## **An APL Subset Interpreter for a New Chip Set**

by

## **James Hoskin**

## **MSc (Physics), University of Calgary**

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## **Approval**

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Name. James D Hoskin

Degree: Master of Science

Examining Committee. Chairperson. Dr. W. S. Luk

> Dr. R. F. Hobson Senior Supervisor

Dr. Jay Weinkam/

Dr. R. D. Cameron External Examiner

Dr. Carl McCrosky External Examiner

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Author:

(signature)

James (Zeke) Hoskin

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## **Abstract**

The APL language provides a powerful set of functions and operators to handle dynamic array data. Current APL interpreters are hampered by excessive interpretive overhead. The Structured Architecture Machine project has produced a machine architecture (SAM) and an intermediate language (ADEL) intended to allow an APL interpreter to execute almost as quickly as good compiled code for most applications

This thesis describes a microcoded interpreter for a subset of APL It differs from earlier programs used for preliminary benchmarks and architecture evaluation in that it contains a complete memory management subsystem and a nearly complete linker and some of the tables have been moved from dedicated hardware to segmented memory

Performance measurements show that SAM APL is two orders of magnitude faster than a good microcomputer APL and suggest that the implementation behaves more like a compiled language than an interpreted one. Worst-case performance is not as good as previously predicted. Several approaches to increasing execution speed are discussed, and it is expected that the ultimate performance will meet the goals of the SAM project

The interpreter developed in this project is intended to become the nucleus of a practical APL system The final system will require code for the primitives. operators. and data types which were omitted in this version, an error handler, and input/output routines tailored to the final hardware design

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Back in the old days, in 1962

A feller named Ken lverson decided what to do. He gathered all the papers he'd been writing for a spell And put 'em in a little book and called it APL

Now, writing jots and squiggles is a mighty pleasant task But it doesn't answer questions that a lot of people ask. Ken needed an interpreter for folks who couldn't read  $So$  . . . . .

> from "APL Blossom Time" J. C. L Guest

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## **Chapter 1**

## **Introduction**

## **1.1. Interpreted Languages**

Program developers are traditionally faced with a clear-cut choice between using a compiled language or an interpreted one. Languages that support dynamic data structures are difficult to compile. Also, it is usually impossible to fix and restart a compiled program at the point where an error was detected

The major disadvantage of interpreted programs has been lack of speed. A traditional interpreter has to do almost as much work per line of code as a compiler would, and does this work every time the line is executed.

One obvious solution is to provide both an interpreter and a compiler for the same language, using the interpreter for development and then compiling the code for production [Bud81 Bud84. GuW78, Per74. Sah78, Wei79. Wgg851 Both IBM [Dr086. Chi86land STSC [PM186] have produced partial compilers for their APL systems. For practical systems, it may suffice to rewrite the most heavily-used code in a compilable language and provide a good mterface to compiled modules as part of the interpreter [Wgg86].

Compilation makes sense for programs that will be run many times unchanged. There are many cases, however, where a program may be written to solve a given problem and discarded, or when a program is undergoing modification throughout its useful life. For

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example, the interpreter developed in this project is a set of APL programs. It is being fixed whenever bugs are discovered and extended as time becomes available to implement more features. It would be a serious nuisance to face a recompilation cycle every time a change is made.

Another approach is "incremental compilation" [Van77]. A line of code is translated into machine code the first time it is encountered during the execution phase, and the machine code is retained for subsequent executions of that line. Incremental compilation introduces a significant trade-off between speed and generality The fastest possible object code for a given operation would include such details as the rank, shape, type, and location of the operands explicitly This kind of code would probably be generated by a good interpreter on any given pass through the corresponding line of source code. The more specific the code produced by an incremental compiler on the first execution of a given line, the greater is the probability that, on a subsequent execution, a property of some operand will have changed too much to allow the code to be used.

The advent of microcoded processors suggested a completely different treatment. If the primitives of the language are written in microcode, then the code generation phase of the interpretive process will be unnecessary. and execution may be speeded up enough to compensate for the time taken by the remaining semantic checking and memory management tasks [HLL73] and [Zak78] describe two interesting microcoded APL interpreters The most important example of a microcoded APL is IBM's APL Assist [HaL76] which supplements an existing APL interpreter with some microcoded routines, giving a speedup ranging from slight to a factor of 8. In general. improvements due to microcoding some or all of an interpreter on existing hardware have not given performance approaching the speed of compiled code

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Hobson [Hob82. HGT861 has chosen an architectural approach to the problem of interpretive overhead In his functionally distributed multiprocessor, one unit handles the interpretation and another does the actual data manipulation. Input/output and user communication are handled by a third processor. Early benchmarks indicate that the performance of an interpreter tailored to such a machine can be almost as good as that of compiled code running on a single-microprocessor system.

The project of developing such an architecture is well under way This thesis deals with the development of an appropriate interpreter for it

## **1.2. Languages and Data Structure**

The function of a high-level computer language is to insulate the programmer from the low-level details involved in making a lightning-idiot machine work on a practical problem. There are two main approaches to this insulation: program structure and data structure.

A "structured" language generally has built-in mechanisms for loops and branches to spare the programmer from the details of counting and jumping past alternatives, and some way of preventing unrelated chunks of code from interfering with each other. More advanced languages have ways to link separately-compiled programs. resolve forward references, and so forth

A data structure, such as a Lisp list or a SNOBOL string, is a way of representing a collection of data objects as a single primitive object. All practical computer languages support some kind of data structures, generally arrays. However, many languages, which will here be called "scalar" languages, do not operate on these structures except in a pieceby-piece manner The languages such as Lisp. SNOBOL, SETL, APL, and Nial which

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actually can operate on data structures as data structures are often called "very high-level languages", and have been found to be more straightforward to use (on problems which suit their data structures) than the scalar languages.

APL is the language chosen for this interpreter APL pioneered the use of "operators" which apply functions to arrays. That is, " $+$ " is the ordinary addition function, and "/" is the "reduction" operator such that  $+/ARRAY$  produces the sum along the rows of the array This approach is more general than simply implementing a "sum" primitive, because the operator can apply to a whole set of primitives producing "rowproduct" with the multiplication function, "rowmaximum" with the maximum function. "columnsum" when used with the "axis" operator, and so on.

The drawbacks of APL. aside from the interpretive overhead addressed by the SAM project, are its unusual character set, its lack of convivial ways to structure code. and its lack of interfaces to external devices. Some approaches to these problems are discussed in in Appendix B.

### **1.3. Previous and Related Work**

At the time this thesis project was begun, the Architecture Support Package (ASP) system already existed in the form in which it was used for developing the current version of the interpreter [Hob83]. Thornburg [Tho831 had already written a model of function call and return using an earlier version of ASP. Gudaitis [Gud85] had written a body of microcode implementing various array accessing algorithms.

The call and return protocol used in the current version of the interpreter is loosely based on Thornburg's. It is not certain to what extent the final set of execution microcode will use Gudaitis' work.

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Outside the SAM project, the most closely related body of work is the series of DELtran experiments of Flynn and Hoevel [FIH83, FIH84J. Their work combines the contour model of [Joh71] with a format-based scheme for encoding and executing instructions The resulting code can be more compact than machine code for traditional architectures, and corresponds more closely to high-level source code. Recent related work by the members of this group includes an execution architecture for Backus' FP language [HHH86] and an extension of the format-based instruction architecture to cover parse subtrees (and directed acyclic subgraphs) of different depths [FJW85].

Several other groups are doing related work McCrosky [McC85, McJ861 has designed an array-theory-based processor for the Nial language The SAM architecture is similar to the "decoupled" architectures described in [Hsi85 Lio85, SWP861

The DEL (Directly Executable Language) stream of computer architecture can be compared and contrasted with DEX (Direct Execution) of source programs [Chu79. ChA811. By working with a linearized and compact translation of the source program, a DEL implementation gets higher speed and simpler hardware at the expense of retranslation during program editing. Hobson [Hob84] defines a DIL (Directly Interpretable Language) as a linearized form from which the source code can be recovered

Another related field is the design of hardware architectures designed to support functional languages [Veg84 RuM861, production systems [Qui86], and Artificial Intelligence languages [Wah87]. The SAM architecture is not highly specialized for a particular source language and should support implementations of any of these languages more efficiently than conventional uniprocessor configurations. Since SAM'S memory architecture is designed to handle arrays efficiently, an array-based A1 language like Nial [MGJ84. JGM86] should run especially well.

## **Chapter 2**

## **The SAM Architecture**

The Structured Architecture Machine is based on the SJ16 microprocessor (See Figure 2-I), a microengine designed to be used as a microprogrammed high-level controller in a multiprocessor environment. The SJ16 has fairly conventional data paths, but the microinstructions include unique control features for high-level microprogramming. These control, features allow several microoperations to be specified and performed in parallel.

A SAMjr (Figure 2-2) system consists of an SJ16 with a segmented memory controller, a segmented memory consisting of at least one megabyte of dynamic RAM, and zero or more Special Function Units (SFUs).

A complete SAM system (Figure 2-3) consists of an Environmental Control Unit (ECU) connected to two or more SAMjr systems by dual-port memories. The ECU handles the external environment, including the user interface and mass memory. One SAMjr acts as the Program Management Unit (PMU). and is responsible for fetching instructions and maintaining the execution environment. The other SAMjr is the Data Management Unit (DMU) which does the actual data manipulations. There may be more than one DMU. The SAMjr systems are connected to each other by a custom pipeline processor which handles instruction and operand verification while queuing instructions decoded by the PMU for execution by the DMU.

In the system currently under development, the ECU is a stand-alone microcomputer



**Figure 2-1: The** SJ16 **Microengine** 

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**Figure 2-2:** A SAMjr system



**Figure 2-3:** A Complete SAM System

workstation (a NEC Advanced Personal Computer), and the only SFU is a Weitek based floating-point processor connected to the SAMjr which is acting as the DMU.

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## **2.1. The ASP Machine Model and Microcode Support Facility**

The ASP (Architecture Support Package) system consists of a set of APL workspaces which interpret, measure the performance of, and compile microprograms written in "MicroAPL", a subset of APL plus a set of SAMjr hardware commands. A MicroAPL program is both a valid APL program and a microassembler program for the SAMjr.

#### **2.1.1. lnterpretation**

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lnterpretation of MicroAPL programs is handled by an ordinary APL interpreter MicroAPL commands such as indexing, branching, and assignment are simply the corresponding APL primitives. SAMjr registers, buffers, and memories are modelled by APL arrays. SAMjr microoperations are represented by APL functions. For example, the function SSN operates on the data structures representing the SAMjr segmented memory, buffers. and registers exactly the same way the compiled microprogram "Segment Source Next" is intended to operate on the hardware devices The next value from the specified segment is placed in the specified register, and the stream buffers are inspected to see if a new memory read operation must be started

Operations which are to be performed in parallel on the hardware are written on the same line of the MicroAPL function, delimited by the  $\Delta$  symbol. The present version of ASP does not encode parallel operations except as specifically written by the programmer, and does not reject lines of code which cannot be run in parallel on the hardware.

#### **2.1.2. Measurement**



#### Figure 2-4: The PLUS Microprogram

The ASP interpreter can keep track of the number of microcycles used by a program. Every model of **a** microoperation includes a STROGEN statement which sets some flags to simulate a **tne** level pipeline and increments a clock unless the operation is on the same line of code ds the preceding operation (and therefore to be executed in parallel on the hardware). The PLUS microprogram in Figure 2-4 is a good example. Here, line [1] calls STROGEN. line [2] sets the ALUMODE flag, line [3] mimics the hardware addition and sets global RESULT, and line [4] calls SETFLAGS which checks the result and both buses for hardware conditions such as overflow and carry.

This basic. clocking scheme has been augmented in various ways. Since a memory cycle takes two machine cycles, the instructions dealing with memory access set a delay flag so that the clock will be incremented an extra tick if the result of a read operation is used before it would be physically available. Trace flags can be set to print out the names and times of functions being executed. And since the model is simply a set of APL programs. arbitrary modifications may be made at any time. At various times, code was added to keep track of ADEL instructions executed, display the contents of the pipe, or suspend execution of selected formats.

Simulation and measurement for the present version of the interpreter were deliberately

done on a version of ASP that models only the SJ16 processors, the SJMC memory controllers, the segmented memories, and the pipe. Facilities for modelling the ECU, the dual-port memories connecting the ECU to the SAMjr systems, and the floating-point processor are being developed separately.

#### **2.1.3. Multitasking**

Ideally, the ASP system should run a separate task for each processor being modelled. An earlier version ran as consecutively loaded tasks on MTS APL, requiring manual intervention every time control passed from one processor to the other. Some experiments were done using multiple tasks in Dyalog APL communicating via a UNlX pipe The version used in this part of the project works by keeping two complete copies of the ddta structures. When control passes from one processor to the other, all of the data structures pertaining to the individual processors are renamed For example, the general register array is named  $R$ : when control passes from the PMU to the DMU. R is renamed to PMUR and then DMUR is renamed to R. Since both processors use the pipeline and a set of dual-port registers. the arrays modelling the pipeline and the dual-port registers are not renamed. On every occasion when one processor must await the results of some operation on the other, its clock is set to the timing as of the end of the wait.

#### **2.1.4. Compilation**

Valid MicroAPL programs are compiled to SAMjr microcode by a compilation program contamed In another APL workspace The compiler is stricter than the simulator about . the inclusion of APL code that is not valid MicroAPL. Compilation therefore provides a useful check on the validity of microprograms.

For reasons of flexibility, the interpreter has made heavy use of named constants, which

are not included in the current definition of MicroAPL. For example. the dual-port register containing the stack pointer is named STKPTR, and only in a single initialization is it mapped to a particular integer. Therefore, the interpreter cannot be compiled in its present form. It will be a straightforward job to transform the source code into a form where such globals are hard-coded as integers, but this job has not yet been done. Possibly it will be feasible to extend the definition of MicroAPL (and alter the compiler) rather than add another step to the translation process.

#### **2.1.5. Modules**

APL does not provide a simple way to partition an application into separate chunks of code larger than individual programs. Initially, the entire model was kept in a single APL workspace, ASPSJ16. Now the memory manager, pipeline routines, executable microcode for format and operation syllables, and symbol table management functions are kept in separate workspaces.

All of these workspaces must be copied into ASPSJl6 to be executed. The only advantage to keeping them separately is that maintenance is easier. A set of utilities had to be written to keep track of where functions in the combined workspace belong, and whether they have been modified since the last time they were copied out.

