Toward the Efficient Compilation of Fair Search

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(joint with Andrew Tolmach)

Outline

- (Brief) FLP intro
- Fair vs unfair search
- Implementing fair search
- First-class stores
- Design constraints
- Data representation
- Garbage collection
- Results, conclusion

Functional Logic Programming

- One paradigm unifying functional and logic programming
- From the functional world (Curry syntax):
 - (Higher-order) functions

s f g x = f x (g x)

Algebraic datatypes, pattern matching
 data T = B Bool T T | L

flatten L = []
flatten (B x l r) = flatten l ++ x:(flatten r)

Haskell
 Curry (or it can be)

Functional Logic Programming

- From the logic world:
 - Logic variables, built-in search

flatten x where x free

instantiates $\mathbf x$ to $\mathbf L$ or (B $_$ $_$ $_$) and looks for solutions

Nondeterminism

```
coin = 0
```

coin = 1

(evaluating coin yields either 0 or 1, nondeterministically – in theory)

Built-in Search

Back to our example:

data T = B Bool T T | L
flatten L = []
flatten (B x l r) = flatten l ++ x:(flatten r)

We load this program in Pakes and run the following goal: test> flatten X Free variables in goal: X Result: ([]) Bindings: X=I. ?

(Capital letters denote free variables in the top level.)

In a nutshell: When a free variable X is demanded, e.g., our operation (flatten X):

1. Find the set of values (constructors) that it can be instantiated to, based on its type.

The type of X is T. Possible values are L and (B $X_1 X_2 X_3$), where X_i s are free.

2. For each value v, instantiate X to v and compute (f X).

flatten L \Rightarrow [] (our solution!) flatten (B X₁ X₂ X₃) \Rightarrow flatten X₂ ++ X₁:(flatten X₃) The latter will spit out an infinite number of solutions.

Fair vs. Unfair Search

How do we find solutions algorithmically? On previous slide:

For each value v, instantiate X to v and compute (flatten X).

Strategy A Instantiate X=L, compute (flatten X) until no more computation can be done. Then instantiate $X=(B X_1 X_2 X_3)$ and repeat.

Strategy B Spawn two parallel computations, one for X=L and one for $X=(B X_1 X_2 X_3)$ and compute (flatten X) in each.

Strategy A (depth-first search) is the norm. Popular FLP implementations, including Pakcs and the Münster Curry Compiler (MCC), search for solutions depth-first.

A well-known defect of depth-first search is that it is *incomplete*. Most simply, when

f (B x r l) = undefined
f L = []

(f X) should yield the solution [] at some point. If we search depth-first and pick the first branch before the second, this never occurs.

Fair Search

Strategy B (breadth-first search) is *complete*. All solutions that can be computed in a finite amount of steps will be discovered in a finite amount of time.

- ♦ We compute all alternatives in (pseudo-)parallel.
- Each alternative reports its result when it is done computing.
- If some alternatives diverge, the program will naturally diverge. But not before reporting all computable solutions.

Implementing Fair Search

Each search alternative computes independently from the others. Therefore, some data involved in the computation *must* be distinct for each alternative.

A High-Level Approach We are computing (flatten X):

- Replicate the term n times, where n is the number of possible values for X.
- 2. In each replica, substitute X for one of the values.
- 3. Proceed with computing all replicas in parallel.

Fair Search: Term Replication

(flatten X)

$$\Rightarrow \begin{bmatrix} (\texttt{flatten L}) \\ (\texttt{flatten } (\texttt{B} X_1 X_2 X_3)) \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} \\ [\\ (\texttt{flatten } X_2) + X_1 : (\texttt{flatten } X_3) \end{bmatrix}$$

 $\Rightarrow \begin{bmatrix} (\texttt{flatten L}) + X_1 : (\texttt{flatten X_3}) \\ (\texttt{flatten (B X_4 X_5 X_6)}) + X_1 : (\texttt{flatten X_3}) \end{bmatrix}$

Fair Search: Term Replication

Possible problem: duplicating large terms can be a big time and space expense.

