Prolog for VLSI Layout: Experiences in the Design and Implementation of Topolog, A Prolog-Based Module Generation and Layout System

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1. Abstract

The Topolog module generator is the major circuit-design component of the ASP silicon compiler. Topolog is an attempt to determine the utility of Prolog specifically and logic programming generally for the programming of solutions to large-scale VLSI circuit design problems. We have verified that Prolog's clause-based programming style permits easy extensibility of VLSI module generators for new technologies and user-written macroblocks. We have demonstrated that Prolog, even without the well-known assert, retract, and write operators is not a pure applicative language. We have devised a method of type definition in Prolog, and have preliminary evidence that our method is superior in efficiency to the general term unification method commonly found in the literature.

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2. Introduction

Topolog is the module generator, layout engine, and circuit database manager for the ASP Silicon Compiler. It takes in a description of a circuit to be generated, constraints on the bounding box and a set of ports, and outputs a sticks-based layout description which can be converted to a fabricatable form by our mask-level design environment, STICKS-PAC [Cheng87].

A module generator is a program which, given a description of a circuit (or module) as a collection of blocks or subcells and a set of parameters, returns a cell, or piece of silicon, which matches the parameters given. The subcells may be modules in their own right, or elementary pieces of silicon called leaf cells. A layout engine is a program which, when given a description of a circuit as either a collection of logical units called gates or as a list of transistors and connections, returns a piece of silicon which implements the circuit.

Topolog combines the functions of a module generator and layout engine in the hope that a combination of these tools may solve problems specific to each. Typical module generation systems manipulate pieces of geometry rather than circuit elements, which means that most module generation programs and parameters simply direct the manipulation of pieces of wire rather than function. Further, if a module consists of submodules, the choice of which submodule to instantiate first has a very large effect on the resultant circuit for purely geometric reasons. Folding a layout program into a module generator permits the generator to concentrate on the functional design of circuits, rather than on their geometry, which in practice yields much more concise module descriptions. Further, if the submodules are expanded as blocks and jointly placed and routed, the second problem disappears.

Typical layout generators are flat: that is, a single long list of transistors is used to describe the function to be generated. This is both tedious from the point of view of a user (who must enter his circuit as a long sequence of logic equations, rather than using circuit hierarchy) and robs the layout engine of inherent partitioning of most logic circuits. This is particularly onerous since most automated placement tools either implicitly or explicitly partition a circuit into connected
subcircuits. The class of placement tools for which do such partitioning is very broad indeed, including particularly clustering, min-cut, force-directed and clique-based placement tools. Even simulated annealing, which specifically does not work by circuit partitioning, derives its name and its original motivation from the formation of metal into disjoint clusters.

3. Description of Topolog

This section will be relatively brief, since we presume that the Prolog community will be more interested in the performance of Topolog and the programming techniques which we used than in the actual algorithms employed.

Topolog is designed around the basic abstraction of a block. A block represents a primitive circuit element, and it is defined by the fields it contains and the routines which generate it. A block has a p-side and an n-side both of which have a max_height and min_height, a set of elements, a set of sticks, and a set of pins. In addition, the block have various fields used only by Topolog itself, a set of netNames, and a max_width and min_width. Topolog's basic function is to group blocks into rows, and to route signals between the blocks. A single routing channel runs between the p- and n-side of any row; a power bar runs above the p-side of every row, and a ground bar runs beneath the n-side of any row. Odd rows are flipped about the horizontal axis so that power and ground bars may be shared between rows. It is tempting to consider Topolog as a standard cell layout program, but this is quite misleading. Since blocks can be anything which shares the characteristics mentioned here, it is more accurate to describe Topolog as a Gate Matrix style layout engine.

Topolog has a six stage pipeline. After inputs are parsed, a preliminary generation of all the blocks is done. In this pass, the max_height and min_height and max_width and min_width of the blocks are fixed. The blocks are then grouped into rows, and placed within rows. During this placement phase, macroblocks (modules) are expanded into their primitive components. Detailed generation of blocks is done; the blocks are fleshed out into a sticks-and-elements description, and the pins for channel routing are defined. The channel is then routed. Finally, numbers are
assigned to each row and the package is output.

3.1. Technology Independence and Extensibility 1: Block Generation

Topolog currently supports five types of blocks: static cmos and-or-invert gates, domino cmos gates, pass and transmission gates, and an experimental circuit style called precharged cascode voltage switches. These five types of blocks are all that we have experimented with.

Topolog, however, is designed to support any circuit style or technology that can be expressed in the style mentioned above. The terms \( p\text{-side} \) and \( n\text{-side} \) refer to \( p \) and \( n \)-diffusion regions, reflecting our primary concern with CMOS technology; however, there is no reason, in principle, to use these regions specifically for these purposes. One can imagine, for example, using Topolog for NMOS designs using the \( p\text{-side} \) for the enhancement device.

The addition of a new circuit type is quite easy, due to Prolog's clause-based programming style. The user must write a clause for the procedure \( \text{buildBlock} \)\( (\text{Input, Block}) \), where \( \text{Input} \) is the input for the block; for example, the clause header for \( \text{aoi} \) blocks is \( \text{buildBlock} \)\( (\text{Output} = \text{aoi}(\text{Expr}), \text{Block}) \). This clause must return a \( \text{Block} \), which is a data structure with the fields mentioned above. Some of these fields (in particular, the \( \text{max}\_\text{height} \) and \( \text{min}\_\text{height} \) fields of the two sides and the \( \text{max}\_\text{width} \) and \( \text{min}\_\text{width} \) field) must be filled in, since these are used by the placement code. In addition, the user probably wishes to store a parse form of \( \text{Expr} \) for later use. The user may use a variety of builtin tools to construct his clause; these will be fully described in the final version of the paper.

\( \text{buildBlock} \) only does the first pass at generation of a block. In the second pass, the block must become an object with a full set of elements and sticks. The procedure \( \text{generate}_\text{block} \)\( (\text{Block}, \text{Prows}, \text{Nrows}, \text{Columns}) \) is called to instantiate a block on the rows and columns given; these columns are guaranteed to be in the range given by height and width. Again, a large set of tools are available to aid in the construction of this routine.
3.2. Existing Blocks and Generation Routines

Our existing logic blocks are all designed by the well-known Uehara-Van Cleemput procedure. The UVC algorithm has been shown to derive near-minimal-width single-diffusion-strip static CMOS arrays.

3.3. Module Generation and Extensibility II

It is convenient for users to define modules as collections of blocks or other modules. As a result, buildBlock has a "catch-all" clause; if it cannot build a block any other way, it calls a procedure defined by its first argument. Specifically:

```
buildBlock(X, Block) :-
    X ==.. [BlockType|BlockArgs],
    concat(BlockArgs, [Block], FunctionArgs),
    Call ==.. [BlockType|FunctionArgs],
    Call.
```

Hence a request in Topolog's input file of the form:

```
alu(x, y, z).
```

would result in a call to the Prolog procedure:

```
alu(x, y, z, Block).
```

Of course, the user would have to define that procedure. buildBlock calls must be used to build the various component blocks (including other modules, which would be invoked by the same mechanism). A final call

```
buildCompositeBlock([Block1,...,Blockn], Block)
```
must appear as the last call in the \texttt{alu} procedure. Here, \texttt{Block1}, \ldots, \texttt{Blockn} are the blocks built by the call to \texttt{buildBlocks} in the \texttt{alu} procedure.

Of course, the \texttt{alu} procedure must be known to \texttt{Topolog} at the time of invocation; the request:

\begin{verbatim}
use(file).
\end{verbatim}

loads the procedures defined in \texttt{file}.

No other clauses are required for module construction, since the placement routines break modules into their component parts before the blocks are actually generated; hence \texttt{generateBlock} clauses need only be supplied for primitive blocks.

4. Types and Type Definitions

\texttt{Topolog} is about 3000 lines of \texttt{Prolog} code. 16 major data types are defined in the program, with a varying number of fields — from 2 to 19. These data types are often widely shared among various procedures (the \texttt{logicBlock} datatype is used by virtually every stage in the pipeline). Moreover, as in all program development, these datatypes often change during the course of program development, as new requirements for the various datatypes are discovered and old requirements discarded.