### **2.2. The ADEL Intermediate Language**

The intermediate language used by the SAM APL project is the ADEL (A Directly Executable Language) developed by R. F. Hobson [Hob82, Hob84]. ADEL is a format-based language related to the FORTRAN DEL of Flynn and Hoevel [FIH83, Fly80]. An ADEL instruction consists of a "format" syllable which specifies the valence of the function, zero or more operand syllables. and zero or more function syllables. For example:

$$
X \leftarrow Y + Z
$$
  
becomes  
DLR X Y Z +  
and  

$$
A \leftarrow B + X C \div D
$$
  
becomes  
SLR C D ÷  
DLS A B + X

APL code is translated into ADEL in a context which includes a local symbol table. If Y is the 16th identifier used in a given function, then the operand syllable for Y would contain the value 16. All syllables are eight-bit bytes, so a given user-defined function 'is restricted to using 255 identifiers, excluding comments and primitives. Note that this does not restrict the number of global symbols. as long as no particular function uses more than 255 of them.



#### Table 2-1: ADEL Formats as of 1984

(See Table 2-2 for Meanings of Characters)



## Table 2-2: Meaning of Format Characters

The ADEL formats defined in [Hob841 are given in Table 2-1. The characters used to describe the formats are defined in Table 2-2. Note that the set of formats is not rigid. Several formats have been added during the development of the front-end translator, and others were added to handle data transfer and program intended to be tailored to fit particular applications. package might reserve a special format for the  $+=\,$  inner product, and a language The set of formats is For example, a crosstabulation extension to include one of the nested array systems would have different operator formats.

Not all of the formats recognized by the translator have been implemented. Table 2-3 contains a list of implemented formats Since the formats have been implemented as APL functions, it has been necessary to rename them.



**Table 2-3:** 

Formats Implemented as of January 1987

## **Chapter 3**

## **The Memory Manager**

The first four months of this software project were devoted to developing the memory management subsystem. About five percent of the subsequent work has consisted of keeping track of the information required for the memory manager to allocate space for results and dear out the space used by reassigned variables, popped stack values and expired immediate execution expressions.

Since APL. objects are dynamically sized and APL functions are recursive by definition. the execution of an APL program involves allocating and freeing space for arrays both during function call/return and during the execution of primitive functions. The management of memory in the PMU is somewhat less critical, since functions are generally created under immediate execution and are seldom erased. For the sake of consistency, the same memory manager is used by both processors.

## **3.1. Background**

Memory managers are an important part of multiprogramming operating systems and dynamic language execution environments. As such, they are the subject of a considerable body of theoretical treatment. The descriptions of the basic problem, paged and segmented memories, and fitting strategies are based on [PeS83, Dei841.

#### **3.1.1. The Basic Memory Management Problem**

An operating system may allow programs to be relocateable, thus allowing memory to be partitioned among several programs. Programs can easily be swapped in and out of fixedsized partitions. Allowing variable partition sizes raises the basic memory management problem of allocating space in the most efficient way. Memory may become "fragmented", so that free memory is distributed among a set of noncontiguous blocks. It then becomes necessary to choose which free block will be used for a given job. and it may also be necessary to relocate existing jobs in order to merge free blocks.

#### **3.1.2. Paged Memory**

One powerful way to deal with the allocation problem is to waive the requirement that a program or a data object be contained in a contiguous chunk of memory. If memory is divided into "pages", and logical addresses consist of a page number and an offset, then a page table can be used to map page numbers into real memory addresses.

The principal differences in simple paged memory systems concern the implementation of the page table: if kept in registers, the table is of limited size; and if kept in main memory, it slows each memory access by a factor of two. The slowdown can be worse in a pipelined machine, since the second memory access cannot be overlapped with the first. A compromise solution is to keep a subsection of the page table in contentaddressable registers The SAM architecture keeps the entire table in high-speed static memory

#### **3.1.3. Segmented Memory**

In a paged system, pages are of fixed size, so that large objects may occupy several pages and several small objects may share a single page. The segmented memory scheme partitions physical memory into variably-sized segments, with one object occupying one segment. While segmented memory is a more natural way to deal with objects of varying size, it is naturally subject to fragmentation and requires the same kind of allocation scheme as variably sized partitions in a job scheduler

A segmented memory may be paged, so that a segment need not be composed of contiguous real memory. This usually faces all memory accesses with two levels of indirection, and must be implemented with registers or cache to give reasonable performance. A paged segmented memory can be implemented with a single level of indirection by keeping physical page addresses in the segment table. This is the approach used by the SAM memory controller. The only major drawback to one- level paged segmented memory is that it makes it somewhat more difficult to implement a virtual memory that swaps individual pages.

#### **3.1.4. Fitting Strategies**

Given a segmented memory or a variable-partition-size swapping scheme, there will be a list of free blocks to be considered when a new block is to be allocated. The two bestknown methods of choosing a free block are first-fit (choosing the first free block big enough to satisfy the request) and best-fit (searching the whole list and choosing the smallest free block big enough to satisfy the request).

In order to make a best-fit strategy perform well, it may be necessary to keep the free block list in a structure amenable to a binary search. If the free list is kept sorted in increasing order of size, the two strategies become identical.

Since searching time grows with list size, it may be advantageous to garbage collect before an unfillable allocation request arrives For example, the implementers of CDC APL\*STAR found that limiting the "Hole Table" to 16 entries caused the proportion of garbage collects triggered by hole table overflow to be much smaller than the proportion triggered by memory fragmentation [Edw87]. Another approach is to ignore freed blocks completely unless they can be merged with the "one big hole". speeding up typical allocations at the expense of more frequent garbage collection.

#### **3.1.5. Buddy systems**

It is possible to eliminate the linear search of free block lists, by keeping a set of lists of free blocks of various sizes and looking only in the appropriate list for a given allocation request Such a system was developed by Knowlton [Kno65] and has since been the subject of continual investigation [PuS70] and development [PaH86].

The basic idea of these "Buddy Systems" is that only a fixed set of block sizes is available. If no free block exists in the size requested, then a bigger block is split, with the left-over part ("buddy") put into the appropriate free block list. When a block is deallocated and its "buddy" is still free. the two blocks may be merged. Buddy systems thus avoid the overhead of list searches, at the expense of block splits and recombinations and of a considerable amount of memory fragmentation, both "internal" (due to the coarseness of the set of block sizes available) and "external" (since the splitting process may leave the free memory as a set of blocks too small to satisfy an allocation request).

The total fragmentation of a buddy system is typically thirty or forty percent [PeN77] of total memory Gwen a demand for quick allocation and the availability of cheap memory, this overhead is tolerable as long as the buddy system outperforms the more memoryefficient alternatives.

## **3.2. The SAM Segmented Memory Model**

Any memory management scheme must reflect the hardware support available. For instance, a demand paging system would be awkward to implement on a machine with no page fault interrupt mechanism. The memory manager developed here is based on the paged segmented memory controller forming part of the SJ16 architecture. (See Figure  $3-1.$ 

The most important feature of the model (See Figure **3-2)** is the address translation table called the SPT, for Segment Page Table A logical address consists of a Segment number and an Offset. The logical page number is found by adding the segment number to the integer part of (Offset  $\div$  Pagesize), and the physical page is given by SPT[Logical page] The rest of the offset is a byte index into the physical page. A data structure may be distributed in widely scattered physical pages, as long as these pages are pointed to by a contiguous block of the SPT. While it would be possible to keep more than one small object in a single page of segmented memory. "workspace full" problems in existing APL systems are almost always caused by a small number of large objects. In the SAM system, one thousand small arrays would use one thousand pages, leaving at least three thousand pages for larger arrays

The result of this policy is that no program "cares" where in physical memory a given object resides When garbage collection is necessary, it is not because physical pages are misplaced but because there is not a large enough contiguous block of free pointers in the SPT and the corrective measure involves moving single pointers instead of complete pages of data.

While this memory scheme is in fact a paged segmented memory, the fact that all the



#### Figure 3-1: The SJMC Memory Controller



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**Figure 3-2:** The SAM Memory Model

pages of a given segment are kept in a contiguous block of the SPT allows memory accesses with one level of indirection instead of two, at the expense of the page table garbage collect described above. Keeping the page table in fast static memory would slow each access slightly compared to the access time for a "hit" where a subset of the page table **15** kept in registers However, in the SAMJr Memory Controller, this address translation can be done during dynamic RAM precharge time, so no performance is lost. This scheme gives the performance of a register-based page table where every access is a "hit" without the special hardware and complex microcode required to deal with "misses".

The other interesting feature of the memory model is that segmented memory is streamed. Once a given address in segmented memory is accessed, the value stored at the next location is read into a buffer and is available to the processor in advance. Memory writes are done to a buffer which is flushed into memory in parallel with other operations. This policy accelerates string and array processing tasks. The tradeoff for any purely array-oriented memory is that scalar accesses require more code than the corresponding accesses on a scalar machine (See Figure 3-3)

The streaming of scalars is handled in microcode by doing the preliminary memory control operations in parallel with other operations. This allows the system to access scalars efficiently without adding hardware and extending the instruction set to include conventional scalar operations



Figure 3-3: Streamed Segmented Memory Access

## **3.3. The Present Model**

The memory manager used by the present interpreter is very nearly the simplest possible manager using the architecture support available. The memory manager table (see Figure 3-4) consists of one doubly linked list of free blocks, with each list item containing a block size and each allocated item containing a pointer back to the main data access table. Allocation begins with a first-fit search of the free block list The block is then either shrunk (if it is more than big enough) or removed from the list (if it is an exact fit). ' Adjacent free blocks are merged immediately at deallocation time. Garbage collection is tedious but reasonably fast. since only SPT items must be moved. A garbage collect is terminated as soon as a free block big enough to satisfy the current allocation request has been created

A commented call tree of the implemented memory management functions is given in Figure 3-5.

It seems likely that garbage collection can be made to occur arbitrarily seldom, by keeping a table somewhat bigger than the actual size of the memory. A practical measurement of this prediction has been deferred.

### **3.4. The Production Model**

Before the microcode was written for the current version of the memory manager, a slightly more complicated manager was written in APL. The production memory manager .may be based on this earlier version which used two linked lists instead of one. Free blocks containing a single page are kept in one list, and multiple-page blocks in the other. Since (again pending measurements) the large majority of data structures will fit into single pages, it is expected that this system will outperform more complex systems such



**Figure 3-4: Memory Management Tables**


#### **Figure 3-5:**  Memory Manager - Commented Call Tree

as a buddy system. Eliminating the common one-page blocks from the searches for larger blocks should improve search times by a factor of better than two and reduce fragmentation

### **3.5. Theoretical Predictions**

While a realistic appraisal of performance must await measurements made with real programs running on a completed implementation, it is not difficult to give a rough but reasonable theoretical analysis of various allocation strategies. In this section, we look at the current single-list strategy. the projected two-list strategy, and a simple binary buddy system. For the sake of simplicity, we will consider only the costs of allocation.

Consider only requests and free blocks larger than a single page. Then the number of blocks searched for a typical request depends on the distribution of sizes in both the free block list and the set of requests Models based on various distributions gave expected search lengths of 3.5 to 6.0 iterations with an infinitely long free block list and from 30 to 3 5 with the free block list limited to 16 elements For the sake of simplicity, an expected search of 3.5 iterations will be assumed for the rest of this analysis.

Let Pm be the probability of an allocation request being for more than one block. The. distribution of free block sizes may differ from the distribution of allocation request sizes due to merges (which lead to larger blocks) and partial allocations (which give smaller free blocks). In particular, the deallocation of a single-page object gives a single-page free block only when neither of its neighbours are free. For the sake of simplicity. assume that this happens about half the time, so that the proportion of multipage free blocks will be 1 - $0.5(1 - Pm)$ , or  $0.5(Pm + 1)$ .

#### **3.5.1. Single Free Block List**

Using a single list for all free blocks, a single-page allocation will be satisfied by the first block on the list A rnultipage allocation will search about 3.5 multipage blocks. Since  $0.5(Pm + 1)$  of the free blocks are multipage blocks, the allocation will search  $7/(Pm + 1)$  blocks. Then a proportion Pm of the allocation requests require  $7/(Pm + 1)$  search iterations, and  $(1-Pm)$  of the requests require 1 iteration, giving an expected average of

1

 $(1 - Pm) + 7Pm/(Pm + 1)$  search iterations.

#### **3.5.2. Two Free Block Lists**

Using a pair of free block lists all allocations require a block size check to determine which list is to be searched. Then the multipage allocation requests will search 3.5 free blocks, giving an expected performance of 1 check and  $(1-Pm) + 3.5Pm$  or  $(1 + 2.5Pm)$ search iterations

#### **3.5.3. Buddy System**

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Assume that freed blocks are NOT merged if doing so would give an empty list for a given block size. ' An allocation will require (log N) checks to choose among N different block sizes by a binary search. (Block sizes are distributed logarithmically, so in principle the correct list can be found by calculation. Unless the hardware supports binary logarithms, it is quicker to do a depth-three binary search.)

For a SAM APL .interpreter. N would be about 9, with the checking biased so the first --, check is'for a single block. A multipage allocation will take 3 additional checks to choose among the 8 larger block sizes. Then the average number of checks is  $1 + 3Pm$  at allocation time. One additional check must be made to ensure that the required list is not empty, giving 2+3Pm checks.

There will be a probability Pe that the list is empty. requiring a split. The list to be split may itself be empty, leading to an expected number of splits of  $Pe/(1-Pe)$  for a given allocation. The value of Pe is dependent on the recombination strategy. For the purposes of comparison, values of 0.1 and 0.25 will be used.

#### **3.5.4. Comparing the Systems**

A search iteration requires one segment startup, two reads, and a comparison, taking altogether about four cycles. (The microcode used in this search is listed in Figure 6-6) A simple comparison requires a test and a branch, costing at least two cycles. Then the .average cost of finding a free block to be allocated is shown in Table 3-1 A rough pass at microcoding a split gives about 10 cycles if all list headers are kept in registers. or 16 cycles if headers are kept in memory



Note: Buddy1 is  $Pe = 0.1$ . Buddy25 is  $Pe = 0.25$ 

Table 3-1: Cost of Finding a Free Block

This simplified analysis suggests that the one-list model may perform as well as the two-list version as long as the proportion of multi-page allocation requests is not more than thirty percent. The buddy system does not outperform the linked-list systems unless the proportion of empty lists can be kept to ten percent.

## **Chapter 4**

## **Symbol Table Management**

The three main management tasks of an APL interpreter are to store APL arrays, to provide a convenient way for executing APL code to access arrays and functions, and to provide a way for a user to get arrays and functions in and out of the system. Each of these tasks is organized around its own table: the Data Access Table for array storage. the Contour Access Table for access by executing functions, and the Global Symbol Table for user access

### **4.1. The Linker**

The linker is responsible for accepting ADEL functions from the front-end system, writing them into segmented memory, and installing the identifiers used by the functions into the Global Symbol Table.