The situation can be improved by avoiding replication of sub-terms which do not contain free variables.

```
E.g., (not X, reallyBigTerm) yields
 (True, E1)
 (False, E2)
```

If reallyBigTerm doesn't contain free variables, E₁ and E₂ will point to the same object, which was obtained by reducing reallyBigTerm *once*.

Fair Search: Store Replication

A Low-Level Approach

- ♦ We compile all functions to fixed machine code.
- We largely abandon notions of "term duplication" and "reduction." (Machine code can't be duplicated anyway.)
- We introduce a concept of *store*: mapping from logic variables to values. Each alternative has its own store that is independent from all other stores.
- When the value of a free variable is demanded, we duplicate the store instead of the term.

Fair Search: Store Replication

 $(\texttt{flatten X})_{\mathtt{s} \equiv \{\mathtt{X} \mapsto \mathtt{Free}\}}$

$$\Rightarrow \begin{bmatrix} (\texttt{flatten } X)_{\texttt{s}' \equiv \texttt{s}[X \mapsto L]} \\ (\texttt{flatten } X)_{\texttt{s}'' \equiv \texttt{s}[X \mapsto (\texttt{B} \ \texttt{X}_1 \ \texttt{X}_2 \ \texttt{X}_3), \texttt{X}_1 \dots \texttt{X}_3 \mapsto \texttt{Free}]} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} \square_{s'} \\ (\texttt{flatten } X_2) + X_1 : (\texttt{flatten } X_3)_{s''} \end{bmatrix}$$

 $\Rightarrow \begin{bmatrix} (\texttt{flatten } X_2) +\!\!\!+ X_1 : (\texttt{flatten } X_3)_{s''[X_2 \mapsto L]} \\ (\texttt{flatten } X_2) +\!\!\!+ X_1 : (\texttt{flatten } X_3)_{s''[X_2 \mapsto (B \ X_4 \ X_5 \ X_6), X_4 \dots X_6 \mapsto \texttt{Free}]} \end{bmatrix}$

First-Class Stores

Logic variable stores are first-class stores (FCS):

- Stores are mappings from store references to values
- Stores are objects that can be passed in and out of functions, stored in variables, etc.
- Stores can be copied (or "checkpointed")
- Store references can be dereferenced in any "valid" store.

FCS: Interface

A C interface for our formulation of first-class stores:

typedef Fcs ...

typedef SRef ...

Fcs fcs_new(void);

Fcs fcs_checkpoint(Fcs);

- SRef fcs_allocate(Fcs, Value);
- void fcs_update(Fcs, SRef, Value);

Value fcs_deref(Fcs, SRef);

Important: If a reference exists in a store s, it also exists in any store $s^2 = fcs_checkpoint(s)$, for as long as s^2 is live.

FCS in FLP

- We use stores to map logic variables to values.
- ♦ A store reference corresponds to a logic variable.
- The code manipulates only store references and is store-agnostic. (This will come in handy!)
- The fcs_checkpoint operation is used whenever stores must be duplicated. (Whenever we spawn parallel search alternatives.)
- Intuitively, two search alternatives running the same code (dereferencing the same references) will have different outcomes if their stores are different.

Goal of This Work

We want an efficient *compiler* for FLC that employs *fair* search via store replication. Pieces involved:

- A compiler from a high level FLP to machine code
- An efficient data representation and implementation for first-class stores
- A runtime system composed of
 - concurrency primitives
 - a garbage collector that correctly and efficiently collects stores

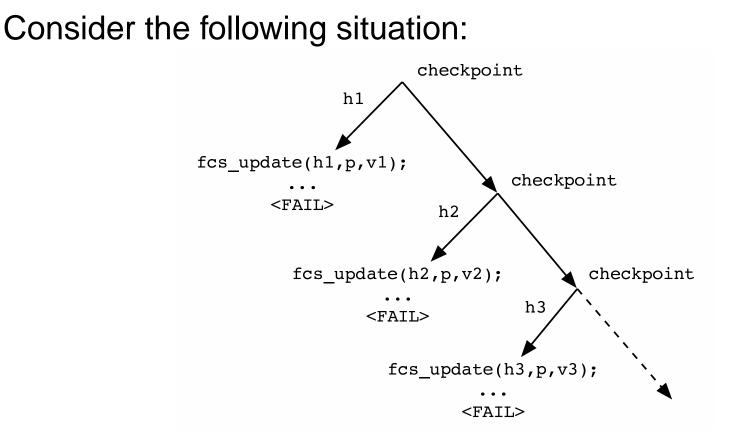
This work addresses items in blue.

