

A Mutable Region System for Equational Reasoning about Pointer Algorithms

by

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Master Dissertation

submitted to the Department of Informatics



SOKENDAI (The Graduate University for Advanced Studies)

June 2019

Abstract

Acknowledgement

The first person I'd want to thank is Josh Ko who guided and trained me in the programming language world and convinced me of the beauty and power of rigorous reasoning. I'd also thank Zirun Zhu for being my best friend in Japan—you are probably the funniest and most heartwarming man I ever knew. Without any doubts, I've spent a lot of wonderful time and had many inspiring discussions with other fellow students and interns of the lab—Yongzhe Zhang, Chunmiao Li, Liye Guo, Vandang Tran, João Pereira.

I also want to thank my supervisors Zhenjiang Hu and Ichiro Hasuo for investing time and funding on my project and giving precious comments on my research. Besides, I want to thank Taro Sekiyama and Jeremy Gibbons for discussing my project with me and giving me much encouragement.

During my days in Japan, all these people mentioned above, my parents in China, and my friends scattered over cities and countries helped me to fight with depression and find a meaning for my life. Thank you all, sincerely.

I'm grateful for all the happiness and sorrow I had in Japan.

Zhixuan Yang
Arai, Ichikawa
On a rainy day, 2019

Contents

1	Introduction	1
2	Limitation of Static Region Systems	5
2.1	Motivating Example: Constant-space <i>foldr</i> for Linked Lists	6
2.2	Verifying <i>foldr</i> _{sw} : First Attempt	7
3	Mutable Region System	11
3.1	Preliminaries: the Language and Logic	11
3.2	Effect System as Logic Predicates	15
3.3	Inference Rules	16
3.4	Semantics	19
4	Program Equivalences	23
4.1	Equational Rules for Separation Guards	24
4.2	Verifying <i>foldr</i> _{sw} Resumed	25
5	Discussions	29
5.1	Related work	29
5.2	Conclusion	31
5.3	Future work	31
	Bibliography	33

1

Introduction

Plotkin and Power’s algebraic effects (Plotkin and Power, 2002, 2004) and their handlers (Plotkin and Pretnar, 2013; Pretnar, 2010) provide a uniform foundation for a wide range of computational effects by defining an effect as an algebraic theory—a set of operations and equational axioms on them. The approach has proved to be successful because of its composability of effects and clear separation between syntax and semantics. Furthermore, the equations defining an algebraic effect are also natural tools for equational reasoning about programs using the effect, and can be extended to a rich equational logic (Plotkin and Pretnar, 2008; Pretnar, 2010). The equations of some algebraic effects are also proved to be (Hilbert-Post) complete, including the effects of global and local state (Staton, 2010).

However, if one is limited to use only equational axioms on basic operations and must always expand the definition of a program to the level of basic operations, this style of reasoning will not be scalable. A widely-studied solution is to use an *effect system* to track

possible operations used by a program and use this information to derive equations (i.e. transformations) of programs. For example, if two programs f and g only invoke operations in sets ϵ_1 and ϵ_2 respectively and every operation in ϵ_1 commutes with every operation in ϵ_2 , then f and g commute:

$$x \leftarrow f; y \leftarrow g; k \quad = \quad y \leftarrow g; x \leftarrow f; k$$

The pioneering work by Lucassen and Gifford (Lucassen and Gifford, 1988) introduced an effect system to track memory usage in a program by statically partitioning the memory into *regions* and used that information to assist scheduling parallel programs. Benton et al. (Benton et al., 2007, 2009, 2006) and Birkedal et al. (Birkedal et al., 2016) gave relational semantics of such region systems of increasing complexity and verified some program equations based on them. Kammar and Plotkin (Kammar and Plotkin, 2012) presented a more general account for effect systems based on algebraic effects and studied many effect-dependent program equations. In particular, they also used the Gifford-style region-based approach to manage memory usage.

There is always a balance between expressiveness and complexity. Despite its simplicity and wide applicability, tracking memory usage by *static* regions is not always effective for equational reasoning about some pointer-manipulating programs, especially those manipulating recursive data structures. It is often the case that we want to prove operations on one node of a data structure is irrelevant to operations on the rest of the structure; thus a static region system requires that we annotate the node in a region different from that of the rest of the data structure. If this happens to every node (e.g. in a recursive function), each node of the data structure needs to have its own region, and thus the abstraction provided by regions collapses: regions should abstract disjoint memory cells, not memory cells themselves. In Chapter 2, we show a concrete example of equational reasoning about a tree traversal program and why a static region system does not work.

The core of the problem is the assumption that every memory cell statically belongs to one region, but when the logical structure of memory is mutable (e.g. when a linked list is split into two lists), we also want regions to be mutable to reflect the structure of

the memory (e.g. the region of the list is also split into two regions). To mitigate this problem, we propose a *mutable region system*. In this system, a region is either (i) a single memory cell or (ii) all the cells reachable from a node of a recursive data structure along the points-to relation of cells. Although this definition is apparently restrictive, we believe it sufficient to demonstrate our ideas in this paper and generalisations to more forms of regions are easy, e.g. regions that are the disjoint union of two subregions. For example, the judgement

$$l : \text{ListPtr } a \vdash t \ l : 1 \ ! \{get_{rc \ l}\} \quad (1.1)$$

asserts the program $t \ l$ only reads the linked list starting from l , where the type data $\text{ListPtr } a = \text{Nil} \mid \text{Ptr } (\text{Ref } (a, \text{ListPtr } a))$ ¹ is either *Nil* marking the end of the list or a reference to a cell storing a payload of type a and a *ListPtr* to the next node of the list. The cells linked from l form a region $rc \ l$ but it is only dynamically determined, and therefore may consist of different cells if the successor field (of type $\text{ListPtr } a$) stored in l is modified.

We also introduce a complementary construct called *separation guards*, which are effectful programs checking some pointers or their reachable closures are disjoint, otherwise stopping the execution of the program. For example,

$$l : \text{ListPtr } a, l_2 : \text{Ref } (a \times \text{ListPtr } a) \vdash [\text{rc } l * l_2] : 1$$

can be understood as a program checking the cell l_2 is not any node of linked list l . With separation guards and our effect system, we can formulate some program equations beyond the expressiveness of previous region systems. For example, given judgement (1.1), then

$$[\text{rc } l * l_2]; \text{put } l_2 \ v; t \ l \quad = \quad [\text{rc } l * l_2]; t \ l; \text{put } l_2 \ v$$

says that if cell l_2 is not a node of linked list l , then modification to l_2 can be swapped with $t \ l$, which only accesses list l . In [Section 4.2](#), we demonstrate using these transformations, we can straightforwardly prove the correctness of a constant-space linked list folding

¹Here *Nil* and *Ptr* are data constructors. *Nil* has type $\text{ListPtr } a$ and *Ptr* has type $\text{Ref } (a, \text{ListPtr } a) \rightarrow \text{ListPtr } a$.

algorithm, which is a special case of the Schorr-Waite traversal algorithm.

The remainder of this paper is structured as follows: in [Chapter 2](#) we show the limitation of static region systems by guiding the reader through an attempt to equationally reason about an interesting constant-space list folding algorithm; in [Chapter 3](#) we present our solution—a mutable region system, its semantics and inference rules, which are our main technical contribution; in [Chapter 4](#) we give a series of program equivalences based on our region system and use them to complete our proof attempt in [Chapter 2](#); in [Chapter 5](#) we discuss related work and future directions.

2

Limitation of Static Region Systems

This chapter we show the limitation of static region systems with a practical example of equational reasoning: proving the straightforward recursive implementation of *foldr* for linked lists is semantically equivalent to an optimised implementation using only constant space. The straightforward implementation is not tail-recursive and thus it uses space linear to the length of the list, whereas the optimised version cleverly eliminate the space cost by reusing the space of the linked list itself to store the information needed to control the recursion and restore the linked list after the process. This optimisation is essentially the Schorr-Waite algorithm (Schorr and Waite, 1967) adapted to linked lists, whose correctness is far from obvious and has been used as a test for many approaches of reasoning about pointer-manipulating programs (Bird, 2001; Butler, 1999; Möller, 1997; Reynolds, 2002).