#### **4.1.1. The User Interface**

The user interface to SAM APL is a text editor and translator which run on the workstation that serves as the ECU. To write an APL program, a user edits the text form and mvokes the TRANSLATE facility. This produces the ADEL code corresponding to the APL source, plus a local symbol table and enough information to link it.

In a complete SAM system, this internal function form (See Figure 4-1) is then written into dual port memory by the ECU, to be accessed by the linking routine in the PMU. At



**Figure 4-1: Function Internal Form** 

the current stage of the model, an ADEL assembler on the computer system supporting the ASP model is used to create an APL variable containing the internal function form. Some of the ADEL functions were written by hand instead of generated by translation of source code. This was done because some of the ADEL formats developed during this project are not yet within the scope of the translator. These include the special branching and looping primitives and the simplified indexed assignment format.

Once the ADEL function has been created, the LlNK program places it into segmented memory, creates symbol table entries for all of the objects in the program, and creates data arrays for all of the nonscalar literals in the program The Global Symbol Access Table  $(SSAT)$  shown in the diagram for the internal function form is created at link time and added to the function segment by the linker.

To run a line of APL immediate execution (and therefore, to start an APL program), the assembler is used to create a dummy function which is flagged as an immediate execution line in the "Type" field. LlNK invokes the execution manager after linking an immediate execution "function", then purges the function from segmented memory after execution is complete A list of the programs used by the linker is shown in Figure 4-2.

The present procedure for testing APL functions is necessary while the editor and the rest of the model run on different computers without a convenient way of transferring data. In the working SAM system, the front-end computer will place the translated function into dual port memory and the LlNK program will read dual-port memory instead of taking an argument

Changes needed to adapt the LlNK program to the finished hardware are minimal, so performance, reliablility, and space figures for the production version of the linker are expected to resemble those of this model very closely

LINK BRINGINFN RUN-SAM STLINKFN CHKNUMV STLINKHDR STLINKFNAME FINDSTPAGE<br>\*\*\*\* HASH\_GNRL STENTWR STLINKID \* \* \* STLINKID / CHKNUMV FTNDSTPAGE GETCONFPG STMATCHID HASH-GNRL HASH-IDL HASHMODULO HASH-NUML HASHMODULO HASH QUOTE HASHMODULO SENDTODMU STENTWR

HALF THE SAM FRONT END NOT uCODE.READ FN INTO PROG SEG THE BOSS PROGRAM FOR IFETCH/IEXEC LINK PGM IN PROG SEG TO S.T. PROG IS AT START OF NUM LIT BIND OBJS IN FN HDR FN NAME UNLIKE OTHER OBJS S.T. PAGE IS NOT EMPTY HASH ID, LBL, LITERAL WRITE S.T. ENTRY LINK 1 ID INTO S. T. LINK 1 ID INTO S. T. PROG IS AT START OF NUM LIT S.T. PAGE IS NOT EMPTY GET NEW S.T.CONFLICT PAGE  $ID$   $AT(PROG, IDOFF) = (STS1, 14)?$ HASH ID, LBL, LITERAL HASH ID OR LBL HASH NUM LITERAL HASH QUOTED LITERAL SEND LITERAL TO DMU WRITE S.T. ENTRY The Linker - Commented Call Tree

### **4.2. The Global Symbol Table**

**Figure 4-2:** 

#### **4.2.1. Symbol Table Pages**

Every identifier in an internal-form function corresponds to an entry in the global symbol table. A sample symbol table entry is shown in Figure **4-3.** 

The linking process for each identifier consists of hashing the identifier and checking the corresponding symbol table entry. If it is free, the identifier is written into it; if it contains



an identical identifier, linking is complete; and if it contains a different identifier, the collision chain is followed until either an identical identifier is found or the end of the chain is reached. If no match is found, the identifier is written into an overflow symbol table page and the new page is linked onto the collision chain.

Once a symbol table entry has been found or created, a pointer to it is written in the internal function segment.

In the usual state of affairs for linking, the system is in immediate execution with no suspended programs. The call stack pointer in the symbol table chain will be pointing to the DMU polnter for the global value, if any. The procedure for function call will be discussed in the following chapter. For now it is sufficient to mention that the linking process is not affected by the state of the call chain. Of course, a newly linked identifier will be tied to the DMU item containing the current value.

#### **4.2.2. Literals**

When a numeric or character literal is found in a source program, it becomes an identifier in the ADEL code, mapping to a symbol table entry that can not be hidden by function call. Thus. if several programs happen to use the value 3.6, they will all map to the same symbol table entry. When a literal is linked. its value is sent immediately to the DMU, and a DMU item pointer for the literal value is written immediately into the symbol table page.

'The exception to this is the integer scalars **0** through 255. The first 256 DAT entries are reserved for these values. so the DAT pointer can be written into the symbol table page kvithout transmitting a value to the DMU. A few special values like pi and e are kept in known DAT locations for the use of system code. and future version of the interpreter may recognize the equivalent APL expressions and treat them as literals.

Carrying this idea to its logical conclusion, it is possible to execute all expressions involvmg no variables at link time and link the results as literals. Statistics on the frequency of execution of such "constant expressions" must be gathered before any decision is made about implementing them as literals.

Ordinarily the symbol table page contains the character string representation of its identifier In the case of literals, this representation may be too long to fit into a page. When the string length is greater that 40 characters. the "id" section of the symbol table page contains a pointer back to the local symbol table of the function owning the literal. Therefore, if two separate functions happen to have identical long literals, each will have a separate symbol table entry

Most identifiers are quite short, and since a symbol table page has room for a call chain which will never exist in the case of a literal, a certain amount of space is wasted. Future versions of the interpreter may keep short literals in smaller symbol table pages.

### **4.2.3. Symbol Table Garbage Collect**

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Except in the case of long literals, symbol table entries do not contain backpointers to the functions that use them This makes it difficult to get rid of symbol table entries when their referents no longer exist. Older APL interpreters (MTS APL is a good example) never attempt to clean up the symbol table: when the symbol table is full. the only recourse is to )COPY everything into a clear workspace. An easy but slow way to deal with a full symbol table is to relink every function and global variable, essentially )COPYing the workspace to itself.

**Tu** allow symbol table entries to be deleted efficiently, a reference counter could be included in each symbol table page. Linking an identifier to an existing page would increment the reference count, and expunging a defined function would decrement the reference counts of all of its identifiers, with a page marked as free when its reference count drops to zero This idea will be investigated when the standard interpreter has been completed.

The collision chain pointers in each symbol table page show the predecessor and successor in a chain of symbol table pages corresponding to identifiers that hash to the same value The forward pointer is used only when linking a new identifier The back pomter was intended to be used during symbol table garbage collection. Keeping the collision chain as a circular list makes back pointers unnecessary, and the field may be used for a reference count in the next version.

#### **4.2.4. Labels**

Function labels are a special case for linking and calling. Unlike ordinary identifiers. they have values and unlike ordinary literals. they have names that may be localized by subsequent function calls. At link time, labels are linked to symbol pages exactly the same way as ordinary identifiers, and their status as labels is indicated by setting the sign bit of the symbol table page address. The item in a function segment following a negative symbol table pointer contains the label value instead of the pointer for the next identifier.

## **4.3. Related Environment Tables**

The environment of an executing program is partly under direct user control but principally a function of the state of execution. In a lexically scoped language, the environment is determined by the block structure of the main program. In a dynamically scoped language like APL, the availability of an identifier is controlled entirely by the most recently called function which localizes the identifier Thus, this section **must** be read **in**  conjunction with the treatment of the CALL and RETURN mechanisms in the following chapters.

#### **4.3.1. The Data Access Table**

Data manipulation is done in the DMU. Each data item, whether or not it is available to the current scope of the program. is represented by an entry in the Data Access Table, or DAT, kept in DMU segmented memory. Scalars and array descriptors are contained in the DAT entry The entire descriptor of a vector fits into the DAT entry. The length fields for the trailing axes of higher-rank arrays, and the element values of all nonempty arrays. are kept in DMU segmented memory. each segment pointed to by its DAT entry When code is actually executed in the DMU, the DAT entries are used by the data manipulation routines.

A physically distinct tag memory contains 4-bit type and shape tags corresponding to the entries in the DAT

As it is currently laid out, the first 256 DAT entries are reserved for the scalar constants 0 through 255. Then a few entries are reserved for constants like " $\pi$ ", "e", and pointers to read-only segments containing the character set and the results of  $\ell$ 256 in both index origins. The next DAT entry is a scratch location indexed by system constant NEWDEST. used to write the result of each primitive operation before erasing the previous value of the identifier being assigned.

The rest of the DAT is divided into two stacks. The stack growing forward from NEWDEST is pointed to by NEXTCG (Next Constant or Global) and contains constants and global values that are never erased except by explicit request. The stack growing backward from the end of the DAT is pointed to by STKPTR, and contains values on the execution stack, iocai variabies, and the arguments and results of functions. NEXTCG and STKPTR are kept as mailbox registers in the pipe processor because both the PMU and the DMU must use them

#### **4.3.2. The Contour Access Table**

The Contour Access Table (CAT) for a given environment consists of a list of words containing syntax tags (see Section 5.2.1) and DAT pointers. When ADEL code is executing each operand syllable is used as a word index into the current segment of the CAT

The relationship among internal function forms, the symbol table. the DAT. and the CAT is diagrammed in Figure 4-4



**Figure 4-4: Relationship Among Interpreter Tables** 

## **Chapter 5**

### **Execution**

The paradigm of user interaction with an interpreted language is the Lisp Read-Eval-Print cycle. APL follows this cycle closely, with the important restriction that the user interaction is built into the system instead of being a user-defined program and, as such, modifiable.

SAM APL user interface is similar to the paradigm. The basic cycle is texteditexecute-print. The details of the editing and printing parts of the cycle are outside the scope of this thesis. The execution part of the cycle begins with linking a line of code as described in the previous chapter. The present chapter deals with the execution of ADEL code once it has been put into program memory and its identifiers have been put into the Global Symbol Table

### **5.1. Instruction Fetch**

The PMU half of program execution is done by the IFETCH routine. An instruction fetch begins by reading the format syllable of an ADEL instruction. As a side effect, the format is sent to the pipeline. Then the PMU EXECs the format syllable -- that is. executes the microprogram beginning at the address given by the format table indexed by the format. This is equivalent to an instruction decode operation in a standard commercial computer. In the model, this consists of executing the MicroAPL program whose name is in the indexed row of the format table

Execution of the format microprogram causes the operand and function syllables to be fetched Syllables denoting primitive APL functions (and also some "hidden primitives" such as  $pp$  and  $+/$ ) are sent directly to the pipeline. Operand syllables are used as word indices into the Contour Access Table. The indexed words, which consist of syntax tags and data pointers, are sent to the pipeline. At the end of the format microprogram. the pipe is released if no errors have been detected and the process is repeated as soon as a new pipe becomes available

Syllables representing user-defined programs are not sent to the pipeline. The PMU microprograms for function call pass the CAT entries indexed by the argument syllables to the pipeline, then do the processing necessary to preserve state information for the current environment and set up a new environment. The details of the CALL and RETURN procedures are described in section 5.4.

### **5.2. Instruction and Operand Verification**

One of the most important features of the SAM architecture is the way instruction verification(syntax checking) and operand **verification(pre1iminary** semantic checking) are handled in hardware. A conceptual view of the verification process is shown in Figure **5-1.**  Each operand is represented in ADEL code as an index into a frame of the Contour Access Table (CAT). This index is conceptually added to a frame pointer (in the current implementation, each frame begins on a page boundary, so the addition is done by the addressing hardware) and the resulting tag/data pointer is read from the CAT. The tags are then used in the instruction verification process and the data pointers are sent to the pipe.

In the DMU, the data pointers are used to index a Data Access Table (DAT). The



**Figure 5-1: Conceptual Diagram of Verification Process** 

DAT **contains data tags, and either the values (of scalars) or partial descriptors(of arrays). The data tags are used for preliminary semantic checking. Finally, the values and/or descriptors are used** by **microprograms for the format and operator syllables** 

### **5.2.1. Instruction Verification**

The Instruction Verify Unit is a compatibility matrix table lookup unit built into the pipeline processor. The tag for each operand is a two-bit value denoting constant. variable. function, or reserved. The table used for instruction syntax verification is shown in Table **5-1.** 

<b>DEST</b>	VARIABLE	CONSTANT	FUNCTION	<b>OTHER</b>
<b>RIGHT</b> <b>ARGUMENT</b> $\lambda$	v c f o a o u t rnnh	v c f o a o u t n h r n	v c f o a <sub>o</sub> u t n h r n	v c f o モ a o u r n b h
<b>LEFT</b> <b>ARGUMENT</b>				
VARIABLE CONSTANT FUNCTION <b>OTHER</b>	11 0 <sub>0</sub> $\mathbf{1}$ $\mathbf{1}$ $\Omega$ -0 0 O 0 -0 O n 0 O	o o O 0 $\Omega$ $\Omega$ O. - 0 $\Omega$ 0 O - 0 <sup>0</sup> 0 O Ω	$\overline{0}$ $\Omega$ - 0 0. $\Omega$ $\Omega$ O. $\Omega$ $\Omega$ <sup>o</sup> O - റ O 0 O 0	O. 0 <b>O</b> - 0 $\Omega$ $\Omega$ O. -0 Ω O. Ω 0 0 O <sup>0</sup> $\Omega$

Table 5-1: Syntax Checker Table

When the microprogram executing a format in the PMU is ready to release the pipe, the output is tested. If the result shows that the instruction was syntactically invalid (the compatibility tdble entry is a zero) an error routine is invoked and the pipe is not released. (At the current stage of the interpreter, the error routines have not been written. When an invalid instruction is detected, the simulator halts and displays an error message )



**Table 5-2:**  Semantic Compatibility Table

### **5.2.2. Operand Verification**

The Operand Verification Unit is driven by the pipeline processor. As each syllable is read from the pipe, a tag describing its type (character, boolean, integer, real) and class  $(s$ calar, array, reserved, no value) is read from the tag memory into the appropriate tag register

The preliminary semantic analysis process consists of checking that both operands are valid data types. The compatibility table used is shown in Table 5-2.

Two table-controlled semantic summary values are available from the OVU hardware in the pipeline. but these cannot be used automatically to accept or reject the semantics of a given operation because different functions have different domains Table 5-3 compares the types of the operands and is used to reject invalid type combinations (bearing in mind that all pairs of types are valid arguments for "equal" and "not-equal") and to select the correct "cast" procedure if the types do not match.

Table **5-4** gives the shape of the operands and the type of the expected result and is used not for semantic checking but for allocating space for the result.



