# DATA TYPES AS FUNCTIONS

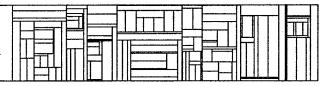
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Brian H. Mayoh

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Computer Science Department
AARHUS UNIVERSITY

Ny Munkegade – DK 8000 Aarhus C – DENMARK Telephone: 06 – 12 83 55



# Abstract

This paper introduces a new, simple definition of what a data type is. This definition gives a precise meaning one possible volution of the theoretical problems: when can one an actual parameter of type T be substituted for a formal parameter of type T'? when can a type T' be implemented as another type T'? The preprint is an extended version of a paper presented at MFCS 78, Zakopane.

#### DATA TYPES AS FUNCTIONS

This preprint is an extended version of [12]. The data types associated with equational and <u>cluster</u> specifications are described in more detail; the theory behind various techniques for verifying that one data type can be substituted for or represented as another is given in full.

### 1. THE PROBLEM

Currently there is much interest in the design of programming languages which allow:

- (1) parametrized types, the construction of new types from old [1];
- (2) operations that are polymorphic in that they have type parameters(e.g. an operation for sorting a vector of any size);
- (3) limiting the operations that can be applied to a type [2];
- (4) types that are abstract in that the type representation cannot be used outside the type declaration [3];

The first two of these raise the theoretical problem:

When can an actual parameter of type T be substituted for a formal parameter of type T!?

This is similar to, but not the same as the theoretical problem for abstract data types:

When can a type T be represented as another type T!?

In this paper we discuss these two theoretical problems, not the design of new programming languages. The literature on the design problem is extensive; those interested should begin by looking at such languages as CLU, ALPHARD, MESA, EUCLID [4] and pondering on the TINMAN requirements [5]. In the remainder of this section we give an example of

(1)-(4); in section 2 we emphasize the abstractness of all type declarations; in section 3 we give a simple, precise definition of data types and relate it to other definitions in the literature; in section 4 we propose a solution of the two theoretical problems.

The usual example of a parametrized type is STACK(EL) with stacks of values of type EL as its values. We can define polymorphic operations for such specifications as:

PUSH : STACK(EL) x EL → STACK(EL)

NEW : STACK(EL)

POP : STACK(EL) → STACK(EL)

TOP :STACK(EL) → EL

in many existing programming languages. In some of these languages we can limit the permissible operations on values of STACK(EL) to PUSH, NEW, POP and TOP. Once a limit has been put on the operations on values of STACK(EL), there may be many useful representations of this parametrized type.

The concept of a parametrized type should be distinguished from the construction of new types from types and constructors that are provided ab initio by the programming language. In PASCAL one can write

TYPE ENTRY = RECORD

identifier: ALFA;

attribute: INTEGER;

#### END

and there are no parameters in this construction of a new type from the primitive types, ALFA and INTEGER. Suppose a language allows both the type ENTRY and the parametrized type STACK(EL) with operations PUSH, NEW, POP and TOP. Then the language may well allow the declaration

TYPE SymbolTable = STACK(STACK(ENTRY))
and the definition of operations for such specifications as

BlockEntry, BlockExit: SymbolTable → SymbolTable

Initialize: SymbolTable

Extend: SymbolTable X ENTRY → SymbolTable

Find, Offset, BlockNumber: SymbolTable x ENTRY → INTEGER.

Presumably the language will also allow one to limit the operations on values of SymbolTable to the seven specified above. If so, there may be many useful representations of the type SymbolTable - some using different representations of stacks, and some not even using stacks.

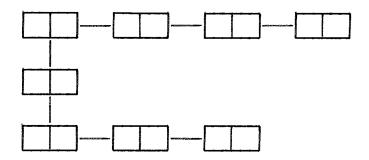


Figure 1. Values of the abstract type SymbolTable

### 2. ABSTRACT TYPES

All types are "abstract" in the sense that a programmer never knows what the values of a type really are. Suppose a language allows the definition of a type

VECTOR = ARRAY [low bound, high bound] OF REAL.

The programmer can read or write values into a variable v of type VECTOR by constructions like

```
v [index] := real value;
some real variable := ... v [index] ...;
or PUT (v, index, real value);
some real variable := ... GET (v, index) ...;
```

The programmer may think of vectors as "tuples of reals", but she would never notice if vectors had been implemented in some quite different way satisfying

```
PUT (v, index, GET (v, index)) = v;

GET (PUT (v, index, real value), index) = real value;
```

Indeed there are good reasons for allowing different implementations of vectors — if the difference between the two bounds is large, but only a few vector components exist at any one time, an intelligent implementation might use a hashing function.

Although the programmer never knows what the values of a type really are, she usually knows how she wants to use them. If her program is to manipulate rational numbers, she should be able to introduce a data type with associated operations: ADD ... EQUALS. She should be able to write a declaration like

#### cluster

RATIONAL = record

N, D: INTEGER;

# end record;

```
procedure Normalize (VAR x,y: INTEGER);

(* code to replace the values of x,y by values such that x \div y^*)

(* is unchanged but x,y have no common divisor and y > 0 *)

within
```

```
function ADD (a, b: RATIONAL): RATIONAL;
          var c, d: INTEGER;
          begin
               c := a.N * b.D + a.D * b.N; d := a.D * d.D;
               Normalize (c, d);
               ADD. N := c; ADD. D := d;
          end function;
     (* other public functions on rationals *)
     function CREATE (a, b: INTEGER): RATIONAL;
          begin
               Normalize (a, b);
               CREATE.N: = a; CREATE.D: = b;
          end function;
     function NUMERATOR(r: RATIONAL): INTEGER;
          NUMERATOR : = r.N;
     function DENOMINATOR (r, RATIONAL): INTEGER;
          DENOMINATOR: = r.D;
     function EQUALS (a, b: RATIONAL): BOOLEAN;
          EQUALS : = (a.N = b.N) AND (a.D = b.D);
(* Note that our declaration format separates private and public
                                                              * )
(* operations, allows mutually recursive definitions or operations *)
(* and permits several types to be defined in the same cluster
                                                              *)
end cluster
```

When the programmer writes a declaration, she reveals that she is thinking of rationals as "fractions in lowest terms as a pair of integers". Nevertheless the type RATIONAL is abstract in that there are many different declarations that give the same result for any computation on rationals that only uses the operations ADD ... EQUALS. An alternative declaration that avoids incessant renormalization is:

```
cluster
     RATIONAL = record
                         N, D: INTEGER;
                   end record;
     procedure Normalize (VAR x, y: INTEGER);
     (*
                as before
                                               * )
<u>wìthin</u>
     function ADD (a, b: RATIONAL): RATIONAL;
          begin
                ADD. N : = a. N \times b. D + a. D \times b. N;
                ADD.D: = a.D \times b.D;
          end function;
     (* other public functions on rationals *)
     function CREATE (a, b: INTEGER): RATIONAL;
          begin
                CREATE, N: = a; CREATE.D: = b;
          end function;
     function NUMERATOR (r: RATIONAL): INTEGER;
          var a, b: INTEGER;
          begin
                a := r.N; b := r.D; Normalize (a, b);
                NUMERATOR: = a;
          end function;
```

function DENOMINATOR (r: RATIONAL):INTEGER;

var a, b: INTEGER;

begin

a: r.N; b: = r.D; Normalize (a, b);