Data Representation

- One crucial observation: For a store value to be live, it must be reachable via a reference and a store.
- ♦ If it is only reachable through a reference, it is garbage.
- ♦ If it is only reachable through a store, it is garbage.
- Intuitively, we need both a store and a reference in order to pull out the value and use it. Otherwise it's out of our reach.

Consequence: Some values may be garbage but still reachable!

Data Representation



If p remains live as the computation proceeds, we must make sure that v1, v2, etc., do not.

FCS: A Simple Approach

We could represent first-class stores as contiguous arrays and references as integer offsets into stores.

Pros: Fast dereferencing.

Cons: Checkpointing is inefficient, especially space-wise.

Serious cons: Proper collection is intractable! When a reference becomes garbage, we need to kill off its corresponding values in all the stores that it can be dereferenced in.

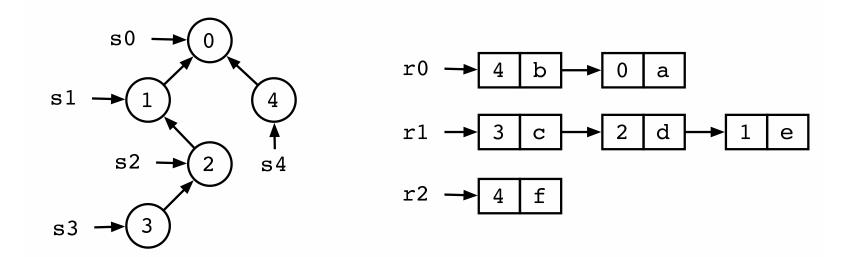
FCS: Getting Trickier

We represent first-class stores as integer lists and references as lists of possible values.

- A reference is a list of (t, v) pairs, where t is a tag that uniquely identifies a store.
- A store s is a list t₁, t₂, ..., t_n of tags, where t₁ is the (unique) tag of s, t₂ is the tag uniquely identifying its "checkpoint parent," and so on.

This trick exploits an important invariant in FLC: after a s is duplicated in the process of spawning two or more search alternatives, s becomes unreachable!

FCS: Linked Lists

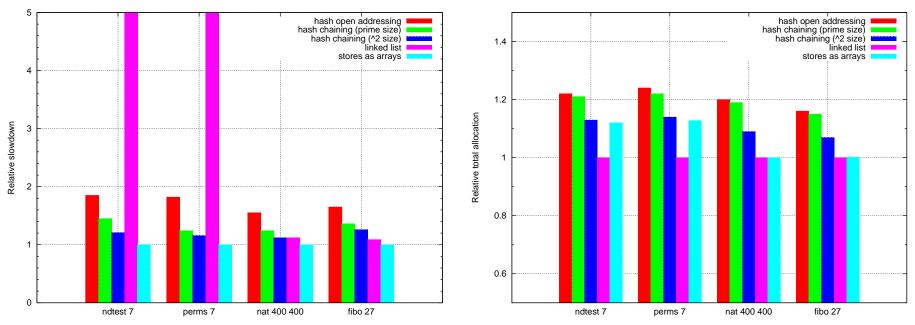


Pros: Makes good use of space and fixes our GC problem. (We will see how in a few slides.)

Cons: Dereferencing is slow. A value may be looked up in several stores before it is found.

FCS: Hash Tables

The linked list approach is inadequate due to very slow dereferencing. Replacing linked lists with hash tables in the reference representation produced significantly better results.