**Table 5-3:** Semantic Summary Table for Types

## **5.3. Instruction Execution**

The basic cycle in the DMU consists of

- **(1)** Read a format syllable from the pipe
- Execute the microprogram whose address is the entry

of the DMU Format table indexed by the syllable

Each format microprogram begins by fetching the relevant number of operand pointers from the pipe and starting a stream reading the DAT at the offset given in the pointer As a side effect, the DMU tags are read into the tag registers. When all of the operands have been read, the format microprogram reads zero or more operator codes from the pipe. Next. the microprogram signals that it is ready to release the pipe. At this point, the semantic checking hardware verifies that both operands have values (See Table 5-2)

The format microprograms resemble one another up to the point of pipe release. The next stage consists of additional semantic checks where possible, casts where necessary. and allocating space for the result. The DAT entry indexed by constant NXTDEST is used to hold a scalar destination or the descriptor for an array destination.



1999年, 1999年, 1999年, 1999年, 1999年, 1999年, 1999年, 1999年, 1999年

Note: Each 4-Bit entry describes Left Shape  $(0 =$  Scalar,  $1 =$  Array) Right Shape " Result Type  $(00 =$  Character,  $01 =$ Boolean)  $(10 =$  Integer.  $11 =$  Real)

**Table 5-4:** Semantic Summary: Shape and Type

Next comes a CASE decision based on rank and/or type For formats that include operator syllables, the next step is to execute the operator microprogram from the operator EXEC table indexed by the op code read from the pipe The primitive microprogram may

be executed once or in a looping structure. Some formats do not have operator syllables. usually because the operator is implicit in the format, as in the INDAS3 format for indexed assignment. (Many ADEL formats without operators never go through the pipe. BRABS. NXTL. NOOP, and so on are examples.)

Finally. the format microprogram calls 'CLR\_WRI\_DST', a microprogram which deletes the current array pointed to by the destination entry of the DAT, and copies the entry at NXTDEST into the destination position. The DMU then waits until a pipe is available to it. .and repeats the process.

The DMU contains two EXEC tables, one for formats and one for functions The format table is different from the PMU format table, since the DMU does not execute such ADEL operators as NOOP and NXTL, but does contain special formats for data The function table contains the APL primitives, and a set of "hidden primitives" transfer currently including " $+/$ " and "-/".

# **5.4. Function Call and Return**

#### **5.4.1. Call**

After a function-calling format such as SLUR or SUS has processed its operands, the next and final syllable of the instruction denotes the function to be called. The tag part of the CAT entry indexed by the syllable is checked to make sure it is really a function. In parallel with this check, the current set of state variables is copied to the end of the current CAT frame and a stream is opened to the segment pointed to by the pointer part of the DAT entry.

The new function segment begins with a table of descriptors used to verify that the

function and format valence match and to set counters for the number of objects of various kinds in the new environment While these checks and initializations are being done, streams are opened into the Global Symbol Access Table (GSAT) in the function segment and into a new CAT frame. For each object referenced by the called function. the GSAT entry is used to find the corresponding Global Symbol Table page.

The first item is always the result. and is handled by pushing a "no value" onto the DAT stack. The new top-of-stack pointer is written into the CAT with type tag "Variable". The new CAT entry is also pushed onto the call chain of the corresponding symbol table page.

Arguments that are not already on the stack are pushed onto the stack. with their (new or current) stack positions copied into the CAT with type "Variable". The new CAT entry is pushed onto the call chain of the corresponding symbol table page If the arguments were not already on the stack, the DMU microcode will make copies of the argument array segments

The identifier of the function itself is handled by writing the segment number of the function into the CAT with type tag "Function". Since a function name is not a local object, nothing is done to the symbol table. (It is legal in APL to have a local variable with the same name as the function This is sometimes done to prevent inadvertent recursion The resulting local variable is usually assigned a character vector containing a message that the function is suspended and should not be rerun without remedial action.)

Local identitiers are listed after the right argument, with the number of locals indicated by an entry in the function descriptor table. For each local identifier, "no value" is pushed onto the DAT stack. and the new top-of-stack pointer is catenated to the "Variable" tag

to arrive at a value that is written into the CAT and pushed onto a symbol table call chain APL semantics allows local functions. but at this stage of development the interpreter assumes that all identifiers in the local list will be variables.

Since APL semantics do not include "own" or "retentive" variables, it is not necessary to initialize local variables beyond instructing the DMU to set their tags to "No Value" and setting the PMU tags to "Variable" so the first statement that assigns a value to a local variable will be recognized as valid syntax.

The rest of the identifiers in the function are global named objects, literals, and line labels, listed in the order that they were recognized by the translator

Labels are distinguished at link time by setting the sign bit of the corresponding entry in the GSAT When a label is encountered during the CALL process, the sign bit is cleared and the symbol table Is **fwnd** as with other identifiers. The value of the iabei is read from the next GSAT entry. The label value is catenated with the "Constant" type tag to get the value which is written into the CAT and pushed onto the symbol table call cham Thus a label is effectively a "local constant".

Literals, global variables and functions are not distinguished in the CALL process. In each case the symbol table page pointed to by the GSAT entry is consulted to find the end of the call chain, and the value there is written into the CAT. Thus the tag and DAT pointer of each item will be the same in the new environment as in the old. In particular since literals cannot be shadowed (because there is no way to get the translator to accept a hteral as an argument, local variable, or label), they will retain their "Constant" type tags and their pointers into the Global/Literal area of the DAT.

#### **5.4.2. Return**

Return from a user-defined function is a simpler process than calling. The DMU is told the number of local objects (including arguments but not labels or the result) and clears the values of that many items in the local/execution stack The process of popping the values is completed by incrementing the stack pointer by the same number. For each local object (including arguments, labels, and the result), the call chain in the corresponding symbol table chain is popped.

The result of popping the name but not the value of labels is simply that the old values become unshadowed. Remember that the value of labels is set at link time, not at call time, so no new values are created by CALL.

The effect of popping the name but not the value of the result is that the old value of the name is unshadowed, and the new object which was created by CALL and given a value during the execution of the called function is now on top of the stack in the calling environment. All of the currently implemented formats involving CALL return the result to the stack this way. Implementing a format like DLUR, which assigns the returned value to a specific variable. would require that the **"D"** syllable would have to be stored during CALL to be used during RETURN.

When the identifiers have been dealt with, the state information at CALL time is read from the end of the CAT frame preceding the called frame. (Note that the size of this . frame **IS** not known yet, so its beginning cannot be located. However, its end is right before the beginning of the called frame.) This information is used to open the old CAT frame. reset the LINENO register, and restart the stream used for executing ADEL code at the syllable following the end of the ADEL instruction which initiated the CALL.

### **5.5. Primitives**

An APL implementation consists of systems to accept input and control the environment, plus the all-important programs that do the work itself. While the primitives are the first thing anybody notices when learning or using a language, in an implementation project they can be the last thing considered.

APL primitive functions can be divided into three groups: the scalar arithmetic and logical functions, the mixed functions, and the array-manipulation functions.

#### **5.5.1. Scalar Functions**

A scalar function like "+" applies to scalars, or to the elements of an array. producing a result of the same shape as its arguments. For example, the result of adding a 3-row, 6column matrix of numbers to another matrix of the same shape will be a new 3-by-6 numeric matrix The principle of "Scalar Extension" allows a scalar to be added to an array producing an array result.

Scalar functions are executed via the DLR family of formats. Arguments cannot be rejected from inspection of the tags except in the case where character arguments are used with a function whose domain is arithmetic The semantic checking consists of matching ranks and shapes.

#### **5.5.2. Mixed Functions**

. The mixed functions produce results of different shape from the arguments For example. "matrix divide" can be used to solve a system of linear equations It takes as arguments a matrix of coefficient values and a vector of constant values, returning a vector of variable values. "Encode" and "decode" are two more mixed arithmetic functions.

24 60 60 decode hours,minutes.seconds

produces the total number of seconds represented by the three values, and

24 60 60 encode numberofseconds

breaks the number of seconds down into hours, minutes and seconds. None of the mixed arithmetic functions have been implemented in the first pass at the interpreter.

Two mixed functions do not deal with arithmetic. Index-Of, a dyadic function symbolized by *L*, finds the position of the elements of its right argument in the vector<sup>3</sup> used as its left argument. For example

23456~34 **1** 2 (in index origin 0)

Membership, vihose symbol is *E,* returns a boolean of the same shape as its left argument, indicating whether or not each element is found anywhere in the right argument.

Neither membership nor index-of have been implemented yet.

### **5.5.3. Array Manipulation Functions**

The array manipulation functions generally take a left argument describing the size and shape of the array to be returned, and a right argument containing the array to start with The most important functions in this group are Reshape. Take, and Index, denoted by  $\hat{I}, \rho$  and  $\hat{I}$ , respectively. Only the Index function is found in most scalar languages.

5 6T ARRAY

returns an array containing the first 6 columns of the first 5 rows of ARRAY, and

5 6 **IJ** ARRAY

returns a 5-by-6 array containing the first 30 items of ARRAY taken in row-major order.

In APL indexing can select arbitrary subarrays as well as scalar elements.

ARRAY **[3 4;~4]** 

returns a two-by-four element matrix containing the elements in the first four columns and rows 3 and 4 of ARRAY.

Reshape and Index have been implemented in the first version of SAM APL. Thesilape and moex have been implemented in the first version of SAM ATL. Both of that the "left" argument is numeric. In order to make indexing consistent with the other  $\pm$ array selection functions, X[A] is translated as if it were written A index X. Both of

Indexed Assignment is complementary to indexing, allowing a specified element (in APC. a specified orthogonal subarray of elements) of an array to be respecified. Only indexed assignment of a scalar element has been implemented in this version.

The other APL array-manipulation functions include Drop, Rotate, Transpose, Select (or Replicate in most current APL implementations) and Expand. None of these have been implemented in the first version of SAM APL.

Two more functions can be included in this group: monadic **L,** the index generator, and catenation. whose symbol is the comma. (Monadic comma, or ravel, is a special case of reshape included in APL for convenience. Its implementation will be easy enough but has been given a lower priority than most of the other primitives.)

The expression **"LN"** generates a vector containing the first N integers, beginning at zero or one depending on the current "index origin". The index generator has been implemented in a restricted form, with a limit of 256 elements and with only origin zero. Source code using the index generator is translated into ADEL code using the DMR family of formats. Semantic checking consists of verifying that the argument is numeric The operator microcode simply generates a DAT entry with length field copied from the argument and a pointer to a read-only segment containing  $\iota$ 256.

Catenation is the most important APL primitive function that has not been implemented in this version of SAM APL. and is the first function scheduled for implementation in the ongoing project.

It should be noted that there need not be a one-to-one correspondence between primitives in the intermediate language and primitives in the source program. For example, the front-end translator is free to recognize common combinations of APL functions like  $\times \iota$ and produce a "hidden primitive" in the intermediate program. In the opposite direction, a different translator might accept a source program containing an unimplemented function, say "stringwise\_index", and produce ADEL code containing a call to a user-defined function or a sequence of ADEL instructions

## **5.6. Derived Functions**

Derived functions **in** APL are formed by applying an operator to functions and data. Examples of derived functions are plus-reduction  $(+/)$ , inner product  $(+. \times)$  and first-axis times-reduction  $(x/1)$ . In the SAM APL project, derived functions are handled in three different **ways** 

#### **5.6.1. Special Case Microcode**

Some derived functions (currently comprising  $+$  and  $-$  ) have the status of "hidden primitives". The APL expression

 $X \leftarrow +/Y$ 

translates to the single ADEL instruction

DFR Y X  $+$ /

and is executed by the DER format microcode and the PLUSRD function microcode. This approach is useful when there is an algorithm that is more efficient than straightforward iteration as in and/ and **or/** Reduction of vectors of length greater than two is rare for the transcendental functions and reduction by these functions will probably be handled by general-purpose microcode or PMU derived functions

It is uncertain whether derived functions using the scan  $(\cdot)$  operator

 $+ \1 2 3 4 5 10$ 1 3 6 10 15 25

will be implemented with DFR, since the current version of the format hard-codes the rank reduction (i.e.. the reduction of a vector is a scalar) associated with the reduction operator.

**5.6.2. PMU Derived Functions** 

Some primitive scalar functions will, at least initially, be implemented as system-resident defined functions. Derived functions which use these primitives will be implemented as defined functions which incorporate calls to the defined "primitives". The consequences of this policy are discussed in the last chapter.

### **5.6.3. General Microcode**

The product operator has a format to itself

$$
X \leftarrow Y + X Z
$$

would be translated into

DLRZYX+ $x$ 

The format microcode then sets up loops and EXECS primitive functions. The scan operator, and some cases of reduction, will also be handled this way.

Note that it is quite possible to implement the commonest products as special cases. It is also possible to use reduction primitives where they exist instead of using only scalar functions in the loops.

## **Chapter 6**

## **Performance Considerations**

## **6.1. Control Primitives**

Many APL programs contain loops, and the standard APL language does not contain an efficient way to deal with them. The typical construction

 $\rightarrow$  **LOOP**  $\times$  **LIMIT**  $>$  **COUNT**  $\leftarrow$  **COUNT** + 1

involves all of the usual overhead in checking the type of COUNT, performing the addition and cornparison, doing whatever arithmetic is involved in the branching idiom used. and finally transferring control.

In the SAM machine. the type checking is done in parallel with the operations, so performance would be improved. However, the transfer of control is a PMU operation, so time would be wasted waiting for the DMU to finish its work and send the result back to the PMU This waste of time is especially pernicious because the PMU is intended to keep ahead of the DMU. For this reason, SAM APL is extended to include control primitives

The control primitives developed in the first implementation consist of a branching primitive and a pair of loop control primitives The ADEL form of the branching primitive is named BIFSTK (Branch IF StacK) The looping primitives consist of a counter initializer with a zero check called INlTBZ and an increment-test-and-branch primitive called IBNZ. The ADEL code tested had the form

**SMR VEC p INITBZ END**   $Stack$   $\leftarrow$  Length Counter  $\leftarrow$  Stack END if Counter=0

**LOOP: body of loop IBNZ LOOP** 

**Increment Counter**  -, **LOOP unless Counter-0** 

**END:** 

The corresponding APL source code would contain a dyadic system function INITCZ (Initialize and check if zero) and a monadic INCCNZ(lncrement and check if not zero), with the syntax

-, **END INITCZp X Start Counter and**   $\rightarrow$  END if Counter = 0 **LOOP: body of loop**<br>
- *INCCNZ LOOP* **Incr Counter and**  $FND$  : → LOOP unless Counter = 0

The looping primitives were developed and tested while writing a user-defined version of plus-reduction The measured results are shown in Table 6-1 Note that the per-element times include the time taken for data manipulation as well as time taken by the control primitives themselves



Table **6-1:** Performance of Control Primitives

Note that using the looping primitive allowed the per-element loop time to be reduced to

**196** cycles. as compared to 297 using the branching primitive. Without the branching primitive, the loop time would be slightly longer This measurement would have required the standard format for computed branch, BSTACK. which has not been implemented yet.