DENOMINATOR: = b;

end function;

function EQUALS (a, b: RATIONAL): BOOLEAN;

EQUALS : =  $(a. N \times b. D = a. D \times b. N)$ ;

end cluster

We have given this rather detailed example to emphasize the theoretical problem – when can a type declaration be replaced by another type declaration without affecting the results of computations. To solve this problem we need a precise definition of what we mean by "type". This definition ought to give precise meanings to parametrized types and polymorphic operations. As both these concepts use types as parameters, this requirement on the definition of "type" can be rephrased: when can an actual parameter type be substituted for a formal parameter type. If our concept of parametrized type allows equations, we must be careful about the meaning of equality signs if we are to avoid the horrors of collapsing types (the authors of [6] were not and one can prove TRUE = FALSE from their definition of signed integer). Neither of our declarations of the type RATIONAL satisfies

NUMERATOR(CREATE(n, d)) = n

even although this equation seems natural.

# 3. DEFINITION AND SPECIFICATION OF DATA TYPES

What is a data type? One answer is: a <u>computational rule</u> that can be applied to <u>expressions of the type</u> to yield <u>values of the type</u>. In many cases we will have:

- (1) values of the type can be stored, printed, passed as parameters ...;
- (2) among the expressions of the type we have identifiers and a constant for each value;
- (3) each value of the type is the result of applying the computational rule to some expression;
- (4) applying the computational rule to an expression always yields a value; but there is no reason to require these properties. However, there is no loss of generality if we insist on (4) in the form:

the values of a type are partially ordered and there is a least element

\_\_\_ in this partial order

and thereby avoid partial function troubles.

<u>Definition</u> A <u>data type</u> is a total function f from a set of expressions E to a partially ordered set of values V with a least element in the partial order.

For us the data type problem is not to say what data types really are, but rather to single out the particular function associated with a program language text that purports to be a data type declaration.

Many data types are <u>syntactic</u> in the sense that their expressions are given by a grammar like

Since Algol 60 the syntax of most programming languages has been given by such grammars so these languages can be regarded as syntactic data types. However, the fact that a data type is syntactic does not help in defining its computational rule, in giving the "semantics of the type". The best one can do if one has no information beyond the grammar is to define the computation rule  $f: E \rightarrow V$  by  $f(e) = \bot$  and take the singleton set consisting of  $\bot$  as V. If one has information like:

The set of values is  $\{TT, FF, \bot\}$  and the computational rule is given by the usual truth tables

then the associated data type becomes more useful.

The simplest way of giving semantic content to a data type is by domain equation [7]. Scott and others have shown how to solve equations like

 $StackR \sim \{T\} + RATIONAL \times StackR$ 

A solution of such an equation  $Z \sim T(Z)$  consists of a particular set V and functions <u>decode</u>:  $T(V) \rightarrow V$ , <u>project</u>:  $V \rightarrow T(V)$  such that <u>decode</u>  $^{\circ}$  <u>project</u> is the identity on V. A solution gives a set of values for a type  $f: E \rightarrow V$  and two functions that are useful in defining the computational rule, but one needs something else to define the set of expressions for a type. Our first declaration for the rationals gives a solution of the domain equation

RATIONAL ~ INTEGER × INTEGER

with CREATE in the rôle of "decode" and NUMERATOR  $\times$  DENOMINATOR in the rôle of "project". The solution does not have the identity as  $\frac{1}{2} \frac{1}{2} \frac{1}{$ 

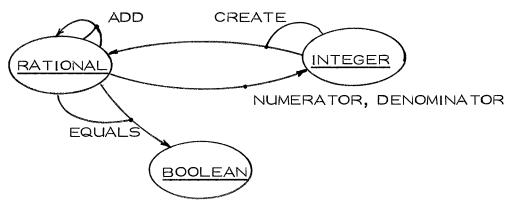
<NUMERATOR (CREATE(n, d)), DENOMINATOR(CREATE(n, d))>

On the other hand we do have the equation

CREATE (NUMERATOR(r), DENOMINATOR(r)) = r corresponding to the fact that  $\underline{\text{decode}}^{\circ}$  project is the identity.

To progress we could follow the tradition in programming language semantics (cf. syntax-driven compiler generators) and give names to the rules in a grammar. It seems more natural to follow the ADJ group [6] and introduce signatures instead of grammars. A signature  $\Sigma$  consists of a set SORT( $\Sigma$ ) and a set of operator symbols  $\Sigma_{W,S}$  for each s in SORT( $\Sigma$ ) and each sequence w of elements from SORT( $\Sigma$ ). This sounds complicated but the following alternative representations of the signature for our data type RATIONAL should make the idea clear:

#### Picture representation:



#### Structure representation:

SORT(TRADITIONAL) = {RATIONAL, BOOLEAN, INTEGER}

ADD: RATIONAL × RATIONAL → RATIONAL

. . . . . . .

CREATE: INTEGER × INTEGER → RATIONAL

NUMERATOR, DENOMINATOR: RATIONAL → INTEGER

EQUALS: RATIONAL × RATIONAL → BOOLEAN

#### Grammar representation:

<RATIONAL> ::= CREATE(<INTEGER>, <INTEGER>)

ADD(<RATIONAL>, <RATIONAL>) ...

<INTEGER> ::= NUMERATOR(<RATIONAL>)

DENOMINATOR(<RATIONAL>)

<BOOLEAN> ::= EQUALS(<RATIONAL>, <RATIONAL>)|

The grammar representation shows that the data types given by a signature  $\Sigma$  are <u>syntactic</u>. If we have a set  $E_s$  for each s in  $SORT(\Sigma)$ , then we can define  $\Sigma[E_s]$  as the set of "words" built from elements of  $E_s$  by using the operator symbols in  $\Sigma$ . If we also have functions  $\sigma_E: E_{s_1} \times E_{s_2} \times \ldots E_{s_n} \to E_{s_0}$  for each operator symbol  $\sigma$  in  $\Sigma_{s_1}$   $s_2$   $\ldots$   $s_n$ ,  $s_0$  (in other words, if we have a  $\Sigma$ -algebra), we can define a function f by

$$f(e) = \underline{if} \ e \ is \ a \ form \ \sigma[e_1, \dots, e_n]$$

$$\underline{then} \ \sigma_{\vdash}(f(e_1), \dots, (f(e_n)) \ \underline{else} \ e.$$

Now we have a data type f with  $\Sigma[E_S]$  as its set of expressions and  $\bigcup (E_S|S)$  in SORT( $\Sigma$ )) as its set of values. This data type f:  $E \to V$  is rather special, it is a "retract", it satisfies:  $V \subset E$ ,  $f^0$  f = f, f is the identity on V. An even more special type  $f_{\Sigma}$  is given by our construction when each  $E_S$  has only one element, the undefined element L of sort s.