In the following, we start with an attempt to an algebraic proof of the correctness of this optimisation—transforming the optimised implementation to the straightforward one

with equational axioms of the programming language and its effect operations. From this attempt, we can see the limitation of static region systems: we want the region partitioning to match the logical structure of data in memory, but when the structure is mutable, static region systems do not allow region partitioning to be mutable to reflect the change of the underlying structure.

2.1 Motivating Example: Constant-space *foldr* for Linked Lists

The straightforward implementation of folding (from the tail side) a linked list is simply

$$\begin{aligned} \text{foldrl} &: (A \rightarrow B \rightarrow B) \rightarrow B \rightarrow \text{ListPtr } A \rightarrow \mathbf{F } B \\ \text{foldrl } f \ e \ v &= \text{case } v \text{ of} \\ &\quad \text{Nil} \rightarrow \text{return } e \\ &\quad \text{Ptr } r \rightarrow \{(a, n) \leftarrow \text{get } r; b \leftarrow \text{foldrl } f \ e \ n; \\ &\quad \quad \text{return } (f \ a \ b)\} \end{aligned}$$

where $\mathbf{F } B$ is the type of computations of B values and the letter l in *foldrl* means linked lists. The program is recursively defined but not tail-recursive, therefore a compiler is likely to use a stack to implement the recursion. At runtime, the stack has one frame for each recursive call storing local arguments and variables so that they can be restored later when the recursion returns. If we want to minimise the space cost of the stack, we may notice that most local variables are not necessary to be saved in the stack: arguments f and e are not changed throughout the recursion, and local variables a , n and r can be obtained from v . Hence v is the only variable that a stack frame needs to remember to control the recursion. Somewhat surprisingly, we can even reduce the space cost further: since v is used to restore the state when the recursive call for n is finished and the list node n happens to have a field storing a *ListPtr* (that is used to store the successor of n), we can store v in that field instead of an auxiliary stack. But where does the original value of that field of n go? They can be stored in the corresponding field of its next node too. The following program implements this idea with an extra function argument to juggle with these pointers of successive nodes.

$$\begin{aligned}
foldrl_{sw} f e v &= fwd Nil v \\
fwd f e p v &= \text{case } v \text{ of} \\
&\quad Nil \rightarrow bwd f e p v \\
&\quad Ptr r \rightarrow \{(a, n) \leftarrow get r; put (r, (a, p)); \\
&\quad \quad fwd f e v n\} \\
bwd f b v n &= \text{case } v \text{ of} \\
&\quad Nil \rightarrow \text{return } b \\
&\quad Ptr r \rightarrow \{(a, p) \leftarrow get r; put (r, (a, n)); \\
&\quad \quad bwd f (f a b) p v\}
\end{aligned}$$

$foldrl_{sw}$ is in fact a special case of the Schorr-Waite traversal algorithm which traverses a graph whose vertices have at most 2 outgoing edges using only 1 bit for each stack frame to control the recursion. The Schorr-Waite algorithm can be easily generalised to traverse a graph whose out-degree is bounded by k using $\log k$ bits for each stack frame, and the above program is the case when $k = 1$ and the list is assumed to be not cyclic.

2.2 Verifying $foldrl_{sw}$: First Attempt

Let us try to prove the optimisation above is correct, in the sense that $foldrl_{sw}$ can be transformed to $foldrl$ by a series of applications of equational axioms on programs that we postulate, including those characterising properties of the language constructs like **case** and function application, and those characterising the effectful operations *get* and *put*.

To prove by induction, it is easy to see that we need to prove a strengthened equality:

$$\{b \leftarrow foldrl f e v; bwd f b p v\} = fwd f e p v \quad (2.1)$$

which specialises to $foldrl_{sw} = foldrl$ when $p = Nil$. When $v = Nil$, the equality can be easily verified. When $v = Ptr r$, we have

$$fwd f e p v = \{(a, n) \leftarrow get r; put (r, (a, p)); fwd f e v n\}$$

Assuming we have some inductive principle allowing us to apply Equation 2.1 to

$fwd\ f\ e\ v\ n$ since n is the tail of list v (We will discuss inductive principles later in ??), we proceed:

$$\begin{aligned}
 fwd\ f\ e\ p\ v &= \{(a, n) \leftarrow get\ r; put\ (r, (a, p)); \\
 &\quad b \leftarrow foldrl\ f\ e\ n; bwd\ f\ b\ v\ n\} \\
 &= [-\text{Expanding } bwd -] \\
 &\quad \{(a, n) \leftarrow get\ r; put\ (r, (a, p)); \\
 &\quad b \leftarrow foldrl\ f\ e\ n; \\
 &\quad (a, p) \leftarrow get\ r; put\ (r, (a, n)); \\
 &\quad bwd\ f\ (f\ a\ b)\ p\ v\} \tag{2.2}
 \end{aligned}$$

Now we can see why the optimisation works: fwd first modifies node v (i.e. $Ptr\ r$) to point to p , and after returning from the recursive call to n , it recovers p from node v and restores v to point to n . Hence we can complete the proof if we show the net effect of those operations leaves node v unchanged.

To show this, if we can prove the two computations in Equation 2.2 commute with $foldr\ f\ e\ n$:

$$\begin{aligned}
 &\{b \leftarrow foldrl\ f\ e\ n; \boxed{(a, p) \leftarrow get\ r; put\ (r, (a, n)); K}\} \\
 &= \{\boxed{(a, p) \leftarrow get\ r; put\ (r, (a, n));} \ b \leftarrow foldrl\ f\ e\ n; K\} \tag{2.3}
 \end{aligned}$$

Then

$$\begin{aligned}
 fwd\ f\ e\ p\ v &= \{(a, n) \leftarrow get\ r; put\ (r, (a, p)); \\
 &\quad (a, p) \leftarrow get\ r; put\ (r, (a, n)); \\
 &\quad b \leftarrow foldrl\ f\ e\ n; \\
 &\quad bwd\ f\ (f\ a\ b)\ p\ v\} \\
 &= [-\text{Properties of } put\ \text{and } get; \text{ See below -}] \\
 &\quad \{(a, n) \leftarrow get\ r; \\
 &\quad b \leftarrow foldrl\ f\ e\ n; \\
 &\quad bwd\ f\ (f\ a\ b)\ p\ v\}
 \end{aligned}$$

$$\begin{aligned}
&= [- \text{Contracting the definition of } foldrl -] \\
&\quad \{b' \leftarrow foldrl\ f\ e\ v; bwd\ f\ b'\ p\ v\}
\end{aligned}$$

which is exactly what we wanted to show (Equation 2.1). The properties used in the second step are

$$\begin{aligned}
\{put\ (r, v); x \leftarrow get\ r; K\} &= \{put\ (r, v); K[v/x]\} \\
\{put\ (r, v); put\ (r, u); K\} &= \{put\ (r, u); K\} \\
\{x \leftarrow get\ r; put\ (r, x); K\} &= \{x \leftarrow get\ r; K\}
\end{aligned} \tag{2.4}$$

Therefore, what remains is to prove the commutativity in Equation 2.3, which is arguably the most important step of the proof. Intuitively, $get\ r$ and $put\ (r, (a, n))$ access cell r , i.e. the head of the list v , while $foldrl\ f\ e\ n$ accesses the tail of list v . Hence if we want to derive Equation 2.3 with a region system, we can annotate cell r with some region ϵ_1 and all the cells linked from n with region ϵ_2 so that Equation 2.3 holds because $get\ r$ and $put\ (r, (a, n))$ access a region different from the one $foldrl\ f\ e\ n$ accesses.