The standard method of data structure creation and access in \texttt{Prolog} is through the mechanism of general term unification. This mechanism makes the definition of data structures quite easy, but spreads the definition of a type among all the clauses that access the type. Naturally, this mechanism makes the modification of type definitions quite onerous. Further, if types are large this tends to degrade the legibility of the code.

The problem of spreading type definitions throughout a program is well known in the \texttt{Lisp} community [Charniak]. In that community, records are defined as fixed-length lists, and some combination of \texttt{car} and \texttt{cdr} is used to access the various fields (this is known as the \texttt{caddadr} problem). Of course, the problem is somewhat worse there, since a \texttt{Lisp} programmer must ask
whether some instance of cdaddr means net.name, or, instead, block.transistors.

There have been two traditional solutions to this problem in the Lisp community. The first has been to define a build procedure for each data structure, and an access procedure for each field. In general, the build procedure and access procedures for each type were maintained in a separate file.

The second solution is the one that we chose to adopt for Prolog. A procedure typedef, which builds the various building and accessing procedures for us.

```
typedef(X) :-
    X == [Name | Args],
    makeStructType(Y, Name, Args),
    assert Macros(Name, Args, Y, 1), !,
    deleteMakeStatement(Name),
    assert(makeStruct(Name, Y)).
```

makeStructType makes a dummy template for the record, so that the unification mechanism does the actual structure creation for us (in other words, the arguments in the structure definition are replaced by unique variables)

```
makeStructType(X, Name, Args) :-
    makeVarList(Args, Vars),
    X == [Name | Vars].

makeVarList([], []) :- !.

makeVarList([_, X], [_, Y]) :-
    makeVarList(X, Y).
```

assert Macros creates a clause in the procedure field for each field of the record. This permits symbolic access to each field of the record. It also deletes old access clauses for this field of this data structure.

```
assert Macros(_, [], _) :- !.

assert Macros(Name, [Arg | Args], Dummy, Count) :-
```
atomic(Arg, 1),
arg(Count, Dummy, Val),
P = field(Dummy, Arg, ),
removeExistingMacro(P),
Q = field(Dummy, Arg, Val),
assert((Q :- !)),
Count' is Count + 1,
assertMacros(Name, Args, Dummy, Count').

removeExistingMacro simply retracts any clause whose head matches the term passed in.

Once this procedure is defined, a call of the form typedef(logicFn(op, args, count, flipped)) will define the following clauses:

makeStruct(logicFn, logicFn(_,_,_,_,_,())),
field(logicFn(_,_,_,_,_,_,_), op, 5) :- !.
field(logicFn(_,_,_,_,_,_,_), args, 5) :- !.
field(logicFn(_,_,_,_,_,_,_), count, 5) :- !.
field(logicFn(_,_,_,_,_,_,_), flipped, 5) :- !.

and a data structure of type logicFn can be accessed and used directly.

Emblishments are possible once this basic tool is in place. For example, we might wish to access a fair number of fields with one call:

fields(Struct, []) :- !.
fields(Struct, [Field = Val | Fields]) :-
    field(Struct, Field, Val),
    fields(Struct, Fields).

And hence fields(LogicFn, [op = Op, args = Args]) digs out both the args and op field of LogicFn if LogicFn is a logicFn. This trick can be used to initialize fields as well:

makeStruct(StructName, Struct, StructFields) :-
    makeStruct(StructName, Struct),
    fields(Struct, StructFields).

and hence makeStruct(logicFn, LogicFn, [op = and, args = [z, y, z], count = 5]) makes LogicFn
a logicFn with the fields set appropriately.

4.1. Generic Types and Access Procedures

Many CAD systems use variants on object-oriented programming. This paradigm is attractive for these applications, since one wishes to define operations over abstract objects. These abstract objects may differ in detail, but they share all the necessary attributes for the appropriate operations.

An example of an abstract data type is given by our block, which we encountered above. As we noted, there are currently five sorts of block defined; of course, there are potentially very many more. The placement routine, however, cares not a whit about anything other than that the block has certain named fields.

This variant on object-oriented programming could have been provided by writing procedures block_height, block_width, and so forth, and demanding that the user or programmer write (generally trivial) clauses for each such procedure. As it is, the programmer must write clauses for only two procedures (buildBlock and generateBlock), and ensure that the appropriate fields are included when the new block is typedef'd. Hence much of the abstraction that we seek is provided by the generic access procedure field. Clearly this approach places a much less onerous burden on the programmer, and thus fulfills the economy of representation that is one of object-oriented programming's principle advantages.

4.2. Efficiency Considerations

typedef, field and makeStruct, as given above, are highly inefficient, if conceptually simple. Here we assume that unifications form the great cost of most Prolog implementations, and hence we count the number of unifications involved in our scheme. If we consider some procedure $P$ that accesses $n$ fields of some structure $Q$ of $m$ fields, we see that there are $O(nm)$ successful unification operations (since $n \leq m$ this is $O(m^2)$), and as many as $O(nl)$ unsuccessful unifications, where $l$ is the total number of fields of all types defined in the program. Hashing techniques (eg,
[Warren] [PLM]) can reduce the latter number to $O(m^2)$ given a compiled environment. If $Q$ were expanded in $P$’s header, as is the usual case, then a total of $O(m)$ unifications would be done.

We can do somewhat better, in general, than this latter number. First, we must only unify those fields that we actually desire, which will reduce the number of successful unifications. Second, we must avoid a single global field procedure with many clauses, which will reduce the number of unsuccessful unifications.

The following field procedure reduces the number of successful unifications by doing a table lookup on the symbolic name of a field to find the corresponding argument number, and then using the builtin arg to get that argument:

\[
\text{field}(\text{Struct, FieldName, Val}) \leftarrow \\
\text{functor}(\text{Struct, Functor, Arity}), \\
\text{fieldNum}(\text{Functor, Arity, FieldName, FieldNum}), \\
\text{arg}(\text{FieldNum, Struct, Val}).
\]

The following typedef and associated code generates a new version of makeStruct, as well as the new procedure fieldNum:

\[
\text{typedef}(X) \leftarrow \\
X = \ldots \text{[Name|Args]}, \\
\text{functor}(X, \text{Name, Arity}), \\
\text{assert Macros}(\text{Name, Args, Arity}), !, \\
\text{delete MakeStatement}(\text{Name}), \\
\text{assert}((\text{makeStruct}(\text{Name, Y}) \leftarrow \text{functor}(Y, \text{Name, Arity}))), \\
\text{assert}((\text{typecheck}(Y, \text{Name}) \leftarrow \text{functor}(Y, \text{Name, Arity}))).
\]

\[
\text{delete MakeStatement}(\text{Name}) \leftarrow \\
\text{remove Existing Macro}(\text{makeStruct}(\text{Name, ...})), \\
\text{remove Existing Macro}(\text{typecheck}(\ldots, \text{Name})).
\]

\[
\text{assert Macros}(\text{Name, Args, Arity}) \leftarrow \\
\text{assert Macros}(\text{Name, Args, I, Arity}).
\]

\[
\text{assert Macros}([], [], [], _) \leftarrow !.
\]

\[
\text{assert Macros}(\text{Name, [Arg|Args], CountIn, Arity}) \leftarrow \\
\text{atomic(Arg), !}, \\
P = \text{fieldNum}(\text{Name, Arity, Arg, ...}),
\]

\[
= \text{makeStruct}(\text{Name, Arity, Arg, ...}).
\]
removeExistingMacro(P),
R = fieldNum(Name, Arity, _, CountIn),
removeExistingMacro(R),
Q = fieldNum(Name, Arity, Arg, CountIn),
assert(Q),
NextCount is CountIn + 1,
assertMacros(Name, Args, NextCount, Arity).

assertMacros(_, [Arg | Args], ProcName, Count) :-
   write('Error -- must be an atom as a field name, not '),
   write(Arg), nl,
   break.