Careful coding is still a useful way of improving speed. Omitting an extra step to ensure that the limit and loop counter are scalar saves **641** cycles of set-up time but raises the per-element time to **391** cycles.

Five different branching techniques are shown in Figure **6-1.** The looping primitive IBNZ is the only one that does not require the PMU to wait for a value from the DMU.

The standard branching primitive BSTACK executes a typical APL test-and-branch as three ADEL instructions. the test itself, the APL instruction which uses the destination and the result of the test to place a line number on the stack, and the branch instruction itse!f The "Branch-frcm-Stack" **primitives** BRANS!? and BRANLS combine **the** branching operation with the preceding operation, so that the APL expression

 $\rightarrow$  (Y < X) / LBL

can generate the ADEL code

 $SLR$  X Y < Stack  $\leftarrow$  Y < X BRANSR LBL /

The present branching primitive BIFSTK could not be generated from APL source code without an IF keyword BRANSR and BRANLS would generate two ADEL instructions, as BIFSTK does rather than three, as BSTACK does. Since they use an APL function such as "/" or "xi" instead of a special IF instruction, their performance should be somewhere between BIFSTK and BSTACK. The BlFLR format is based on the realization that the destination syllable of a branching instruction could be used to contain the destination of the branch itself. By condensing the test, the branch instruction, and the destination into a single format, BlFLR would run slightly faster than BIFSTK.

All of the branching primitives discussed here will be implemented for testing purposes. -- ADEL Code although only a subset can be kept in the finished interpreter.



Note: Semicolon is used here as an instruction Separator **Figure 6-1:** Some Possible Branching Techniques

The control primitives in the final implementation may differ from the constructs implemented in this version, which are efficient but not necessarily in keeping with the theme of APL (In fact. they are not available as APL source but have to be hand-coded into an ADEL assembler)

One common APL idiom that can be handled by a control structure is execution involving a constant string. The expression

**execute**  $(Y = 3) / Y \leftarrow X \div 2'$ 

might be compiled to a test-and-branch equivalent to the (illegal) APL expression

 $Y \leftarrow X \div 2 \rightarrow (Y=3) \downarrow NEXTLINE$ 

## **6.2. Call/Return Mechanisms**

Function call and return is one of the principal sources of overhead for programs in mast languages This is particularly true of APL, which allows applications to be built of dozens or hundreds of independent functions, with the dynamic nature of the language making function call expensive on conventional architectures The SAM architecture has been designed to address the problem of efficient function call and return. The process is fairly similar to DELtran practice [FIH83] and was described in detail in Section 5.4.

To recapitulate. each identifier in a user-defined function is mapped to a symbol table entry. At call time, each entry is inspected, and the corresponding data table pointer is written into a Contour Access Table which corresponds exactly to the numbers used to encode objects (other than primitives) referenced by the user-defined function. if the item is a local variable, the DAT pointer used is top-of-stack after a "no value" has been pushed onto the DAT stack. If the item is a label, the corresponding line number is written into the CAT. For both labels and variables, the CAT entry is pushed onto the call chain for the symbol table entry, shadowing the previously accessible value.

The arguments. if any, are currently handled by copying the DAT entries and the array segments on function call. Array segments for arguments that are from the stack do not have to be copied. Later investigation will determine whether it is worthwhile to implement a "protected-call-by-reference" mechanism to avoid the overhead of making copies of the arguments

Function return consists of deallocating all of the local variables which have received values and marking the corresponding tags "undefined" The corresponding symbol table page call chains are popped, and corresponding values are popped from the DAT stack
By popping the result name (in the symbol table page) but not the result value (in the DAT stack) on function return, the value is available as the top-of-stack in the calling environment.

Function call overhead is still considerable compared to that for a simple branch. Several improvements are possible.

- $\bullet$  Implementing the formats like DLUR which put the result of a called function into a specific variable instead of the stack will save one pipe cycle by allowing a sequence like SLUR followed by DGETSS to be replaced by a single ADEL instruction.
- **<sup>0</sup>**It is not necessary to perform a symbol table access for literals. since the DAT pointer is known at link time.
- ADEL keeps literals in the same CAT as named identifiers. It is feasible to keep a separate contour for literal values **[FJW85]** To make this work with the current system, new formats would be needed, so  $X \leftarrow X + 31$ " would translate to "LLC X **31** +". Note that this would require some way to handle applications with more than 256 literals.
- An edit-time check could determine whether arguments are modified. so the DAT pointer could simply be copied for an unmodified argument, without any need for reference counting and protection flagging in the DMU.
- Special areas of the CAT and DAT could- be reserved for "leaf" functions which do not themselves call other functions, allowing a call to be a matter of resetting the environment pointer.
- $\bullet$  If the identifiers used by a function are listed in order of use, function execution could be begun before the new environment is complete. More identifiers would be set up as the PMU gets ahead of the DMU, or as they are needed.

**0** Finally, hardware could be designed to assist the calling mechanism. For example, the CAT and DAT were kept in hardware in an earlier version of the SAM design **[HGT86].** Moving these two tables to segmented memory slows down call/return performance by a factor of about two.





**COST is measured clock values. Values in parentheses are estimates.** 

#### **Table 6-2:** Measurements of Ackermann's Function

A good way to exercise the calling mechanism is Ackerrnann's Function, a doublyrecursive function with the number of calls growing more than exponentially with its arguments. The APL version of Ackermann's Function is shown in Figure **6-2** The

number of calls, branches. and failed branches for various arguments is shown in Table **6-2.** The cost of execution is given in the final column. The simulator is too slow to run 3 ACKER 3, but a linear regression of cost versus number of calls, successful branches, and failed branches gives agreement to within 0.1% for all cases up to 2 ACKER 2, and has been used to extrapolate to 3 ACKER 3. The extrapolated speed is compared with various other implementations in Table **6.3.** A hardware speed of 8 MHz is assumed.

. [HGT86] give a simulator measurement of 1.5 ms for **2'** ACKER 2 on 5 MHz hardware. This works out to 7500 machine cycles. over twice as fast as the current simulator. Most of the difference is due to the present implementation of the CAT and DAT in segmented memory instead of in hardware.



**Table 6-3:** Comparison of 3 ACKER 3 on Various Systems

### **6.2.1. The ICKER Function**

Since Ackermann's function exercises the calling mechanism of a language implementation, it seemed that it would be interesting to write a purely iterative version of the function. A comparison of the performance between ACKER and ICKER (shown in Figure 6-3) should provide a comparison between calling and branching performance of the implementation. A table of performance measurements is shown in Table 6-4.

$$
\nabla \qquad R \leftarrow X \text{ ICKER } Y; \text{STACK}; \text{PTR}
$$
\n[1]\n
$$
\begin{aligned}\nSTACK &\leftarrow 256 \rho - 1 \\
2] &\quad PTR \leftarrow 0 \\
3] &\quad TOP; \rightarrow BIGX \text{ IF } X > 0 \\
4] &\quad R \leftarrow Y + 1 \\
5] &\rightarrow END \\
6] &\quad BIGX; \rightarrow BIG2 \text{ IF } Y > 0 \\
7] &\quad Y \leftarrow 1 \\
8] &\quad X \leftarrow X - 1 \\
9] &\rightarrow TOP \\
10] &\quad BIG2; \text{STACK}[PTR] \leftarrow X - 1 \\
11] &\quad PTR \leftarrow PTR + 1 \\
12] &\quad Y \leftarrow Y - 1 \\
13] &\rightarrow TOP \\
14] &\quad END; \rightarrow 0 \text{ IF } PTR = 0 \\
15] &\quad PTR \leftarrow PTR - 1 \\
16] &\quad Y \leftarrow R \\
17] &\quad X \leftarrow \text{STACK}[PTR] \\
18] &\rightarrow TOP\n\end{aligned}
$$

Figure 6-3: Iterative Ackermann's Function

According to the table. ACKER is more efficient than ICKER on every APL implementation except SAM APL. The measurements for LISP and compiled C were made later in the project and will be discussed later

My first conjecture was that either SAM APL calls are relatively slow, or its branches are relatively efficient. To check whether the effect was caused by something in ACKER



**Table 6-4:** Comparison of 3 ACKER 2 and 3 ICKER 2

or something in ICKER, execution timings with different sets of arguments were compared with the timing on MTS APL. The results of this comparison are shown in Table 6-5.



**Table 6-5:** Comparison of ACKER and ICKER on SAM and MTS for Various Arguments

As seen in Table 6-5, the relative performance of SAM and MTS APL for Ackermann's function is fairly constant. Also, the ratio between the costs of ICKER and ACKER on MTS APL was not dependent on the arguments But ICKER on SAM APL performed much better than ACKER with larger arguments. Figure 6-4 plots MIS and SAM costs for ACKER and ICKER against MTS ACKER cost on a logarithmic scale.



To test the effect of number of instructions (and to make sure that nothing in the simulated run was compromising the integrity of the clock), the simulations were rerun with an **ADEL** instruction counter added to the simulator. Clock values were plotted against instruction counts in Figure 6-5. It is apparent that ICKER uses slightly more instructions that ACKER, but the average instruction costs less, and the graph of cost versus instructions for ICKER does not pass through the origin. Since the instruction mix for ICKER contains a single function call, while the proportion of calls to other instructions

in ACKER is nearly constant, it is evident that the performance advantage of ICKER is due to replacing function calls with array references and branches.





At this point the pressing question was why other APLs did not show the same behaviour. Either the greater number of high-level instructions executed in ICKER must outweigh the function calls in ACKER, or the array references found in ICKER but not ACKER must cost almost as much as function calls. A comparison of the costs of array references, function calls, and branches is shown in Table 6-6.



**Table 6-6:** Costs of Array References, Function Calls, and Branches

Notice that on MTS APL, test-and-branch, function call, and array references all take the ~ same order of time, 50 to 80 microseconds. In SAM APL, array references are twice as fast as test-and-branch, and function calls are more than twice as slow. This behaviour is more typical of compiled code than interpreted, suggesting that interpretation on SAM will resemble the execution of compiled code in detail as well as in raw performance. This is supported by the figures for LISP and compiled C in Table 6-4. In interpreted LISP, the list references for ICKER outweigh the function calls for ACKER, but in compiled C. ICKER is faster than ACKER by about the same ratio as in SAM APL.

At this stage in the development of the interpreter. only a very rough guess can be made at the ultimate performance of the call and return mechanisms.

At a very rough guess, then, the speed of function call and return might be doubled. The microroutines in this version of the interpreter have not been fine-tuned for performance (see Section 6.6). Implementing the CAT and DAT in hardware and tuning up the microcode would certainly bring the performance up to the predictions in [HGT86] which is double the current speed. Several suggestions for improving the performance of call and return are listed earlier in this section. An interpreter using all of these ideas and with fine-tuned microcode may be twice as fast as the current model without hardware assistance Finally, the technology used in the 68020 chip is at least a factor of two faster than the technology of an 8MHz SAM system. Raising hardware speed to 16 MHz, moving all tables to hardware, using the fastest algorithms possible, and fine-tuning every line of microcode should speed up call and return by an order of magnitude, giving a speed for Ackermann's Function about the same as compiled C on a 68020 machine

I have drawn two tentative conclusions from the ACKER-ICKER investigation. First: the poor performance of ICKER on other APLs is due to inefficient array accesses. Second. part of the advantage of ICKER over ACKER on the SAM system may be due to the fact that function call and return are still in their initial correct-but-not-fast incarnations In the final version of SAM APL, ICKER will run faster than ACKER, but probably by about half of the current margin.

It should be mentioned here that the indexed assignment instruction used was specialcased for scalar indices of vectors. (See Section 6.4 for more detail on this instruction.) A more general indexed assignment would be slower. However, there is no reason to suppose that the translator will have any difficulty recognizing scalar indices of vectors and generating the appropriate ADEL instruction even when more general cases are supported. It is unlikely that future developments will make ICKER run more slowly.

## **6.3. Array Arithmetic**

## **6.3.1. Dragalong and Beating**

APL often does operations that appear to be unnecessary. For example, in the expression  $A \leftarrow 10$   $\uparrow$   $B + C$ , where  $B$  and  $C$  are 1000-element vectors, a naive implementation would perform 1000 additions, then discard 990 of the results. In his PhD thesis, Abrams [Abr70] suggested that a more sophisticated APL interpreter should "beat" the addition by "dragging along" a temporary expression until the take operation is complete for both arguments. This approach has a few problems: for instance, suppose B and C were of different lengths. This error could easily be caught. But suppose the operation to be beaten were division instead of addition, and C contained a zero after the 10th position. It is not clear whether an interpreter which did not catch the domain error could be considered to be correct.

In many cases, there is no possibility of an error being introduced by an optimization of this kind. For example, the result of  $A \leftarrow B + C + D$  is the same whether two passes are used to create a temporary  $(C + D)$  and then add B to it or both additions are done in **d** single pass. For vectors of length N. 6N memory references are needed for the twopass algorithm and 4N memory references for the one-pass algorithm. Hobson [Hob841 points out that, given an efficient vector machine, the two vector additions will run faster than the sequence of alternating scalar additions. The implementation of some repeated operations is discussed in more detail in Section 6.5.

#### **6.3.2. Derived Functions**

Derived functions can be optimized using the distributive properties of the primitives involved. For example. **SCALAR** +.x **VECTOR** is identical to **SCALAR** x +/ **VECTOR**  but **SCALAR+.+ VECTOR** has no such optimization. The proportion of optimizable cases of this kind in APL applications programs is probably too low to justify the overhead that would be needed to recognize them.

On the other hand, some optimizations are not dependent .on special cases. For example, the AND reduction of a Boolean vector

 $X \leftarrow \text{and}/ \text{BOOL}$ 

reduces to the question of whether or not there is a zero anywhere in the result, and efficient APL implementations do not go on testing elements of the vector past the first zero.

#### **6.3.3. Storage States**

One interesting approach to the mixture of arithmetic and selection operations is to use a separate processor for each kind of operation [Sny82]. Snyder's MAPLE architecture handles selection operations as descriptor manipulations whenever possible. This would handle an expression like  $A \leftarrow 10$  f  $B[1:2]$  by generating three successively modified descriptors for B The array manager would then use the final descriptor to fetch elements of A as needed for arithmetic and logical operations.

Using such storage states speeds up selection expressions at the expense of making data fetches somewhat complicated. A well-designed address calculating mechanism would allow array data fetches at the same speed as ordinary fetches on a scalar machine. The kind of streaming memory access used in the SAM project would be too complicated to justify with current technology.

