Typed Adapton: Refinement types for nominal memoization

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Abstract. Nominal memoization combines memoized functional programming with a controlled form of imperative cache effects. By leveraging these imperative effects, nominal memoization can dramatically outperform traditional ("structural") memoization. However, the nominal memoization programming model is error-prone: when the programmer unintentionally misuses names, their incremental program ceases to correspond to its purely functional semantics.

This paper develops a refinement type system for nominal memoization that enforces a program's correspondence with its purely functional semantics. Our type system employs set-indexed types in the style of DML (Xi and Pfenning 1999), extended with polymorphism over kinds and index functions. We prove that our type system enforces the dynamic side conditions required by Hammer et al. (2015a). Past work shows that these conditions suffice for realistic uses of nominal memoization while also guaranteeing from-scratch consistency of the incremental programs. Furthermore, we propose a decidable form of name-level computations for expressing generic naming strategies in type-generic incremental code.

1 Introduction

Memoization is an evaluation technique that records a mapping from computation identities to computation results, enabling low-cost *reuse* of computations' results. For example, in simple cases, the identity of a function application is the function name and its argument.

Structural memoization is defined by comparing the input and output structures of memoized computations based on their structural identity, which is commonly implemented in O(1) time per pointer comparison by using a variant of hash-consing. Hash-consing is an allocation strategy that hashes and shares identical structures, assigning them identical pointers. This allocation strategy allows memoization implementors to use pointer identity to efficiently test structural identity.

Nominal memoization uses a notion of pointer name identity to identify both computations and data structures (input and output structures). Unlike structural memoization, the users of nominal memoization explicitly provide first-class names to each allocation site, and in turn, these names orchestrate an explicit, deterministic *naming strategy*, with direct implications for incremental computations, where names help isolate the impact of input changes.

Hammer et al. (2015a) give a library of incremental programs that mix first-class names and nominal memoization into otherwise conventional programs that compute with inductive input and output structures. The first-class names determine the identity of output structures and the computations that compute them; when these associations change (via an external "mutator"), change propagation repairs the demanded computation's output. In Sec. 2, we give a detailed example that maps a list of input elements to a list of output elements. Hammer et al. (2015a) showed that, using an explicit naming strategy via first-class names, the listmap algorithm can respond to an imperative input change such as element insertion with worst-case O(1) re-executions and fresh allocations, rather than the O(n) required with structural memoization, which rebuilds and rehashes the prefix of the changed output.

The practical performance of nominal memoization stems directly from how the programmer designs a naming strategy. To implement it, they instrument an otherwise-functional program with names that *explicitly relate input and output structures*. This instrumentation is error-prone, and making a mistake may change the meaning of the program, turning it into an imperative program whose from-scratch behavior no longer corresponds to a purely functional program, and whose incremental behavior may be unsound (that is, "glitchy").

This paper presents a refinement type system for nominal memoization of purely functional programs. Our type system enforces that all programs behave as though they are purely functional, excluding behavior that overwrites prior allocations with later ones. Such store objects are *one-shots*: they are written once and never mutated within the from-scratch run. Instead, changes occur at the *meta level* of the incremental computation, where input changes from the external environment induce output changes via a general-purpose change propagation algorithm. In this work, we limit our focus to the semantics of one-shot store objects; in turn, prior work showed how the soundness of nominal memoization, including change propagation, follows directly from this semantic property (Hammer et al. 2015b).

Compared with prior work, we extend the expressivity of naming strategies, permitting programs to be generic in how names of output structures are chosen, a pattern we refer to as *name parametricity*. This generality allows programmers to safely author and compose generic library code.

Contributions:

- We develop a type system for a variant of Nominal Adapton (Hammer et al. 2015a). We use types to *statically* enforce that nominal memoization behaves as though the program is purely functional, a condition modeled after the dynamic soundness criteria from Hammer et al. (2015a).
 - In this work, unlike prior work, we only consider non-incremental runs. This focus is justified, since the property of having a functional semantics is a

property of non-incremental runs. For functional correctness, we want to reason about incremental runs as non-incremental runs, and to reason about non-incremental runs as purely-functional runs, whose use of the store to name objects is inconsequential. Our type system enforces this latter property, whereas prior work established the former and *merely assumed* the latter for all programs.

- We formalize a type system that permits writing code that is generic in a naming strategy. In particular, we show various forms of name- and namefunction parametricity, and illustrate through examples its importance for expressing nominal memoization in composable library code.
- To encode the necessary invariants on names, our type system uses setindexed types in the style of DML (Xi and Pfenning 1999), extended with polymorphism over kinds and index functions; the latter was inspired by abstract refinement types (Vazou et al. 2013).

2 Overview: listmap example series

Consider the higher-order function listmap that, given a function over integers, maps an input list of integers to an output list in a pointwise fashion. Below, we consider several versions of listmap that only differ in their naming strategy: The names given to allocations differ, but nothing else about the algorithm does. In each case, we show the naming strategy in terms of both its program text and type, each of which reflect how names flow from input structures into output structures. In particular, the refinement type structure exposes the naming strategy of each version, and it creates a practical static abstraction for enforcing a global naming strategy with composable parts.

The type below defines incremental lists of integers, (List[X;Y] Int). This type has two conventional constructors, Nil and Cons, as well as two additional constructors that use the two type indices X and Y. The Name constructor permits names from set X to appear in the list sequence; it creates a Conscell-like pair holding a first-class name from X and a sub-list for the rest of the sequence. The ref constructor permits injecting (incrementally changing) references holding sub-lists into the list type; these pointers have names drawn from set Y.

```
\begin{array}{lll} \text{Nil}: & \forall X,Y. & \text{List}[X;Y] \text{ Int} \\ \text{Cons}: & \forall X,Y. & \text{Int} \rightarrow \text{List}[X;Y] \text{ Int} \rightarrow & \text{List}[X;Y] \text{ Int} \\ \text{Name}: \forall X_1,X_2,Y. & \text{Nm}[X_1] \rightarrow & \text{List}[X_2;Y] \text{ Int} \rightarrow & \text{List}[X_1 \perp X_2;Y] \text{ Int} \\ \text{Ref}: \forall X,Y_1,Y_2. & \text{Ref}[Y_1] \text{ (List}[X;Y_2] \text{ Int}) \rightarrow & \text{List}[X;Y_1 \perp Y_2] \text{ Int} \\ \end{array}
```

For both of these latter forms, the constructor's types enforce that each name (or pointer name) in the list is disjoint from those in the remainder of the list. For instance, we write $X_1 \perp X_2$ for the disjoint union of X_1 and X_2 , and similarly for $Y_1 \perp Y_2$. However, the names in X and pointer names in Y may overlap (or even coincide); in particular, we use the type List[X;X] Int in the output type of listmap1, below.

The Nil constructor creates an empty sequence with *any* pointer or name type sets in its resulting type. For practical reasons, we find it helpful to permit types to be *over-approximations* of the names actually used in each instance. In light of this, the type of the constructor for Nil makes sense.

2.1 listmap1: Naive nominal memoization

First, we consider the simplest naming strategy for listmap, where input and output names coincide; statically, both are drawn from an arbitrary name set X. Meanwhile, the names of the input reference cells are drawn from a name set Y. Notice how the output type list type communicates that the output names and pointer names are drawn from the same set X, the name set of the input list (that is, its first type index).

```
listmap1 : \forall X, Y : \mathsf{NmSet.} (Int \rightarrow Int) \rightarrow (List [X;Y] Int) \rightarrow (List [X;X] Int) listmap1 = \lambda f. fix rec. \lambda l.

match 1 with

Nil \Rightarrow Nil,

Cons(h,t) \Rightarrow Cons(f h, rec t)

Ref(r) \Rightarrow rec (get r)

Name(n,t) \Rightarrow Name(n, Ref(memo_ref n [ rec t ]))
```

Turning to the program text for listmap1, the Nil and Cons cases below are conventional. The latter two cases handle pointers and names in the list. In the Ref case, the programmer observes the contents of the reference cell holding the remainder of the input, and recurs. In the Name case, the programmer injects the name n into the output list, but also uses this name in the construct memo_ref to to allocate named a reference cell, and a memoized recursive call on the list tail. We use memo_ref to combine these steps into one step that consumes a single name. In fact, this construct decomposes into a special combination of simpler steps (viz., reference allocation, thunk allocation, and thunk forcing). Sec. 3 presents a formalism for these decomposed steps.