Unfortunately, this strategy does not quite work for two reasons: First, since the argument above for r also applies to n and all their successors, what we finally need is one region ϵ_i for every node r_i of a linked list. This is unfavourable because the abstraction of regions collapses—we are forced to say that $foldrl\ f\ e\ n$ only accesses list n 's first node, second node, etc, instead of only one region containing all the nodes of n . The second problem happens in the type system: Now that the reference type is indexed by regions, the type of the i -th cell of a linked list may be upgraded to $Ref\ \epsilon_i\ (a \times ListPtr_{i+1}\ a)$. But this type signature prevents the second field of this cell pointing to anything but its successor, making programs changing the list structure like $foldrl_{sw}$ untypable. This problem cannot be fixed by simply change the type of the second field to be the type of references to arbitrary region, because we will lose track of the region information necessary for our equational reasoning when reading from that field. {Zhixuan: Clarify more} {Zhixuan: unnecessary for region polymorphism in typing recursive definitions}

The failure of static region systems in this example is due to the fact that a static region system presumes a fixed region partitioning for a program. While as we have seen

in the example above, in different steps of our reasoning, we may want to partition regions in different ways: it is not only because we need region partitioning to match the logic structure of memory cells which is mutable—as in the example above, a node of a list is modified to points to something else and thus should no longer be in the region of the list. Even when all data are immutable, we may still want a more flexible notion of regions—in one part of a program, we probably reason at the level of lists and thus we want all the nodes of a list to be in the same region; while in another part of the same program, we may want to reason at the level of nodes, then we want different nodes of a list in different regions.

3

Mutable Region System

Our observations in the last section suggest us to develop a more flexible region system. Our idea is to let the points-to structure of memory cells *determine* regions: a region is either a single memory cell or all the cells reachable from one cell along the points-to structure of cells. We found it is simpler to implement this idea in a logic system rather than a type system: we introduce effect predicates $(\cdot) ! \epsilon$ on programs of computation types where ϵ is a list of effect operations in the language and two ‘virtual’ operations get_r and put_r where r is a region in the above sense. The semantics of $t ! \epsilon$ is that program t only invokes the operations in ϵ . Inference rules for effect predicates are introduced.

3.1 Preliminaries: the Language and Logic

As the basis of discussion, we fix a small programming language with algebraic effects based on Levy’s call-by-push-value calculus (Levy, 2012). For a more complete treatment

Base types:	$\sigma ::= \text{Bool} \mid \text{Unit} \mid \text{Void} \mid \dots$
Value types:	$A, B ::= \sigma \mid \text{ListPtr } D \mid \text{Ref } D \mid A_1 \times A_2 \mid A_1 + A_2 \mid \mathbf{U}\underline{A}$
Storable types:	$D ::= \sigma \mid \text{ListPtr } D \mid \text{Ref } D \mid D_1 \times D_2 \mid D_1 + D_2$
Computation types:	$\underline{A}, \underline{B} ::= \mathbf{F}A \mid A_1 \rightarrow \underline{A}_2$
Base values:	$c ::= \text{True} \mid \text{False} \mid () \mid \dots$
Value terms:	$v ::= x \mid c \mid \text{Nil} \mid \text{Ptr } v \mid (v_1, v_2) \mid \text{inj}_1^{A_1+A_2} v \mid \text{inj}_2^{A_1+A_2} v \mid \text{thunk } t$
Computation terms:	$t ::= \text{return } v \mid \{x \leftarrow t_1; t_2\} \mid \text{match } v \text{ as } \{(x_1, x_2) \rightarrow t\}$ $\quad \mid \text{match } v \text{ as } \{\text{Nil} \rightarrow t_1, \text{Ptr } x \rightarrow t_2\}$ $\quad \mid \text{match } v \text{ as } \{\text{inj}_1 x_1 \rightarrow t_1, \text{inj}_2 x_2 \rightarrow t_2\}$ $\quad \mid \lambda x : A. t \mid t \ v \mid \text{force } v \mid \text{op } v \mid \mu x : \underline{A}. t$
Operations:	$\text{op} ::= \text{fail} \mid \Omega \mid \text{get} \mid \text{put} \mid \text{new} \mid \dots$

Figure 3.1: Syntax of the language.

of such a language, we refer the reader to Plotkin and Pretnar's work (Plotkin and Pretnar, 2008). The syntax of this language are listed in Figure (3.1). The language has two categories of types: value types, ranged over by A , and computation types, ranged over by \underline{A} . Value types excluding thunk types ($\mathbf{U}\underline{A}$) are called storable types, ranged over by D . In this language, only storable types can be stored in memory cells. Furthermore, we omit general recursively-defined types for simplicity and restrict our treatment to only one particular recursive type: type $\text{ListPtr } A$ is isomorphic to

$$\text{Unit} + \text{Ref } (A \times \text{ListPtr } A)$$

We assume that the language includes the effect of failure (*fail*), non-divergence (Ω) and *local state* (Staton, 2010). Failure has one nullary operation *fail* and no equations.

Local state has the following three operations:

$$\begin{aligned} get_D &: Ref\ D \rightarrow F\ D \\ put_D &: Ref\ D \times D \rightarrow F\ Unit \\ new_D &: D \rightarrow F\ (Ref\ D) \end{aligned}$$

and they satisfy (a) the three equations in (2.4) and

$$\{x \leftarrow get\ r; y \leftarrow get\ r; K\} = \{x \leftarrow get\ r; K[x/y]\}$$

(b) commutativity of *get* and *put* on different cells, for example,

$$\{put\ (l_1, u); put\ (l_2, v); K\} = \{put\ (l_2, v); put\ (l_1, u); K\} \quad (l_1 \neq l_2)$$

(c) commutativity laws for *new*, that is,

$$\begin{aligned} \{l \leftarrow new\ v; put\ (r, u); K\} &= \{put\ (r, u); l \leftarrow new\ v; K\} \\ \{l \leftarrow new\ v; x \leftarrow get\ r; K\} &= \{x \leftarrow get\ r; l \leftarrow new\ v; K\} \\ \{l_1 \leftarrow new\ v; l_2 \leftarrow new\ u; K\} &= \{l_2 \leftarrow new\ u; l_1 \leftarrow new\ v; K\} \end{aligned}$$

and (d) the following *separation law*: for any D ,

$$\begin{aligned} \{l_1 \leftarrow new_D\ v_1 \quad &= \quad \{l_1 \leftarrow new_D\ v_1; t_1\} \\ \text{match } l_1 \equiv l_2 \text{ as} & \\ \{False \rightarrow t_1 & \\ True \rightarrow t_2\} & \} \end{aligned} \quad (AX\text{-SEP})$$

which is a special case of the axiom schema $B3_n$ in (Staton, 2010) but is sufficient for our purposes. {Zhixuan: In the standard treatment, these equations come with a context of free variables like l_2 in the last one. Do you find it strange if I omit them?}

{Zhixuan: Give a brief description of the semantics... }

We also need an equational logic for reasoning about programs of this language. We refer the reader to the papers (Plotkin and Pretnar, 2008; Pretnar, 2010) for a complete

treatment of its semantics and inference rules. Here we only record what is needed in this paper. The formulas of this logic are:

$$\begin{aligned} \phi ::= & t_1 = t_2 \mid \forall x : A. \phi \mid \exists x : \underline{A}. \phi \\ & \mid \exists x : A. \phi \mid \exists x : \underline{A}. \phi \mid \phi_1 \wedge \phi_2 \mid \phi_1 \vee \phi_2 \\ & \mid \neg \phi \mid \phi_1 \rightarrow \phi_2 \mid \top \mid \perp \end{aligned}$$

The judgement of this logic has form $\Gamma \mid \Psi \vdash \phi$ where Γ is a context of the types of free variables and Ψ is a list of formulas that are the premises of ϕ . The inference rules of this logic include:

1. Standard rules for connectives in classical first order logic and structural rules for judgements (e.g. weakening and contraction on premises),
2. standard equivalences for language constructs including sequencing, thunking, function application, case analysis, etc, as in call-by-push-value (Levy, 2012). For example,

$$\begin{array}{c} \frac{}{v : A, t : A \rightarrow \underline{A} \mid \vdash \{x \leftarrow \mathbf{return} \ v; t\} = t \ v} \quad \frac{\Gamma \vdash (\lambda x. t) : A \rightarrow \underline{A} \quad \Gamma \vdash v : A}{\Gamma \mid \vdash (\lambda x. t) \ v = t[v/x]} \\[10pt] \frac{}{t_1 : \underline{A}, t_2 : \underline{A} \mid \vdash \mathbf{case} \ True \ \mathbf{of} \ \{ \ True \rightarrow t_1; \ False \rightarrow t_2 \} = t_1} \end{array}$$

3. rules inherited from the effect theories, for example,

$$\frac{}{l_{1,2} : \mathit{Ref} \ D, v_{1,2} : D, t : \underline{A} \mid \vdash \{ \mathit{put} \ (l_1, v_1); \mathit{put} \ (l_2, v_2); t \} = \{ \mathit{put} \ (l_2, v_2); t \}}$$

4. algebraicity of effect operations, the inductive principle over computations and the universal property of computation types.