Another storage state is the arithmetic progression vector, which can be represented by an initial value, a step size, and a length. In the absence of a combinational multiplier built into the addressing mechanism, this representation will slow the use of its elements in any nonsequential fashion. [HaL76] point out that arithmetic progressions are used very heavily in array subscripting operations. It may be desirable to allow the subscripting microcode to generate and use arithmetic progression vectors even if the rest of the interpreter does not recognize them.

There is one place in the SAM interpreter where a storage state approach is used. The index generating primitive saves both space and time, as  $\iota$ N simply generates a descriptor with length field N, pointing to a read-only array segment containing the elements of  $\mu$ 256. The principal drawback of this approach is the necessity to check whether a variable is an index vector before deallocating its old value. This problem has been handled by declaring a set of segments to be read-only. The deallocator simply checks whether the segment number is greater than a global READ-WRITE-BEGIN. At the present stage of the interpreter development, this slows down all deallocations with no benefit but a speed-up of the index generator. However, now that read-only segments exist, one part of a nestedarray implementation is in place.

A second problem with the descriptor-only approach to the index generator is the existence of two separate index origins. The answer to this problem will be to maintain two separate index-vector segments, setting the pointer to the value active at execution time

Measurements taken from actual use will be needed to decide whether this approach is optimal.

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#### **6.3.4. Arithmetic Coprocessor**

For any application involving multiplication, division. or any floating-point operations, an arithmetic coprocessor is necessary for good performance. John Gudaitis [Gud85] shows that the performance of array algorithms is sensitive to the protocol used to access the arithmetic unit(s). If the output of the coprocessor is buffered. it is possible to overlap the bus cycle sending the next arguments with the operation time of the previous cycle.

Inner product requires an accumulation operation in its inner loop. In the general case, it may or may not be faster to use two arithmetic units. The multiply-and-add sequence of ordinary inner product is a processor primitive for the Weitek chip set used in the current SAM implementation. Consider a different inner product, say  $\times$  +. Whether or not it is faster to do the vector additions on a different processor than the multiplicative accumulations depends on whether the bus cycles for the extra data transfer outweigh the multiplication time for the processor chosen. In practice,  $+ \times$  is expected to be the only common inner product not involving a logical function. The matrix comparison products like and. $=$  might use the coprocessor if floating-point numbers are involved.

Two other cases where multiple arithmetic operations can be 'combined are matrix division and monadic  $-x$  (the determinant function.) It is important that an APL program dealing with heavily numeric operations such as eigenvectors. Fourier transforms. and the like should be interpreted in such a way as to optimize the streaming through the arithmetic coprocessor Probably several cycles of implementation will be necessary to get really good mathematical performance. The transcendental APL primitives must be implemented somehow, and it is not at all obvious that the published algorithms will be optimal for the SAM architecture.

#### **6.3.5. Performance Measurements for Array Arithmetic**

Problems like Ackermann's Function, with a high proportion of calls and branches to data manipulation operations, are a worst case for an interpreter. Here we will look at the performance of SAM APL in a simple array operation, finding the sum of 1000 integers. In Table 6-1, the microcoded  $+/$  primitive was shown to run with a set-up time of 423 cycles and a per-element time of 2 cycles, giving a predicted time for the addition of 2423 cycles or 0 3 milliseconds. Table 6-7 compares this with some other implementations.



**Table 6-7:** Time to Add 1000 Integers

For this benchmark, the advantage of compilation is very much reduced. For example. MTS APL is now twice as fast as compiled C instead of seventy times as slow. SAM APL is about six times as fast as MTS APL in both cases.

## **6.4. Selection Functions**

For most applications, the set of APL functions which are used to create, select subsets from and alter the shape and orientation of arrays are used more heavily than the array arithmetic functions A considerable body of theory has been developed to combine the selection functions into a set of stepper generators that are used to delay the selection process until all of the steps in the selection have been specified [HaL72. GuW78, TrB82, Ben841. For example,

4 5 **TARRAY** [2 4 3;5+ $\iota$ 6]

can be treated by generating a "Stepper" or "Grid Selector" for the indexing operation,

**apply the "Take" operation to the stepper, and finally apply the resulting stepper to the array itself.** 

As in the example of combining two additions into a single-pass operation (discussed in Section 6.3.1). this kind of optimization is only useful on a scalar machine. Hobson [Hob84] points out that, given an underlying vector machine, the extra memory references for the straightforward approach are not likely to outweigh the fact that a large number of scalar instructions will be generated instead of a small number of vector instructions. This makes it difficult to justify the extra interpretation needed to form a syntax tree for each expression, combine different operations in the ways discussed above, and attempt to execute the entire tree of operations so formed in one pass.

In the present version of SAM APL, only shape, reshape, indexing, and indexed assignment have been implemented. Indexing and indexed assignment have been restricted to scalar selections into vectors only. With this restriction, an interesting simplification of indexed assignment can be made. Since only three values are used, a single INDASN format can take array, index, and value as its D. L. and R syllables. One possible consequence of this simplification is that the performance of indexed assignment may be unrealistically fast giving misleading times for the iterative version of Ackermann's Function discussed earlier. At this stage, it does not seem to be too much to expect that the front-end translator can recognize simple scalar-vector indexed assignments and generate the single- operation format, even if the general case will take multiple ADEL operations.

## **6.5. Sequences of Operations**

The use of variable-length instructions with a specific format syllable allows an interesting improvement to repeated executions of the same operation in APL source code. The example most found in application programs is repeated catenation.

RESULT  $\leftarrow$  X  $\rightarrow$  Y  $\rightarrow$  Z

In a case like this, especially where X and Y are scalars and Z is a large array. the cost of creating an intermediate result  $(Y,Z)$  and then catenating X to this result is much greater than the cost of combining the operations in an intelligent way. The design of ADEL includes formats for repeated operations of this type. The current interpreter does not include any such formats, and their development must be given a lower priority than the many missing features necessary to a robust and complete APL implementation.

Formats that implement repeated executions of a given function are a special case of formats that correspond to more than two levels of a syntax analysis tree **[FJW85].** As the authors point out, the number of separate formats needed for all combinations of functions and operands make it impractical to attempt to implement a full set of formats beyond the two-level case (a single function with its arguments and results). Often the result of an expression is used more than once.

 $X \leftarrow Y \leftarrow 0$ <br>  $X \leftarrow Y \leftarrow W + Z$ <br>  $R \leftarrow X[Y+Z] - W[Y+Z]$  $[1]$  $\lceil 2 \rceil$  $[3]$ 

Example 1 above and other examples of repeated assignment can easily be special-cased. Handling Example 2 in a single ADEL instruction would require a large set of new formats and a redesign of the pipe hardware to handle extra values. A compromise would be to implement a  $D \leftarrow D$  instruction, so the example would be translated into the ADEL code

#### DLR Z W  $X +$  ; DGETSD Y

This format could be implemented easily because a copy of the last value assigned can always be found in the NXTDEST entry of the DAT. Thus multiple assignments are possible without using the stack.

Example 3 is the most interesting, as an example of the kind of optimization that is ' expected from a good compiler. It would be feasible to implement ADEL formats that would do the job (See Section 5.5.3 for a description of the subscript function):

SLR Y Z  $+$ DLMT W R [ SLMS x [  $LSL$  R -

where "T" in an argument syllable means that the top element of the stack is used but not popped. There are currently about 100 formats that use S as an operand, and a fully general implementation of this idea would require one or more new formats for most of these. It would not be simple to produce a translator that could recognize that a common subexpression is free of side effects and generate the appropriate code.

## **6.6. Fine-Tuning the Microcode**

SAMjr microcode requires very careful attention to detail in order to use the maximum amount of parallel execution. For example, the microcode used for a search iteration in the memory manager is shown in the first part of Figure **6-6.** This code requires 6 cycles per iteration and omits and IDLE instruction needed for proper timing of the memory controller hardware. A more efficient way to microcode the search uses the "mailbox" registers in the pipe to hold temporary results. This code takes 5 cycles the first iteration and 4 cycles per iteration thereafter

**Old Version** 

#### **SEARCH:**

**MMSl SRW R[Pl Start stream R[SZl** + **SSN MMSl Read size**   $SF R[SZ]$  MINUS  $R[N] \triangle R[SUC] \leftarrow$  SSN MMS1 Compare size and read successor FOUND IF  $\sim$ NEG  $\triangle$  R[PRED]  $\leftarrow$  R[P] **Test-and-branch, prepare next iteration SF ARSHIFT RISUCI End-of-list?** + **SEARCH IF -SB A R[PI** + **R[SUCI Begin next iteration unless end-of-list** 

**New Version** . **SEARCH: MMS1 SRW R[P]** Start stream for 1st iter **SEARCHl: IDLE**   $SF \ R[M]$  MINUS M $[SAVE] \leftarrow$  SSN MMS1 Read and save size, compare with N → **FOUND IF NEG A SF ARSHIFT M[SUC]** ← SSN MMS1 Test-and-branch, read successor.check for end  $\rightarrow$  **SEARCH1 IF**  $\tilde{S}B \triangle R[P] \leftarrow MMS1$  **SRW M[SUC]** Start stream for next iter unless end-of-list

**Figure 6-6:** Speeding Up List Search Microcode

This search code takes 5 cycles the first time through and 4 cycles per iteration thereafter

The best way to apply detailed microcode optimizing techniques to this interpreter will be to obtain some measurements and identify the bottlenecks. Certainly the search loop shown above will be one place where a cycle saved will have a large effect on the performance of the interpreter as a whole.

## **6.7. Summary of Performance Considerations**

The present interpreter gives simulated results about six times as fast as MTS APL on an IBM mainframe and 100 times as fast as STSC APL on a 10-MHz **80286** machine. Compared to compiled C on a **68020** machine, speed ranges from an order of magnitude slower (for Ackermann's Function) to an order of magnitude faster (for vector arithmetic). Neither of these benchmarks is a realistic model of a real application. For most applications. the performance of the completed interpreter on the present hardware design should more than satisfy the project goal of performance comparable to good compiled code.

## **Chapter 7**

## **Discussion**

## **7.1. Hardware Feedback**

One of the primary purposes of this project was to provide feedback to the team developing the SAM hardware. It would be unreasonably optimistic to expect a hardware designer to anticipate all of the requirements of software. Writing software for a machine that exists only as a simulator leads to the opposite problem, as the software wish list may well include items that are prohibitively difficult to implement in hardware. By having the hardware people inspect and pass on each request as it is made, we hope to avoid both of these problems.

#### **7.1.1. Pipe Clear**

The simplest hardware modification to arise from a software request was a facility to clear out the pipeline between the PMU and the DMU.

The PMU fetch-and-execute cycle puts the format syllable into a pipe as a side effect of fetching it. Then argument and operator syllables are put into the pipe by the format microcode. But several PMU format syllables, such as no-op and branch-to-next-line, do not require any DMU action It is simple enough to write DMU format microcode that does nothing except release the pipe when it encounters these formats. However, this uses up two microcycles of DMU time, and also uses up space in the limited DMU format table. By allowing the PMU to clear out the pipe when it executes a format that does not

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require DMU action, the DMU processing is speeded up and the extra formats become available for operations like data transfer, which are not initiated by PMU format fetch.

#### **7.1.2. Triadic Addition**

As a contrast to the simple and useful pipeline clear facility. I will mention the Triadic Addition request. After generating a considerable quantity of microcode, I noticed that address calculations and the like often required three values to be added together. Performing two separate additions required two microcycles. Since the major time requirement of addition is carry propagation. I reasoned that three registers could be added, via a carry-save adder in front of the standard adder, in one microcycle.

This particular request turned out to be neither necessary nor feasible. The microcode format is sufficiently parallel that most extra additions can be done in tandem with other operations. The infeasibility derives not so much from the added complexity of the adder as from the redesign of the bus system that would be needed to use it.

#### **7.1.3. Memory Mapping Table Size**

Garbage collection is required when there is no sufficiently large block of contiguous free pointers available in the memory mapping table. Of course, garbage collection is useless if there are not enough pages of free memory left to hold the required object. Therefore, by making the number of entries in the mapping table larger than the number of pages in segmented memory, garbage collection should arise very seldom

It is easy to see that garbage collection could be required using a table less than on the order of the square of the number of memory pages, by executing a long sequence of commands

comment  $N =$  total number of pages comment SMALLEST  $=$  smallest page to be allocated  $SIZE = SMALLEST$ while there is room in memory ALLOCATE (SIZE) ALLOCATE (1) FREE(first allocation of this iteration) INCREMENT (SIZE) FRE<br>FRE<br><u>end</u>

Let K be the number of iterations before there is not enough room in the memory. The next allocation will be for SMALLEST+K pages, and there will be K single pages already allocated Then the size of table required to map all of these free and allocated blocks without garbage collection will be  $(SMALLESTXX)+(K\times (K-1))+2$  and choosing the worst possible value of N/3 for SMALLEST, we find that the mapping table is roughly  $N^2 \div 6$ items long.

In practice. of course, a much smaller memory mapping table will be required to make garbage collection exceedingly rare. The size of mapping table needed will be determined by experiments that will be performed when the interpreter has been developed to the state that it can run real APL code. The practical suggestion arising from this theoretical consideration is that the mapping table should be twice the size needed to map all of the real pages of segmented memory.

## **7.2. A Minimal APL Subset**

It is immediately obvious that the APL language contains redundancies. For instance, the ten dyadic logical functions found in APL can be implemented as the NAND function. with preprocessing to translate expressions involving the other nine dyadic logical functions to equivalent expressions using NAND.

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Some APL primitives are tempting candidates for replacement by defined APL functions. Matrix divide is the obvious first choice. Encode and decode, grade up and grade down. factorial/gamma and binomial coefficient. and the trigonometric functions could also be replaced. Some derived functions could also be recognized by the preprocessor and implemented as calls to defined APL functions, especially if monadic -.x (determinant) is implemented

Since it is much easier to write APL functions than microcode, it is of considerable interest to be able to define a minimal subset of APL sufficient for implementing the rest of APL. Some caution is needed here. since a strictly minimal subset would not include such things as addition and negation, since  $(-X) = (0 - X)$  and  $(X + Y) = (0 - (0 - X))$ Y) What we are after is not a provably minimal subset, but a provably sufficient subset which is comfortable to work with.

#### **7.2.1. Arithmetic and Logical Functions**

The arithmetic functions consist of addition, subtraction, multiplication, and division. The transcendental functions can all be implemented as series, especially if the values of e and pi are available. Encode, Decode. Determinant (if it is to be included at all), and Matrix Divide can be implemented as user-defined functions in a straightforward way.

All ten nontrivial dyadic logical functions are included in the subset. Although it is simple enough to get along without most of them, it is even simpler to implement one, i then make slight changes in the microcode to arrive at microcoded versions of all ten. It may be necessary to omit complete microcode for some of the logical functions in order to conserve control store.

