# **Apple Array Allocation**

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Array languages like J and APL suffer from a lack of embedability in implementations. Adroit memory management can make embedding easier; one would like to avoid thinking about ownership across two garbage collectors. Here we present statically determined memory allocation used in the Apple array system, a compiler for a high-level, expression-oriented functional language. Arrays are usable as pointers, with no dependence on a garbage collector. Ownership is simple and Apple code does not constrain memory management in the host language. Two embeddings—one in Python and one in R—attest to the feasibility of this method.

#### 1 INTRODUCTION

Array languages such as APL and J suffer from undeserved obscurity [1]. In order to make wider use of array languages, one might hope to embded them. So we look to C—we would like an array language that does not impose on its host language [3]. C is almost vulgar in offering pointers as arrays; the typical APL implementation uses reference-counting [2, p. 47] for performance reasons. But this means that arrays depend on a garbage collector running alongside to behave sensibly. By compiling high-level array expressions to a C procedure with all allocations and frees predetermined, we reduce the demands of calling Apple code in other environments.

### 2 METHOD

### 2.1 Flat Array Types

Apple arrays are completely flat in memory, consisting of rank, dimensions, and data laid out contiguously.

A 2x4 array:

0	1	2	3	4	5	6	7	8	9	10
2	2	4	$a_{00}$	$a_{01}$	$a_{02}$	$a_{03}$	$a_{10}$	$a_{11}$	$a_{12}$	$a_{13}$

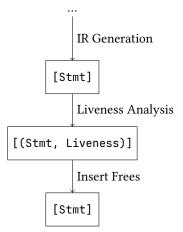
All data are flat; there are only two primitive types: 64-bit integers and 64-bit floats. Arrays of tuples are supported but not arrays of tuples of arrays; there are no user-defined types.

Such constraints, with extra bookkeeping information added during IR generation, are responsible for the surprising fact that memory allocation can be completely determined at compile time. Though they may seem spartan to a functional programmer, this is enough to stand toe-to-toe with NumPy.

## 2.2 Compiler Pipeline

Our work is based on established liveness algorithms; we annotate statements with uses and defs and thence compute liveness intervals for arrays.

2 Anon.



The IR used in the Apple compiler is sequences of statements and expressions, viz.

Expressions can read from memory via At and Stmts make use of memory access for writes, allocations etc. A function

```
aeval :: Expr -> IRM (Temp, Lbl, [Stmt])
```

translates an array expression, assigning a Lbl for subsequent access and associating the labeled array with a Temp to be used to free it.

All accesses to an array use the ArrayPtr type, which specifies the Lbl for tracking within the compiler.

```
-- register, offset, label
data ArrayPtr = ArrayPtr Temp (Maybe Exp) Lbl
   Then one has
defs :: Stmt -> IntSet
defs (Malloc l _ e) = singleton l
defs _ = empty

uses :: Stmt -> IntSet
uses (Write (ArrayPtr _ _ l) e) = insert l (defsE e)
uses
```

Note that with such labels one can access the array from different Temps; this is necessary for generating efficient code in the compiler.

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```
With defs and uses thus defined, we compute liveness intervals for Stmts [4].
```

```
data Liveness =
  Liveness { done :: IntSet, new :: IntSet }
```

For each (Stmt, Liveness), when we encounter a label in the done set, we look up the Temp to free it and insert a free statement.

#### 2.3 Generation

apple\_0:

(movtemp r\_0

(free r<sub>2</sub>)

apple\_1:

For instance, we do not write:

We must take care when arranging loops; the loop should exit by continuing rather than jumping to an exit location. This ensures that the frees inserted at the end of live intervals are always reachable.

 $(condjump (< (reg r_2) (int 2)) apple_1)$ (movtemp r\_0 @(ptr r\_1+(+ (asl (reg r\_2) (int 3)) (int 16)))) (jump apple\_0) apple\_1: But rather: (condjump ( $\geq$  (reg r\_2) (int 2)) apple\_1) apple\_0: (movtemp r\_0 @(ptr r\_1+(+ (asl (reg r\_2) (int 3)) (int 16)))) (condjump (< (reg r\_2) (int 2) apple\_0)</pre> apple\_1: Had we written the former, the free would be unreachable, viz. apple\_0:  $(condjump (< (reg r_2) (int 2)) apple_1)$ (movtemp r\_0 @(ptr r\_1+(+ (asl (reg r\_2) (int 3)) (int 16)))) (jump apple\_0) (free  $r_2$ ) apple\_1: This is not the case with the latter: (condjump ( $\geq$  (reg r\_2) (int 2)) apple\_1) apple\_0:

@(ptr r\_1+(+ (asl (reg r\_2) (int 3)) (int 16))))

 $(condjump (< (reg r_2) (int 2) apple_0)$ 

4 Anon.

Since jumps are only generated by the compiler in a few cases, we can guarantee that the generated code is correct by being careful.

#### 3 EMBEDDINGS

Apple has been embedded in Python and R. Observe the following interactions:

```
>>> import apple
>>> area=apple.jit('''
λas.λbs.
    \{ \Sigma \in [(+)/x] \}
     ; 0.5*abs.(\Sigma ((*)`as (1\Thetabs)) - \Sigma ((*)`(1\Thetaas) bs))
    }
111)
>>> import numpy as np
>>> apple.f(area,np.array([0,0,3.]),np.array([0,4,4.]))
> source("./apple.R")
> shoelace<-jit("
λas.λbs.
    \{ \Sigma \in [(+)/x] \}
     ; 0.5*abs.(\Sigma ((*)`as (1\Thetabs)) - \Sigma ((*)`(1\Thetaas) bs))
")
> run(shoelace,c(0,0,3),c(0,4,4))
```

Thus we can run Apple code on NumPy arrays and R vectors without frustrating the garbage collector.

Apple code can be called from C on its own terms, unlike NumPy or J, viz.

```
#include <stdio.h>
#include <stdlib.h>
#define R return
#define DO(i,n,a) {I i;for(i=0;i<n;i++){a;}}
typedef double F;typedef int64_t I; typedef void* U;
typedef struct Af {I rnk; I* dim; F* xs;} Af;
U poke_af (Af x) {
    I rnk = x.rnk;
    I t = 1;
    D0(i,rnk,t*=x.dim[i]);
    U p = malloc(8+8*x.rnk+8*t);
    I* i_p = p; F* f_p = p;
    *i_p = rnk;
    D0(i,rnk,i_p[i+1]=x.dim[i]);
    D0(i,t,f_p[i+1+rnk]=x.xs[i]);
    R p;
```

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```
extern F shoelace(U, U);

int main(int argc, char *argv[]) {
    F xs[] = {0,4,4};
    F ys[] = {0,0,3};
    I d[] = {3};
    Af a = {1,d,xs};
    Af b = {1,d,ys};
    U x = poke_af(a);U y = poke_af(b);
    printf("%f\n", shoelace(x,y));
    free(x);free(y);
}
```

### 4 CODA

Apple has the potential to be far more efficient; one could consolidate allocations, e.g.

```
irange 0 20 1 ++ irange 0 10 1
```

performs one allocation for each irange but this could be consolidated into one—irange 0 20 1 is inferred to have type Vec 20 int in the existing compiler; with liveness and size information one could do something like linear register allocation in which arrays are assigned to memory slots.

This vindicates flat arrays; while banning arrays of functions &c. may not be obvious to a functional programmer, studying this particular set of constraints has potential; one could use it to implement a compiler for a high-level language that produces performant and portable code.

### REFERENCES

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