Example demanded computation graph (DCG). Suppose that we execute listmap1 on an input list with two integers, two names n_1 and n_2 , and two reference cells p_1 and p_2 , as shown in Fig. 1. Recall that listmap1 employs a naming strategy that uses the names n_1 and n_2 from input list inp as the names and pointers in output list out, shown below the input list, connected by a complex of DCG structure. Bold boxes denote references in the input and output lists, and bold circles denote memoized thunks that compute with them. Each bold store object (thunk or reference) is associated with a name.

The two thunks, shown as bold circles named by n_1 and n_2 , follow a similar pattern: each carries outgoing edges to the Name list structure to which they correspond (north- and south-west edges), the reference that they dereference (northeast), the recursive thunk that they force, if any (due west), and the reference that they name and allocate (southeast), either named n_1 or n_2 .

The benefit of using names to identify these memoized thunks and their output list references is that imperative O(1) changes to the input structure can be reflected into the DCG with only O(1) changes to its named content. Further, the names also provide a primitive form of *cache eviction*: Memoized results are *overwritten* when names are re-associated with new content. By contrast, structural memoization generally requires rebuilding recursive structures whose content changes; in this case, an O(1) change to inp will generally require rehashing and rebuilding a linear number of output cells, most of which duplicate the content of an existing cell, modulo the O(1) change. This contrast is detailed in prior work (Hammer et al. 2015b).

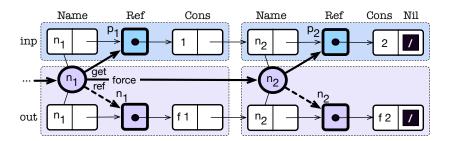


Fig. 1. Example DCG, from running listmap1 on inp, a two-integer, two-name input list. Its names are drawn from set $X = \{n_1, n_2, \ldots\}$, and its pointer names are drawn from the set $Y = \{p_1, p_2, \ldots\}$. The two lower purple sections consist of store objects named by n_1 and n_2 ; they result from applying listmap1 to this input list. The DCG consists of these objects and their relationships, shown as directed edges (the undirected edges are just for illustration).

Unintended imperative effects: Unintended feedback and churn. Conceptually, the program above captures the idea of mapping a list, with several additional nominal steps in the Name case: the program maps input names to output names, output reference cells, and a corresponding memoized recursive call; all are identified with the same name. Conceptually, these are the right steps; but, on closer inspection, however, two problems arise quickly with this naive naming strategy:

- 1. Unintended feedback: If the input pointer names are drawn from the same set as those in the Name cells, then the allocations that listmap1 performs will overwrite these input pointers with output pointers. Note that above, if X = Y, this is indeed possible.
- 2. **Unintended churn**: If the same input list is mapped twice in the same incremental program with two different element map functions, each alternation from one to the other will overwrite the prior with the latter.

```
let xs = listmap1 abs inp
let ys = listmap1 sq inp
```

For example, the mapping of sq will overwrite the mapping of abs, since the output pointers of each mapping coincide.

Likewise, if another function, e.g., listfilter, uses the same naming strategy as listmap1, and these functions are composed to create a larger incremental function pipeline, the first function will overwrite the output of the second computation, rather than retaining incrementally-changing versions of both simultaneously.

Both feedback and churn arise because the names equip the program with a dynamic semantics that overlays an *imperative* allocation strategy atop an otherwise *purely functional* algorithm. Generally, as programmers, we want to exploit this imperative allocation strategy to enable efficient incremental computations, via a clever use of imperative overwriting, memoization, and dependency tracking. However, in the cases where the imperative, incremental changes are unintended, they lead to unintended behavior, either in terms of performance, or correctness, or both.

2.2 listmap2: Name parametricity for the output list

Fortunately, first-class names allow us to begin addressing both problems described above:

```
\label{eq:listmap2} \begin{array}{ll} \texttt{listmap2} & : & \forall X,Y : \mathsf{NmSet.} & \forall Z : \mathsf{Nm.} & \mathsf{Nm[Z]} \rightarrow (\mathsf{Int} \rightarrow \mathsf{Int}) \rightarrow (\mathsf{List}[X;Y] \; \mathsf{Int}) \\ & \rightarrow (\mathsf{List}[X;(Z,X)] \; \mathsf{Int}) \\ \\ \texttt{listmap2} & = \; \lambda \mathsf{m.} \; \; \lambda \; \; \mathsf{f.} \; \; \mathsf{fix} \; \; \mathsf{rec.} \; \; \lambda \; \; \mathsf{1.} \\ & \; \; \mathsf{match} \; \; \mathsf{l} \; \; \mathsf{with} \; \; \cdots \\ & \; \; \mathsf{Name}(\mathsf{n,t}) \; \Rightarrow \; \mathsf{Name}(\mathsf{n}, \; \mathsf{Ref}(\mathsf{memo\_ref}(\mathsf{m,n}) \; [\; \mathsf{rec} \; \mathsf{t} \; ])) \end{array}
```

We adapt the original program by adding a new parameter, first-class name m, of type Nm[Z] and in the Name case of the pattern match, we name the output structure by the name pair (m,n) instead of just the name n. Statically, the index of the output list's reference cells is (Z,X) instead of merely X, allowing us to use different names in set Z to distinguish different, simultaneous uses of the function.

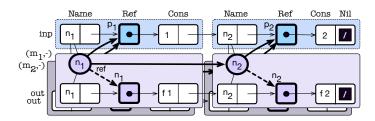


Fig. 2. Using disjoint names $m_1 \perp m_2$ to create namespaces $(m_1, -)$ and $(m_2, -)$

Because it is parametric in the function **f** as well as the distinguished name **m**, a calling context may use **listmap** to instantiate different integer mappings, as follows:

```
let x = listmap2 m1 abs inp
let y = listmap2 m2 sq inp
```

Assuming that the names $\mathtt{m1} \in \mathsf{Z}$ and $\mathtt{m2} \in \mathsf{Z}$ are disjoint (or "apart"), which we write as $\mathtt{m1} \perp \mathtt{m2}$, we have that the reference cells and memoized calls for \mathtt{x} and \mathtt{y} are distinct from each other, and co-exist without overwriting one another. Fig. 3 illustrates this scenerio graphically, with two disjoint layers of output.

Otherwise, if we had that m1 = m2, the code would exhibit the churn pattern described above, which may not be intended. The type system proposed by this paper rules out this behavior, as well as all other cases of churn or feedback.

2.3 listmap3: Name-function parametricity for the output List

When writing reusable library code, the programmer may want to avoid choosing the specifics of how to name the output list structure, and its memoized computation. For instance, when allocating the reference cell in the Name case above, rather than choose to pair the names $\mathfrak m$ and $\mathfrak n$, she may want to be even more generic by defering even more decisions to the calling context. To express this generality, she parameterizes listmap with a name-transforming function nmf. She names the reference cells of the output list type by applying nmf to the names of the input list. In terms of program text, this generalization universally quantifies over the name function nmf, and it applies this function in the Name case.

```
listmap3 : \forall X, Y : \mathsf{NmSet}. \ \forall \mathsf{nmf} : \mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}. \ (\mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}) [\mathsf{nmf}]
\to (\mathsf{Int} \to \mathsf{Int}) \to (\mathsf{List}[X;Y] \ \mathsf{Int}) \to (\mathsf{List}[X; \, \mathsf{nmf} \ X] \ \mathsf{Int})
listmap3 l = \lambda \mathsf{nmf}. \lambda f. fix rec. \lambda l.

match l with
...

Name(n,t) \Rightarrow Name(n, Ref (memo_ref (nmf n)) [rec t]))
```

Compared to the version above that is parametric in a distinguished name m, this version is more general; to recover the first version, let the name function nmf be one that pairs its argument with the distinguished name m, viz., $\lambda a.(m,a)$, as asserted below:

```
listmap2 m f l = listmap3 (\lambda a.(m,a)) f l
```

2.4 listmap4: Naming strategy for higher-order parametricity

Suppose we want a map function that works over any type of lists, not just lists of integers. To do so with maximum generality, we may also want the computation to be generic in the naming relationships of the input and output list elements.

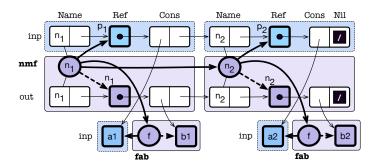


Fig. 3. listmap4 permits higher order composition, with a non-trivial element mapping function f, and output element namespace fab. Though drawn as single references and thunks, the DCG of the element mapping f could be any generic structure, including the mapping of another inductive data structure.