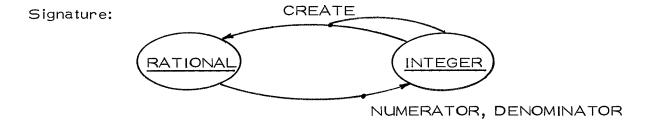
The two declarations of the type RATIONAL in the last section were disguised presentations of (ADD ... EQUALS)-algebra with carriers:

Einteger = the usual integers,

Eboolean = the usual booleans,

Erational = pairs of usual integers.

If the data type  $f: E \to V$  given by our construction is to reflect the intended representation independence of the type declarations, the "words" built from  $E_{\rm rational}$  must be removed from the expression set E. The imprecision of the last two sentences highlights a crucial difficulty in the theory of abstract data types: we need to use predefined types when declaring new types, but these declarations may introduce new elements and undesirable properties in predefined types. As always, the most undesirable property is inconsistency and this can easily occur with <u>equational</u> specifications of types given be a signature.



Equation set:

Figure 2. Equational specification of a type

This is a convenient way of specifying the data type but the current agitation about error algebras [7], maximal equivalence classes [8], and final algebras [9] shows that it is not easy to say exactly what data type is being specified by a set of equations. Suppose we have an equational specification

$$L_i = R_i$$
 for i in I

of a type given by a signature  $\Sigma$ . If we have a set  $E_s$  for each s in SORT( $\Sigma$ ), we can define an equivalence relation  $e \sim e^t$  on  $\Sigma[E_s]$ :

the equation  $e=e^t$  can be derived from the specification by substitution of elements of  $\Sigma[E_s]$  for variables and the laws of equality.

Suppose we also have functions  $\sigma_E: E_{s_1} \times E_{s_2} \times \ldots E_{s_n} \to E_{s_0}$  for each operator symbol  $\sigma$  in  $\Sigma_{s_1}$   $s_2$   $\ldots$   $s_n$ ,  $s_0$ . Suppose j is an interpretation, a map from variables to  $E_s$ . By substitution this map can be extended to words built from variables and operator symbols. The usual requirement for our  $\Sigma$ -algebra to be a model is

$$j(L_i)$$
 is the same element as  $j(R_i)$ 

for every interpretation j and every equation  $L_i = R_i$ .

Because  $e \sim e^{1} \Rightarrow$  by substituting and then applying the laws of equality we can derive  $e = e^{1}$ 

the requirement gives

$$e \sim e' \Rightarrow f_{\digamma}(e)$$
 is same element as  $f_{\digamma}(e')$ .

We generalize this.

<u>Definition</u> A <u>model</u> of an equation specification with signature  $\Sigma$  consists of a set  $E_s$  for each sort s in SORT( $\Sigma$ ) and an equivalence relation  $\Xi$  on  $\Sigma[E_s]$  such that

(COMPATIBLE) 
$$e \sim e^{t} \Rightarrow e \equiv e^{t}$$
  
 $\equiv is identity on E_{s}$ 

The model is <u>operational</u> if we also have  $e_1 \equiv e_1^1 \& \dots \& e_n \equiv e_n^1 \Rightarrow \sigma[e_1, \dots e_n] \equiv \sigma[e_1^1, \dots e_n^1]$  for all operator symbols  $\sigma$ .

Any equational specification of a type given by a signature  $\Sigma$  has the model given by  $f_{\Sigma}$  - each  $E_s$  has just one element, and distinct elements are not equivalent. If the equational specification is to give new types from old, we must have  $E_s$  for some subset OLD of SORT( $\Sigma$ ). An element e of  $\Sigma[E_s]$  is <u>determined</u> if its sort s is in OLD and there is an e¹ in  $E_s$  such that  $e \sim e¹$ .

Any equivalence relation  $\equiv$  satisfying (COMPATIBLE) agrees with  $\sim$  on determined elements, and gives a candidate for the data type specified by the equations:

the set of expressions is  $\Sigma[\mathsf{E_S}]$  the set of values is the quotient set  $\Sigma[\mathsf{E_S}]/\Xi$  the computation rule is f(e) = equivalence class of e.

There are equivalence relations satisfying (COMPATIBLE) if and only if for every e in  $\Sigma[E_S]$  there is at most one e' in  $E_S$  satisfying e $\sim$  e'. The finest of these equivalence relations is  $\sim$ ; the coarsest is  $\div$  defined by: for any two elements of the same sort s we have  $e_1 \div e_2$ , unless there is no e' in  $E_S$  such that we have one of  $e_1 \sim e'$  and  $e_2 \sim e'$  but not both. The finest operational model is given by  $\sim$ , the model given by  $\div$  may not be operational, the coarsest operational model is the final algebra of Wand [9].

Before we can explain parametrized types we must solve the theoretical problem

When can an actual parameter of type T be substituted for a formal parameter of type T!?

A partial solution is: if all formal expressions of type T' occur in E then  $f: E \to V$  can be substituted for T'; we give a complete solution in the next section.

Let us begin with polymorphic operations. Suppose we want to write a function LOOK to test if an element of an arbitrary type EL is in a vector v whose components are of type EL. Assuming "formal functions":

EQUALS : EL x EL → BOOLEAN

LowerBound, UpperBound: VEC → INTEGER

GET: VEC x INTEGER → EL

we can write the declaration:

function LOOK (v: VEC; e: EL): BOOLEAN;
var i: INTEGER;
begin LOOK : = FALSE;

for i : = LowerBound(v) to UpperBound(v)
do if EQUALS(e,GET(v,i))
then LOOK : = TRUE;

#### end function

Our declaration of the polymorphic operations LOOK does not convey semantic information directly. Assume that the formal expressions occurring in the declaration of a polymorphic operation LOOK belong to a "formal expression set  $E^{II}$ . The declaration of LOOK gives a data type, if we bind the parameters by giving some data type  $f: E \rightarrow V$ .

Consider our first cluster specification for the rationals. The functions ADD ... EQUALS are declared as polymorphic operations.

Not only does the part of the specification between <u>cluster</u> and <u>within</u> determine

 $E_{INT}$  = values of type INTEGER

 $E_{BOO}$  = values of type BOOLEAN

 $E_{RAT}$  = pairs of values of type INTEGER,

but it also binds the parameters of the polymorphic operations — it gives values to formal expressions like

$$a.N \times b.D + a.D \times b.N$$

in the declarations of ADD ... EQUALS. The binding of the parameters gives functions

and these give a data type f':  $\Sigma[E_{RAT}, E_{INT}, E_{BOO}] \rightarrow E_{RAT} + E_{INT} + E_{BOO}$  by  $f'(\sigma[e_1 \dots e_m]) = \sigma'_{E}(f'(e_1, \dots f'(e_m))$ . If the declarations of

ADD ... EQUALS had been mutually recursive, we would have used a slightly different approach:

Given a function  $f^{II}$ :  $\Sigma[E_{RAT}, E_{INT}, E_{BOO}] \rightarrow E_{RAT} + E_{INT} + E_{BOO}$  the declarations give functions for each operator symbol, and these functions can be combined into a new function  $F(f^{II})$ :

$$\Sigma[E_{RAT}, E_{INT}, E_{BOO}] \rightarrow E_{RAT} + E_{INT} + E_{BOO}$$
. The function f' is the least fix point of F.