It can be seen that the number of successful unifications in P is \(O(n)\) (assuming \(arg\) does \(O(1)\) unification, as it will in any rational implementation). The number of unsuccessful unifications remains; of course, using indexing, this is already \(O(1)\), but we must assume a dumb execution environment. For the moment we satisfy ourselves with a two-level procedure, as defined by the following (last) version of typedef:

typedef([X] :-
   X =..[Name | Args],
   functor(X, Name, Arity),
   assertMacros(Name, Args, Arity), !,
   deleteMakeStatement(Name),
   assert(makeStruct(Name,Y) :- functor(Y, Name, Arity)),
   assert((typecheck(Y, Name) :- functor(Y, Name, Arity))).

deleteMakeStatement(Name) :-
   removeExistingMacro(makeStruct(Name, _)),
   removeExistingMacro(typecheck(_, Name)).

assertMacros(Name, Args, Arity) :-
   gensym(ProcName),
   P = fieldNum(Name, Arity, ArgName, ArgNum),
   removeExistingMacro(P),
   Q =.. [ProcName,ArgName,ArgNum],
   assert([ P : Q, !]),
   assertMacros(Name, Args, ProcName, 1).

assertMacros(_, [], _, _) :- !.

assertMacros(Name, [Arg | Args], ProcName, CountIn) :-
   atomic(Arg), !,
   Q =.. [ProcName, Arg, CountIn],
   assert([ Q : !]),
NextCount is CountIn + 1,
assert Macros(Name, Args, ProcName, NextCount).

assert Macros(_, [Arg] Args), ProcName, Count) :-
    write('Error -- must be an atom as a field name, not '),
    write(Arg), nl,
    break.

In this case, a new procedure is generated for each datatype (the procedure gensym generates a new atom). This procedure has one clause for each field, and fieldNum has one clause per datatype. It can readily be seen that the worst-case number of unsuccessful unifications is $O(tn+mn)$, where $t$ is the total number of datatypes defined in the program.

In an interpreted environment, this is still not competitive with the standard method of type creation and access, but we believe that the robustness and concision of the resulting code is worth the performance penalty. This is particularly true since, in a compiled environment, the penalty for unsuccessful unifications largely disappears. If $n << m$, as it is in practice, in a compiled environment our last version would run faster than the standard method. In practice, we have found that the last version of typedef improves both our runtime and global stack usage by about 60% over the naive version first used.

The idea of generating a special-purpose procedure for each datatype may be taken to an extreme by currying the typename, arity, and fieldname together to obtain a special-purpose one-clause procedure for each field in each type; for example, logicFn5flipped(5), thus faking the hash function that a compiler would provide. The Cprolog built-in name was used to turn atoms into lists and vice-versa as intermediate stages in the currying. We found that our storage costs grew dramatically, largely as a result of the computation of the procedure name from the given datatype and fieldname. Further, performance was only slightly better than the early naive version of typedef, and uncompetitive with two-level lookup. The code for field is given here:

field(Struct, FieldName, Val) :-
functor(Struct, Functor, Arity),
name(Functo, FName),
name(Arity, A1),
name(FldName, ArgName),
concat(A1, ArgName, Tmp),
concat(FName, Tmp, Tmp2),
name(ProcName, Tmp2),
Proc =.. [ProcName, FieldNum],
Proc,
arg(FieldNum, Struct, Val).

We are indebted to Peter Vanroy for suggesting the two-level table lookup.

5. Destructive Assignment

5.1. Introduction

The Aquarius Project [Despain85a] at Berkeley is developing high-performance computers [Despain85b] for the execution of Prolog. Part of the evaluation effort that we are making is to understand the advantages and disadvantages of Prolog for the implementation of programs to solve challenging problems in difficult domains of discourse. In particular, we have engaged in the design and implementation of a suite of Prolog CAD tools for VLSI design [Despain86] [Pincus86] [Bush87] [Cheng87] [McGeer87].

In the course of implementing a VLSI layout program in Prolog during the summer and fall of 1985, we experienced difficulties in implementing standard routing and transistor placement algorithms. After discussions with other groups that had used Prolog for Computer-Aided Design of Integrated Circuits programs, we concluded that the difficulties we experienced were common among Prolog CAD programmers. We investigated the nature and source of our difficulties, and concluded that the principal problem lay in Prolog's lack of a destructive assignment operator akin to Lisp's rplaca or rplacd. We then investigated the addition of such an operator to Prolog. This paper presents the results of that study.
This chapter is organized into seven sections. Section II gives examples of the algorithms that we could not implement in nominal time in pure Prolog. Section III gives a general graph-theoretic argument to explain the difficulty of data-structure manipulation without destructive assignment. Section IV defines the destructive assignment operator \texttt{rplacarg} that we require and its operational and semantic characteristics. Section V describes an method for implementing the \texttt{rplacarg} in C-Prolog, or any implementation of Prolog that supports the \texttt{var} builtin. Section VI describes the features that must be added to a Warren Abstract Machine (WAM) [Warren83] to implement \texttt{rplacarg} for both the structure-sharing and structure-copying case. In particular, we show that a highly-efficient \texttt{O(1)} \texttt{rplacarg} primitive may be added to our WAM-based Programmed Logic Machine (PLM). Section VII describes a multidimensional array implementation based on the \texttt{rplacarg} construct. In an appendix, we show that any implementation of Prolog that supports the \texttt{!, fail} implementation of negation supports \texttt{var} as well; hence we conclude that \texttt{rplacarg} is semantically implied by \texttt{cut} and \texttt{fail}.

5.2. II-Algorithms We Couldn't Implement Efficiently In Prolog

The central art of computer science is performing computations in the most time-efficient manner possible. Without efficiency concerns, all of computer science is trivia.

Concern for efficiency leads us to create data structures. Data structures are ways of storing intermediate results of computation, so that these computations need not be re-performed. Indeed, one might argue persuasively that all of computer science is the design of data structures that have the property that the amount of computation required to solve a given problem is minimized.

The core of our argument is that the implementation of some operations over some data structures is difficult and inefficient in Prolog, that these data structures are relatively familiar objects in some application domains, and that these difficulties arise precisely because of the applicative nature of Prolog. We have a general argument to explain this phenomenon, but our case can best be understood in light of a few examples.
We have not been able to implement the algorithms given below in nominal time in pure Prolog. By pure Prolog we mean Prolog without the well-known *assert/retract* primitives, which are known to be non-applicative (or, in the Prolog parlance, non-logical) or the *var* primitive, which we can show below is semantically equivalent to the non-applicative *replace* primitive we advocate. Further, we can show that the well-known *cut*, *fail* construction for negation is equivalent to *var*, so we do not consider implementations using *cut*, *fail*.

5.2.1. Kernighan-Lin Min-Cut Algorithm

The Kernighan-Lin min-cut algorithm [Ullman82] is a greedy procedure to partition hypergraphs into two equal-sized sets so that the *cut* — the number of hyperedges that connect the two sets — is minimized. It has been shown that the min-cut problem for hypergraphs is NP-complete [Garey79]. However, the Kernighan-Lin algorithm is an excellent approximation procedure.

The Kernighan-Lin algorithm begins with the nodes of the graph partitioned arbitrarily into the two sets, called *left* and *right*. On each iteration, a pair of nodes *(l, r)* is selected for interchange; the pair selected is that creating the greatest decrease or smallest increase in the *cut*. The pair are not immediately interchanged; but are merely marked as selected, treated as interchanged, and removed from *left* and *right*. When *left* and *right* are empty (there are no more unselected nodes), the total summed cost of the interchanges are computed in order. The largest negative total is taken, if there is any, those pairs of nodes are interchanged, and the selection process begins on the new *left* and *right*; if no negative total is found, the algorithm terminates.