The programmer achieves this generality by abstracting over several structure-specific choices: The input and output list's element type structure, call it A and B, respectively, as well as the name structure of these types, and their relationship. In particular, we introduce another name function fab to abstract over the ways that element map function f uses the names in structures of type A to name of the output element structures of type B. To classify the types A and B, we use the kinds $k_A \Rightarrow \star$ and $k_B \Rightarrow \star$, respectively; the index sorts k_A and k_B classify the name-based indices of these types, and the index sort $k_A \rightarrow_{\mathsf{Nm}} k_B$ classifies the name function fab, which relates names of type A to names of type B.

```
\begin{array}{l} \forall X,Y: \mathsf{NmSet}. \, \forall \mathsf{nmf}: \mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}. \\ \forall A: k_A \Rightarrow \star . \, , \forall B: k_B \Rightarrow \star . \quad \forall \mathsf{fab}: k_A \to_{\mathsf{Nm}} k_B. \quad \forall Z: k_A. \\ (\mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}) \, [\mathsf{nmf}] \\ \to (A \, [Z] \to B \, [\mathsf{fab} \, Z] \, ) \\ \to (\mathsf{List} \, [X;Y] \, A \, [Z] \, ) \to (\mathsf{List} \, [X; \mathsf{nmf} \, X] \, B \, [\mathsf{fab} \, Z] \, ) \end{array}
```

In the transition from listmap3 to listmap4, neither the program text nor the algorithm change, only the type, given above.

Fig. 4 summarizes the four versions of listmap that we toured above. Each highlights differences from the prior, with listmap1 indicating the type indices that decorate the "usual" type for listmap: $(Int \rightarrow Int) \rightarrow List Int \rightarrow List Int$. Next, listmap2 is parameteric in a name, which it pairs with the input names; listmap3 generalizes this pattern to any name function nmf, not just pair-producing ones; listmap4 generalizes the mapping function to relate any two types, A and B, which any naming strategy fab.

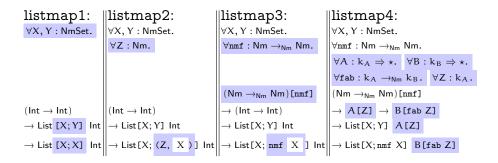


Fig. 4. Comparison of types for listmap, including differences.

2.5 Monadic scopes hide the plumbing of a naming strategy

In the final version of listmap above, we see that being fully parametric in both the type and the naming strategy means introducing a lot of type structure that can be viewed as "plumbing". Specifically, for each name-index function, we must instrument the code to abstract over and apply the name function when the corresponding allocations occur (e.g., as with fab and nmf above). This adds some bookkeeping to both the types and the programs that inhabit them.

In some cases, it is useful to obey a more restrictive monadic scoping discipline, which helps hide the plumbing of generic programs in the style of listmap4, above. Specifically, we use the original, simplist version listmap1, along with caller-named scopes that disambiguate different uses of the names in the common input list:

```
let x = scope m1 [ listmap1 abs inp ]
let y = scope m2 [ listmap1 sq inp ]
```

Intuitively, the use of scope merely hides the explicit name function parameter (either m1 or m2, above), threading it behind the scenes until an allocation or memoization point occurs, where the name function maps the given name into a particular *namespace*. In this sense, scope controls a monadic peice of state that is implicitly carried throughout a dynamic scope of computation.

Limitations of disjoint dynamic scopes: No sharing. While permitting economical sizes of code and types, the disadvantages of scope relate to that of monads generally: Sometimes the "plumbing" hidden by a monadic computation cannot express the generic composition pattern that the programmer intends. For instance, differently-scoped sub-computations will never share any allocations or sub-computations since, by definition, their allocated names will be disjoint. Meanwhile, the programmer may want to permit (safe) forms of sharing within their naming strategy.

For instance, suppose two lists share common elements and we wish to map both of them with the same element-mapping function. In this case, we may use listmap4 to share the allocation of mapped elements, but not the allocation of list cells (which could do strange things, and generally would not preserve the structure of the input lists). Meanwhile, using listmap1 and different scopes would fail to share the allocation of mapped elements, making the two subcomputations completely disjoint.

We should note that the type system presented here does not permit any sharing, though we imagine that with special annotations, we could make such effects explicit, and rely on the human programmer to check these uses for safety. As future work, we sketch another annotation-based approach in Sec. 8 for expressing controlled forms of feedback.

3 Program Syntax

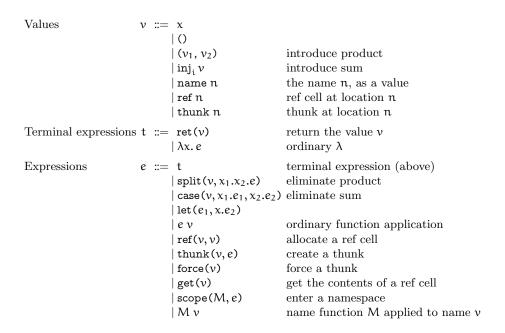


Fig. 5. Syntax of expressions

Fig. 5 gives the abstract syntax for expressions e and values v. We use call-by-push-value (CBPV) conventions in this syntax, and in the type system that follows. There are several reasons for this: CBPV can be interpreted as a "neutral" evaluation order that corresponds with call-by-value or call-by-name, but prefers neither in its design. Further, since we make the unit of memoization a thunk, and since CBPV explicates the creation of all closures and thunks, it also exposes exactly the structure that we wish to extend as a general-purpose

abstraction for incremental computation. In particular, thunks are the unit by which we cache results and track dynamic dependencies. Finally, focusing on thunks helps us model a demand-driven incremental computation, e.g., incremental, demand-driven sorting algorithms.

Expressions consist of elimination forms for pairs and sums (split and case, respectively); intro and elimination forms for producing values (ret and let, respectively); intro and elimination forms for function types (lambda and application, respectively); intro and elimination forms for thunks, and intro and elimination forms for value pointers (reference cells that hold values).

The syntax in these figures closely follows that of prior work on Adapton, including Hammer et al. (2015a). We generalize that work with the notion of a name function, which generalizes the idea of a namespace and also operations over names such as "forking". The construct scope controls monadic state for the current name function, adding a function to its function composition for the dynamic extent of its subexpression e. The name function application form permits programs to compute with names and name functions that reside within the type indicies. Since these name functions always terminate, they do not affect a program's termination behavior.

The syntax for values consists of the unit value, pairs of values, tagged injected values (one of the two "halves" of a sum type), first-class names, value pointers and thunk pointers.

We do not distinguish syntactically between value pointers (for reference cells) and thunk pointers (for suspended expressions); the store maps pointers to either of these possibilities.

Desugaring memo_ref. We can desugar the syntax memo_ref into the following construction of primitive operations: the creation and elimination of a named thunk that allocates a named reference cell. We distinguish the two name uses with the tags 1 and 2:

```
memo_ref n [e] = force (thunk n.1 (let x = e in ref(n.2,x)))
```

4 Type System

The structure of our type system is inspired by Dependent ML (Xi and Pfenning 1999; Xi 2007). In DML—unlike a fully dependently typed system—the system is separated into a program level and a less-powerful index level. The classic DML index domain is integers with linear inequalities, making type-checking decidable. Our index domain includes names, sets of names, and functions over names. Such functions constitute a tiny domain-specific language that is powerful enough to express useful transformations of names, but preserves decidability of type-checking.

In DML, indices usually have no direct computational content. For example, when applying a function on vectors that is indexed by vector length, the length

```
Kinds
                                     \mathsf{K} \ ::= \ \star
                                           |\star\Rightarrow K
                                                                  type argument (binder space)
                                            \mid \gamma \Rightarrow K
                                                                 index argument (binder space)
Name terms
                                    M \; ::= \; n
                                                                 name
                                            ()
                                                                 \quad \text{unit} \quad
                                            \mid \lambda \alpha.\, M
                                                                 {\bf abstraction}
                                                                  variable
                                            | a
                                            | M(M)
                                                                 application
                                            | M.1
                                                                 append 1 to name
                                            | M.2
                                                                 append 2 to name
                                            |(M_1, M_2)|
                                                                 tuple
Index sorts
                                     \gamma ::= Nm
                                                                 name
                                            | NmSet
                                                                 name set sort
                                            | 1
                                                                 unit index sort
                                            | \gamma * \gamma
                                                                 product index sort
                                            \mid \gamma \to_{\mathsf{Nm}} \gamma
                                                                 index-level function
Symbols
                                     s
Names
                           n, m, p, q := s
                                                                 \operatorname{symbol}
                                            root
                                                                 {\rm root\ name}
                                            | n.1 | n.2
                                                                 append bit (disjunctive names)
                                            |(n_1, n_2)|
                                                                 pairing (conjunctive names)
Index variables
Index expressions i, X, Y, Z, R, W := a
                                            | \{n\}
                                                                 singleton set
                                            | X \perp Y
                                                                 separating union
                                            | () | (i, i) | M[i]
```