In this paper, we will only use the first three kinds of rules.

3.2 Effect System as Logic Predicates

Unlike existing type-and-effect systems, our mutable region system is defined as logic predicates on computation terms in the logic. Let op range over possible effect operations in the language. We extend the term of the logic:

$$\begin{aligned}\phi &::= \dots \mid t ! \epsilon \\ \epsilon &::= \emptyset \mid \epsilon, op \mid \epsilon, get_{rc\ v} \mid \epsilon, put_{rc\ v} \mid \epsilon, get_v \mid \epsilon, put_v\end{aligned}$$

The new term is well-formed when

$$\begin{array}{c} \frac{\Gamma \vdash t : \mathbf{FA}}{\Gamma \vdash t ! \cdot : \mathbf{form}} \qquad \frac{\Gamma \vdash t ! \epsilon : \mathbf{form}}{\Gamma \vdash t ! \epsilon, op : \mathbf{form}} \quad (op \notin \{get, put\}) \\[10pt] \frac{\Gamma \vdash t ! \epsilon : \mathbf{form} \quad \Gamma \vdash v : Ref\ D}{\Gamma \vdash t ! \epsilon, o_v : \mathbf{form}} \quad (o \in \{get, put\}) \\[10pt] \frac{\Gamma \vdash t ! \epsilon : \mathbf{form} \quad \Gamma \vdash v : ListPtr\ D}{\Gamma \vdash t ! \epsilon, o_{rc\ v} : \mathbf{form}} \quad (o \in \{get, put\}) \end{array}$$

Although ϵ is formally a list of comma-separated operations, we will regard it as a set and thus use set operations like inclusion and minus on it.

Example 3.2.1. Let Γ be $\{l : Ref\ D, r : Ref\ (D \times ListPtr\ D)\}$,

$$\Gamma \vdash \{(a, n) \leftarrow get\ r; put\ (l, a)\} ! \{put_l, get_{rc\ (Ptr\ r)}\} : \mathbf{form}$$

is derivable.

Although our effect predicate is defined only on first-order computations, we can work with higher order functions by using quantification in the logic. For example, if $\Gamma \vdash t : ListPtr\ D \rightarrow \mathbf{FA}$

$$\Gamma \mid \vdash \forall(l : ListPtr\ D). t\ l ! \{get_{rc\ l}\}$$

is well-formed and it expresses that function t only reads l when it is applied to list l .

The intended meaning of effect predicate $t ! \epsilon$ is: provided that the regions mentioned in ϵ are disjoint, the computation t only applies a finite number of operations in ϵ . Before giving a formal definition of this semantics, we present the inference rules first in the rest of this section, which may provide more intuition, and then in [Section 3.4](#) we give the formal semantics of effect predicates.

3.3 Inference Rules

An advantage of tracking effects in the equational logic is that we only need to design inference rules for effects-related language constructs—**return**, sequencing and operation application. Other language constructs like case-analysis are handled by the equational logic as we will see in the example below. Our inference rules are:

- two structural rules

$$\frac{\Gamma \mid \Psi \vdash t ! \epsilon \quad \epsilon \subseteq \epsilon'}{\Gamma \mid \Psi \vdash t ! \epsilon'} \text{R-SUB} \qquad \frac{\Gamma \mid \Psi \vdash t = t' \wedge t' ! \epsilon}{\Gamma \mid \Psi \vdash t ! \epsilon} \text{R-EQ}$$

- rules for **return** and sequencing

$$\frac{}{x : A \mid \vdash \text{return } x ! \emptyset} \text{R-PURE} \qquad \frac{\Gamma \mid \Psi \vdash t_1 ! \epsilon \quad \Gamma, x : A \mid \Psi \vdash t_2 ! \epsilon}{\Gamma \mid \Psi \vdash \{x \leftarrow t_1; t_2\} ! \epsilon} \text{R-SEQ}$$

- rules for effect operations, for any operation of type $A \rightarrow \underline{B}$ in the language,

$$\frac{}{v : A \mid \vdash \text{op } v ! \{\text{op}\}} \text{R-OP}$$

and specially for get_l and put_l (Formally, they are not operation of the language so

these rules are needed)

$$\frac{}{l : \text{Ref } D \mid \vdash \text{get } l ! \{ \text{get}_l \}} \quad \frac{}{l : \text{Ref } D, a : D \mid \vdash \text{put } (l, a) ! \{ \text{put}_l \}}$$

- rules for $\text{get}_{\text{rc } l}$ and $\text{put}_{\text{rc } l}$

$$\frac{\Gamma \mid \Psi \vdash t ! \epsilon \setminus \{ \text{get}_{\text{rc Nil}}, \text{put}_{\text{rc Nil}} \}}{\Gamma \mid \Psi \vdash t ! \epsilon} \text{R-NIL}$$

$$\frac{\Gamma, a, n \mid \Psi \vdash k ! \epsilon[l/x] \cup \epsilon[\text{rc } n/x]}{\Gamma \mid \Psi \vdash \{ (a, n) \leftarrow \text{get } l; k \} ! \epsilon[\text{rc } (\text{Ptr } l)/x]} (\text{get}_x \in \epsilon) \text{R-GETRC}$$

$$\frac{\Gamma \mid \Psi \vdash k ! \epsilon[l/x]}{\Gamma \mid \Psi \vdash k ! \epsilon[\text{rc } (\text{Ptr } l)/x]} (\text{put}_x \in \epsilon) \text{R-PUTRC}$$

These rules deserve some explanation: The rule R-NIL means that $\text{get}_{\text{rc Nil}}$ and $\text{put}_{\text{rc Nil}}$ cannot be used by the program; the rule R-GETRC means that if a program has the permission to read the list from l , it can read the cell l and its permission on $\text{rc } (\text{Ptr } l)$ is split into the same permission on cell l and the rest of the list (i.e. $\text{rc } n$);

Example 3.3.1. By R-GETRC, if $\Gamma, a, n \mid \vdash k ! \{ \text{get}_l, \text{put}_l, \text{get}_{\text{rc } n}, \text{put}_{\text{rc } n} \}$ is derivable, then

$$\Gamma \mid \vdash \{ (a, n) \leftarrow \text{get } l; k \} ! \{ \text{get}_{\text{rc } (\text{Ptr } l)}, \text{put}_{\text{rc } (\text{Ptr } l)} \}$$

is derivable.

the rule R-PUTRC says that if a program has the permission to write the list from l , then it has the permission to write the cell l itself. This rule seems not very useful, but it reflects the fact that even if a program can write $\text{rc } \text{Ptr } l$, if it cannot read l , its accessible cells are restricted to l only.

- Finally, we introduce a rule that may not be valid in more general settings but is safe in our context because it assumes all lists in memory are finite. The rule is,

under side condition $get_{rc} v \in \epsilon$,

$$\frac{\Gamma, v \mid \Psi, v = Nil \vdash t ! \epsilon \quad \Gamma, v, r \mid \Psi, v = Ptr r \vdash t = \{(a, n) \leftarrow get\ r; k\} \\ \Gamma, v, r, a, n \mid \Psi, v = Ptr r, t[n/v] ! \epsilon[n/v] \vdash k ! \epsilon[n/v] \cup \epsilon[r/rc\ v]}{\Gamma, v : ListPtr\ D \mid \Psi \vdash t ! \epsilon}$$

Without this rule, a recursive program defined with μ can only satisfy $\cdot ! \epsilon$ if $\Omega \in \epsilon$ (See Scott-induction in Chapter 9 of paper (Pretnar, 2010)). This rule allows a program that is a structural recursion along some linked list to satisfy predicate $\cdot ! \epsilon$ without including Ω in ϵ , as if it is not a recursive program.