The minimum requirement of this algorithm is that the cost of each interchange be rapidly computed. This in turn implies that each hyperedge have a pointer to each node upon which it is incident. Similarly, once a node is selected, it must be marked as selected; the selection, or not, of a node affects future cost computations on hyperedges incident upon that node. If marking a node as selected involves regenerating the node (as it does if neither *var* nor some form of destructive assignment is used), each hyperedge incident upon that node must be regenerated. There are potentially $2^{n-1}$ such hyperedges on an *n* node hypergraph, and hence this is quite an expensive
operation. Similarly, when the nodes are interchanged, if an interchange requires regeneration of each node, then every hyperedge must be regenerated. There are at most $2^n$ hyperedges on an $n$-node hypergraph.

We provide our Prolog implementation in an appendix, using our $rplacarg$ primitive, to be discussed in section IV.

5.2.2. $O(n)$ Average-Case Sorting

Jon Bentley [Bentley84] has posed a puzzle in sorting. Given two integers $N,M$, with $N < M$, generate $N$ distinct random numbers in the range $[0,M]$ and print them out, sorted, in average-case time $O(N)$.

Clearly this problem cannot be solved in worst-case time better than $O(N \log N)$. However, Knuth [Knuth86] has posted a solution to this puzzle with average-case behavior $O(N)$, and worst-case behavior $O(N^2)$.

The core of Knuth's method is the use of a hash table of size $N$, which is simply a vector. Implementation of vectors in Prolog has proven quite troublesome, and there have been a number of proposals. In section VII we show that the central difficulty in the implementation of vectors is the avoidance of copying the entire vector when a single argument is set. This is precisely the problem that we are trying to solve, and so it is unsurprising that our proposal here makes the implementation of arrays quite easy. For the moment, we just note that we can use the builtin $\text{functor}$ to get storage, and assume that in any rational Prolog implementation $\text{arg}$ is $O(1)$, and hence can be used for indexing.

Knuth uses a monotone-increasing function to hash each random number into the hash table, and uses an insertion-sort to resolve collisions. Clearly the hash table remains sorted; if there are a very small number of collisions, then time of the algorithm is $O(N)$; in fact, the probability of collision is very small. It is possible, however, for all numbers to hash to the same bucket, in which case Knuth's time is $O(N^2)$. 
We have devised a Prolog algorithm, using var, which matches Knuth's performance. Instead of maintaining only one item per bucket and using an insertion sort to avoid collisions, we maintain a bucket as a list and sort the list on output. In order to avoid having to replace the entire array, we maintain an unbound cdr on each bucket. The algorithm appears here.²

```prolog
:- consult(rand_int).

table(Seed):- M = 24, N = 96, M² is 2×M, functor(S,s,M²), % Initialize.
    fill_table(M,N,S,Seed,M),
    S =.. [1,S],
    print_sort(SS),nl,
    print(' DONE !'),nl.

/* fill the hash table */

fill_table(M,N,S,Seed,0).
fill_table(M,N,S,Seed,1) :-
    rand_int(Seed,Seed₁,1,N,T),
    H is 1+2×M×(T-1)//N,
    arg(H, S, V), insert(T,V,I,J),
    fill_table(M,N,S,Seed₁,J).

/* Insert an element into the table, maintaining the unbound cdr */

insert(X, Y, I, J) :- var(Y), !, Y = [X|_], J is I-1. % Insert element.
insert(X, [X|_], _, _) :- !. % eliminate dupes
insert(X, [H|T], I, J) :- insert(X,T,I,J). % Skip down list.

/* If a bucket is empty, it is unbound, ar! hence unifies to kruft (or, for that matter, any atom). If it is nonempty, it is a list, and hence won't unify */

print_sort([]).
print_sort([kruft\T]) :- !, print_sort(T). % Strip empty lists
print_sort([H|T]) :- sort(H,C), lprint(C), pr_st(T). % Print 1th bucket

pr_st([]) :- print(\), nl. % Terminate the printout
pr_st([kruft\T]) :- !, pr_st(T). % Strip out empty lists
pr_st([H|T]) :- print(\), sort(H,C), lprint(C), pr_st(T). % Print the bucket

lprint([H]) :- print(H). % Print last item
lprint([H|T]) :- print(H), print(\), lprint(T). % Print list element
```

² If we use a random number generator with period > M, then all generated samples are distinct and we need not check for duplicates. If there are no duplicates, then the only entry in a bucket which might unify to a list containing a new entry is the unbound cdr, and hence we do not need var.
Now, in the average case there are a small number of collisions, and hence our algorithm, like Knuth's, is $O(N)$. In the worst case, where every number hashes to the same bucket, the cost of sorting the bucket is $O(N \log N)$ but the worst-case time is determined by the cost of adding items to the end of the list. This, of course, is $\sum_{i=1}^{N} = O(N^2)$

Of course we could do somewhat better than this if either we could prepend to the list or maintain a form of balanced tree rather than a list with an unbound cdr. Unfortunately, either balancing a tree or prepending items to the list involves generating a new tree or list, and thus changing the appropriate entry in the hash table. But changing the appropriate entry in the hash table without copying the entire hash table (an $O(N)$ cost for every new random number, giving us a worst-case time of $O(N^2)$) requires some form of destructive assignment. It is quite easy to see that if some form of destructive assignment is employed, the worst-case time of the algorithm goes to $O(N \log N)$, which is nominal for this problem.

If we did not use var, then this hash-table algorithm would require copying the hash table in the event of a collision. This gives a worst-case time complexity of $O(N^2)$, which is the same as the implementation using var. The space complexity of the algorithm using var is $O(N)$, however, and the space complexity without var is $O(N^2))$.

5.3. III -- A Graph-Theoretic View

In order to understand why the above examples are difficult to solve efficiently in a purely applicative manner, we need an abstract view of the data structures created and used by programs. We picture a program's data structure as a dynamic graph, whose nodes are the records used to instantiate the structure and the atoms and constants in use by the program, and whose edges represent pointers to substructures. For example, the data structure `foo(1,2,3)` is represented as the graph:
Atoms and constants are nodes of outdegree 0. Unbound variables have indegree = outdegree = 1, and their sole edge points to themselves.

Only nodes with indegree = 0 are accessible at the top level of the program; we call these nodes the roots of the program's data structure.

In general, we assume that it costs no more to generate a node and set its outbound edges than it does to visit it. Hence, we are concerned with the number of visits that we must make to a node.

We may create or destroy nodes, but the nodes themselves hold no value; here we are unconcerned with the internal value of a node (in the case of an internal node, its type – e.g. the name of its functor; in the case of an atom or constant, its name or value). All we are concerned with here is reassigning the edges of the graph. If a constant field changes in a structure, we depict this by changing the edge representing the field from the old constant node to the new.

5.3.1. Principles of Modification

Now that we have a graph-theoretic model of data structures, we can turn our attention to the principles that we wish a modification operation to have. First, we have a Principle of Consistency: if a node in the graph is modified, then either the modification must be invisible to all ancestors in the graph or those ancestors must be modified to be ancestors of the modified node. Second, we have a Principle of Atomic Modification: if a clause $C$ modifies a node $N$, it should
not have to modify any other node in the graph solely to maintain the principle of consistency.

The principle of consistency is clearly just correctness in a more specific guise. The principle of atomic modification is a consequence of various principles of structured programming and language design, principally abstraction and information hiding. There is no reason to believe that an arbitrary program data structure graph is homogenous (in other words, all nodes are of the same type). Clearly, then, if a clause is forced to modify an arbitrary number of nodes in the graph, it is potentially forced to modify nodes of any type. Clearly this contradicts any reasonable definition of modularity or information hiding. Indeed, one can argue that an important consequence of the structured programming revolution is the notion that a procedure should operate on only a finite number of types, independent of the number of types defined by the entire program.

A weaker form of the principle of atomic modification may be derived on complexity grounds. In general, the number of nodes in a program's data structure graph at any time is polynomial in the size of the input. We can certainly devise programs in which the number of reassignments of graph edges is of the same order as the complexity of the program. Hence if the number of modifications a clause must make in order to maintain the principle of consistency is not bounded above by some integer \( k \geq 0 \) independent of the size of the input, then the complexity of the program will not be nominal. From these considerations we derive the Principle of Bounded Modification: if a clause \( C \) modifies a node \( N \), it should not have to modify more than \( k \) other nodes in the graph, for \( k \) an integer \( \geq 0 \), independent of the size of the input.