Fig. 6. Syntax for indices

```
Effects
                             \epsilon ::= \langle W; R \rangle
Type constructors
                            d
Type variables
                            α
                         A,B \ ::= \ \alpha
Value types
                                                               type variable
                                   | d
                                                               type constructor
                                    |A + B|
                                                              \operatorname{sum}
                                    |A \times B|
                                                               product
                                    unit
                                                              named reference cell
                                    | Ref [i] A
                                    | Thk[i] E
                                                              named thunk (with effects)
                                    | A[i]
                                                              application of type to index
                                    A B
                                                              application of type constructor to type
                                    | Nm [i]
                                                              name type (name contained in name set i)
                                    \mid (Nm \rightarrow_{Nm} Nm)[M] name function type (singleton)
Computation types C, D := FA
                                                              liFt
                                   |A \rightarrow E|
                                                              function
\dots with\ effects
                            \mathsf{E} \ ::= \ \mathsf{C} \rhd \varepsilon
                                                              {\it effects}
                                    | \forall \alpha : K. E
                                                              type polymorphism
                                    \mid \forall \alpha : \gamma \text{. E}
                                                              index\ polymorphism
Typing contexts
                            \Gamma \ ::= \ \cdot
                                    | Γ, p : A
                                                              ref pointer
                                    | Γ, p : E
                                                              thunk pointer
                                    |\Gamma, x:A
                                                              value variable
```

Fig. 7. Syntax for types

Fig. 8. Sorts statically classify both name terms M, as well as the name indices i that index types

Fig. 9. Kinds statically classify types and effects

Fig. 10. Value typing

index is not directly manipulated at run time. However, indices do reflect properties of run-time values. The simplest case is that of an indexed *singleton type*, such as Int[k]. Here, the ordinary type Int and the index domain of integers are in one-to-one correspondence; the type Int[3] has one value, the integer 3.

While indexed singletons work well for the classic index domain of integers, they are less suited to names—at least for our purposes. Unlike integer constraints, where integer literals are common in types—for example, the length of the empty list is 0—literal names are rare in types. Many of the name constraints we need to express look like "given a value of type A whose name in the set X, this function produces a value of type B whose name is in the set f(X)". A DML-style system can express such constraints, but the types become verbose:

$$\forall \alpha : \mathsf{Nm.} \ \forall X : \mathsf{NmSet.}$$

 $(\alpha \in X) \supset (A[\alpha] \to B[f(\alpha)])$

The notation is taken from one of DML's descendants, Stardust (Dunfield 2007). The type is read "for all names a and name sets X, such that $a \in X$, given some A[a] the function returns B[f(a)]".

We avoid such locutions by indexing single values by name sets, rather than names. For types of the shape given above, this cuts the number of quantifiers in half, and obviates the \in -constraint attached via \supset :

$$\forall X : \mathsf{NmSet.}$$

 $A[X] \to B[f(X)]$

This type says the same thing as the earlier one, but now the approximations are expressed within the indexing of A and B. Note that f, a function on names, is interpreted pointwise: $f(X) = \{f(x) \mid x \in X\}$.

(Standard singletons will come in handy for index functions on names, where one usually needs to know the specific function.)

$$\begin{split} & (\varepsilon_1 \operatorname{then} \varepsilon_2) = \varepsilon} \\ & ((W_1; R_1) \operatorname{then} \langle W_2; R_2 \rangle) = \langle W_1 \cup W_2; R_1 \cup R_2 \rangle & \text{if } W_1 \perp W_2 \text{ and } R_1 \perp W_2 \\ \hline ((E \operatorname{after} \varepsilon) = E') \\ & ((C \rhd \varepsilon_2) \operatorname{after} \varepsilon_1) = (C \rhd (\varepsilon_1 \operatorname{then} \varepsilon_2)) & ((\forall \alpha : \gamma, E) \operatorname{after} \varepsilon) \\ & ((C \rhd \varepsilon_2) \operatorname{after} \varepsilon) = (C \rhd (\varepsilon_1 \operatorname{then} \varepsilon_2)) & ((\forall \alpha : \gamma, E) \operatorname{after} \varepsilon) \\ & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd (\varepsilon_1 \operatorname{then} \varepsilon_2)) & ((\nabla \circ \varepsilon_1) \operatorname{after} \varepsilon) \\ \hline & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd (\varepsilon_1) \operatorname{after} \varepsilon) \\ \hline & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) \\ \hline & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) \\ \hline & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) \\ \hline & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & ((C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after} \varepsilon) \\ \hline & (C \rhd \varepsilon_1) \operatorname{after} \varepsilon) = (C \rhd \varepsilon_2) \operatorname{after$$

Fig. 11. Computation typing

For aggregate data structures such as lists, indexing by a name set denotes an *overapproximation* of the names present. That is, the proper DML type

$$\forall Y : \mathsf{Nm.} \ \forall X : \mathsf{NmSet.}$$
 $(Y \subseteq X) \supset (A[Y] \to B[f(Y)])$

can be expressed by

$$\forall X : \mathsf{NmSet.}$$
 $(A[X] \to B[f(X)])$

Following call-by-push-value (Levy 1999, 2001), we distinguish *value types* from *computation types*. Our computation types will also model effects, such as the allocation of a thunk with a particular name.

4.1 Index Level

We use several meta-variables for index expressions. By convention, X, Y, Z, R and W are sets of names; i is any index (perhaps a pair of indices).

Names Names are the value form of name terms. A name n is either a symbol s, the root name root, the "halves" n.1 and n.2, or a pair of two names (n_1, n_2) .

Name terms Name terms model names and functions over names; they correspond to terms in a λ -calculus extended with a type of names. A name term M is either a name (value) n, the unit name (), a function $\lambda a.M$ or argument a, application M_1 M_2 , the tuple (M_1 , M_2), or a "half" (M.1 or M.2). Name terms do *not* allow recursion or destruction: a name function cannot case-analyze its argument.

We write $M \downarrow_M V$ for big-step evaluation of name terms; the rules are given in Figure 12.

We write $M \equiv M'$ when name terms M and M' are convertible, that is, applying a series of β -reductions and/or β -expansions to one term results in the other.

Name sets Supposing we give a name to each element of a list. Then the entire list should carry the set of those names. We write $\{n\}$ for the singleton name set, and $X \perp Y$ for a union of two sets X and Y that requires X and Y to be disjoint.

Indices An index i (also written X, Y, \ldots when the index is a set of names) is either an index-level variable a, a name set $\{n\}$ or $X \perp Y$, the unit index (), a pair of indices (i_1, i_2) , or a name term M applied to an index M[i]. For example, if $M = (\lambda a. a.1.2)$ then M[root] = root.1.2.

Sorts The index level has its own type system; to reduce confusion, types at this level are called index *sorts*. The sort Nm classifies indices that are single names, and the sort NmSet classifies name sets. To combine index sorts γ_1 and γ_2 , use the product sort $\gamma_1 * \gamma_2$. Finally, the sort \rightarrow_{Nm} classifies index-level functions: Nm \rightarrow_{Nm} Nm takes a name and returns a name.

Entailment We assume an entailment relation $\Gamma \vdash P$, where P is a set-theoretic proposition such as $\mathfrak{n} \in X$ or $X \subseteq Y$ or $X \perp Y$; the latter is interpreted as $(X \cap Y) = \emptyset$.

We assume that entailment is closed under conversion: for example, if $M(n) \equiv n'$, then $\Gamma \vdash M(n) \in N$ iff $\Gamma \vdash n' \in N$. We also assume that weakening holds: if $\Gamma_1, \Gamma_3 \vdash P$ then $\Gamma_1, \Gamma_2, \Gamma_3 \vdash P$.

4.2 Kinds

We use a relatively simple system of *kinds* K to classify the different animals in the type system:

- The kind ★ classifies value types, such as unit and (Thk[i] E).
- The kind $\star \Rightarrow K$ classifies type expressions that are parametrized by a type. Such types are called *type constructors* in some languages; for example, list by itself has kind $\star \Rightarrow \star$ (and list unit has kind \star).
- The kind $\gamma \Rightarrow K$ classifies type expressions that are parametrized by an index. For example, the List type constructor from Section 2 takes two name sets and the type of the list elements, e.g. List[X;Y] Int. Therefore, List has kind NmSet \Rightarrow (NmSet \Rightarrow ($\star \Rightarrow \star$). (The inner kind ($\star \Rightarrow \star$) is an instance of the $\star \Rightarrow K$ kind, not the $\gamma \Rightarrow K$ kind.)