However we define f' we can extend  ${\rm ADD'_E}\dots {\rm EQUALS'_E}$  by

$$ADD_{E}(a,b) = ADD_{E}(f'(a), f'(b))$$

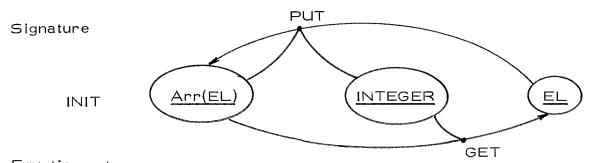
$$EQUALS_{E}(a,b) = EQUALS_{E}(f'(a), f'(b))$$

The data type f defined by the cluster specification is the restriction of f¹ to  $\Sigma[\emptyset, E_{\text{INT}}, E_{\text{BOO}}]$  where  $\emptyset$  is the empty set. The equivalence relation on  $\Sigma[\emptyset, E_{\text{INT}}, E_{\text{BOO}}]$ :

f(e) is the same element as f(e)

shows the connection with equational specifications.

Now for the declaration of parametrized types. Figure 3 shows an equational specification of the parametrized type UnboundedArray of (EL).



Equation set

GET (PUT(a,i,e) = e for 
$$i \neq j$$
 PUT (PUT(a,i,e),j,e') = PUT (PUT(a,j,e'),i,e)

Figure 3. Equational specification of a parametrized type

Suppose  $E_{INT}$  is the set of integer expressions and  $E_{EL}$  is the set of expressions of the type EL. Let  $E_{\underline{ARR}(\underline{EL})}$  be the empty set so the set of expressions for the type specified by the equations are the words built from  $E_{INT}$  and  $E_{EL}$  using PUT, GET and INIT. We can show (COMPATIBLE) for the  $\sim$  given by the equations in Figure 3. The only non-determined expressions are those of sort  $\underline{ARR}(\underline{EL})$ . All these expressions are  $\div$  -equivalent whereas

- (a)  $PUT(a, i, e_1)$  and  $PUT(a, i, e_2)$  are not equivalent in any operational model when  $e_1$  and  $e_2$  are different,
- (b) PUT(PUT(a, i,  $e_1$ ), i,  $e_2$ ) and PUT(a, i,  $e_2$ ) are equivalent in the coarsest operational model, but they are not  $\sim$  equivalent.

Since there is considerable debate about what is actually defined by an equational specification, it seems better to define this parametrized type from the functions decode and project given by a solution of the domain equation

$$Z \sim Z \times INTEGER \times EL + (T)$$

The functions for the operator symbols PUT and GET are given by

function PUT(a:ARR; i:INTEGER; e:EL):ARR;

PUT : =  $\frac{\text{decode}}{\text{decode}}$  (a, i, e);

function GET(a:ARR; i:INTEGER):EL;

case project (a) of

 $\langle a', i', e \rangle$ : if i = i' then e else GET(a', i)

otherwise undefined;

The resulting data type can be different from that given by our equations in that there is no reason why <u>decode</u> should give the same value for the arguments  $PUT(PUT(a, 1, e_1), 2, e_2)$  and  $PUT(PUT(a, 2, e_2), 1, e_1)$ .

This remains true even if the second equation is replaced by GET(PUT(a, i, e), j) = GET(a, j) for  $i \neq j$  because <u>project</u>  $\circ$  <u>decode</u> need not be the identity.

However we declare the parametrized type UnboundedArray of (EL), it can be used in such declarations as:

#### cluster

STACK(EL) = record

contents: UnboundedArray of (EL);

pointer: INTEGER;

### end record;

# within

function NEW: STACK;
 NEW.pointer: = 0;

procedure PUSH (s: STACK, e: EL);

 begin PUT (s. contents, s. pointer, e);

 s. pointer: = s. pointer + 1;

 end procedure;

procedure POP (s: STACK);

 s. pointer: = s. pointer - 1;

function TOP (s: STACK); EL;

TOP: = GET (s. contents, s. pointer);

#### end cluster;

Here NEW, PUSH, POP and TOP are polymorphic operations with formal types STACK and EL. Like LOOK they lose their polymorphism when these formal types are bound to actual types. The actual type to be bound to the formal type STACK is given between  $\underline{record}$  and  $\underline{end}$   $\underline{record}$ . As soon as we have a set  $\underline{E}_{EL}$  we get a set  $\underline{E}_{arr}$  and a data type g that gives

a meaning to the PUT and GET expressions in the cluster declaration. If we take  $E_{arr} \times E_{integer}$  as  $E_{stack}$ , we get data types:  $NEW_g$ ,  $PUSH_g$ ,  $POP_g$ ,  $TOP_g$ . These data types induce a data type for the cluster as a whole by:

$$f(\sigma[\,e_{\scriptscriptstyle 1}\,\,\ldots\,\,e_{\scriptscriptstyle m}\,]\,)\ =\ \sigma_{\scriptscriptstyle Q}(f(\,e_{\scriptscriptstyle 1}\,)\,\,\ldots\,\,f(\,e_{\scriptscriptstyle m}\,))$$

This data type should be compared with that given by the analogous construction for the declaration

# cluster

STACK(EL) = record

if not empty

then (head: EL; tail: STACK(EL))

end record; (\* see Figure 4 \*)

#### within

function NEW: STACK;

NEW. not empty: = FALSE;

function PUSH (s: STACK; e: EL): STACK;

begin

PUSH. not empty: = TRUE;

PUSH.head: = e; PUSH.tail: = s;

end function;

function POP (s: STACK): STACK;

POP : = if s. not empty then s. tail else undefined;

function TOP (s: STACK): EL;

TOP: =  $\underline{if}$  s. not empty  $\underline{then}$  s. head  $\underline{else}$   $\underline{undefined}$ ; end cluster.

If we want to allow for errors and other stack operations, this declaration is easier to modify than the equivalent equational specification

Our declaration used a recursive data type [10, 11, 1] for presenting a solution of the domain equation  $Z \sim (T) + Z \times EL$ ; it can be reformulated using the equation functions <u>decode</u> and <u>project</u> instead of record field selection.