It is very unlikely that any modification discipline that guarantees consistency over a range of programs and data structures may violate the principle of atomic modification and nevertheless uniformly respect the principle of bounded modification. Hence it seems very likely that these two principles are in fact equivalent.

---

For a thorough discussion of such principles, see, eg, [MacLennan83]
The only form of assignment permitted by Prolog is that an unbound variable's sole edge may be assigned to point to anything (or, equivalently, the variable's node may be replaced in the graph by any subgraph). A more general form of assignment, not permitted by pure Prolog, permits edges to be reassigned once assigned.

Prolog's form of assignment raises the possibility of conflict between the principle of consistency and the principle of atomic modification. If a node $N$ is to be modified in Prolog, then the node must be regenerated, and a new node $N'$ created. All ancestor nodes to $N$ must be modified to point to $N'$ in order to maintain the principle of consistency; the principle of atomic modification forbids the procedure that generates $N'$ from modifying the ancestor nodes.

We immediately observe that there is no conflict between the two principles under Prolog's form of assignment if the program's data structure graph is a forest of trees. Let $N$ be modified by clause $C$ to $N'$. Now, either $N$ is a root or it is not. If $N$ is a root, then it has no ancestors and hence no other nodes need be regenerated in order to maintain the principle of consistency, and hence the principle of atomic modification is not violated. If $N$ is not a root, then it has a set of ancestors say $N_1, \ldots, N_k$, and the set has been traversed by a set of clauses $C_1, \ldots, C_k$, where clause $C_i$ traversed node $N_i$, $N_i$ is the parent of $N_{i+1}$ in the program's data structure graph and $C_i$ is the parent of $C_{i+1}$ in the program's proof tree (or, if you prefer, calling tree). Hence $C_i$ may generate $N_i'$, where $N_i'$ is identical to $N_i$ save that it is the parent of $N_{i+1}'$ rather than $N_{i+1}$. Since each clause modifies one and only one node in the data structure graph, the principle of atomic modification is upheld.

If the program's data structure graph contains networks or more general graphs, then the principles are in conflict indeed. The difficulty is that node $N$ in a network has several parents, only one of which is known to be an argument to a clause in the program's proof tree. In the case of a tree above, the graph could be easily modified since the set of nodes which had to be regenerated were visited in the natural course of satisfying the program's proof tree. In the case of a network, this is not the case. The principle of consistency cannot be maintained simply by
upward traversal of the program's current proof tree. Rather, the set of parents must be found by explicit traversal of the program's data structure graph and directly modified. Since this procedure is recursive, potentially the program's entire data structure graph must be immediately regenerated, which is a clear, and serious, violation of the principle of atomic modification. It is also, in general, a violation of the principle of bounded modification.

Prolog programmers therefore have three choices. First, we may use only trees or simplifications of trees (such as lists, simply a special case of a binary tree); second, we may violate the principle of atomic modification, which in practice makes many programs expensive and difficult to write; or we may choose to embrace a more general form of modification.

5.4. IV -- Requirements for a General Form of Modification

The preceding argument shows the general requirements for a general form of modification. First, any such operation must follow the two principles laid down in the preceding section. Second, such an operation permit atomic traversal of any edge in the program data structure graph. Third, values of variables and structure components form part of the state of the program at any time; backtracking restores program state, and hence must restore variable values. Therefore assignments must be undone automatically on backtrack. Fourth, fully general assignment such as Lisp's setq is not required; all that is required is some method of manipulating arguments of structures atomically.

5.4.1. Methods of Representation of Data Structures.

For obvious reasons, the method of modifying data structures is bound up in their representation. We examine three options:

5.4.1.1. Use of the Prolog Database, and Modifications using Assert/Retract

This has been a popular choice among Prolog CAD programmers [Hill85a], but we find it unsatisfactory for several reasons. First, we find that one of the strengths of Prolog is its ability to equate several variables without assigning any of them to values; an assignment to any one
therefore assigns to them all. Assert destroys such links between logical variables. Second, links between nodes in the data structure graph must be maintained through some form of keys, and the Prolog database search mechanism employed to search for the successor nodes. This search may appear to be \( O(1) \) to the programmer, but an actual \( O(1) \) search on a procedure that changes during the course of a program's execution requires an adaptive hashing scheme beyond that employed by most Prolog execution environments. On another level, the use of such keys is really a form of explicit pointers, and one of the major motivations for symbolic programming languages has historically been the desire to avoid explicit pointers. Third and most important, such modifications are not undone on backtrack, which we (and most Prolog programmers) find unacceptable.

5.4.1.2. Use of Secondary Storage Structures with Explicit Keys

In this method, rather than storing the actual pointers to successor nodes, nodes store keys and search a secondary structure which may be easily modified for the value. We have two objections to this. First, structures which may be modified easily are trees, and hence the cost of any modification is bounded below by \( \log n \), and above by \( n \). Second, the objection to explicit pointers cited above applies here. Third, additional storage structures unnecessarily complicate the code.

5.4.1.3. Use of Implicit Pointers and an Explicit Assignment Mechanism, rplacarg

We prefer to manipulate pointers implicitly, in the manner of classic Prolog and Lisp programs. In order to do this, we need an explicit mechanism to reset pointers.

The mechanism we choose is a generalization of Lisp's rplaca and rplacd mechanisms. Our mechanism, rplacarg(Term, ArgNum, Value), sets the argument ArgNum of term Term to Value. No unification is done on Term, other than to determine that it has at least ArgNum arguments.

---

See, e.g. [Bratko86]
and to determine the address of \textit{ArgNum}. The value \textit{Value} is then written into the appropriate location, and the old value and the address trail.

Notice that \textit{rplacarg} when called on a list with \textit{ArgNum} = 2 is equivalent to \textit{rplacd}; when \textit{ArgNum} = 1 it is equivalent to \textit{rplaca}.

When \textit{rplacarg} is used to manipulate edges, both the principles enumerated in the previous section are respected; consistency is maintained, since the assignment is transparent to all other nodes in the graph, and atomic modification is maintained since only one memory location (and hence only one node) is affected. Further, structures are represented naturally, without explicit indices; no secondary data structures are required, and hence pointer traversal is \(O(1)\).

5.5. V -- Implementation of Rplacarg in Quasi-Pure Prolog

Quasi-pure Prolog is Prolog code that does not use \textit{assert}, \textit{retract} or \textit{write}, but that does use \textit{cut}, \textit{fail} and other built-in meta-logical primitives such as \textit{var}. In this section, we demonstrate an implementation of \textit{rplacarg} using the \textit{var} primitive.

Conceptually, what we want to do here is permit programs written in Prolog to behave as if Prolog was a language that permitted multiple assignments, when in fact it permits only a single assignment. In order to do this, we must store rather more than the value of some component of a structure in its slot; we must store a data structure, containing at least the current value of the slot and an unbound variable; the unbound variable is reserved to be bound to future values of the component. Both an inductive view of this requirement and the need to save old values against backtracking indicate that all old values of the component must be stored in this structure.

The simplest structure which performs these tasks for us is a list, whose last element is an unbound variable and whose remaining elements are past values of the component, in order; the first element of the list is the first value of the component, and the last (but one) is the current value of the component. Accessing the current value involves traversing the list until the last bound element is reached, and returning that value; setting the current value involves traversing
the list until the last element is found, and then binding that element to a list consisting of the current value followed by an unbound variable.

To avoid semantic confusion when either unbound variables or lists become values of the component, we use an equivalent data structure, which we call a \textit{valStruct}; a \textit{valStruct} has two components, \textit{value} and \textit{futureValues}. The equivalence of a \textit{valStruct} to a list is easily seen if it is remembered that the Prolog list operator is merely syntactic sugar for the binary operator \\\n
\texttt{,} which was the list operator in early Prolog implementations.