4.3 Effects

Effects are described by $\langle W; R \rangle$, meaning that the associated code may write names in W, and may read names in R.

Effect sequencing is a (meta-level) partial function over a pair of effects: if ϵ_1 then ϵ_2 is defined and equal to ϵ , then ϵ describes the combination of having effects ϵ_1 followed by effects ϵ_2 . Sequencing is a partial function because the effects are only valid when (1) the writes of ϵ_1 are disjoint from the writes of ϵ_2 , and (2) the reads of ϵ_1 are disjoint from the writes of ϵ_2 . Condition (1) holds when each cell or thunk is not written more than once (and therefore has a unique value). Condition (2) holds when each cell or thunk is written before it is read.

A second meta-level partial function, effect coalescing—written "E after ε "—combines "clusters" of effects. For example:

$$(C \rhd \langle \{n_2\}; \emptyset \rangle)$$
 after $\langle \{n_1\}; \emptyset \rangle = C \rhd (\langle \{n_1\}; \emptyset \rangle \text{ then } \langle \{n_2\}; \emptyset \rangle) = C \rhd \langle \{n_1, n_2\}; \emptyset \rangle$

Coalescing goes under quantifiers:

$$(\forall \alpha : \star. C \rhd \langle \emptyset; \emptyset \rangle) \text{ after } \langle \emptyset; \{n_3\} \rangle = \forall \alpha : \star. C \rhd (\langle \emptyset; \emptyset \rangle) \text{ after } \langle \emptyset; \{n_3\} \rangle) = \forall \alpha : \star. (C \rhd \langle \emptyset; \{n_3\} \rangle)$$

4.4 Types

The value types, written A, B, in Figure 7 include standard sums + and products \times , a unit type, the type Ref[i] A of references named i containing a value of type A, the type Thk[i] E of thunks named i whose contents have type E (see below), the application A[i] of a type to an index, the application A B of a type A (e.g. a type constructor d) to a type B, the type Nm[i], and a singleton type (Nm \rightarrow_{Nm} Nm)[M] where M is a function on names.

As usual in call-by-push-value, computation types C and D include a connective F, which "lifts" value types to computation types: FA is the type of computations that, when run, return a value of type A. (Call-by-push-value usually has a connective dual to F, written U, that "thUnks" a computation type into a value type; in our system, Thk plays the role of U.)

Computation types also include functions, written $A \to E$. In standard CBPV, this would be $A \to C$, not $A \to E$. We separate computation types alone, written C, from computation types with effects, written E; this decision is explained below.

Computation types-with-effects E consist of $C \triangleright \varepsilon$, which is the bare computation type C with effects ε , as well as universal quantifiers (polymorphism) over types $(\forall \alpha : \mathsf{K}. \, \mathsf{E})$ and indices $(\forall \alpha : \gamma. \, \mathsf{E})$.

Why distinguish computation types from types-with-effects? Can we unify computation types C and types-with-effects E? Not easily. We have two computation types, F and \rightarrow . For F, the expression being typed could create a thunk, so we must put that effect somewhere in the syntax. For \rightarrow , applying a function is (per call-by-push-value) just a "push": the function carries no effects of its own (though its codomain may need to have some). However, suppose we force a thunked function of type $A_1 \rightarrow (A_2 \rightarrow \cdots)$ and apply the function (the contents of the thunk) to one argument. In the absence of effects, the result would be a computation of type $A_2 \rightarrow \cdots$, meaning that the computation is waiting for a second argument to be pushed. But, since the act of forcing the thunk has the effect of reading the thunk, we need to track this effect in the result type. So we cannot return $A_2 \rightarrow \cdots$, and must instead put effects around $(A_2 \rightarrow \cdots)$. Thus, we need to associate effects to both F and \rightarrow , that is, to both computation types.

Now we are faced with a choice: we could (1) extend the syntax of each connective with an effect (written next to the connective), or (2) introduce a "wrapper" that encloses a computation type, either F or \to . These seem more or less equally complicated for the present system, but if we enriched the language with more connectives, choice (1) would make the new connectives more complicated, while under choice (2), the complication would already be rolled into the wrapper. We choose (2), and write the wrapper as $\mathsf{C} \rhd \varepsilon$, where C is a computation type and ε represents effects.

Where should these wrappers live? We could add $C \triangleright \epsilon$ to the grammar of computation types C. But it seems useful to have a clear notion of *the* effect associated with a type. When the effect on the outside of a type is the only

effect in the type, as in $(A_1 \to FA_2) \rhd \varepsilon$, "the" effect has to be ε . Alas, types like $(C \rhd \varepsilon_1) \rhd \varepsilon_2$ raise awkward questions: does this type mean the computation does ε_2 and then ε_1 , or ε_1 and then ε_2 ?

We obtain an unambiguous, singular outer effect by distinguishing types-with-effects E from computation types C. The meta-variables for computation types appear only in the production E := $C \triangleright \varepsilon$, making types-with-effects E the "common case" in the grammar. Many of the typing rules follow this pattern, achieving some isolation of effect tracking in the rules.

5 Dynamics

Name term values
$$V ::= n \mid \lambda a. M \mid a \mid () \mid (V, V) \mid V.1 \mid V.2$$

$$\boxed{M \Downarrow_M V} \text{ Name term } M \text{ evaluates to name term value } V$$

$$\boxed{\frac{M_1 \Downarrow_M V}{V \Downarrow_M V}} \text{ teval-value} \qquad \frac{M_1 \Downarrow_M \lambda a. M \qquad M_2 \Downarrow_M V_2 \qquad [V_2/a]M \Downarrow_M V}{(M_1 M_2) \Downarrow_M V} \text{ teval-app}}$$

$$\boxed{\frac{M_1 \Downarrow_M V_1 \qquad M_2 \Downarrow_M V_2}{(M_1, M_2) \Downarrow_M (V_1, V_2)}} \text{ teval-tuple} \qquad \frac{M \Downarrow_M V}{M.1 \Downarrow_M V.1} \text{ teval-append}$$

$$\boxed{\frac{M \Downarrow_M V}{M.1 \Downarrow_M V.1}} \text{ teval-append}$$

Fig. 12. Name term evaluation

```
Graphs G,H := \epsilon empty graph  | G,p:\nu | p \text{ points to value } \nu   | G,p:e @ M p \text{ points to thunk } e \text{ in namespace } M \text{ (no cached result)}
```

Fig. 13. Graphs without dependency or cache structure

5.1 Dynamics of name terms

Fig. 12 gives the dynamics for evaluating a name term M into a name term value V. Because name terms lack the ability to perform recursion and pattern-matching, it is easy to see that they always terminate.

Fig. 14. Dynamics for non-incremental evaluation

5.2 Dynamics of program expressions

Fig. 13 defines graphs, the mutable state that names control dynamically. In this work, we are only interested in showing a correspondence to a purely functional run, so we need not model the cache and dependency-graph structure of prior work. As a result, our "graphs" are merely stores that map pointer names to values and expressions.

Fig. 14 defines the big-step evaluation relation for expressions, relating an initial and final graph, as well as a namespace and "current node" to a program. Because we do not build the dependency graph, the "current node" is not relevant here; the current namespace is used in rules ref and thunk to help name the allocated pointer. The shaded rules (including those two, and rules get and force for accessing their contents) change in the incremental version of the semantics, building and using the dependency graph structure that we omit here, for simplicity.

6 Statics Metatheory: Type Soundness and "Pure" Effects

In this section, we prove that our type system and dynamics agree. Further, we show that the type system enforces a purely functional effect discipline, whose dynamics is codified formally by Def. 1, below. Our main theorem establishes that a well-typed, terminating program produces a terminal computation of the program's type, and that the actual dynamic effects are consistent with a purely functional allocation strategy. Specifically, sequenced writes never overwrite one another.

