

# Bottom-Up Enumerative Synthesis of Recursive Programs using Generalized Folds

RAFAELLO SANNA and NADA AMIN, Harvard University, USA

WILLIAM BYRD, University of Alabama at Birmingham, USA

We present DAN, a new technique for extending bottom-up enumerative synthesis to recursive procedures. Bottom-up synthesis has been shown its ability to generate non-trivial programs in many interesting domains, and exploits the problem's overlapping subproblems and optimal substructure. However, these properties are not available in the cases of recursive programs, where subproblems are often not defined. One solution to this problem is to build up recursive programs from smaller sub-problems, exploiting the structure of recursive data structures to guide the recursion itself, often by using a library of recursive functions known to be terminating. We extend this work to arbitrary data structures, using techniques from the generic programming and structured recursion literature.

Additional Key Words and Phrases: Program Synthesis, Recursion Schemes, Generic Programming

## 1 INTRODUCTION

Program synthesis has shown the potential to revolutionize the task of programming. It allows programmers to express the program they want in a simple, but often ambiguous manner, and have the machine do the work of finding a program that fits their specification. A canonical example of this has been the task of programming-by-example (PBE), in which a user specifies a general transformation by demonstrating it on a series of concrete input/output pairs (IO pairs). The role of a programming-by-example system is to learn the underlying transformation from the examples and render it as program code which can be applied to any object.

### 1.1 Bottom-Up Synthesis

One of the most successful methods of program synthesis has been bottom-up synthesis, which enumerates possible values in the language by the size of the smallest expression which produces them. We start first with the constants of the language and the inputs from the user, each of these with size zero. We then grow our set of values by applying the language's operators to each of the values we've seen so far, then keeping all the new values we find, along with the programs that created them. As the set of possible values grows, we eventually reach the values we want to generate. The program which generates these values is then the one specified by the user. The major advantage of bottom-up synthesis is that we work with values, not programs. This is a classic example of overlapping substructure - there are many programs which generate the same value, so we can share repeated work by keeping track only one canonical program for every value.

$$\begin{aligned}
 t &\rightarrow 0 \mid 1 \mid \dots \\
 &\mid x \\
 &\mid t + t \mid -t \mid t * t \mid (t) \\
 &\mid Y(x.t) \\
 &\mid \text{if } t \text{ then } t \text{ else } t
 \end{aligned}$$

Fig. 1. A Simple Language with the Fix-Point Combinator

## 1.2 Partial Operators and the Fix-Point Combinator

The major upside of bottom-up synthesis is also its downside. While all values have programs which evaluate to them, in many languages, not all programs evaluate to values. Operators in these language are called partial - they can only sometimes be meaningfully evaluated. One operator which is of this kind is the fix-point combinator of language 1, written as the term  $Y(x.t)$ , which evaluates its body  $t$  with the variable  $x$  bound to  $Y(x.t)$ . This operator expands the set of possible programs which can be expressed, but also introduces many programs which do not meaningfully terminate. For example, the only possible reduction step in the evaluation of the term  $Y(\text{rec}. \text{rec})$  is  $Y(\text{rec}. \text{rec})$ . When combined with a conditional operator such as `if-then-else`, it makes the language Turing complete, meaning we cannot meaningfully check if a program terminates.

## 2 RELATED WORK

The most common solution to the problem of non-terminating programs is to introduce a full recursive operator such as  $Y$ , but restrict its use only to certain special cases[1][2]. For instance, in the ESCHER[2] tool, the  $Y$  operator can only be generated when the recursion is shown to be structural, as in many languages with totality checkers. The advantage of techniques such as this one is that the language comes more naturally to humans. Most programmers do not worry about totality checking in their day-to-day work, so the ability to express non-terminating procedures is natural (for this reason, many papers reformat their examples to remove the  $Y$  combinator altogether, showing a more traditional recursive definition form). The major disadvantage is that we lose the overlapping substructure of the original problem. The value of a term (indeed, whether or not it even *has* a value) is a function not only on its free variables, but on its location within the program as a whole. This requires the generation of partial programs which take on different (or no) values when evaluated in different areas of the program.

The alternative method to this is to break down the recursion into smaller parts, each of which can be solved using non-recursive code[3]. An illustrative example of this method is the  $\lambda^2$ [3] project. Rather than using a language with a possibly non-terminating fix-point combinator, it enumerates terms using higher-order procedures typical of functional programming, such as `map`, `filter` and `fold`. These procedures allow for natural, inductive transformations on functional data structures, while still ensuring the language is terminating. The major downside of this approach is that it requires a complete set of such combinators in advance. It isn't possible, for example, to define new data structures or new operators on existing data structures.

We expand prior work by performing true bottom-up synthesis on arbitrary data structures. To achieve this, we use generic programming and recursion schemes to derive operators such as `fold` automatically.

## 3 GENERALIZED FOLDS

Recall the definition of Lists and their fold:

```
data List a = Cons a (List a)
           | Nil
```

```
foldList : (onCons : el -> acc -> acc) -> (onNil : acc) -> List el -> acc
foldList onCons onNil = rec
  where
    rec : List el -> acc
    rec Nil = onNil
    rec (Cons x xs) = onCons x (rec xs)
```

There's a clear correspondence between the parameters of the fold and the constructors in the datatype. For each constructor, the fold takes a procedure which takes the fields of the constructor as arguments and returns a new accumulated value. We essentially define our own interpretation of each constructor, then recursively apply it throughout.

Consider the definition of the following algebraic datatype for representing compound objects:

```
data Obj = Zero
        | Succ Obj
        | Nil
        | Cons Obj Obj
```

We can try writing a fold in the same way: mapping each constructor to a parameter, then combining them.

```
foldObj : (onZero : Obj) ->
         (onSucc : Obj -> Obj) ->
         (onNil : Obj) ->
         (onCons : Obj -> Obj -> Obj) ->
         Obj -> Obj
foldObj onZero onSucc onNil onCons = rec
  where
    rec : Obj -> Obj
    rec Zero = onZero
    rec (Succ x) = onSucc (rec x)
    rec Nil = onNil
    rec (Cons x y) = onCons (rec x) (rec y)
```

## 4 FUTURE WORK

Immediate next steps forwards for the Dan project would be expanding the implementation. Currently, it can only generate dynamically typed programs - all terms must be of a single type. This works, but limits the opportunities for pruning possible programs and makes expressing problems in the implementation language awkward. Another major improvement would be expressing new kinds of recursion other than fold, and defining multiple implementations of fold for different datatypes.

It would also be helpful to further develop the connection between Dan and its category-theoretic notions. Currently, the only scheme provided is a catamorphism, but providing options such as a paramorphism or hyglomorphism would extend its synthesis abilities significantly.

Lastly, Dan needs to be evaluated more thoroughly, both in isolation and compared to other tools, such as ESCHER and  $\lambda^2$ . Becoming competitive with these projects may require some performance engineering, as the current implementation was written primarily for ease of reading and experimentation.

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