Example 3.3.2. *Without this rule, we can only derive $foldrl\ f\ e\ l ! \{get_{rc}\ l, \Omega\}$ using Scott-induction. With this, we can derive*

$$\frac{foldrl\ f\ e\ Nil = \mathbf{return}\ () \quad \mathbf{return}\ () ! \{get_{rc}\ l\}}{f, e, l \mid l = Nil \vdash foldrl\ f\ e\ Nil ! \{get_{rc}\ l\}} \quad \textcircled{1} \quad \textcircled{2}$$

$$f, e, l \mid \vdash foldrl\ f\ e\ l ! \{get_{rc}\ l\}$$

where $\textcircled{1}$ is

$$f, e, l, r \mid l = Ptr\ r \vdash foldrl\ f\ e\ l = \{(a, n) \leftarrow get\ r; K\}$$

$$K =_{\text{def}} \{b \leftarrow foldrl\ f\ e\ n; \mathbf{return}\ f\ a\ b\}$$

and $\textcircled{2}$ is

$$\frac{\dots}{f, e, l, r, a, n \mid l = Ptr\ r, foldrl\ f\ e\ n ! \{get_{rc}\ n\} \vdash K ! \{get_r, get_{rc}\ n\}}$$

An obvious difference of our effect predicates from existing type-and-effect system is that we only have inference rules for effect related language constructs, because other language constructs can be handled by R-EQ and corresponding elimination rules for the construct of the logic.

Example 3.3.3. Letting P denote $\text{case } b \text{ of } \{ \text{True} \rightarrow op_1; \text{False} \rightarrow op_2 \}$, we can derive

$$\frac{\frac{\textcircled{1} \quad \textcircled{2}}{b : \text{Bool} \mid b = \text{False} \vee b = \text{True} \vdash P ! \{op_1, op_2\}}}{b : \text{Bool} \mid \vdash P ! \{op_1, op_2\}} (EXM-BOOL)$$

where $EXM-BOOL$ is the rule in the logic saying that $b : \text{Bool}$ is either True or False , $\textcircled{1}$ is

$$\frac{b : \text{Bool} \mid b = \text{True} \vdash P = op_1 \quad \frac{\mid \vdash op_1 ! \{op_1\}}{b : \text{Bool} \mid b = \text{True} \vdash op_1 ! \{op_1, op_2\}}}{b : \text{Bool} \mid b = \text{False} \vdash P ! \{op_1, op_2\}}$$

and similarly $\textcircled{2}$ is

$$\frac{b : \text{Bool} \mid b = \text{False} \vdash P = op_2 \quad \frac{\mid \vdash op_2 ! \{op_2\}}{b : \text{Bool} \mid b = \text{False} \vdash op_2 ! \{op_1, op_2\}}}{b : \text{Bool} \mid b = \text{False} \vdash P ! \{op_1, op_2\}}$$

{Zhixuan: An example demonstrating the assumption of disjointness of regions by these rules.}

3.4 Semantics

Now let us formalise our intuitive semantics of effect predicate $t ! \epsilon$: when regions mentioned in ϵ are disjoint and have finite cells, t is a computation only using the operations contained in ϵ and t is also finite. Recall that the semantics of $t : \text{FA}$ is an equivalence class of trees whose internal nodes are labeled with operation symbols and leaves are labeled with **return** v for some $v \in \llbracket A \rrbracket$. Trees in $\llbracket t \rrbracket$ are equal in the sense that anyone of them can be rewritten to another by the equations of the effect theory (Bauer, 2018). Therefore if we can define a denotational semantics $\llbracket \epsilon \rrbracket$, presumably to be the set of operations available to t , then $\llbracket t ! \epsilon \rrbracket$ can be defined to mean that $\llbracket t \rrbracket$ has some element

T whose operations is a subset of $\llbracket \epsilon \rrbracket$ and T is a well-founded tree.

However, how to interpret ϵ in the framework of algebraic effects is not straightforward. For op , get_l and put_l in ϵ , they can be easily interpreted by corresponding operations op , $get_{\llbracket l \rrbracket}$ and $put_{\llbracket l \rrbracket}$. For $get_{rc\ l}$ (and $put_{rc\ l}$), we want to interpret it as a set of operations $\{get_{\llbracket l \rrbracket}, get_{r_1}, get_{r_2}, \dots\}$ where $\llbracket l \rrbracket$ points to r_1 , r_1 points to r_2 in the memory, etc. However, in the semantics of algebraic effects, there is no explicit representation for the memory so that we do not immediately know what r_1, r_2, \dots are. (For comparison, the semantics of t in other approaches is usually a function $Mem \rightarrow (\llbracket A \rrbracket, Mem)$ which has an explicit Mem .)

This problem may be tackled by the coalgebraic treatment of effects (Plotkin and Power, 2008), but here we adopt a simple workaround: Although we do not have an explicit representation of memory to work with, we do have an operation get to probe the memory—if $get\ r$ returns v , we know memory cell r currently stores value v . Hence if $[rc\ v]$ is a program traversing linked list v and returns the set of references to the nodes of the list, then we can interpret $t ! \{get_{rc\ v}\}$ in this way: in program $\{r \leftarrow [rc\ v]; t\}$, t only reads the references in r . And as we mentioned in Section 3.3, predicate $t ! \{get_{r_1}, put_{r_2}\}$ implicitly assumes that r_1 and r_2 are disjoint, so to interpret this predicate, we want $[r_1 * r_2]$ not only returns the references in r_1 and r_2 but also checks they are disjoint.

Now let us define such programs more formally, which collect references to cells of memory regions and check disjointness. Following the notation of separation logic (Reynolds, 2002), we write $\phi = l_1 * l_2 * \dots * l_n$ to denote a list of separate regions. Here l_i is either an expression of type $Ref\ D$ or expression $rc\ v$ for some v of type $ListPtr\ D$. We add a new kind of term $[\phi]$ in the language, which we call *separation guards*. It has type $FMemSnap$, where type $MemSnap$ abbreviates

$$FinMap\ (Ref\ (a \times ListPtr\ a))\ (Set\ (Ref\ (a \times ListPtr\ a)))$$

Thus $x : MemSnap$ is finite map from references to sets of references. The semantics of $[\phi]$ is the computation denoted by the following program

$$\begin{aligned} \llbracket [\phi] \rrbracket &= sepChk\ \phi\ \emptyset\ \emptyset \\ sepChk\ []\ x\ rcs &= \text{return}\ rcs \\ sepChk\ (v * \phi)\ x\ rcs &= \text{if } l \in x \text{ then fail else } sepChk\ \phi\ (x \cup l)\ rcs \end{aligned}$$

```

sepChk (rc v *  $\phi$ ) x rcs = {  $x' \leftarrow \text{tvsList } v$ ;
                             if  $x' \cap x \neq \emptyset$ 
                             then fail
                             else sepChk  $\phi$  ( $x' \cup x$ ) ( $rcs \cup \{v \mapsto x'\}$ ) }

tvsList Nil = return  $\emptyset$ 
tvsList (Ptr r) = { ( $a, n$ )  $\leftarrow \text{get } p$ ;  $rs \leftarrow \text{tvsList } n$ ; return ( $r \cup rs$ ) }

```

Thus $[\phi]$ traverses each region $r \in \phi$ one by one and checks their cells are disjoint, otherwise it calls *fail*. When it terminates, it returns a finite map *rcs* mapping every region r in ϕ to the set of its cells, which can be thought as a snapshot of the current memory.