We formalize these notions in two procedures: \textit{accessVal} and \textit{setVal}. \textit{accessVal} accesses the current value of such a nested \textit{valStruct}; \textit{setVal} sets a nested \textit{valStruct} to a new value.

\[
\text{accessVal(valStruct}(X, U), X) :- \\
\text{var}(U).
\]

\[
\text{accessVal(valStruct}(\_\_, Y), X) :- \\
\text{accessVal}(Y, X).
\]

\[
\text{setVal}(U, X) :- \\
\text{var}(U), \\
U = \text{valStruct}(X, \_\_).
\]

\[
\text{setVal(valStruct}(\_\_, Y), X) :- \\
\text{setVal}(Y, X).
\]

Once this construct is adopted it is relatively easy to write \textit{rplacarg}:

\[
\text{rplacarg(Term, ArgNum, Value)} :- \\
\text{arg}(Term, ArgNum, Arg), \\
\text{setVal}(Arg, Value).
\]

It is relatively easy to see that this implementation of \textit{rplacarg} meets our criteria; in particular, old values are restored on backtrack. It does, however, create three problems:

(1) Since components of data structures no longer contain only the value of the component, programs cannot use the unification mechanism of Prolog to examine structures directly; rather, they must use the analog to \textit{rplacarg}:
accessarg(Term, ArgNum, Value) :-
    arg(Term, ArgNum, Arg),
    accessVal(Arg, Value).

This is not a major problem for us, since we prefer access procedures and type definition code to
unification in any case: it makes modification of the definition of data structures easier during pro-
gram development. Many Prolog programmers, however, find the unification mechanism
extremely helpful.

(2) Access times can no longer be bounded by $O(1)$; rather, each access (or set) consumes time
proportional to the number of times a component is set during the course of an algorithm; of
course, this number may be proportional to the time complexity of the algorithm, though in gen-
eral it is $O(1)$. Hence this implementation can in a pathological case square the running time of
an algorithm.

(3) This method stores all old values of every component, which is extremely space-inefficient. We
shall show below that an old value need be stored only in a subset of the cases where the address
of the component would need to be stored if bound as an unbound variable. As shown by Warren
and others [Tick86] experimentally, this is only a small percentage of the cases. Hence most of the
storage used by this algorithm is garbage, and, worse, garbage that cannot be collected by most
garbage collection algorithms.

In sum, this method permits the development of programs using networked data structures
in current Prolog implementations; it also serves to show that rplacarg is no worse a corruption of
pure Prolog than var.

5.6. VI -- Implementation In a Warren Abstract Machine

The Warren Abstract Machine [WAM] is a three-stack architecture for the execution of Pro-
log. Virtually every Prolog implementation assumes some variant of the WAM, or implements
one, all the way from interpreters through dedicated hardware.
In most respects, the WAM is a conventional stack-based Von Neumann architecture. The WAM’s local stack resembles the stack on most conventional machines. The stack contains two types of data structures, environments (analogous to and closely resembling stack frames in conventional architectures), and choice points. These are required to support the non-determinacy of Prolog programs. They save the register values and form a "cap" on the stack which cannot be removed until this choice point is either exercised or removed by a cut. The second stack, the heap, is precisely analogous to the heap in Algol-60, and performs the same function. The third stack, the trail, has no analogue in non-WAM machines. Its purpose is to save the addresses of variables which have been set, so that these variables may be unset upon backtrack.

Clearly not all values need be reset upon backtrack. In particular, variable locations above the top of the heap when the last choice point was laid down will disappear on backtrack, and hence need not be reset; similarly, variables above the top choice point on the stack need not be reset. WAM architectures perform both these optimizations.

5.6.1. Structure-Copying Machines

On structure-copying machines, rplacarg is an extremely simple operation to implement. In such machines, an n-field structure takes up n+1 consecutive locations on the heap. The first location contains the functor and arity information; the remaining n contain the n arguments, in order. Hence implementing rplacarg requires only finding the base address of the structure on the heap, indexing to the argument to be written, and writing it directly; no unification is involved.

Of course, the rplacarg operation must be undone on backtrack, so if the location written must be trailed as if written originally, and its original value trailed with it. The usual optimizations apply; if this location will disappear in any case on the next backtrack, then the trailing need not be done.

The need to trail values as well as locations means that trail entries must become a pair rather than a single entry. Strictly speaking, trail entries need only be a pair if the previous entry was a value, rather than the special value unbound; however, we suspect that the penalty for
making each entry on the trail a pair rather than discriminating on this basis is too small to warrant the additional implementation complexity.

On a side note, this implementation adds some garbage to the trail. Suppose some location $k$ is written twice after some choice point has been laid down and before the next one is laid down; $k$ will be written twice on the trail, and on backtrack will have two values restored, only the second of which is at all relevant to future computation. Touati\cite{Touati86}, however, has demonstrated that it is a small matter to garbage-collect the trail.

5.6.2. Structure-Sharing Machines

Structure-sharing implementations of the WAM do not directly represent a structure on the heap in the straightforward manner of structure-copying implementations. Rather, a structure is represented on the heap by $k+1$ consecutive locations, where $k$ is the number of variables appearing in the skeleton of the structure, that is, the instance of the structure appearing at some location in the program. This practice saves some heap space when constants appear in structures in the program, since the structures' constant arguments are not copied onto the heap.

In a structure-sharing implementation\cite{Warren77}, the first of the $k+1$ heap locations contains a pointer to the skeleton in code space, and the remaining $k$ arguments provide values of the variables referenced in the skeleton. For example, the structure $\text{foo}(1, X, 2, Y)$ would be represented as:
foo(1, X, 2, Y): Structure-Sharing Implementation

_n in the diagram refers to an offset of n locations from the base of the heap entry.

Space-saving is achieved since the skeleton, which is at least as large as the heap entry, appears only once, while the heap entry is created as often as the structure based on this skeleton is instantiated. In a structure-copying implementation, the heap entry is the same size as the skeleton.

Unification is more complex in a structure-sharing environment, and for obvious reasons. rplacarg is more complex in a structure-sharing environment as well. First, the skeleton must be referenced to determine which heap location must be written. It may be that the appropriate argument in a structure-sharing environment is not a heap location (for example, arguments 1 or 3 in the above example), in which case the replacement should not take place, since the replacement would occur in every instance of this skeleton on the heap; clearly not what is desired. rplacarg must fail, or, better, signal an error.

More subtle bugs may occur in a structure sharing environment. Consider the skeleton foo(X, X). The diagram appears below
foo(X, X): Structure-Sharing Implementation

Now, suppose it was desired to replace the first argument of some instance of this structure: if it had been 1, suppose it was written to 2. replacarg would access the first argument in the skeleton, find that it was a heap offset (offset of 1), would find the location in the heap and then write it with the value 2. Subsequent accesses to the second argument would also find that its value was 2, because the first and second argument reference the same memory location.

Note, incidentally, that this bug afflicts the implementation of replacarg given in the previous section. This has not affected the programs we have written using this mechanism (including a 3000-line circuit layout package) since our data structure definition packages do not create skeletons containing either constants or repeated variables.

The solution to this bug is to attach a non-writeable protection flag (it need only be a single bit) to each argument of a skeleton in Prolog. If the flag is FALSE, then the argument can be written; if it is TRUE, then the argument cannot be written. This is not a difficult task, since the writeability of any argument is determined when the skeleton is created, and is quite easy to determine: the only writeable arguments are those which are variables which only appear once in the skeleton.
5.7. VII -- A Note Concerning Arrays

A number of array implementations have been proposed for Prolog in recent years. Most such implementations use the \texttt{assert/retract} primitives of Prolog, or propose new data areas to contain the array, or some combination of these effects.

If \texttt{rplacarg} is admitted, arrays fall quite naturally into standard Prolog as just another form of structure. The principle difficulty that people have in forming arrays is that the necessary relationship between the addresses of the various elements means that the graph of the array data structure is, in some sense, complete; each element of the array has an implicit pointer to every other element of the array. Hence any modification of any element of an array under a purely applicative model of computation requires copying the entire array, as discussed in section III. Once the applicative model is disposed of -- and in section IV we see it does not apply to Prolog, in any case -- array implementation becomes quite easy.

