 2009*, pages 171–182, Coimbra, Portugal, September 2009. ACM.
- 14 Paul Tarau. Declarative Combinatorics: Isomorphisms, Hylomorphisms and Hereditarily Finite Data Types in Haskell, January 2009. Unpublished draft, <http://arXiv.org/abs/0808.2953>, updated version at <http://logic.cse.unt.edu/tarau/research/2010/ISO.pdf>, 150 pages.
- 15 Paul Tarau. Isomorphisms, Hylomorphisms and Hereditarily Finite Data Types in Haskell. In *Proceedings of ACM SAC’09*, pages 1898–1903, Honolulu, Hawaii, March 2009. ACM.
- 16 Wikipedia. Combinatorial number system — wikipedia, the free encyclopedia, 2011. [Online; accessed 21-March-2012].
- 17 Wikipedia. Pairing function — wikipedia, the free encyclopedia, 2011. [Online; accessed 23-March-2012].
- 18 Wikipedia. Lagrange’s four-square theorem — wikipedia, the free encyclopedia, 2012. [Online; accessed 22-March-2012].