6.1 Store typing

$$\frac{G \vdash \Gamma : A}{ \Gamma \vdash p \text{ has-store-type } A} \text{ ref}$$

$$\frac{G \vdash \Gamma \qquad \Gamma \vdash p \text{ has-store-type } A}{\vdash (G, p : \nu) : \Gamma} \text{ ref}$$

$$\frac{\Gamma \vdash e : E}{\vdash (G, p : e) : \Gamma} \text{ thunk}$$

Fig. 15. Store typing

6.2 Read and Write Sets

Join and merge operations. We also define a merge $H_1 \cup H_2$ that is defined for subgraphs with overlapping domains, provided H_1 and H_2 are consistent with each other. That is, if $\mathfrak{p} \in \mathsf{dom}(H_1)$ and $\mathfrak{p} \in \mathsf{dom}(H_2)$, then $H_1(\mathfrak{p}) = H_2(\mathfrak{p})$.

```
\mathcal{D} by Eval-term() reads \emptyset writes \emptyset
 \mathcal{D} by Eval-app(\mathcal{D}_1, \mathcal{D}_2) reads R_1 \cup R_2 writes W_1 \perp W_2 if
                                                                                        \mathcal{D}_1 reads R_1 writes W_1
                                                                                 and \mathcal{D}_2 reads R_2 writes W_2
\mathcal D by \operatorname{Eval-bind}(\mathcal D_1,\mathcal D_2) reads R_1\cup R_2 writes W_1\perp W_2 if
                                                                                        \mathcal{D}_1 reads R_1 writes W_1
                                                                                 and \mathcal{D}_2 reads R_2 writes W_2
                        \mathcal{D} by Eval-scope(\mathcal{D}_0) reads R writes W if
                                                                                        \mathcal{D}_0 reads R writes W
     \mathcal{D} by Eval-fix(\mathcal{D}_0) reads R writes W if
                                                                    \mathcal{D}_0 reads R writes W
  \mathcal{D} by Eval-case(\mathcal{D}_0) reads R writes W if
                                                                    \mathcal{D}_0 reads R writes W
  \mathcal{D} by Eval-split(\mathcal{D}_0) reads R writes W if
                                                                   \mathcal{D}_0 reads R writes W
          \mathcal{D} by Eval-ref() reads \emptyset writes q where e = \text{ref(name } n, v) and q = t n
      \mathcal{D} by Eval-thunk() reads \emptyset writes q where e = \text{thunk(name } n, e_0) and q = t n
          \mathcal{D} by Eval-get() reads q writes \emptyset where e = \text{get}(\text{ref q}) and G(q) = v
\mathcal{D} by Eval-force() reads q, R' writes W' where e = \text{force(thunk q)}
                                                                   \mathcal{D}' reads R' writes W'
                                                         and
                                                         where \mathcal{D}' is the derivation that computed t (see text)
```

Fig. 16. Read- and write-sets of a non-incremental evaluation derivation.

Definition 1 (Reads/writes). The effect of an evaluation derived by \mathcal{D} , written \mathcal{D} reads R writes W, is defined in Figure 16.

This is a function over derivations. We write " \mathcal{D} by Rulename~(Dlist) reads R writes W" to mean that rule Rulename concludes \mathcal{D} and has subderivations Dlist. For example, Eval-scope(\mathcal{D}_0) by R reads W writes provided that \mathcal{D}_0 reads R writes W where \mathcal{D}_0 derives the only premise of scope.

6.3 Lemmas

Lemma 1 (Index-level weakening).

```
    If Γ ⊢ M : γ then Γ, Γ' ⊢ M : γ.
    If Γ ⊢ i : γ then Γ, Γ' ⊢ i : γ.
    If Γ ⊢ A : K then Γ, Γ' ⊢ A : K.
```

Proof. By induction on the given derivation.

Lemma 2 (Weakening).

```
1. If \Gamma \vdash e : A then \Gamma, \Gamma' \vdash e : A.
2. If \Gamma \vdash^M e : C then \Gamma, \Gamma' \vdash^M e : C.
```

Proof. By induction on the given derivation, using weakening on index-level entailment (for example, in the case for the value typing rule 'name') and Lemma 1 (Lemmas) (for example, in the case for the computation typing rule 'AllIndex-Elim').

Lemma 3 (Substitution).

```
1. If \Gamma \vdash \nu : A and \Gamma, x : A \vdash e : C then \Gamma \vdash ([\nu/x]e) : C.
2. If \Gamma \vdash \nu : A and \Gamma, x : A \vdash \nu' : B then \Gamma \vdash ([\nu/x]\nu') : B.
```

Proof. By mutual induction on the derivation typing e (in part 1) or v' (in part 2).

Lemma 4 (Canonical Forms). *If* $\Gamma \vdash \nu : A$, *then*

- 1. If A = 1 then v = ().
- 2. If $A = (B_1 \times B_2)$ then $v = (v_1, v_2)$.
- 3. If $A=(B_1+B_2)$ $\qquad \qquad \textit{then } \nu=\text{inj}_{\mathfrak{i}}\nu \qquad \textit{where } \mathfrak{i} \in \{1,2\}.$
- 4. If A = (Nm[X]) then v = name n where $\Gamma \vdash n \in X$.
- $\text{5.} \ \ \textit{If} \ A = (\text{Ref} \, [X] \ A_0) \qquad \qquad \textit{then} \ \nu = \text{ref} \ n \qquad \textit{where} \ \Gamma \vdash n \in X.$
- 6. If A = (Thk[X] E) then v = thunk n where $\Gamma \vdash n \in X$.
- 7. If $A = (Nm \rightarrow_{Nm} Nm)[M]$ then $v = nmtm \ M_v$ where $M \equiv (\lambda a. M')$ and $\cdot \vdash (\lambda a. M') : (Nm \rightarrow_{Nm} Nm)$ and $M_v \equiv M$.

Proof. In each part, exactly one value typing rule is applicable, so the result follows by inversion. \Box

Lemma 5 (Application and membership commute). If $\Gamma \vdash \mathfrak{n} \in \mathfrak{i}$ and $\mathfrak{p} \equiv M(\mathfrak{n})$ then $\Gamma \vdash \mathfrak{p} \in M(\mathfrak{i})$.

Proof. The set M(i) consists of all elements of i, but mapped by function M. The name p is convertible to the name M(n). Since $n \in i$, we have that p is in the M-mapping of i, which is M(i).

6.4 Main proof

Theorem 1 (Subject Reduction for Reference Semantics). If

- $-\Gamma_1 \vdash M : Nm \rightarrow_{Nm} Nm$
- $\ \mathcal{S} :: \Gamma_1 \vdash^M e : C \rhd \langle W; R \rangle$
- $\, \vdash G_1 : \Gamma_1$
- $-\mathcal{D}:: G_1 \vdash_{\mathfrak{m}}^{M} e \Downarrow G_2; \mathfrak{t}$

there exists $\Gamma_2 \supseteq \Gamma_1$ such that

- $\vdash G_2 : \Gamma_2$
- $-\Gamma_2 \vdash t : C \rhd \langle \emptyset; \emptyset \rangle$
- \mathcal{D} reads $R_{\mathcal{D}}$ writes $W_{\mathcal{D}}$
- $\langle W_{\mathcal{D}}; R_{\mathcal{D}} \rangle \preceq \langle W; R \rangle$

Proof. By induction on the typing derivation S.

$$- \ \, \mathbf{Case} \ \, \frac{\Gamma \vdash \nu : A}{\Gamma \vdash^{M} \ \mathrm{ret}(\nu) : \big((FA) \rhd \langle \emptyset; \emptyset \rangle \big)} \ \mathrm{ret}$$

$$(e=t) \text{ and } (G_1=G_2) \qquad \text{Given}$$

$$(R_{\mathcal{D}}=W_{\mathcal{D}}=R=W=\emptyset) \text{ "}$$

$$(\Gamma_2=\Gamma_1) \qquad \text{Suppose}$$

$$\vdash G_2:\Gamma_2 \qquad \text{by above equalities}$$

$$\vdash G_2 \vdash t:C \rhd \langle \emptyset,\emptyset \rangle \qquad \text{"}$$

$$\mathcal{D} \text{ reads } R_{\mathcal{D}} \text{ writes } W_{\mathcal{D}} \qquad \text{By the corresponding rule in Def. 1}$$

$$\langle W_{\mathcal{D}};R_{\mathcal{D}}\rangle \preceq \langle W;R \rangle \qquad \text{All are empty}$$

$$- \ \, \mathbf{Case} \ \, \frac{\Gamma \vdash \nu : \mathsf{Ref} \, [X] \, \, A}{\Gamma \vdash^{M} \, \mathsf{get} (\nu) : \big(\mathsf{F} \, A \big) \rhd \langle \emptyset; X \rangle} \, \, \mathrm{get}$$