An array is merely a data structure with two fields -- a \textit{dope vector}, which describes how a given element may be found, and a one-dimensional vector of storage which we call a \textit{hunk}, which contains the elements. Now, under any reasonable Prolog implementation data structures will be stored contiguously in memory, so we use the built-in CProlog primitive \textit{functor}, which creates a term of arbitrary size.

We give the code to make and access arrays here. Note that arrays here are structures of four components; the fields \textit{Dimension} and \textit{DimensionVector} are included merely for error-checking.

The code is relatively straightforward, and should be easy to follow. \texttt{makeArray(DimensionVector, Array)} makes a multi-dimensional array of size indicated by \textit{DimensionVector}, which should be a list of positive integers; \texttt{accessElement(Array, IndexVector, Value)} returns the appropriate element of \textit{Array} in \textit{Value}; of course, \textit{IndexVector} should be a list.

\footnote{See, eg [Cohen84], [Touati86]}
of positive integers of the appropriate size of appropriate values. `setElement(Array, IndexVector, Value)` sets the appropriate element of `Array` to `Value`. The other routines appearing here are required for support.

The actual implementation of arrays in CProlog 1.5 was a little more complex than this, since CProlog only permits terms of size 100; readers who wish the array package should write the authors. The point of this section is merely to demonstrate that, once `rplacarg` is admitted in Prolog, then the implementation of arrays is quite natural in Prolog, and requires no other extension to the language.

```prolog
/* Code to make an array; the dimension and Dimension vectors are unnecessary; in fact, dimension is so far unused. Dimension vectors are good for error checking during access... */

makeArray(DimensionVector, array(Dimension, DimensionVector, DopeVector, Elements)) :-
    makeDopeVector(DimensionVector, Dimension, Size, DopeVector),
    allocateStorage(Size, Elements).

/* Make the dope vector for the array; the idea is to make address calculation simple...ie., if the index vector is i[1], i[2], i[3] and the dope vector is d[1], d[2], d[3], the address is:
*/

makeDopeVector([], 0, 1, []), !. /* Size of 1 is a hack for the usual case.. */

makeDopeVector([Dim | _], Dim, Dim, Dim) :-
    Dim =< 0,
    write('Error -- size =< 0 in a dimension of this array'), nl, !,
    fail.

makeDopeVector([Dim | Rest], Dimension, Size, [Size1 | DopeVec]) :-
    makeDopeVector(Rest, Dim1, Size1, DopeVec),
    Size is Dim * Size1,
    Dimension is Dim1 + 1, !.

allocateStorage(N, Storage) :-
    functor(Storage, hunk, N).

/* Dig the value of an element out */

accessElement(array(Dimension, DimensionVector, DopeVector, Elements), IndexVector, Val) :-
```
calculateArg(DopeVect, DimensionVector, IndexVector, Arg2),
Arg is Arg2 + 1,
accessarg(Arg, Elements, Val).

/* Set an Element */

setElement(array(Dimension, DimensionVector, DopeVector, Elements), IndexVector, Val) :-
calculateArg(DopeVect, DimensionVector, IndexVector, Arg2),
Arg is Arg2 + 1,
rplacarg(Arg, Elements, Val).

/* Calculate the arg (offset) of an element from its dope vector. The Index Vector is given merely for error-checking */

calculateArg([], [], [], 0) :- !.
calculateArg([], [], _, _) :- !,
    write('error -- too many dimensions in access'), nl, !,
    fail.
calculateArg(_, _, [], _) :- !,
    write('error -- too few dimensions in access'), nl,
    fail.
calculateArg(IV, RestDops, [IV | RestIV], [Index | RestIndices], Arg) :-
    (IV < Index ->
        write('Error -- Index greater than possible in one dimension'), nl,
        fail;
        calculateArg(RestDops, RestIV, RestIndices, Arg1),
        Arg is (DopeElts * (Index - 1)) + Arg1).

5.8. Integration with Type Definition

The procedures setField and accessField are obvious extensions to the code given here and in the previous section.

6. Circular Data Structures

A circular data structure is any data structure where some individual node n may be accessed through a pointer chain that begins at n. Prolog is not designed to support such structures, largely because the unification algorithm can run to exhaustion chasing the "infinite" pointer chain.
Most Prolog implementations, while theoretically forbidding such structures, do not explicitly perform a check for them: this check is known in the Prolog community as an occurs check. Occurs checks are not done since the performance of an occurs check would greatly reduce efficiency of the unification algorithm. It is virtually never argued that occurs checks should not be done since circular data structures should be entirely legal.

Nevertheless, the class of circular data structures is very broad, and includes some of the most elementary structures in computer science; in particular, doubly-linked lists, circular queues, and chained-and-threaded B-trees are all examples of circular data structures. Hence we argue that an occurs check is not merely inefficient, but contrary to the desired semantics of a complete programming language.

Even in the absence of an occurs check, Prolog implementations do not handle circular data structures well. In our case, we implemented an algorithm that manipulated circuit elements, called blocks, and their connections, called nets. It was clear that the data structure representing a net should contain a list of all blocks incident upon the net, and that each block should contain a list of all nets incident upon it. Here, clearly, is a circular data structure.

In CProlog, however, every attempt to create this structure resulted in an infinite loop in the unification routine; eventually, we gave up, and stored only the net names in the blocks, and looked up the actual nets in a balanced tree sorted by net name -- a cost of $O(\log n)$ for each pointer traversal, and exceedingly clumsy and inelegant.

We conclude that this difficulty is caused because the unification algorithm is too powerful and complex. We suspect that this difficulty will not occur in a Warren Abstract Machine, due to the lazy nature of the WAM unification instructions.

Solutions to this problem are currently being explored. An easy method is to observe that the unification algorithm continues only so long as the structures match; i.e., as long as no error has been found. Clearly if the unification algorithm proceeds to such a point that the maximum depth of any structure in the entire program space has been attained, then we have a case of two
circular, but consistent, structures.

Now, the current depth of the largest structure in the program data space is in general hard to compute, but it is certainly easy to bound above by the total size of the heap before unification began. Hence we suggest that if the unification algorithm takes this many steps, it should terminate (at least on this substructure) with success.

7. Modularity

Most Prolog implementations have a flat namespace. This is a severe problem in any symbolic programming language when programs become sufficiently large; it is a particularly severe problem in Prolog, since programmers are encouraged to write many small procedures.

Some Prologs, such as BIM-Prolog, offer a modules with specified public and private procedures. A public procedure is defined everywhere, a private procedure only within the module. The key point is that every procedure, either public or private, is defined entirely within a single module.

This paradigm is inadequate, in our judgement. Prolog’s clause-based programming encourages, as we mentioned above, a variant on object-oriented programming. Under this style, it is natural to define associate a module with each type. But a procedure under this style is made up of one clause for each type, and hence one clause per module.

Hence we suggest a third procedure type, a shared procedure. A shared procedure is visible everywhere; it is also defined everywhere. Adding a module to a program does not invalidate existing clauses of the shared procedures save those previously defined by this module.

8. Performance

Topolog can place, generate, and route a single bit of an adder in about 40 CPU seconds under CPProlog on a VAX 11/785 running 4.2 BSD. In a compiled VAX environment, we would expect to see the adder laid out in something under 5 CPU seconds. A cifplot of the adder appears in appendix one.
9. Status and Suggestions For Further Work

The relational database inherent in Prolog and the logical variable permitted us to extend a standard layout generator easily and naturally into a powerful functional-level module generator, which gives us a unique CAD tool. We found that Prolog's semantics provide the basis for a form of data-driven programming which subsumes both the functional and object-oriented paradigms.