NEW PUSH(NEW, 
$$e_0$$
) PUSH(PUSH(NEW,  $e_0$ ),  $e_1$ )
$$= \underline{\text{decode}}(T) = \underline{\text{decode}}(\underline{\text{decode}}(T), e_0) = \underline{\text{decode}}(\underline{\text{decode}}(\underline{\text{decode}}(T), e_0), e_1))$$

$$\boxed{\begin{array}{c} e_1 \\ e_0 \end{array}}$$

Figure 4. Value of Stack Type

As another example of the use of recursive data types we give a cluster specification for unbounded arrays

### cluster

UnboundedArray of (EL) = record

if not empty

then (rest: UnboundedArray of (EL)

index: INTEGER

contents: EL)

<u>within</u>

function PUT (a: ARR; i: INTEGER; e: EL): ARR

begin

PUT.not empty := FALSE;

PUT.rest := a;

PUT.index := i;

PUT.contents := e;

end

function GET (a:ARR; i: INTEGER): EL

begin

if a.index = i then a.contents

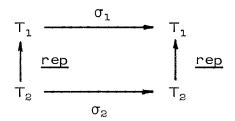
else if a.not empty then GET(a.rest, i)

else undefined

end cluster;

### 4. IMPLEMENTATION

Now that we have a precise definition of a data type, we can tackle the problem – when can one data type  $T_1$  be represented as another  $T_2$  without affecting the results of computations. The first solution of this problem was given for another definition of data types [3]. The key idea is: operations on  $T_1$  can be simulated by operations on  $T_2$ . For operations  $\sigma_1$  from  $T_1$  to T, simulation means the existence of functions  $\underline{rep}$  and  $\sigma_2$  such that the following diagram commutes



Analogous diagrams can be given for any operation on  $T_1$ .

The strong tendency to change the direction of  $\underline{rep}$  arrows in such diagrams is a sign of the subtlety of the data type problem. Much discussion was provoked by an arrow reversal in a recent description [14] of how one might generate  $\sigma_2$  automatically from  $\underline{rep}$  and  $\sigma_1$ .

How can we use the idea of simulation, if we accept the definition of data type in the last section? The natural suggestion is to say that the data type  $f_1: E_1 \to V_1$  can be represented by  $f_2: E_2 \to V_2$  if there are functions encode and decode such that the following diagram commutes:

This is not quite adequate and becomes less so if one does not resist the temptation to reverse the bottom arrow (the counterpart of the <u>rep</u> arrow in the earlier diagram). The reason why it is not adequate is that it forces:

an undefined computation remains undefined

if one type is represented by another; and this seems too stringent. It seems better to give a more general definition which reduces to the commuting of our diagram when  $f_1$  is always defined and the partial order on  $V_1$  is identity.

<u>Definition</u> A function <u>encode</u>:  $E_1 \rightarrow E_2$  represents a data type  $f_1: E_1 \rightarrow \bigvee_1$  as a data type  $f_2: E_2 \rightarrow \bigvee_2$  if and only if there is a monotone function <u>decode</u> such that

$$(1) f_1 \subseteq \underline{\text{decode}} \circ f_2 \circ \underline{\text{encode}}$$

A data type  $f_1$  can be represented as a data type  $f_2$  (abbreviated  $f_1 << f_2$ ) if there is a function encode that represente  $f_1$  as  $f_2$ .

# Lemma << is transitive and reflexive.

<u>Proof:</u> Taking the identity for both <u>encode</u> and <u>decode</u> gives reflexivity. Suppose we have <u>encode</u><sub>1</sub>:  $E_1 \rightarrow E_2$ , <u>encode</u><sub>2</sub>:  $E_2 \rightarrow E_3$ , and monotone  $\underline{\text{decode}}_1: V_2 \rightarrow V_1$ ,  $\underline{\text{decode}}_2: V_3 \rightarrow V_2$  such that

$$f_1 \subset \underline{\text{decode}}_1 \circ f_2 \circ \underline{\text{encode}}_1$$

$$f_2 \subset \underline{\text{decode}}_2 \circ f_3 \circ \underline{\text{encode}}_2$$

From the second inclusion we get

$$f_2 \circ \underline{encode}_1 \subseteq \underline{decode}_2 \circ f_3 \circ \underline{encode}_2 \circ \underline{encode}_1$$
 .

From the first inclusion and the monotonicity of decode, we get

$$\mathsf{f_1} \subseteq \underline{\mathsf{decode}}_1 \, \circ \, \underline{\mathsf{decode}}_2 \, \circ \, \, \mathsf{f_3} \, \circ \, \, \underline{\mathsf{encode}}_2 \, \circ \, \, \underline{\mathsf{encode}}_1 \, .$$

The transitivity of << is proven.

### Theorem Representations do not collapse types.

<u>Proof</u>: Suppose we have data types  $f_1: E_1 \rightarrow V_1$ ;  $f_2: E_2 \rightarrow V_2$ , and functions <u>decode</u> and <u>encode</u> satisfying (1). Suppose  $e_1$  and  $e_2$  are elements of  $E_1$  such that

$$f_1(e_1) \subseteq \vee \rightarrow f_1(e_1) = \vee$$

$$f_1(e_2) \subset \vee \rightarrow f_1(e_2) = \vee$$
.

From (1) we infer

$$f_1(e_1) = \underline{decode} \circ f_2 \circ \underline{encode} (e_1)$$

$$f_1(e_2) = \underline{decode} \circ f_2 \circ \underline{encode} (e_2)$$

If we do not have  $f_1(e_1) = f_1(e_2)$ , we cannot have  $f_2 \circ \underline{\text{encode}}(e_1) = f_2 \circ \underline{\text{encode}}(e_2)$  and our proof is complete.

How can we represent a data type  $f_1$  that is specified by a set of equations over a signature  $\Sigma$ ? Suppose we have a set  $E_S$  for each s in SORT( $\Sigma$ ). If we also have a type  $f_2: E_2 \to V_2$  and a function encode:  $\Sigma(E_S) \to E_2$ , we can define an equivalence relation  $\Xi$  on  $\Sigma[E_S]$  by

e  $\equiv$  e' if and only if  $f_2 \circ \underline{encode}$  (e) =  $f_2 \circ \underline{encode}$  (e'). In the case when  $\equiv$  is a model of the equational specification, we can define  $\underline{decode}$  by

 $\underline{\operatorname{decode}} \; (\vee) = \; \operatorname{the} \; \sim \; \operatorname{equivalence} \; \operatorname{class} \; \operatorname{containing} \; \operatorname{e} \; \\ \qquad \qquad \qquad \qquad \qquad \operatorname{for} \; \operatorname{all} \; \operatorname{e} \; \operatorname{such} \; \operatorname{that} \; \operatorname{f}_2 \; \circ \; \underline{\operatorname{encode}} \; (\operatorname{e}) = \vee \\ \\ \operatorname{and} \; \operatorname{we} \; \operatorname{have} \; \underline{\operatorname{encode}} \; \operatorname{representing} \; \operatorname{the} \; \operatorname{data} \; \operatorname{type} \; \operatorname{f}_1 \; \operatorname{specified} \; \operatorname{by} \; \operatorname{the} \\ \\ \operatorname{equations} \; \operatorname{as} \; \operatorname{f}_2 \; . \; \operatorname{Suppose} \; \operatorname{the} \; \operatorname{type} \; \operatorname{f}_2 \; : \; \operatorname{E}_2 \; \rightarrow \; \bigvee_2 \; \operatorname{is} \; \operatorname{also} \; \operatorname{specified} \; \operatorname{by} \; \operatorname{a} \; \operatorname{set} \\ \\ \operatorname{of} \; \operatorname{equations} \; \operatorname{over} \; \operatorname{signature} \; \Sigma \; . \; \text{If} \; \operatorname{we} \; \operatorname{have} \\ \\ \\$ 

e = e' can be derived from the first equation set

 $\Rightarrow$  e = e<sup>1</sup> can be derived from the second equation set then any model of the second equation set will be a model of the first equation set. By the argument above identity represents the data type for the first equation set as the data type for the second equation set. These data types exist if the second equation set is consistent:

 $e = e^{1}$  can be derived for at most one  $e^{1}$  in  $E_{s}$ .