$$- \ \, \mathbf{Case} \ \, \frac{\Gamma \vdash \nu : \mathsf{Thk} \, [X] \, \left(C \rhd \varepsilon \right)}{\Gamma \vdash^{M} \, \mathsf{force}(\nu) : \left(C \rhd \left(\langle \emptyset; X \rangle \, \mathsf{then} \, \varepsilon \right) \right)} \, \, \mathsf{force}$$

```
(W = \emptyset) and (R = X)
                                                                                   Given
                    \Gamma_1 \vdash \nu : \mathsf{Thk}[X] \mathsf{E}
                                                                                   Given
                    \exists \mathfrak{p}.\ (\nu = \mathtt{thunk}\ \mathfrak{p})
                                                                                   Lemma 4 (Lemmas)
                    \Gamma_1 \vdash \mathfrak{p} \in X
                   \Gamma_1 \vdash p has-store-type E
                                                                                   By inversion of value typing rule
                                                                                   Inversion of \vdash G_1 : \Gamma_1 with p has-store-type E
                  \exists e_{p}. G_{1}(p) = e_{p}
      S_0 :: \Gamma_1 \vdash e_p : C \rhd \epsilon
     \mathcal{D}_0 :: \ G_1 \vdash^M_{\mathfrak{m}} e_p \Downarrow G_2; t
                                                                                  Inversion of \mathcal{D}
                        \vdash G_2 : \Gamma_2
                                                                                  By IH on \mathcal{S}_0 and \mathcal{D}_0
                    \Gamma_2 \vdash t : C \rhd \langle \emptyset, \emptyset \rangle
                             \mathcal{D}_0 reads R_{\mathcal{D}} writes W_{\mathcal{D}}
                             \langle W_{\mathcal{D}_0}; R_{\mathcal{D}_0} \rangle \leq \langle W; R \rangle
                             \mathcal{D} reads R_{\mathcal{D}_0} writes W_{\mathcal{D}_0} By the corresponding rule in Def. 1
                                                                                  by above equality W_{\mathcal{D}} = W = \emptyset,
                             \langle W_{\mathcal{D}_0}; R_{\mathcal{D}_0} \rangle \leq \langle W; R \rangle
                                                                                   ... and inequality for (R_D = \{p\}) \subseteq (X = R).
- \  \, \mathbf{Case} \ \frac{\Gamma_1 \vdash M' : \mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm} }{\Gamma_1 \vdash^{M} \mathsf{scope}(M', e_0) : C \rhd \langle W; R \rangle} \ \mathrm{scope}
       S_0 :: \Gamma \vdash^{M \circ M'} e_0 : C \rhd \langle W; R \rangle
                                                                                                        Subderivation 2 of S
                                         G_1 \vdash_m^M \text{scope}(M', e_0) \Downarrow G_2; t \text{ Given } G_1 \vdash_m^{M \circ M'} e_0 \Downarrow G_2; t \text{ By inv}
        \mathcal{D} ::
       \mathcal{D}_0 ::
                                                                                                         By inversion (scope)
                               \Gamma_1 \vdash M : \mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}
                                                                                                         Assumption
                               \Gamma_1 \vdash M' : \mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}
                                                                                                        Subderivation 1 of S
                \Gamma_1, x : Nm \vdash M'x : Nm
                                                                                                         By rule t-app
                \Gamma_1, x : \mathsf{Nm} \vdash M(M'x) : \mathsf{Nm}
                                                                                                        By rule t-app
                               \Gamma_1 \vdash \lambda x. M (M'x) : Nm \rightarrow_{Nm} Nm
                                                                                                        By rule t-abs
                               \Gamma_1 \vdash M \circ M' : Nm \rightarrow_{Nm} Nm
                                                                                                         By definition of M \circ M'
                                     \vdash G_2 : \Gamma_2
                                                                                                         By IH on \mathcal{S}_0
     13
                               \Gamma_2 \vdash t : C \rhd \langle \emptyset; \emptyset \rangle
                                                                                                         "
                                          \mathcal{D}_0 reads R_{\mathcal{D}_0} writes W_{\mathcal{D}_0}
                                          \langle W_{\mathcal{D}_0}; R_{\mathcal{D}_0} \rangle \leq \langle W; R \rangle
                                          \mathcal D reads R_{\mathcal D} writes W_{\mathcal D}
                                                                                                         By the corresponding rule in Def. 1
                                          \langle W_{\mathcal{D}}; R_{\mathcal{D}} \rangle = \langle W_{\mathcal{D}_0}; R_{\mathcal{D}_0} \rangle
                                          \langle W_{\mathcal{D}}; R_{\mathcal{D}} \rangle \prec \langle W; R \rangle
                                                                                                        By above equalities
      B
```

```
- Case
               \frac{\Gamma_1 \vdash \nu : \mathsf{Nm}\left[X\right] \qquad \Gamma_1 \vdash e : \mathsf{E}}{\Gamma_1 \vdash^{M} \mathsf{thunk}(\nu_1, \nu_2) : \mathsf{F}\left(\mathsf{Thk}\left[M(X)\right] \mathsf{E}\right) \rhd \langle M(X); \emptyset \rangle}
                  C = F(Thk[M(X)] E) and R = \emptyset and W = M(X) Given from S
                  \Gamma_1 \vdash \nu : Nm[X]
                                                                                                                Subderivation
                  (v = name n) and (n \in X)
                                                                                                                Lemma 4 (Lemmas)
                  M n \Downarrow p \text{ and } R_{\mathcal{D}} = \emptyset \text{ and } W_{\mathcal{D}} = \{p\}
                                                                                                                 Given from \mathcal{D}
                  G_2 = (G_1, p : e)
                  \Gamma_2 = (\Gamma_1, p : \mathsf{Thk}[p] E)
                                                                                                                Suppose
                 \vdash G_2 : \Gamma_2
                                                                                                                By application of store typing rule (Fig. 15)
     137
                  \Gamma_2 \vdash p has-store-type E
                                                                                                                By inversion of value typing rule
                  \Gamma_2 \vdash \text{ref } p : \text{Ref } [p] E
                                                                                                                By rule thunk
                  \Gamma_2 \vdash^M \operatorname{ret}(\operatorname{thunk} p) : \operatorname{ret}(\operatorname{Thk}[p] A) \rhd \langle \emptyset; \emptyset \rangle
                                                                                                                By rule ret
                  \mathcal{D} reads R_{\mathcal{D}} writes W_{\mathcal{D}} and W_{\mathcal{D}} = \{p\}
                                                                                                                By the corresponding rule in Def. 1
                  n \in X
                                                                                                                 Above
                  M(n) \in M(X)
                                                                                                                Name term application is pointwise
                  M(n) \in W
                                                                                                                By above equality
                  M(n) = p
                  {p}\subseteq W
                                                                                                                By set theory
                  \langle W_{\mathcal{D}}; R_{\mathcal{D}} \rangle \preceq \langle W; R \rangle
- Case
               \frac{\Gamma_1 \vdash \nu_1 : \mathsf{Nm}\hspace{0.5mm}[X] \qquad \Gamma_1 \vdash \nu_2 : A}{\Gamma_1 \vdash^M \mathsf{ref}(\nu_1, \nu_2) : \mathsf{F}\left(\mathsf{Ref}\hspace{0.5mm}[M(X)]\hspace{0.5mm}A\right) \rhd \langle M(X);\emptyset\rangle} \ \mathrm{ref}
                  C = F(Ref[M(X)] A) and R = \emptyset and W = M(X) Given from S
                  \Gamma_1 \vdash \nu_1 : \operatorname{Nm}[X]
                                                                                                                Subderivation
                  (v_1 = \mathtt{name}\; \mathfrak{n}) \; \mathrm{and} \; (\mathfrak{n} \in X)
                                                                                                                Lemma 4 (Lemmas)
                  M \mathfrak{n} \Downarrow \mathfrak{p} \text{ and } R_{\mathcal{D}} = \emptyset \text{ and } W_{\mathcal{D}} = \{\mathfrak{p}\}
                                                                                                                Given from \mathcal{D}
                  G_2 = (G_1, p : v_2)
                  \Gamma_2 = (\Gamma_1, \mathfrak{p} : \mathsf{Ref} [\mathfrak{p}] A)
                                                                                                                Suppose
                 \vdash G_2 : \Gamma_2
                                                                                                                By application of store typing rule (Fig. 15)
                  \Gamma_2 \vdash p has-store-type A
                                                                                                                By inversion of value typing rule
                  \Gamma_2 \vdash \text{ref } p : \text{Ref } [p] A
                                                                                                                By rule ref
                  \Gamma_2 \vdash^M \operatorname{ret}(\operatorname{ref} \mathfrak{p}) : \operatorname{ret}(\operatorname{Ref} [\mathfrak{p}] A) \rhd \langle \emptyset; \emptyset \rangle
                                                                                                                By rule ret
                  \mathcal{D} reads R_{\mathcal{D}} writes W_{\mathcal{D}} and W_{\mathcal{D}} = \{\mathfrak{p}\}
                                                                                                                By the corresponding rule in Def. 1
                  \mathfrak{n} \in X
                                                                                                                Above
                  M(\mathfrak{n})\in M(X)
                                                                                                                Name term application is pointwise
                  M(n) \in W
                                                                                                                By above equality
                  M(n) = p
                  {\mathfrak{p}}\subseteq W
                                                                                                                By set theory
                  \langle W_{\mathcal{D}}; R_{\mathcal{D}} \rangle \leq \langle W; R \rangle
     E C
```