By probing the memory with separation guards, we can define the semantics of effect predicates now. For any effect set ϵ , let $R_\epsilon = \{l \mid \text{put}_l \in \epsilon\} \cup \{\text{rc } p \mid \text{get}_{\text{rc } p} \in \epsilon\}$. Then take ϕ_ϵ to be an arbitrary $*$ -sequence of all the elements of R . We define the semantics of judgement $\Gamma \vdash t ! \epsilon$ to be the set of $\gamma \in \llbracket \Gamma \rrbracket$ such that $\llbracket \{\phi_\epsilon; t\} \rrbracket_\gamma$ has an element T , which is a tree satisfying:

- there is some $T_1 \in \llbracket \phi_\epsilon \rrbracket_\gamma$ and for any leaf node of T_1 label with **return** x , there is a computation tree T_x , such that T is equal to the tree obtained by replacing every leaf node **return** x of T_1 with corresponding T_x ;
- every T_x is well-founded and every operation in T_x is either: (i) $op \ v$ for some $op \in \epsilon$, (ii) $\text{get } v$ for some $\text{get}_l \in \epsilon$ and $v = \llbracket l \rrbracket_\gamma$, (iii) $\text{put } (v, d)$ for some $\text{put}_l \in \epsilon$ and $v = \llbracket l \rrbracket_\gamma$, (iv) $\text{get } v$ for some $\text{get}_{\text{rc } r} \in \epsilon$ and $v \in x(\llbracket r \rrbracket_\gamma)$, and (v) $\text{put } (v, d)$ for some $\text{put}_{\text{rc } r} \in \epsilon$ and $v \in x(\llbracket r \rrbracket_\gamma)$.

Proposition 3.4.1 (Soundness). *If $\Gamma \mid \psi_1, \dots, \psi_n \vdash \phi$ is derivable from the rules in Section 3.3, then*

$$\bigcap_{1 \leq i \leq n} \llbracket \Gamma \vdash \psi_i \rrbracket \subseteq \llbracket \Gamma \vdash \phi \rrbracket$$

4

Program Equivalences

With effect predicates and separation guards, we can formulate the program transformation we wanted in [Section 2.2](#). For any effect predicate ϵ , let $R_\epsilon = \{r \mid \text{put}_r \in \epsilon \vee \text{get}_r \in \epsilon\}$ be the regions used in ϵ , and ϕ_ϵ be an arbitrary sequence of the elements of R_ϵ joined by ‘*’. For two effect predicates ϵ_1 and ϵ_2 , if their operations (excluding get_r and put_r) are pairwise commutative, i.e. any $op_1 \in \epsilon_1$ and $op_2 \in \epsilon_2$ that are not get_r or put_r satisfy

$$\{x \leftarrow op_1 \ u; y \leftarrow op_2 \ v; k\} = \{y \leftarrow op_2 \ v; x \leftarrow op_1 \ u; k\}$$

then we have:

$$\frac{\Gamma \mid \Psi \vdash t_i ! \epsilon_i \quad (i = 1, 2)}{\Gamma \mid \Psi \vdash [\phi_{\epsilon_1} * \phi_{\epsilon_2}] \langle \{x \leftarrow t_1; y \leftarrow t_2; k\} = \{y \leftarrow t_2; x \leftarrow t_1; k\} \rangle} \quad (\text{Eq-Com})$$

where $c \langle a = b \rangle$ abbreviates $\{c; a\} = \{c; b\}$.

Proof. {Zhixuan: This should easily follow the semantics of $t_i ! \epsilon_i$ }. □

4.1 Equational Rules for Separation Guards

The consequence of [Eq-Com](#) has separation guards serving as the precondition for the equality, and therefore to finally use this equality in equational reasoning, we need to know when this precondition is satisfied. This is accomplished by the following equational rules for separation guards.

$$\frac{}{\Gamma \mid \vdash [\phi] \langle \text{new } t = \{ l \leftarrow \text{new } t; [\phi * l]; \text{return } l \} \rangle} \text{SEP-REFINTRO}$$

$$\frac{\Gamma \mid \Psi, l_1 \neq l_2 \vdash t_1 = t_2}{\Gamma \mid \Psi \vdash [l_1 * l_2] \langle t_1 = t_2 \rangle} \text{SEP-REFELIM}$$

SEP-REFINTRO adds a cell into the separation guards, provided that the cell is newly generated. The validity of this rule comes from [AX-SEP](#) saying that the result of *new* is always different previous values. Conversely, SEP-REFELIM says that the separation guard $[l_1 * l_2]$ (for l_1 and l_2 of type *Ref D*) provides an assumption $l_1 \neq l_2$ for further equational reasoning.

$$\frac{}{\Gamma \mid \vdash [\phi] = [\phi * \text{rc Nil}]} \text{SEP-RCINTRO1}$$

$$\frac{}{\Gamma \mid \vdash [\phi * \text{rc } p * l] \langle \text{put } (l, p) = \{ \text{put } (l, p); [\phi * \text{rc } (\text{Ptr } l)] \} \rangle} \text{SEP-RCINTRO2}$$

SEP-RCINTRO1 and SEP-RCINTRO2 introduce a reachable closure in the separation guards, and then it can be eliminated by the following inductive principle for linked lists.

$$\frac{\Gamma \mid \Psi, p = \text{Nil} \vdash [\phi] \langle t_1 = t_2 \rangle \quad \text{InductiveCase}}{\Gamma \mid \Psi \vdash [\text{rc } p * \phi] \langle t_1 = t_2 \rangle} \text{LISTIND}$$

where `InductiveCase` is

$$\Gamma, l \mid p = \text{Ptr } l, \text{hyp} \vdash \{ (a, n) \leftarrow \text{get } l; [l * \text{rc } n * \phi] \} \langle t_1 = t_2 \rangle$$

and `hyp` is $[\text{rc } n * \phi] \langle t_1 = t_2 \rangle$. And we have some simple structural rules for separation guards:

$$\begin{array}{c} \frac{}{\Gamma \mid \vdash [\phi_1 * \phi_2] = [\phi_2 * \phi_1]} \qquad \frac{}{\Gamma \mid \vdash [(\phi_1 * \phi_2) * \phi_3] = [\phi_1 * (\phi_2 * \phi_3)]} \\[10pt] \frac{}{\mid \vdash [] = \text{return } ()} \qquad \frac{}{\Gamma \mid \vdash [\phi_1 * \phi_2] = \{[\phi_1 * \phi_2]; [\phi_1]\}} \\[10pt] \frac{}{\Gamma \mid \vdash \{[\phi_1]; [\phi_2]\} = \{[\phi_2]; [\phi_1]\}} \end{array}$$

and the commutativity of separation guards with `get` and `put`:

$$\begin{array}{c} \frac{}{\Gamma \mid \vdash \{a \leftarrow \text{get } l; [\phi]; k\} = \{[\phi]; a \leftarrow \text{get } l; k\}} \\[10pt] \frac{}{\Gamma \mid \vdash \{\text{put } (l, a); [l * \phi]; k\} = \{[l * \phi]; \text{put } (l, a); k\}} \end{array}$$

At last, we have the following rule corresponding to the frame rule of separation logic:

$$\frac{\Gamma \mid \Psi \vdash t ! \overline{\phi_1} \quad \Gamma \mid \Psi \vdash [\phi_1] \langle t = \{x \leftarrow t; [\phi_2]; \text{return } x\} \rangle}{\Gamma \mid \Psi \vdash [\phi_1 * \psi] \langle t = \{x \leftarrow t; [\phi_2 * \psi]; \text{return } x\} \rangle}$$

Proposition 4.1.1. *The inference rules above are sound with respect to the semantics.*

4.2 Verifying $foldrl_{sw}$, Resumed

Now we have enough weapons to complete our equational proof for $foldrl_{sw}$ in [Section 2.2](#)—effect predicates for proving commutativity of non-interfering computations and an inductive principle `LISTIND` for finite linked list.