<u>Lemma</u> For any data type  $f: E \rightarrow V$  such that f(E) = V there are data types f and  $\overline{f}$  such that

- (a) f is a retract
- (b) the value set of  $\overline{\underline{f}}$  is the quotient of  $\underline{E}$  by an equivalence relation
- (c)  $f_1 \ll f_2$  for  $f_1, f_2$  in  $\ll f$ ,  $\frac{\wedge}{f}$ ,  $\frac{1}{f}$

<u>Proof</u>: Define the equivalence relation  $\sim$  on E by

$$e \sim e^{\dagger} \Leftrightarrow f(e) = f(e^{\dagger})$$

Define  $\overline{f}$  (e) as the equivalence class containing e. Pick representatives of the equivalence classes. Define  $\widehat{f}$  (e) as the representative of  $\overline{\underline{f}}$  (e). Clearly (a) and (b) are satisfied. Take the identity on E as encode. As f(E),  $\widehat{f}(E)$  and  $\overline{\underline{f}}(E)$  are order isomorphic we have decode for each choice of  $f_1$  and  $f_2$  in (c).

Comment This lemma shows the connection between our definition of a data type, the Scott definition [7] and the algebraic definition [6]. The tedious requirement "f(E) = V" can be satisfied by "adding expressions to E" - this is most artificial when f is totally defined but we have to add an undefined expression because the partial ordering on V has a least element  $\bot$ .

<u>Lemma</u> Let  $f_1: E_1 \to V_1$  and  $f_2: E_2 \to V_2$  be data types. If  $E_1$  is a subset of  $E_2$  and  $f_1$  is the always undefined function, then  $f_1 << f_2$ .

<u>Proof:</u> Take identity as <u>encode</u> and the constant function as <u>decode</u>.  $\Box$ 

Comment In section 3 we defined a data type  $f_{\Sigma}$  for each signature  $\Sigma$ . Our lemma gives  $f_{\Sigma} << f_{E}$  for any data type  $f_{E}$  given by a  $\Sigma$ -algebra. If  $E_{S}$  is both the carrier of such an algebra and the expression set of a type  $f_{S}$ :  $E_{S} \to V_{S}$ , then we have  $f_{S} << f_{E}$  because we can take  $f_{S}$  as decode and identity as encode.

In the last section we gave a partial solution to the theoretical problem: When can an actual parameter of type T be substituted for a formal parameter of type T'? The similarity of the partial solution to the premiss of the last lemma suggests T' << T as the complete solution. If we bind an actual parameter  $f: E \rightarrow V$  to a formal parameter of type T', then the value of a formal expression e is  $f \circ \underline{encode}$  (e). Polymorphic operations like LOOK and parametrized types like STACK(EL) become data types, when formal expressions acquire values. Programming languages should only allow declarations of polymorphic operations and parametrized types that satisfy

(MONOTONICITY) 
$$f_1 \ll f_2 \rightarrow G(f_1) \ll G(f_2)$$

where G(f) is the data type given by the declaration for actual parameter f. Suppose G'(f) is the data type given by some other declaration for the same actual parameter f. If the two declarations accept the same actual parameters, and we have

(UNIFORMITY)  $G(f) \ll G'(f)$  for all actual parameters f we can prove such statements as

$$G(G(f)) \ll G'(G'(ff)).$$

In this way we can verify that such declarations as

SymbolTable = STACK(STACK(ENTRY))

do not depend on the way STACK and ENTRY are implemented.

### 5. VERIFICATION

In this section we describe three useful techniques for proving that a function encode:  $E_1 \to E_2$  represents a data type  $f_1 \colon E_1 \to V_1$  as a data type  $f_2 \colon E_2 \to V_2$ : structural induction, arrow reversal, generator induction. For most data types that occur in practice  $f_1$ ,  $f_2$  are least fix points of functionals  $F_1 \colon (E_1 \to V_1) \to (E_1 \to V_1)$ ,  $F_2 \colon (E_2 \to V_2) \to (E_2 \to V_2)$ .

We say that f is the least fix point of the functional  $F: (E \rightarrow V) \rightarrow (E \rightarrow V)$  if for every e in E we have

$$\forall$$
 n.  $\vdash$  n( $\bot$ L) e  $\subseteq$  g(e)  $\Leftrightarrow$  f(e)  $\subseteq$  g(e)

where  $\bot$  is the always undefined function. When the partial order  $\subset$  on  $\lor$  is such that directed chains have least upper bounds we can write this condition as

$$f(e) = \bigcup F^{n}(\bot\!\!\!\bot) e$$

and continuity gives F(f)e = f(e) for all e. What do we mean by continuity here? The usual definition is in terms of partial orderings on sets like  $E \rightarrow V$ . In computing practice only the following special case arises

(CONTINUITY) 
$$\forall$$
 e.  $\exists$  n.  $F^{n}(\bot\!\!\!\bot)$  e = f(e)

and we can infer this when F has the property

$$g(e) = \bot$$
 or  $F(g)(e) = g(e)$ 

for any e in E and any  $g = F^{n}(\bot\!\!\!\bot)$ .

# Example

Let A be any  $\Sigma$ -algebra. The corresponding data type is the least fix point of the functional

$$F(g)(\sigma[e_1 \ldots e_m]) = \sigma_A(e_1) \ldots g(e_m)).$$

Assume the value of  $\sigma_A$  is undefined if any of its arguments is undefined. The value  $F(\bot\!\!\!\bot)$  e is defined if and only if e has operator depth < n+1; it is the same as  $F^n(\bot\!\!\!\bot)$  e if e has operator depth < n.