#### **7.2.2. Array Manipulation Functions**

Since an APL interpreter can be written in a scalar language with no array facilities except scalar indexing, it follows that all of the other array manipulation functions in APL are redundant. In fact, **[Zak78]** implemented transpose, take, rotate, membership, index-of, gradeup, gradedown, expand, drop, and compress as defined functions. SAM APL will implement all of these in microcode, with the possible exceptions of membership. gradedown. and drop, which can be implemented via the identities

 $X \in Y \leftarrow \rightarrow (Y \cup X) \leq pY$  (where Y is a vector) GRADEDOWN  $X \leftarrow \rightarrow$  GRADEUP  $(-X)$  $L \downarrow X \leftarrow \rightarrow ((-\times L) \times (\rho X) - L) \uparrow X$ 

#### *P.2.3.* **Array Ranks**

 $\nabla$   $R \leftarrow X$  PLUS Y: RANKX: RANKY [1]  $RANKX \leftarrow 1 \uparrow X$ <br>[2]  $RANKY \leftarrow 1 \uparrow Y$  $[2]$  RANKY  $\leftarrow$  11Y<br> $[3]$   $\rightarrow$  RANKOK  $\times$  *y* [3]  $\rightarrow$  RANKOK  $\times$  L RANKX = RANKY [4]  $\rightarrow$  (RANKX=0)/SCALX  $[4] \rightarrow (RANKX=0)/SCALX$ <br>  $[5] \rightarrow (RANKY=0)/SCALY$  $[5] \rightarrow (RANK \, Y=0)/SCAL \, Y$ [6]  $SIGNAL' RANKER ROR$ 161 SIGNAL'RANKERROR' [7]  $SCALX: R \leftarrow (RANKY + 1) \uparrow Y$ [8]  $R \leftarrow R$ .  $(1 \downarrow X) + (RANKY + 1) \downarrow Y$ <br>[9]  $\rightarrow 0$  $\rightarrow 0$  $[10]$ SCAL Y: R  $\leftarrow$  (RANKX + 1) 1 X  $[11]$   $R \leftarrow R$ ,  $(1 \downarrow Y) + (RANKX + 1) \downarrow X$  $[12] \rightarrow 0$  $[13]$ RANKOK: XSHAPE  $\leftarrow$  RANKX 11  $\downarrow$  X  $[14]$  YSHAPE  $\leftarrow$  RANKY 11 | Y [15]  $\rightarrow$  (and / XSHAPE = YSHAPE) ) SHAPEOK [16] SIGNAL 'LENGTHERROR'  $[17]$ SHAPEOK:  $R \leftarrow (RANKX + 1)$ T X [18]  $R \leftarrow R \cdot ((RANKX + 1) \downarrow X) + (RANKY + 1) \downarrow Y$  $\nabla$ 

> **Figure 7-1:** Vector APL Function to do Scalar Arithmetic on an Array of Arbitrary Rank

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The initial subset contains only scalars and vectors. This restriction requires a demonstration that a user-defined protocol and a set of user-defined functions can mimic all of the APL primitives and derived functions as applied to arrays of rank 2 or higher. The following protocol is one possibility,

The programmer is responsible for keeping track of which values are arrays of rank greater than one. All such arrays are modelled as vectors, with the first (rank  $+1$ ) elements containing rank and shape. For character arrays, twice as many elements are used to encode rank and shape as 16-bit integers.

Then all operations involving these arrays are done by defined functions instead of by APL primitives. It is immediately obvious that this can be done, since it is equivalent to writing a subset of an APL interpreter in vector-only APL. The easiest way to establish the feasibility is to show a few examples of such functions. Figures **7-1. 7-2** and **7-3**  show how scalar arithmetic, reduction, and inner product can be modelled.



**Figure 7-2:** Vector APL Function to do Reduction on an Array of Arbitrary Rank

 $\nabla R$   $\leftarrow$  X PLUSTIMES Y  $[1] \qquad \bullet$  PlusDotTimes InnerProduct<br> $[2] \qquad \begin{array}{cc} 5 \text{TRIP} & R \text{ S} \end{array}$ [2]  $STRIP\_R\_S$  oset RANK & SHAPE<br>[3]  $\rightarrow$   $(RANKX = 0 \t1)/SCALX, VECX$  $[3] \rightarrow (RANKX = 0 \ 1)/SCALX, VECX$ <br> $[4] \rightarrow (RANKY = 0 \ 1)/SCALY, VECY.$ [4]  $\rightarrow$  (RANK Y = 0 1)/SCAL Y, VECY.<br>[5]  $COL \leftarrow -1$  $[5]$   $COL \leftarrow -1$ <br> $[6]$   $\rightarrow$   $LERRj$ [6]  $\rightarrow$  LERR <u>if</u> (W  $\leftarrow$  1 T SHAPEY)  $\neq$  (-1)T SHAPEX<br>[7]  $\rightarrow$  NROWS  $\leftarrow$   $\times$ / SHAPEX  $\leftarrow$  (-1) I SHAPEX<br>[8]  $\rightarrow$  NCOLS  $\leftarrow$   $\times$ / SHAPEY  $[P]$  NROWS  $\leftarrow \times /$  SHAPEX  $\leftarrow (-1) \downarrow$  SHAPEX<br>  $[8]$  NCOLS  $\leftarrow \times /$  SHAPEY  $\leftarrow 1 \downarrow$  SHAPEY<br>  $[P]$   $R \leftarrow (NROWS \times NCOLS) \rho 100$ [9]  $R \leftarrow (NROWS \times NCOLS) \rho 100$ <br>[10] CLOOP:  $\rightarrow$  END if NCOLS  $\leqslant$  CO [10] CLOOP:  $\rightarrow$  END if NCOLS  $\le$  COL  $\leftarrow$  COL + 1<br>[11] ROW  $\leftarrow$  -1  $ROW \leftarrow -1$ [12]  $CVAL \leftarrow Y[COL + NCOLS \times L W]$ <br>[13]  $RLOOP: \rightarrow CLOOP$  if NROWS  $\leq R$ [13] RLOOP: -> CLOOP <u>if</u> NROWS  $\leq$  ROW  $\leftarrow$  ROW + 1<br>[14]  $\leq$  R[COL+NCOLs+ROW]  $\leftarrow$  +/CVAL  $\times$  X[(W  $\times$  ROV  $[14] \rightarrow R[COL+NCOLs+ROW] \leftarrow +/ CVAL \times X[(W \times Row) + I+W]$ <br>[15]  $\rightarrow RLOOP$  $[15] \rightarrow RLOOP$ <br>[16] SCALX:  $\rightarrow$  SC.  $SCALK: \rightarrow SCALK$  if RANKY  $\in$  0 1 [17]  $R \leftarrow PLUSREDLA RANK Y.SHAPE Y. X \times Y$ <br>[18]  $\rightarrow 0$  $\frac{[18]}{[19]}$  $[19]$   $VECK: \rightarrow SCAIR$  if  $RANKY \in 0 1$ <br> $[20]$   $X \leftarrow X[(floor(pY) \times (pX) + pY]$  $[21] \rightarrow 0, R \leftarrow \text{PLUSREDLA RANK Y, SHAPE Y, X \times Y}$ <br> $[22] \quad VECY: R \leftarrow \text{PLUSRED RANK X, SHAPEX, X \times } (\rho X) \rho$  $[22]$  *VECY: R* ← *PLUSRED RANKX,SHAPEX, X* ×  $(\rho X)\rho Y$ <br> $[23]$  → 0 [23]  $\rightarrow$  0<br>[24] *SCALY*:  $[24]$  SCALY:  $R \leftarrow$  PLUSRED RANKX SHAPEX X x Y<br> $[25]$   $\rightarrow$  0  $\rightarrow 0$ [26]  $SCALR: \rightarrow 0$   $R \leftarrow +/X \times Y$  $[27]$  LERR: errorprocessing... [28]  $END: RANKR \leftarrow RANKX + RANKY - 2$ <br>[29]  $R \leftarrow RANKR. ((-1) \downarrow SHAPEX). (1 \downarrow SI)$  $R \leftarrow RANKR.((-1) \downarrow SHAPEX).(1 \downarrow SHAPEY).R$  $\nabla$ 

> **Figure 7-3:** Vector APL Function to do Inner Product on an Array of Arbitrary Rank

#### **7.2.4. Operators**

As soon as APL primitives are implemented as user-defined functions, such functions must be included in the domain of operators. Since an APL operator is essentially a frame which determines how a function will be applied to an array, one way to implement an operator that will handle a user-defined function is to write a user-defined function (that is, an ADEL function which runs in the PMU instead of a microcode function which runs in the DMU) which accepts a user-defined function as an argument. This leads to ALL user-defined functions being in the domain of operators, and with a very small additional amount of work, user-defined operators become available.

In the initial subset, reduction is implemented as hard microcode for all of the microcoded primitives All other operators, and reduction when applied to source-level primitives, are written as APL functions.

## **7.3. Microcode Size**

The present interpreter contains almost ail of the environment maintenance routines, about one-third of the executable formats, and on the order of a tenth of the executable operators. DMU format code has been restricted to vectors and scalars only, and the arithmetic operators to integers only.

Altogether, there are 2275 lines of microcode. excluding comments and headers. distributed as follows:



Since both the PMU and the DMU contain a copy of the memory manager, there are altogether 2492 lines of microcode in the system. However, the PMU and DMU do not share control memory, so the individual totals are more important than the combined total.

The microcode for the PMU is complete except for the executable formats. The final version of the PMU microcode should be about 2000 instructions.

The DMU will require a complete set of formats, extended to deal with floating-point numbers and arrays of arbitrary rank, plus code for all of the primitives that have not yet been written. Writing the entire DMU part of the interpreter in microcode would probably require about 8000 instructions. It would be interesting to see if a sufficient subset of APL could be squeezed into 2K DMU instructions, making possible a cheaper machine with most of the primitives and array code written as ADEL routines.

## **7A. Conclusions**

The implementation of the first subset of SAM APL has raised a number of interesting theoreticdl and practical problems. While there was no intention of implementing the entirety of APL in this version, the need for operators acting on user-defined versions of primitives made it necessary to do much of the groundwork for important extensions. The initial performance of some control primitives is very encouraging, whether or not they can be expressed comfortably in traditional APL. Difficulties arising from a half-thought-out speed-up to the index generator led to the necessity for read-only segments, which may prove useful in the implementation of nested arrays.

The simulated performance of SAM APL has been very satisfactory, a factor of more than 100 faster than the fastest microcomputer APL tested. Function call and return are

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still bottlenecks in SAM execution, and several feasible ways of speeding up call and return must be investigated. Performance measurement of different ways of expressing Ackermann's Function has apparently shown serious bottlenecks in the way other APL implementations handle array references.

It is worth noting that it took about one year's work before the first APL expression

 $X \leftarrow 1$ 

could be executed. Extensions and improvements have been much quicker to define and test given a working nucleus upon which to build It is probably premature to guarantee that a complete working APL interpreter will be easy to develop from the current state of the project, but at the very least there are good grounds for optimism.

# **Appendix A**

# **Glossary of Acronyms**





## **Appendix B**

## **Extensions to APL**

The present project has been limited to implementing a subset of the APL language. with a single small extension in the form of control primitives. The next step will be to implement the entire APL language, or at least a sufficient set of primitives to allow the remainder of the language to be written in APL itself.

The implementation of real extensions to the language must be postponed until standard APL is finished. However. all existing commercial APL interpreters contain significant extensions, and it is proper to discuss them here, if only to get some idea about which shouid be considered for SAM APL.

## **B.1. Nested Arrays**

Nested arrays are by far the most important extension to standard APL. A standard array consists of a rectangular block of characters or of numbers of a given type. While this is a simple and natural way to represent financial tables and engineering matrices, it is much less flexible than, for example, a Lisp list. Suggestions for "generalized" or "nested" arrays where each element of an array is (or contains) another array. have been circulating since the first useable APL implementations. [Fal73, GhM73]

After about a decade and a half of discussion, three major implementations of nested arrays were completed The first to be released was I. P. Sharp's system. Then came the NARS system of STSC APL, and finally IBM's APL2 Another important nested array APL is the Dyalog APL available for UNlX systems.

Sharp's system differs from the other commercially available systems by being "grounded". that is, an the enclosure of a simple<sup>1</sup> scalar object is distinguishable from the object itself. The other systems are "floating", so that an enclosed simple scalar is identical to the original simple scalar.

The obvious consequence of this distinction is that "floating" APL requires heterogeneous arrays, since the catenation of an enclosed character and an enclosed number is identical to the catenation of the character and number. "Grounded" **APL** may permit heterogeneous arrays (although no implementations do so), but does not require them.

A less obvious consequence is the pervasiveness of arithmetic and logical functions. Since in a floating system there is no way to distinguish a number from an element of a nested array containing a number. then nested arrays are in the domain of arithmetic and logical functions. In a grounded system, nested arrays may or may not be in the domain of these functions In practice, a set of composition operators was defined, such that

 $(X+with$  disclose Y)

means "Go through arrays X and Y element by element. For each pair of elements, disclose them, apply the plus function, and enclose the result as an element of the result array "

The simple provision of a nested array data type, plus the enclose and disclose functions, is only the first step in providing a useful nested array language. Providing a way to apply mathematical and logical functions to the contents of nested arrays is the second step Many more functions and operators must be defined to allow the

<sup>&#</sup>x27;A **"simple" scalar is an ordinary nurnbet or character, but not the scalar result of an enclose operation** 

manipulation of "nestedness" as simply and powerfully as standard APL operators deal with l'arrayness'l

Writing the microcode to support nested array operations would be a large but fairly straightforward project Consider the DMU microcode for the DLR format (Figure B-1):

DLR\_D Read Args from Pipe Basic Semantic Check(Neither Arg Undefined). Cast if types different Read 2 words from L and 2 words from R ! CASE on argument ranks j-(Both Scalar) EXEC the function /-(Left Scalar) CALL 'DLR-SV' i-(Right Scalar) CALL 'DLR-VS' ;-(Both Vector) CALL **'DLR-VV'** 

CALL 'CLR\_WRI\_DST'

**Figure B-1:** Broad Outline of DLR Microcode

To handle nested arrays, assume that a floating system is to be implemented, so that  $X \leftarrow Y + Z$  might refer to a scalar Y and a nested array Z. Then the DLR routine in the PMU would not have to be changed at all. In the DMU, a test would be inserted after the basic semantic check, as in Figure (B-2). Assume that a nested array has a DMU tag of 1100.