First, we change our proof goal [Equation 2.1](#) to have a precondition that v is a finite list:

$$\forall p. [\mathbf{rc} \ v] \langle \{ b \leftarrow \text{foldrl } f \ e \ v; \text{bwd } f \ b \ p \ v \} = \text{fwd } f \ e \ p \ v \rangle$$

Then we use LISTIND to prove it by induction on v . The base case is still straightforward. The inductive case is to prove

$$\{ (a, n) \leftarrow \text{get } r; [l * \mathbf{rc} \ n] \} \langle \{ b \leftarrow \text{foldrl } f \ e \ v; \text{bwd } f \ b \ p \ v \} = \text{fwd } f \ e \ p \ v \rangle \quad (4.1)$$

under the assumption that $v = \text{Ptr } r$ and inductive hypothesis

$$\forall p. [\mathbf{rc} \ n] \langle \{ b \leftarrow \text{foldrl } f \ e \ n; \text{bwd } f \ b \ p \ n \} = \text{fwd } f \ e \ p \ n \rangle$$

This is shown by equational reasoning:

$$\begin{aligned} & \{ (a, n) \leftarrow \text{get } r; [r * \mathbf{rc} \ n]; \text{fwd } f \ e \ p \ v \} \\ = & \text{[- Expanding fwd -]} \\ & \{ (a, n) \leftarrow \text{get } r; [r * \mathbf{rc} \ n]; (a, n) \leftarrow \text{get } r; \dots \} \\ = & \text{[- Commutativity of get and } [l * \mathbf{rc} \ n] \text{ -]} \\ & \{ (a, n) \leftarrow \text{get } r; (a, n) \leftarrow \text{get } r; [l * \mathbf{rc} \ n]; \dots \} \\ = & \text{[- Two consecutive get r -]} \\ & \{ (a, n) \leftarrow \text{get } r; [r * \mathbf{rc} \ n]; \text{put } (r, (a, p)); \text{fwd } f \ e \ v \ n \} \\ = & \{ (a, n) \leftarrow \text{get } r; \text{put } (r, (a, p)); [r * \mathbf{rc} \ n]; \text{fwd } f \ e \ v \ n \} \\ = & \text{[- } [r * \mathbf{rc} \ n] = \{ [r * \mathbf{rc} \ n]; [\mathbf{rc} \ n] \} \text{ -]} \\ & \{ (a, n) \leftarrow \text{get } r; \text{put } (r, (a, p)); [r * \mathbf{rc} \ n]; [\mathbf{rc} \ n]; \text{fwd } f \ e \ v \ n \} \end{aligned}$$

Now we can apply our inductive hypothesis to $\{ [\mathbf{rc} \ n]; \text{fwd } f \ e \ v \ n \}$ by instantiating $p = v$, and we get:

$$\begin{aligned} & \text{Above program} \\ = & \{ (a, n) \leftarrow \text{get } r; \text{put } (r, (a, p)); [r * \mathbf{rc} \ n]; [\mathbf{rc} \ n]; b \leftarrow \text{foldrl } f \ e \ n; \text{bwd } f \ b \ p \ n \} \\ = & \text{[- Expanding bwd -]} \\ & \{ (a, n) \leftarrow \text{get } r; \text{put } (r, (a, p)); [r * \mathbf{rc} \ n]; b \leftarrow \text{foldrl } f \ e \ n; \\ & (a, p) \leftarrow \text{get } r; \text{put } (r, (a, n)); \text{bwd } f \ (f \ a \ b) \ p \ v \} \end{aligned}$$

This time we can use [Eq-Com](#) to show that $\text{foldrl } f \ e \ n$ and $\{(a, p) \leftarrow \text{get } r; \text{put } (r, (a, n))\}$ are commutative. It is straightforward to derive

$$r, n \mid \vdash \{(a, p) \leftarrow \text{get } r; \text{put } (r, (a, n))\} ! \{\text{get}_r, \text{put}_r\}$$

and in 3.3.2 we have derived $f, e, n \mid \vdash \text{foldrl } f \ e \ n ! \{\text{get}_{\text{rc } n}\}$, so by [Eq-Com](#) we can proceed:

Above program

$$\begin{aligned} &= \{(a, n) \leftarrow \text{get } r; \text{put } (r, (a, p)); [r * \text{rc } n]; (a, p) \leftarrow \text{get } r; \text{put } (r, (a, n)); \\ &\quad b \leftarrow \text{foldrl } f \ e \ n; \text{bwd } f \ (f \ a \ b) \ p \ v\} \\ &= [- \text{Commutativity of separation guards with get and put -}] \\ &\quad \{(a, n) \leftarrow \text{get } r; \text{put } (r, (a, p)); (a, p) \leftarrow \text{get } r; \text{put } (r, (a, n)); \\ &\quad [r * \text{rc } n]; b \leftarrow \text{foldrl } f \ e \ n; \text{bwd } f \ (f \ a \ b) \ p \ v\} \\ &= [- \text{Simplifying get and put -}] \\ &\quad \{(a, n) \leftarrow \text{get } r; [r * \text{rc } n]; b \leftarrow \text{foldrl } f \ e \ n; \text{bwd } f \ (f \ a \ b) \ p \ v\} \\ &= [- \text{Contracting foldrl and f a b -}] \\ &\quad \{(a, n) \leftarrow \text{get } r; [r * \text{rc } n]; b' \leftarrow \text{foldrl } f \ e \ v; \text{bwd } f \ b' \ p \ v\} \end{aligned}$$

This completes our equational proof for $\text{foldrl}_{\text{sw}}$.

5

Discussions

5.1 Related work

The correctness of the Schorr-Waite algorithm has been proved by different approaches: relational algebra (Möller, 1997), data refinement (Butler, 1999), separation logic (Reynolds, 2002) and equational reasoning (Bird, 2001). Among them, Bird’s approach is most related to ours. The fundamental difference between our work and his is that he worked with a fixed (purely-functional) model of memory, whereas we followed the axiomatic approach for equational reasoning (Gibbons and Hinze, 2011) so our reasoning only depends on algebraic axioms of effect operations. Our work extends the approach by Gibbons and Hinze (2011) in the sense that we use an effect system for proving some equivalences instead of solely relying on equational axioms. As we were developing the equational proof for the Schorr-Waite algorithm, we tried a proof using only equational axioms. But we found the proof complicated and many of its steps too low-level if we could only work

with primitive operations. Thus we turned to use effect systems to prove important steps—commutativity of two non-interfering computations—in a more intuitive way.

Our separation guards are borrowed from separation logic (Reynolds, 2002) with extreme restriction on the forms of assertions, but we expect an extended version of our system may use a wider family of assertions as in separation logic. Our work is different from separation logic in the way that our goal is to show two programs are observationally equivalent while separation logic shows a program establishes a post-condition described by a logic language. The fundamental difference on the proof goal makes these two approaches useful in different settings. However, the work by Nishimura (2008) resembles our approach: they derived inequalities of separation assertions and program constructs on the semantic foundation of predicate transformers of separation logic, whereas our semantic foundation is algebraic effects. Another closely related approach is relational separation logic (Yang, 2007), which aims to show two programs executed respectively on two states satisfying a pre-relation produce two states satisfying a post-relation. It is an interesting question to compare and establish the connection between our algebraic-effects-based approach and separation-logic-based approaches in the future, possibly through the connection between monads and predicate transformers established by Hasuo (2015).

Our work followed Kammar and Plotkin (2012) to use an effect system to validate program transformations. Their results are general to algebraic effects while we almost focused on the effect of mutable state, but as discussed in the paper, our mutable region system is more flexible when dealing with mutable data structures. Unlike theirs and most existing region systems, our mutable region system is defined as predicates in an logic for the programming language instead of within the type system for the language. The advantage of our choice is that our inference rules only need to deal with language constructs related to effects and we get the ability to handle higher order functions almost for free.

5.2 Conclusion

Our work started from an attempt to prove the correctness of the Schorr-Waite algorithm by equational reasoning, and as in many previous research works, we observed that the key is to prove two computations do not interfere and thus can be executed in any order. From the aspect of algebraic effects, non-interference means that these two computations use commutative effect operations, so the problem is reduced to track operations used by a computation, which is usually done with type-and-effect systems. However, existing static-region-based effect systems are inadequate for the Schorr-Waite algorithm because the mutable nature of the algorithms demands different region partitioning at different stages.