Suppose we have encode:  $E_1 \rightarrow E_2$ , decode  $V_2 \rightarrow V_1$  such that

(1) 
$$F_1$$
 (decode o foencode) (e)  $\subseteq$  decode o  $F_2$  (f) o encode (e)   
 (MONOTONICITY)  $f(e) \subseteq f'(e) \Rightarrow F_1$  (f) (e)  $\subseteq F_1$  (f') (e)

From 
$$\bot = \bot_1(e) \subseteq \underline{\text{decode}} \circ \bot_2 \circ \underline{\text{encode}} (e)$$
 we can infer  $F_1^K ( \bot_1 ) e \subseteq F_1^K (\underline{\text{decode}} \circ \bot_2 \circ \underline{\text{encode}}) e$  
$$\subseteq F_1^{K-1} (\underline{\text{decode}} \circ F_2( \bot_2 ) \circ \underline{\text{encode}}) (e)$$
 .....

$$\subseteq \underline{\text{decode}} \circ F_2^K(\bot\!\!\!\bot_2) \circ \underline{\text{encode}} (e)$$

By (CONTINUITY) we have K such that

$$f_1(e) = F_1^K(\underline{\parallel}_1) e$$
  
 $f_2(e) = F_2^K(\underline{\parallel}_2) e$ 

for any e in  $E_1$ . Substitution now gives

$$f_1 \subset \underline{\text{decode}} \circ f_2 \circ \underline{\text{encode}}$$

We have proved  $f_1 << f_2$  by structural induction.

### Example

Suppose A and B are  $\Sigma$ -algebras. The corresponding data types are the least fix points of the functionals

$$F_{1}(f)(\sigma [e_{1} \ldots e_{m}]) = \sigma_{A}(f(e_{1}) \ldots f(e_{m}))$$

$$F_{2}(f)(\sigma [e_{1} \ldots e_{m}]) = \sigma_{B}(f(e_{1}) \ldots f(e_{m}))$$

Requirement (1) becomes

$$\sigma_{A}(\underline{\text{decode}} \circ f \circ \underline{\text{encode}}(e_{1}) \dots \underline{\text{decode}} \circ f \circ \underline{\text{encode}}(e_{m}))$$
 $\subset \underline{\text{decode}} \circ \sigma_{B}(f \circ \underline{\text{encode}}(e_{1}) \dots f \circ \underline{\text{encode}}(e_{m})).$ 

We can infer this from

(2)  $\sigma_{A}(\underline{\text{decode}}(b_{1}), \dots \underline{\text{decode}}(b_{m})) \subset \underline{\text{decode}} \circ \sigma_{B}(b_{1} \dots b_{m})$ Although we can also infer (2) from

 $\sigma_{A}(\underline{\text{decode}}(b_{1}), \ldots \underline{\text{decode}}(b_{m})) = \underline{\text{decode}} \circ \sigma_{B}(b_{1} \ldots b_{m})$ It is usually much easier to prove (2) directly - to require  $\underline{\text{decode}}$  to be a  $\Sigma$ -homomorphism is to require too much.

# Example

In section 2 we had two declarations of types for rationals. Let  $f_1 \colon E_1 \to V_1$  be the data type for the declaration inwhich rationals were "fractions in lowest terms", and  $f_2 \colon E_2 \to V_2$  be the data type for the declaration that avoided incessant renormalization. Define decode by

decode (v) =  $\underline{if}$  v is a rational  $\underline{then}$  Normalize(v)  $\underline{else}$  v.

Requirement (2) becomes

 $ADD_1$  (Normalize(a), Normalize(b))  $\subseteq$  Normalize  $\circ$   $ADD_2$  (a, b)

CREATE, (a,b) ⊂ Normalize ∘ CREATE, (a,b)

 $NUMERATOR_1$  (Normalize(r))  $\subseteq NUMERATOR_2$  (r)

 $DENOMINATOR_1$  (Normalize(r))  $\subseteq$  DENOMINATOR<sub>2</sub> (r)

 $EQUALS_1$  (Normalize(a), Normalize(b))  $\subseteq EQUALS_2$  (a,b)

The cluster definitions give  $V_1 = V_2 = Integer^2 + Integer$  where Integer = ( ... -1,0,1,2...). If  $a^1 = Normalize(a)$  and  $b^1 = Normalize(b)$ , there are integers k and I such that

$$a.N = k \times a!.N$$
  $a.D = k \times a!.D$ 

$$b \cdot N = I \times b^{\dagger} \cdot N$$
  $b \cdot D = I \times b^{\dagger} \cdot D$ 

Since Normalize(x,y) = Normalize( $k \times l \times x$ ,  $k \times l \times y$ ) for all integers x, y, k, l we have

Normalize(a'.N $\times$  b'.D + a'.D $\times$  b'.N, a'.D $\times$  b'.D)

= Normalize(a.N  $\times$  b.D + a.D  $\times$  b.N, a.D  $\times$  b.D) and this is ADD<sub>1</sub> (Normalize(a), Normalize(b)) = Normalize o ADD<sub>2</sub> (a,b). Because the representation of a rational in lowest terms is unique, we have

a'.N = b'.N AND a'.D = b'.D  $\Leftrightarrow$  a.N  $\times$  b.D = a.D  $\times$  b.N and this is EQUALS<sub>1</sub> (Normalize(a), Normalize(b)) = EQUALS<sub>2</sub> (a,b). The inclusions for CREATE, NUMERATOR, DENOMINATOR are

Normalize(a,b)  $\subseteq$  Normalize(a,b)

Normalize(r). $N \subseteq Normalize(r).N$ 

Normalize(r).D  $\subset$  Normalize(r).D

so they require no proof. We have verified that  $\mathbf{f_1}$  can be represented by  $\mathbf{f_2}$  .

Our next verification technique is arrow reversal: to infer  $f_{\rm l} << f_{\rm g}$  from

- (3) <u>project</u>  $\circ$   $f_1 \subseteq f_2 \circ \underline{\text{encode}}$  and  $\underline{\text{decode}} \circ \underline{\text{project}}(v) = v$ . Suppose we have  $\underline{\text{encode}} : E_1 \to E_2$ ,  $\underline{\text{project}} \lor_1 \to \lor_2$  such that
- (4)  $project \circ f(e) \subset f' \circ encode(e) \Rightarrow project \circ F_1(f)(e) \subset F_2(f') \circ encode(e)$ We can infer

$$\underline{\mathsf{project}} \circ \mathsf{F}_1^{\mathsf{k}} \left( \underline{\mathbb{I}}_1 \right) \mathsf{e} \subset \mathsf{F}_2^{\mathsf{k}} \left( \underline{\mathbb{I}}_{\mathsf{e}} \right) \circ \underline{\mathsf{encode}} (\mathsf{e})$$

from  $\underline{\text{project}}(\underline{1}) = \underline{\text{project}}(\underline{1}_1) e \subseteq \underline{1}_e \circ \underline{\text{encode}}(e)$ .