In Figure B-2, the additional test in DLR is shown, followed by a rough outline of the

IF L tag is 1100 CASE on R tag |-1100: CALL 'DLR NN' /-lOxx CALL 'DLR-NV' 1-Olxx. CALL 'DLR-NS' ELSE IF R tag is 1100 CASE on L tag /-IOxx: CALL 'DLR-VN' I-Olxx: CALL 'DLR-SN' ELSE (Rest of flowchart as before)

DLR SN (DMU Microcode for DLR. Scalar L and Nested R) Initialize a count for length of R Open segment containing ptrs for R

> LOOP: Read one pointer for R CASE on tag for the pointer 1-SCALAR. EXEC the function '-VECTOR CALL 'DLR-SV' NESTED: CALL 'DLR-SN'

> > **Figure 8-2:** Change to DLR Microcode to Support Nested Arrays

DLR\_SN microcode that would be called where the left argument is scalar and the right argument is nested One serious problem that would have to be solved is the recursive nature of nested arrays If an element of Z is itself a nested array, the DLR-SN function simply calls itself. Since the nature of microcode is not recursive, a different programming style would be needed

The shape- manipulation functions could operate on nested arrays exactly as if they were arrays of integers. In fact. there is no reason that any of the shape-manipulation code should 'know' that its argument is a nested array. The problem of 'pad' elements for expand and overtake would disappear if we set up the tables so the value pointed to by 0 is the fill value for a nested array. Presently, the DAT element in position **0** is in fact scalar zero, which is a suitable fill element.

The treatment of nested arrays in this section has necessarily been superficial, and the fact that the examples have been simple is not meant to suggest that the implementation would not be difficult. As the interpreter is developed. I expect to include a very restricted set of nested array operations. possibly limited to Enclose and Disclose. Such a limited nested array system is of little practical use, but it should at least help to prevent implementation decisions that would interfere with a real nested array system in the future.

## **8.2. Operators**

The use of operators to apply functions to arrays is the keystone of APL. Standard APL contains only four such operators: reduction, axis, scan, and product. New operators have been suggested almost as often as new functions. In particular, new operators have been necessary for the manipulation of nested arrays.

Recently, [IPS841 published the beginnings of a formal calculus of operators The completion of such a calculus might allow the development of a new programming language based upon it, as of nested array theory led to the development of Nial. Such a language might resemble APL only loosely. The extended version of APL proposed by Kenneth lverson is based on the consistent application of operators [lve86].

The ultimate operator extension is provided in APL2. user-defined operators Together with allowing operators to apply to user-defined functions, this completes the process of

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using the APL operator to create a language that is totally different in flavour from nonoperator languages.

The ADEL intermediate language developed for the SAM APL system encodes the standard APL operators within its format syllables. At present there are no formats for unspecified operators. Adding new operators to an APL implemented via translation into ADEL would require either a new set of formats for each new operator or a set of unspecified-operator formats. New operators could be also realized either by translating them into nested loops during the front-end APL-to-ADEL translation phase, or by trapping to error routines which then call APL programs which do the control operations represented by the new operator. The implementation of operators which apply to sourcelevel representations of some of the APL primitives will make this extension straightforward

# **B.3. File Systems**

Traditional APL does not have a file system. Early implementations had no way at all of keeping data outside of an APL workspace: everything that was entered or generated in a given session would be stored when the active workspace was saved, but that information could not be used by non-APL tasks, and no more of it could be retained than would fit in an APL workspace. Some ways of using APL workspaces as data bases are discussed in [Soo84]

It was very early obvious to most APL theoreticians that the "right" way to implement APL files was as extra-workspace nested arrays. An early feasibility study (lve831 suggested that a practical nested array system was ten years and hundreds of man-years away in the early 1970's. but a restricted subset could be implemented in about a year. The result of this effort was the Sharp/STSC component file system.

An APL component file a file is essentially a vector of APL arrays, kept outside of the workspace and accessed by a set of special file functions instead of standard APL array manipulation functions. Component file systems make practical commercial applications feasible, and have been very successful despite their stopgap nature.

The other approach to files in APL was to create a "shared variable" system, wherein a non-APL file AP (Auxiliary Processor) could trade information with an APL task and perform the job of accessing various proprietary file systems and translating data into and out of APL format. Like component file systems, AP-based file systems are commercially practical but clumsy to use. One very useful offshoot of the AP file system concept has been the shared variable facility, which can be used for inter-task communication within and outside of APL.

Since APL will be used on real computers which support other applications. some thought should be given to file interfacing. For example, an APL designed to run on an IBM PC-compatible computer should have some way of dealing with native DOS files. One attractive way to get data base facilities would be to write a useable interface to one of the commercial data base management systems.

## **B.4. Primitives**

Despite the richness of the set of operations created by applying standard APL operators to standard APL functions, it is not hard to define a problem that is most efficiently . solved by extending the set of primitive functions. This can be done either by adding new primitives or extending the definitions of existing primitives.

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#### **B.4.1. New Primitives**

"String-Index" is one example of a function that can easily be written in standard APL but is used heavily enough that an optimized primitive might be desirable. APL supports character vectors, not strings, so that

**'THIS IS A SENTENCE 'L 'IS'**  2 **3 'THAT WAS A SENTENCE'STRINDX 'AT'**  2

A primitive String\_Index function (and a companion String\_Membership function) would simplify and speed up text processing applications, particularly if the underlying hardware includes string-handling facilities

Various other primitives that have been suggested from time to time include eigenvalue and eigenvector, functions to convert back and forth between matrices and delimited vectors an assortment of graphics primitives, and functions to handle nested arrays and other data structures.

#### **B.4.2 Extending Existing Primitives**

Many features of contemporary APL systems are extensions of functions found in standard APL For example the standard sorting primitives do not apply to matrices or to character vectors. Most contemporary APLs have extended the domain of the sorting primitives to matrices and higher-rank arrays, such that

### $RESULT \leftarrow MATRIX[gradeup MATRIX!]$

sorts MATRIX with the first column in ascending numeric order, with ties broken by the second column and so on An optional left argument contains a collating sequence, so the function can be used to sort a character array.

Another very common extension is the Replicate function. In classical APL.

01011/12345

returns 2 4 5. That is, elements of the right argument corresponding to 1's in the left argument are selected. Another way to look at it is the number of copies of a given element of the right argument is equal to the corresponding element of the left argument. either 0 or 1 Extending this second definition gives the Replicate function:

10203/56789

returns 5 7 7 9 9 9.

# **8.5. Strands and Syntactic Variations**

It is convenient to be able to denote a single literal consisting of a vector of numbers in the form

2 4 6 7 12

instead of

 $(2.4.6.7.12)$ 

Strand notation extends this convenience to all APL variables instead of just numeric constants. For example:

 $X \leftarrow 5$  6 7  $Y \leftarrow '$  These are characters'  $R \leftarrow$  **FOO X Y** 

applies the function FOO to a two-element (nested) vector consisting of X as the first element and Y as the second element. More radically. "strand assignment" can be used to give a value to more than one identifier in one statement. For example  $X Y \leftarrow Y X$ swaps the values of  $X$  and  $Y$ . All APL implementations with floating nested arrays support

strand notation Implementations with no nested arrays, or with grounded nested arrays, do not support it

Strand notation leads to a syntax based on "binding strength" [Bro85] The resulting syntax is close to standard APL except for the existence of strands and such curiosities as

 $1 \ 2 \ 3[2] \leftarrow \rightarrow \ 1 \ 2 \ (3[2])$ 

As imported into APL, strands are at best a mixed blessing. The Nial language is based on array theory [Mor73] with strand notation as the basic input format, and it may turn out that stranded dialects of APL will evolve into a language more like Nial than conventional APL.

## **B,6. New Environment Features**

The standard APL environment is cramped **and** difficult **to work** with APC does not communicate easily with other systems is not easy to interrupt, and does not provide any easy way to run multiple environments and protect subtasks from each other Many proposals have been made for minor and major modifications to the APL environment

#### **B .6.1. Scoping**

One very useful proposal was to extend APL's scoping rules [SAL78]. The standard APL name scoping is dynamic with default global. If a function  $FOO$  calls a function FEE and **FEE** happens to have a loop counter. I which the programmer has neglected to "localize" by including it in the function header, then when  $FEE$  exits.  $FOO$  is left with the value of  $I$  set by  $FEE$ . If  $FOO$  also happens to use  $I$  for some variable, a bug has been created

Seeds et, al proposed that the rules be modified to the programmer could choose the default scoping. and that a "strictly local" scope be added such that a local variable could be made invisible to both called functions and the calling function. This simple change to APL would have eliminated about half of the serious errors in my own code over the last seven years

#### **8.6.2. Event Trapping**

The largest extension to the standard APL environment which has actually been implemented on a commercial scale is Event Trapping. When a condition arises which cannot easily be handled by straightforward coding, a trapped APL allows the programmer to specify what action should be taken. Event Trapping has already been implemented Sharp APL and to a lesser extent on the other timesharing systems. In the Sharp event handler, any function could have a local variable quadTRAP, which contains event codes and APL expressions to be executed when the event arises. For example. "Workspace Full" has event code 1, so if quadTRAP contains "1 RemoveGarbage" and a function encounters Workspace Full, the RemoveGarbage function is invoked. A companion quadSIGNAL system function allows event codes to be raised under program control

Event trapping makes a large number of problems much easier to solve. For example. if a user has managed to enter bad data into an application program. the program can flush the input report the situation to the program maintenance file, give the user a more-orless friendly message instead of a cryptic (to the user) APL error message, and restart in a convenient place.

Extending event trapping to handle interrupts generated from outside the APL task would lead to the possibility of real-time device control programs, more reliable multiple-input-task systems. and more accurate models of computer systems A simplified version of this idea was implemented by [ATC77] for sensor-based applications on the IBM System/7 computer.

#### **B.6.3. Namespace Extensions**

The global scoping of APL names makes it difficult to hide and protect information in a large system. For example, when a given problem can be solved by using sets of programs in two separate workspaces, the two workspaces can not be copied together without checking for name conflicts and renaming objects to avoid conflicts. A very relevant example is the machine model on which the ADEL interpreter is being developed. The model encompasses two separate processors, and to switch back and forth between them it is necessary to rename all of the data objects in the workspace.

The straightforward way to implement multi-namespace systems in APL is to use an APL which supports multiple tasks, and pass information between them using shared variables. All of the important commercial mainframe APL interpreters (and none of the currently available microcomputer APLs) include a full shared variable facility

MCM APL [MCMxx] included a one-level structuring system. Objects from one or two of 256 collections of programs and arrays could be accessed at any one time. Using this system. tasks could be integrated with little programming effort, although the performance of the diskette-based hardware was quite slow.

A more flexible system was proposed in [BeD82]. A new set of system commands is used to create, load, and save heirarchical "segments". Whenever an executing function encounters a name which is not in the current workspace, the segment with that name is loaded and the reference is resolved in the new context (which means the segment name must be identical to the name of a new object within that segment)

The Analogic APL machine ( [Ber84]) handles name partitioning by threading through multiple "instances" of a workspace with one "prototype". Each workspace is partitioned into "contexts". If a name in a given context is explicitly exported, it may be references in another context, with the value being either a private copy (if the context is an instance) or a permanent stored value (if it is a prototype).

The most drastic namespace handling facility proposed to date is [TaW84]. All APL objects and tasks are subsumed under "The Tree" A complete set of naming and binding procedures is defined, after which different tasks can communicate by using qualified names

The completed SAM APL may include a modest namespace extension suitable for separating multiple instances of a set of programs while allowing controlled communication between them The content of the extension will not be worked out at this time The mechanics of the implementation should be fairly simple. Each context would contain its own set of symbol table pointers, so that the linker but not the execution supervisor must distinguish between different objects with the same name The details of the extension will require a careful look at what is most useful and easiest to implement.

# **8.7. New Data Types**

#### **8.7.1. Complex Numbers**

The numeric domain of Sharp APL and APL2 include complex numbers. It is fairly easy to model complex numbers as arrays with an extra axis of length 2, but it is more convenient to have a complex primitive data type which is in the domains of all of the relevant primitives A further extension related to complex numbers would be quaternions, 4-tuples of real numbers encompassing complex numbers and 3-space vectors as special

cases. Although quaternions have never been implemented as primitive data types, [Gus861 points out that they be useful for some graphics operations.

#### **8.7.2. Beyond Floating Point Numbers**

The APL scheme of booleans extending automatically to integers which extend automatically to floating point numbers handles most mathematical problems conveniently. One common extension is multiple integer types, to increase speed and reduce storage costs for small integers An "infinity" value can be handled by current hardware and would be useful for catching such problems as the DOMAIN ERROR in (arctan(Nxtan(X))) where X approaches 90 degrees. Financial applications would benefit from a fixed decimal storage type. Other possibilities are large integers, rational numbers. and multiple floating-point types to handle overflow/underflow from the default floatingpoint type

#### **8.7.3. Graphics**

Graphics primitives will certainly be necessary in any new general-purpose language. Efficient graphics applications may require special data types which do not map neatly into arrays At very least. a set of more than 256 characters will be needed.

## **8.8. Control Structures**

The operator-based approach to arrays make most ALGOL-type nested loops unnecessary in APL. However, instances requiring loops and branches are not supported by anything but the branching and execution primitives The branch primitive suffers from both high cost and a lack of safeguards against branching into loops. spaghetti code, and "driftwood" (floating branches). Execution is too powerful and general a mechanism to use for a simple IF statement, and implementing it efficiently requires a translation into a test-andbranch or equivalent.

The version of ADEL developed for the first SAM APL supports a BIFSTK branch and a pair of looping primitives to initialize-and-test and increment-and-test The translator does not recognize corresponding objects in the source code: at the current stage, they are only available in hand-assembled programs.

The idea of introducing ALGOL-style control structures into APL has often been raised and has never been accepted APLgol [Ke173] is the best-known such implementation. MCM APL contained a way of editing control structures into APL programs Neither of these implementations has convinced the mainstream of APL programmers that such control structures are useful in APL.

The "EACH" operator found in floating nested array systems is a control structure with an APL flavor.

#### FOO EACH X

applies the function FOO to each element of the array X. replacing an explicit loop with an implicit loop. Other operators for program control are described in [Eu85A, Eu85B].

## **B .9. Arrays of Functions**

Once APL arrays have been generalized to contain arrays as elements, a further generalization would be to allow arrays to contain functions One possible application of arrays of functions would be as a structured environment, with sets of functions needed for a particular purpose kept together in a particular subarray.

A branch of APL theory has developed dealing with the application of function arrays to data arrays (Ben86, Lan86). At first thought, this kind of operation is too far from the original conception of APL to be integrated smoothly into the language. Arrays of **functions resemble a generalization of the Nial "Atlas" as in the following example from [JGM86].** 

**Average IS div[sum.tally]** 

**As in the case of strand notation and a calculus of operators, it will probably be necessary to develop a new language to apply arrays of functions in a general way.** 

# **References**