To address this problem, we proposed a mutable region system in which regions are determined by the points-to structure of memory cells, so that regions partitioning naturally follows when the points-to structure in memory is modified. Our system is formalised as effect predicates and separation guards. Semantics and sound inference rules for them are given and they allow us to formulate and prove statements like: this program only reads the cell linked from pointer p , and the linked lists from p_1 and p_2 are disjoint. With these tools, we can give an equational proof for the Schorr-Waite algorithm restricted to linked lists, which we think is intuitive and elegant.

5.3 Future work

The system described in this paper is very restrictive and needs future development in many aspects:

- We neglected effect handlers in this paper and it is important to incorporate them into the mutable region system in the future.
- It is also very beneficial to generalise our system from local-state to other dynamically creatable effects.
- Although we presented our system with only linked lists, it should be able to be generalised to arbitrary tree-like data structures easily. However, how to deal

with graphs in memory seems much more difficult. To prove the correctness of the Schorr-Waite algorithm on graphs using the method in [Section 4.2](#), we need to formulate a statement that *traverse g* only reads the nodes reachable from node *g* and not marked visited. Therefore the possible effect operations used by *traverse g* not only depends on the points-to structure of memory cells but also some other mutable state (keeping track of which nodes are visited), so an important question is how to upgrade effect predicates to be more expressive to describe the possible effects used by programs like this. If we aim at generality, it seems that eventually we need some expressive programming language to describe effect usage of programs precisely.

Bibliography

- Andrej Bauer. 2018. What is algebraic about algebraic effects and handlers? *CoRR* abs/1807.05923 (2018). arXiv:1807.05923 <http://arxiv.org/abs/1807.05923>
- Nick Benton, Andrew Kennedy, Lennart Beringer, and Martin Hofmann. 2007. Relational Semantics for Effect-based Program Transformations with Dynamic Allocation. In *Proceedings of the 9th ACM SIGPLAN International Conference on Principles and Practice of Declarative Programming (PPDP '07)*. ACM, New York, NY, USA, 87–96. <https://doi.org/10.1145/1273920.1273932>
- Nick Benton, Andrew Kennedy, Lennart Beringer, and Martin Hofmann. 2009. Relational Semantics for Effect-based Program Transformations: Higher-order Store. In *Proceedings of the 11th ACM SIGPLAN Conference on Principles and Practice of Declarative Programming (PPDP '09)*. ACM, New York, NY, USA, 301–312. <https://doi.org/10.1145/1599410.1599447>
- Nick Benton, Andrew Kennedy, Martin Hofmann, and Lennart Beringer. 2006. Reading, Writing and Relations. In *Programming Languages and Systems*, Naoki Kobayashi (Ed.). Springer Berlin Heidelberg, Berlin, Heidelberg, 114–130.
- Richard S. Bird. 2001. FUNCTIONAL PEARL: Unfolding Pointer Algorithms. *J. Funct. Program.* 11, 3 (May 2001), 347–358. <https://doi.org/10.1017/S0956796801003914>
- Lars Birkedal, Guilhem Jaber, Filip Sieczkowski, and Jacob Thamsborg. 2016. A Kripke Logical Relation for Effect-based Program Transformations. *Inf. Comput.* 249, C (Aug. 2016), 160–189. <https://doi.org/10.1016/j.ic.2016.04.003>

- Michael Butler. 1999. Calculational Derivation of Pointer Algorithms from Tree Operations. *Science of Computer Programming* 33 (03 1999), 221–260. [https://doi.org/10.1016/S0167-6423\(98\)00016-1](https://doi.org/10.1016/S0167-6423(98)00016-1)
- Jeremy Gibbons and Ralf Hinze. 2011. Just Do It: Simple Monadic Equational Reasoning. In *Proceedings of the 16th ACM SIGPLAN International Conference on Functional Programming (ICFP '11)*. ACM, New York, NY, USA, 2–14. <https://doi.org/10.1145/2034773.2034777>
- Ichiro Hasuo. 2015. Generic Weakest Precondition Semantics from Monads Enriched with Order. *Theor. Comput. Sci.* 604, C (Nov. 2015), 2–29. <https://doi.org/10.1016/j.tcs.2015.03.047>
- Ohad Kammar and Gordon D. Plotkin. 2012. Algebraic Foundations for Effect-dependent Optimisations. In *Proceedings of the 39th Annual ACM SIGPLAN-SIGACT Symposium on Principles of Programming Languages (POPL '12)*. ACM, New York, NY, USA, 349–360. <https://doi.org/10.1145/2103656.2103698>
- Paul Blain Levy. 2012. *Call-by-push-value: A Functional/imperative Synthesis*. Vol. 2. Springer Science & Business Media.
- J. M. Lucassen and D. K. Gifford. 1988. Polymorphic Effect Systems. In *Proceedings of the 15th ACM SIGPLAN-SIGACT Symposium on Principles of Programming Languages (POPL '88)*. ACM, New York, NY, USA, 47–57. <https://doi.org/10.1145/73560.73564>
- Bernhard Möller. 1997. *Calculating With Pointer Structures*. Springer US, Boston, MA, 24–48. https://doi.org/10.1007/978-0-387-35264-0_2
- Susumu Nishimura. 2008. Safe Modification of Pointer Programs in Refinement Calculus. In *Proceedings of the 9th International Conference on Mathematics of Program Construction (MPC '08)*. Springer-Verlag, Berlin, Heidelberg, 284–304. https://doi.org/10.1007/978-3-540-70594-9_16
- Gordon Plotkin and John Power. 2002. Notions of Computation Determine Monads. In *Foundations of Software Science and Computation Structures*, Mogens Nielsen

- and Uffe Engberg (Eds.). Springer Berlin Heidelberg, Berlin, Heidelberg, 342–356. https://doi.org/10.1007/3-540-45931-6_24
- Gordon Plotkin and John Power. 2004. Computational Effects and Operations: An Overview. *Electron. Notes Theor. Comput. Sci.* 73 (Oct. 2004), 149–163. <https://doi.org/10.1016/j.entcs.2004.08.008>
- Gordon Plotkin and John Power. 2008. Tensors of Comodels and Models for Operational Semantics. *Electron. Notes Theor. Comput. Sci.* 218 (Oct. 2008), 295–311. <https://doi.org/10.1016/j.entcs.2008.10.018>
- G. Plotkin and M. Pretnar. 2008. A Logic for Algebraic Effects. In *2008 23rd Annual IEEE Symposium on Logic in Computer Science*. 118–129. <https://doi.org/10.1109/LICS.2008.45>
- Gordon Plotkin and Matija Pretnar. 2013. Handling Algebraic Effects. *Logical Methods in Computer Science* 9, 4 (Dec 2013). [https://doi.org/10.2168/lmcs-9\(4:23\)2013](https://doi.org/10.2168/lmcs-9(4:23)2013)
- Matija Pretnar. 2010. *Logic and handling of algebraic effects*. Ph.D. Dissertation.
- John C. Reynolds. 2002. Separation Logic: A Logic for Shared Mutable Data Structures. In *Proceedings of the 17th Annual IEEE Symposium on Logic in Computer Science (LICS '02)*. IEEE Computer Society, Washington, DC, USA, 55–74. <http://dl.acm.org/citation.cfm?id=645683.664578>
- H. Schorr and W. M. Waite. 1967. An Efficient Machine-independent Procedure for Garbage Collection in Various List Structures. *Commun. ACM* 10, 8 (Aug. 1967), 501–506. <https://doi.org/10.1145/363534.363554>
- Sam Staton. 2010. Completeness for Algebraic Theories of Local State. In *Foundations of Software Science and Computational Structures*, Luke Ong (Ed.). Springer Berlin Heidelberg, Berlin, Heidelberg, 48–63.
- Hongseok Yang. 2007. Relational Separation Logic. *Theor. Comput. Sci.* 375, 1-3 (April 2007), 308–334. <https://doi.org/10.1016/j.tcs.2006.12.036>