GC: General Guidelines

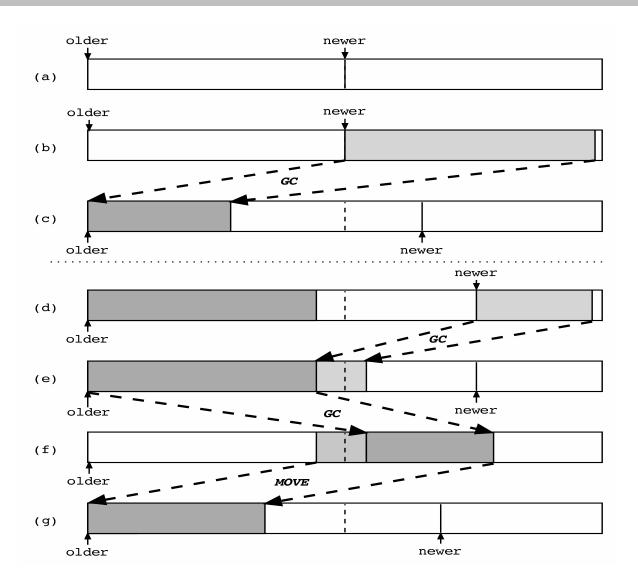
- Most data die young. Approx. 80% of the data is garbage by the time a collection should happen.
- Allocation must be very fast. The allocation routine holds the highest call percentage in the runtime.
- ♦ Frequent allocation ⇒ frequent collections. Since most data die young, we would like to avoid collecting "too early."

Generational GC

We decided on a generational copy collector, following Appel's well known model [Appel '89].

- Implicitly compacting, so allocation happens in a contiguous free space. I.e., allocation is fast.
- Generational means collection is divided in two phases:
 - Minor: collect a little data at a time, frequently.
 - Major: collect the entire heap, much less frequently.
- Gives data more time to die.

GC: Mechanics



GC: Liveness Adjustment

- We represent references as hash tables from store tags t to elements (p,v), where p is a *pointer* to the store object uniquely identified by t.
- Before doing a full collection, we first forward all live stores to the "to" space.
- During collection, when scanning live references, for each (p,v), we can quickly determine if p is garbage: p is in the "from" space and lacks a forwarding pointer.
- If p is garbage, we remove (p, v) from the table and fail to scan it.

Some Numbers

	ndtest 7	perms 7	nat 400 400	fibo 27
none	1.51	1.53	2.15	4.75
coll	1.98	2.31	5.93	13.90
augm	2.21	2.63	1.87	10.81

Micro-benchmark speed comparisons (seconds):

- none: no collector at all
- coll: generational copy collector
- augm: generational copy collector with adjusted notion of liveness

Some Numbers

	ndtest 7	perms 7	nat 400 400	fibo 27
coll	0.30/0.54	0.69/0.48	3.54/0.51	5.25/1.19
augm	1.22/0.02	0.82/0.48	0.02/0.35	4.30/0.09

Time spent in the collector (seconds), accompanied by time spent in the dereference operation.

	ndtest 7	perms 7	nat 400 400	fibo 27
coll	75.83	74.77	76.33	73.12
augm	21.07	22.89	7.43	71.54

Percentage of live data (calculated at majors).

Future Work

- Speed everything up!
- We are working to improve the GC code that does the explicit unlinking, as it is central to GC performance.
- Strictness analysis to avoid thunking and to unbox integers.
- Mode analysis/inference to eliminate superfluous checks. E.g., if a variable is known to be *bound*, avoid unnecessary store dereferencing operations and unnecessary concurrency.
- Serious comparisons with Pakcs, MCC, FLVM.

Related Work

Most notably:

- The Münster Curry Compiler: a native code (unfair) compiler for Curry. http://danae.uni-muenster.de/~lux/curry/
- The FLVM: a (fair) virtual machine for FLP. http://redstar.cs.pdx.edu/~antoy/flp/vm/
- See ICFP'04 paper by Tolmach *et al.* for details on an interpreter closely related to FLC.

The End

Full paper is at http://www.cs.pdx.edu/~marius/.