When we began this research, we were skeptical that the logic programming paradigm was powerful enough to represent conveniently the large data structures and complex algorithms of a modern CAD system. We have discovered this initial view to be quite false; indeed, the language is powerful enough that the apparent lack of semantic structure is easily extended by procedures written in the language and its builtins. This is not true of many programming languages; for example, it is impossible in Lisp to write an efficient array package using the intrinsic data structures of Lisp. We have shown that it is easy in Prolog.

Nevertheless, our Prolog mimicry of powerful semantics is often too inefficient to be of practical benefit. Hence we are currently engaged in the process of modifying an existing Prolog interpreter to implement internally rplacarg and a partitioned namespace.

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11. Appendix -- var is implied by cut, fail

We now show that var need not be a meta-logical primitive of Prolog, but can be written using pure Prolog and the !, fail implementation of not. The idea is that a variable may be
forced to unify with two separate constants (with a failure in between unifications), and that no other construct can do this.

/* not_a(X) succeeds iff X cannot unify to a */
not_a(a) :- !, fail.
not_a(_).

/* not_b(X) succeeds iff X cannot unify to b */
not_b(b) :- !, fail.
not_b(_).

/* X is a variable iff it can unify to both a and b, ie if both not_a(X) and not_b(X) fail */
var(X) :- not(not_a(X)), not(not_b(X)).
not(X) :- X, !, fail.
not(_).

Of course, this variant of negation is somewhat controversial in the Prolog community, especially when it is applied to non-ground terms (as it is here)[Flanagan86]. However, we suspect that we could write a such a var procedure in most reasonable forms of negation; moreover, since we immediately backtrack over the bindings we make, we are not troubled by inconsistent bindings.

12. Appendix -- Implementation of Min-Cut Algorithm in Prolog  Following is the code for the min-cut algorithm, implemented using rplacarg in Prolog. We use setField and accessField as symbolic synonyms for rplacarg and accessarg.

% min-cut algorithm. Given a partition of the graph, find a new partition
% so that the cut is minimized.

min_cut(U, V, NewU, NewV) :-
turn_off_selections(U),
turn_off_selections(V),
min_cut_loop(U, V, Selections),
min_cut_move_list(Selections, Moves),
min_cut_check(Moves, U, V, NewU, NewV).
turn_off_selections([], []).  

turn_off_selections([Block | Blocks]) :-  
  setField(Block, selected, false),  
  turn_off_selections(Blocks).  

% End of algorithm, or try again?  If Moves are [], can't improve placement.

min_cut_check([], U, V, U, V) :- !.

min_cut_check(Moves, U, V, NewU, NewV) :-  
  make_moves(Moves, U, V, NewU, NewV),  

% make moves.  Looks weird, but I swear it's faster this way  O(8n) instead  
% of O(n^2).

make_moves([], U, V, NewU, NewV) :-  
  concat(U, V, L),  
  partitionOntoSides(L, NewU, NewV).

make_moves([cost(U0, V0, _) | Moves], U, V, NewU, NewV) :-  
  setField(U0, side, right),  
  setField(V0, side, left),  
  make_moves(Moves, U, V, NewU, NewV).

partitionOntoSides([], [], []).  

partitionOntoSides([Block | Blocks], [Block | Lefts], Rights) :-  
  accessField(Block, side, left),  
  !,  
  partitionOntoSides(Blocks, Lefts, Rights).

partitionOntoSides([Block | Blocks], Lefts, [Block | Rights]) :-  
  partitionOntoSides(Blocks, Lefts, Rights).

% main loop.  Trivial Cases.

min_cut_loop([], [], []).  

min_cut_loop([], [], []).  

min_cut_loop(U, V, [Selection | Selections]) :-  
  infinity(Inf),  
  min_cut_select(U, V, cost(_, _, Inf), Selection),  
  Selection = cost(U1, V1, Cost),  
  (Cost = Inf -> write('Selection unbound!'), nl, break; true),  
  setField(U1, selected, true),  
  setField(V1, selected, true),  
  delete(U, U1, Up),  
  delete(V, V1, Vp),  
  min_cut_loop(Up, Vp, Selections).
% trim the selections made by min_cut_loop down to a movelist.

min_cut_movelist(Selections, RealSelections) :-
    find_min_point(Selections, 0, 0, 0, 0, N),
    trim_selections(Selections, N, RealSelections).

% find the point where the sum is minimum.

find_min_point([], N, N, N) :- !.

find_min_point([cost(_, _, Cost)|Sels], CostIn, CurMin, LastPt, MinPt, N) :-
    ThisCost is CostIn + Cost,
    ThisPt is LastPt + 1,
    (ThisCost < CurMin ->
        find_min_point(Sels, ThisCost, ThisCost, ThisPt, ThisPt, N)
    ;
        find_min_point(Sels, ThisCost, CurMin, ThisPt, MinPt, N)
    ).

% Now trim selections, guided by N.

trim_selections([], N, []) :- !.

trim_selections([Sel|Selections], N, [Sel|RealSelections]) :-
    N1 is N - 1,
    !,
    trim_selections(Selections, N1, RealSelections).

% Inner loop for the min-cut algorithm. Select a pair to be interchanged.

% Really a double do-loop. min_cut_select is outer do -- Saux is inner do

min_cut_select([], _, CostStruct, CostStruct) :- !.

min_cut_select([U0|RestU], V, CostIn, Cost) :-
    min_cut_select_aux(V, U0, CostIn, NextCost),
    min_cut_select(RestU, V, NextCost, Cost).

min_cut_select_aux([], _, Cost, Cost) :- !.

min_cut_select_aux([V|RestV], U, cost(_, _, Cost), CostOut) :-
    computeCost(U, V, Cost1),
    Cost1 < Cost, !,
    min_cut_select_aux(RestV, U, cost(U, V, Cost1), CostOut).

min_cut_select_aux([V|RestV], U, Cost, CostOut) :-
    min_cut_select_aux(RestV, U, Cost, CostOut).

computeCost(U, V, Cost) :-
    accessField(U, nets, UNets),
    accessField(V, nets, VNets),
    ordered_set_intersection(UNets, VNets, netOrder, Nets),
set_difference(U Nets, Nets, UNets1).
set_difference(V Nets, Nets, VNets1).
computeCostIncrement(U Nets1, U, 0, CostU).
computeCostIncrement(V Nets1, V, 0, CostV).
Cost is CostU + CostV.

% successful if name of X less than name of Y

netOrder(X, Y) :-
    accessField(X, name, NameX),
    accessField(Y, name, NameY),
    X =@< Y.

computeCostIncrement([], _, Cost, Cost) :- !.

computeCostIncrement([Net | Nets], Block, CostIn, CostOut) :-
    partitionBlocks(Net, LeftBlocks, RightBlocks),
    computeIncrement(LeftBlocks, RightBlocks, Block, Inc),
    NextCost is CostIn + Inc,
    computeCostIncrement(Nets, Block, NextCost, CostOut).

partitionBlocks(Net, LeftBlocks, RightBlocks) :-
    accessField(Net, blocks, Blocks),
    splitBlocks(Blocks, LeftBlocks, RightBlocks).

splitBlocks([], [], []) :- !.

splitBlocks([Block | Blocks], LeftBlocks, RightBlocks) :-
    accessField(Block, side, Side),
    accessField(Block, selected, Selected),
    (Side = left, Selected = false; Side = right, Selected = true),
    !,
    splitBlocks(Blocks, LeftBlocks, RightBlocks).

splitBlocks([Block | Blocks], LeftBlocks, [Block | RightBlocks]) :-
    splitBlocks(Blocks, LeftBlocks, RightBlocks).

% How to compute the increment? If either side is null, block must be on the
% other side and hence moving it to this side will increase cost by 1.

computeIncrement([], _, _, 1) :- !.
computeIncrement(_, [], _, 1) :- !.

% If block is the only one on one side, moving it to the other removes this
% net from the cut. Cost decreased by 1.

computeIncrement([U], _, U, -1) :- !.
computeIncrement(_, [U], U, -1) :- !.

% Otherwise no effect on cost.

computeIncrement(_, _, 0) :- !.
13. References


[Hill85] Hill, D., *private communication*


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