By (CONTINUITY) we have k such that

$$f_1(e) = F_1^k(\perp \!\!\!\perp) e$$

$$f_2(e) = F_2^k(\underline{\mathbb{I}})e$$

for any e in  $E_1$ . Substitution now gives

$$project \circ f_1(e) \subseteq f_2 \circ \underline{encode}(e).$$

#### Example

Suppose A and B are  $\Sigma$ -algebras. Define  $F_1$  and  $F_2$  as before. Requirement (4) becomes

$$\begin{array}{ll} \underline{project} \circ \ f(e) \subset f' \circ \ \underline{encode}(\sigma[\,e_1 \ \ldots \ e_m \,\,]\,) \\ \Rightarrow \underline{project} \circ \ \sigma_{\mbox{$A$}}(f(e_1) \ \ldots \ f(e_m \,\,)) \subset \sigma_{\mbox{$B$}}(f'(\underline{encode}(e_1) \ \ldots \ f'(\underline{encode}(e_m \,\,)). \end{array}$$

This is satisfied when each  $\sigma_{\mbox{\scriptsize p}}$  is monotone and

$$\underline{\text{project}} \circ \sigma_{\text{A}}(f(e_{_{1}}) \ldots f(e_{_{m}})) \subset \sigma_{\text{B}}(\underline{\text{project}} \circ f(e_{_{1}}) \ldots \underline{\text{project}} \circ f(e_{_{m}}))$$
 for all f,  $\sigma$ ,  $e_{_{1}} \ldots e_{_{m}}$ . Thus we can infer (4) from

(5)  $\underline{\text{project}} \circ \sigma_{A}(a_{1} \ldots a_{m}) \subseteq \sigma_{B}(\underline{\text{project}}(a_{1}) \ldots \underline{\text{project}}(a_{m}))$  or the unnecessarily strong:  $\underline{\text{project}}$  is a  $\Sigma$ -homomorphism.

# Example

Once again let  $f_1$  and  $f_2$  be the data types for the two declarations of the rationals. Define <u>project</u> as the identity, and <u>decode</u> as before. We can verify  $f_1 << f_2$  by proving <u>decode</u> o <u>project(r) = r</u> and

$$ADD_1$$
 (a,b)  $\subseteq ADD_2$  (a,b)

. . . .

 $CREATE_1 \subset CREATE_2$ ,  $NUMERATOR_1 \subset NUMERATOR_2$ ,

 $\mathsf{DENOMINATOR}_1 \subset \mathsf{DENOMINATOR}_2$ ,  $\mathsf{EQUALS}_1 \subset \mathsf{EQUALS}_2$ 

for the appropriate partial order on  $V_2$  .

We define the partial order  $\subset$  on  $V_2$  by

$$V \subseteq V' \le V$$
 and  $V'$  are the same  $V \subseteq V' = V$ 

We have to prove Normalize(r) = r and the inclusions

Normalize(a.N 
$$\times$$
 b.D + a.D  $\times$  b.N, a.D  $\times$  b.D)

$$\subset$$
 (a.N  $\times$  b.D + a.D  $\times$  b.N, a.D  $\times$  b.D)

Normalize(a', b')  $\subset$  (a', b')

r.N = Normalize(r).N

r.D = Normalize(r).D

(a.N = b.N) AND  $(a.D = b.D) = (a.N \times b.D = a.D \times b.N)$  where a, b, r are normalized. As not all pairs of integers are in  $f_1(E_1)$ , this set of inclusions is easier to prove than our earlier set.

Our third verification technique is generator induction. If we define  $project^m$  by

 $\underline{project}^{m}(a_{1} \ldots a_{m}) = < \underline{project}(a_{1}) \ldots \underline{project}(a_{m}) >$  we can write requirement (5) as

(6)  $project \circ \sigma_A \subset \sigma_B \circ project^m$ , and this is none other than (3) with  $project^m$  in the rôle of encode. If we define  $decode^m$  by

 $\underline{\text{decode}}^{\text{m}}\left(b_{1} \ldots b_{\text{m}}\right) = <\underline{\text{decode}}(b_{1}) \ldots \underline{\text{decode}}(b_{\text{m}})>$  we can write requirement (2) as

(7) 
$$\sigma_{A} \circ \underline{\text{decode}}^{\text{m}} \subset \underline{\text{decode}} \circ \sigma_{B}.$$

Since decode o project is the identity, this gives

$$\sigma_A \subseteq \underline{\text{decode}} \circ \sigma_B \circ \underline{\text{project}}^{\text{m}}$$
.

The operations  $\sigma_A$  and  $\sigma_B$  are data types in their own right and we have  $\sigma_A \subset\subset \sigma_B$  from either (6) or (7). Generator induction is our name for proving (6) or (7) by induction on the "structure of  $V_1$ ".

# Example

In the last section we had two declarations for STACK[EL]. Choose any data type as EL. Let  $f_2: E_2 \to V_2$  be the data type for the declaration in which stacks were unbounded arrays, and  $f_1: E_1 \to V_1$  be the data type for the other declaration. Define decode:  $V_2 \to V_1$ , project:  $V_1 \to V_2$ 

```
and the partial order \subseteq on V_2 by (see Fig. 5)
       project (PUSH( ... (PUSH(NEW, e_0 ), e_1 ) ... e_{p-1} )
             = < a,p>
                   where GET(a, i) = if 0 \le j < p then e_p else \perp
       project(L_{ST}) = L_{ARR}
       \underline{\text{decode}} (a,p) = PUSH(... PUSH(NEW, e_0), e_1) ... e_{p-1})
                  where e_i = GET(a, i)
       \frac{\text{decode}}{\text{decode}} \left( \bot_{ABR} \right) = \bot_{ST}
       \langle a, p \rangle \subset \langle a^{\dagger}, p^{\dagger} \rangle \langle z \rangle \langle a, p \rangle = \langle a^{\dagger}, p^{\dagger} \rangle
                OR GET(a, j) = if 0 \le j < p then GET(a', j) else \bot
Clearly we may not have project \circ decode(a,p) = \langle a,p \rangle, but we do have
decode \circ project(s) = s and project \circ decode(a,p) \subseteq \langle a,p \rangle.
\underline{\text{decode}} \ ( \ \underline{\underbrace{\text{e}_{\text{O}} \ | \ e_{\text{1}} \ \dots \ e_{\text{p}-1} \ | \ e_{\text{p}} \ | \dots }} \ ) = \text{PUSH}(\dots \text{PUSH(NEW}, e_{\text{O}}), e_{\text{1}}) \dots e_{\text{p}-1}
for i = 0, 1, ...p-1
            \leq \equiv > e_1 = d_1
```

Fig. 5 Maps between stacks and unbounded arrays

To prove that the data type  $f_{\text{\tiny 2}}$  can be represented as the data type  $f_{\text{\tiny 2}}$  , we can prove

Since  $project(NEW) = \langle \bot, 0 \rangle$ , we have the first inclusion from:

 $< \bot$ ,0> $\subseteq <$ a,p> for all a,p. Now suppose

$$s = PUSH(...PUSH(NEW, e_0), e_1)...e_{p-1})$$
 and  $a, p > = project(s)$ .

For p > 0 we have

$$\begin{array}{lll} & & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\$$

Our inclusions become

$$e_j$$
 = GET(PUT(a,p,e\_p),j) for  $0 \le j \le p$   
 $e_j$  = GET(a,j) for  $0 \le j < p-1$   
 $e_{p-1}$  = GET(a,p-1).

In the case when s is NEW, or  $\bot_{ST}$ , the inclusions are also satisfied because  $\underline{project} \circ POP_1(s)$  and  $TOP_1(s)$  are undefined.

Since we can formulate the proofs of these inclusions without information about EL, they ensure that stacks can indeed be represented by unbounded arrays.

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