$$- \begin{array}{c} - \text{Case} & \frac{ \Gamma_1 \vdash^M e_1 : (\text{FA} \rhd e_1) \quad \Gamma_1, x : \text{A} \vdash^M e_2 : (\text{C} \rhd e_2) }{ \Gamma_1 \vdash^M \text{let}(e_1, x.e_2) : \text{C} \rhd (e_1 \text{then } e_2) } \text{ let} \\ & \Gamma_1 \vdash^M \text{let}(e_1, x.e_2) : \text{C} \rhd (e_1 \text{then } e_2) \\ & \vdash G_1 : \Gamma_1 & \text{Given} \\ & \mathcal{S}_1 :: \Gamma_1 \vdash^M e_1 : \text{FA} \rhd e_1 & \text{Subderivation } 1 \text{ of } \mathcal{S} \\ & \mathcal{D}_1 :: G_1 \vdash^M e_1 \Downarrow G_{12}; t_1 & \text{Subderivation } 1 \text{ of } \mathcal{D} \\ & \text{exists } \Gamma_{12} \supseteq \Gamma_1 \text{ such that } G_{12} : \Gamma_{12} \text{ By IH on } \mathcal{S}_1 \\ & \Gamma_{12} \vdash t_1 : \text{FA} \rhd (\emptyset; \emptyset) & \text{"} \\ & \mathcal{U}_{D_1}; R_{D_1} \rangle \preceq e_1 & \text{"} \\ & \langle W_{D_1}; R_{D_1} \rangle \preceq \langle W_1, R_1 \rangle & \text{"} \\ & \langle W_{D_1}; R_{D_1} \rangle \preceq \langle W_1, R_1 \rangle & \text{"} \\ & \Gamma_{12} \vdash v : \text{A} & \text{inversion of typing rule ret, for terminal computation } t_1 \\ & \mathcal{S}_2 :: \Gamma_1, x : \text{A} \vdash^M e_2 : \text{C} \rhd e_2 & \text{Subderivation } 2 \text{ of } \mathcal{S} \\ & \Gamma_{12}, x : \text{A} \vdash^M e_2 : \text{C} \rhd e_2 & \text{Lemma } 2 \text{ (Lemmas)} \\ & \Gamma_{12} \vdash^M [v/x] e_2 : \text{C} \rhd e_2 & \text{Lemma } 3 \text{ (Lemmas)} \\ & \Gamma_{12} \vdash^M [v/x] e_2 : \text{C} \rhd e_2 & \text{Lemma } 3 \text{ (Lemmas)} \\ & \Gamma_{12} \vdash^M [v/x] e_2 : \text{C} \rhd e_2 & \text{Lemma } 3 \text{ (Lemmas)} \\ & \Gamma_{12} \vdash^M \text{terminal computation } t_1 \\ & \text{is} \vdash \text{G}_2 : \Gamma_2 \\ & \text{exists } \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{exists } \Gamma_2 \supseteq \Gamma_1 \geq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_1 \geq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_1 \geq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_1 \supseteq \mathcal{S}_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_1 \supseteq \mathcal{S}_1 \text{ such that} & \text{By IH on } \mathcal{S}_2 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_1 \text{ such that} & \text{Exists } \Gamma_2 \supseteq \Gamma_1 \\ & \text{Exists } \Gamma_2 \supseteq \Gamma_$$

Similar to the case for let, using Lemma 4 (Lemmas).

- Case
$$\begin{array}{c} \Gamma, x_1 : A_1 \vdash^M e_1 : E \\ \frac{\Gamma \vdash^M \nu : (A_1 + A_2)}{\Gamma \vdash^M \mathsf{case}(\nu, x_1.e_1, x_2.e_2) : E} \end{array} \text{case}$$

Similar to the case for let, using Lemma 4 (Lemmas).

```
- Case
                    \Gamma_1 \vdash \nu_M : (\mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}) \, [M]
                    \Gamma_1 \vdash \nu : \mathsf{Nm}[i]
               \frac{\Gamma_1 \vdash (\nu_M \; \nu) : F\left(\mathsf{Nm}\left[\mathsf{M}(\mathfrak{i})\right]\right) \rhd \langle \emptyset;\emptyset \rangle}{\mathsf{name-app}}
              \Gamma_1 \vdash \nu_M : (\mathsf{Nm} \to_{\mathsf{Nm}} \mathsf{Nm}) \, [M]
                                                                                            Given
                       v_M = nmtm M_v
                                                                                            Lemma 4 (Lemmas)
                       M \equiv (\lambda a. M')
                                                                                             "
                       \cdot \vdash \lambda \alpha. M' : (Nm \rightarrow_{Nm} Nm)
                                                                                             "
                       M_{\nu} \equiv M
              \Gamma_1 \vdash \nu : Nm[i]
                                                                                            Given
                       \nu = \mathtt{name} \; n
                                                                                            Lemma 4 (Lemmas)
                       \Gamma \vdash \mathfrak{n} \in \mathfrak{i}
                       M \Downarrow_M (\lambda a. M')
                                                                                            By inversion on \mathcal{D} (name-app)
                       [n/a]M' \downarrow_M p
               p \equiv [n/a]M'
                                                                                            By a property of \downarrow_M
                   \equiv (\lambda a. M')(n)
                                                                                            By a property of \equiv
                   \equiv M(n)
                                                                                            By a property of \equiv
                       (\Gamma_2 = \Gamma_1), (G_2 = G_1)
                                                                                            Suppose
                       \vdash G_2 : \Gamma_2
                                                                                            By above equalities and G_1 \vdash \Gamma_1
              \Gamma_1 \vdash n \in \mathfrak{i}
                                                                                            Above
               p \equiv M(n)
                                                                                            Above
              \Gamma_1 \vdash p \in M(i)
                                                                                            Lemma 5 (Lemmas)
              \Gamma_1 \vdash \mathtt{name} \ p : \mathsf{Nm} \ [M(\mathfrak{i})])
                                                                                            By rule name
              \Gamma_1 \vdash \mathtt{ret}(\mathtt{name}\ p) : \mathsf{F}(\mathsf{Nm}[M(\mathfrak{i})]) \rhd \langle \emptyset; \emptyset \rangle \ \mathrm{By} \ \mathrm{rule} \ \mathsf{ret}
                       \mathcal{D} by Eval-name-app reads \emptyset writes \emptyset By the corresponding rule in Def. 1
                       (R_{\mathcal{D}} = R = \emptyset), (W_{\mathcal{D}} = W = \emptyset)
                                                                                            By above equalities
     13
```

$$- \ \, \mathbf{Case} \ \, \frac{\Gamma_1,\alpha:\gamma\vdash^M\,t:E}{\Gamma_1\vdash^M\,t:(\forall\alpha:\gamma.\,E)} \ \, \mathrm{AllIndexIntro}$$

$$\begin{array}{c} \mathcal{S}_0 :: \Gamma_1, \alpha : \gamma \vdash^M t : E \\ \mathcal{D}_0 :: G_1 \vdash^M_m e \Downarrow G_2 ; t \\ \exists \Gamma_2 \subseteq \Gamma_1 \\ \mathcal{D}_0 \text{ reads } R_{\mathcal{D}_0} \text{ writes } W_{\mathcal{D}_0} \\ & \Gamma_2 \vdash t : E \\ \forall R_{\mathcal{D}_0} ; W_{\mathcal{D}_0} \rangle \preceq \langle R ; W \rangle \\ & \Gamma_2 \vdash^M t : (\forall \alpha : \gamma \cdot E) \\ \exists \mathcal{F}_0 :: \mathcal{F}$$

7 Related Work

Similar to the AllIndexElim case.

DML (Xi and Pfenning 1999; Xi 2007) is an influential system of limited dependent types or *indexed* types. Inspired by Freeman and Pfenning (1991), who

created a system in which datasort refinements were clearly separated from ordinary types, DML separates the "weak" index level of typing from ordinary typing; the dynamic semantics ignores the index level.

Motivated in part by the perceived burden of type annotations in DML, liquid types (Rondon et al. 2008; Vazou et al. 2013) deploy machinery to infer more types. These systems also provide more flexibility: types are not indexed by fixed tuples of indices.

To our knowledge, Gifford and Lucassen (1986) were the first to express effects within (or alongside) types. Since then, a variety of systems with this power have been developed. A full accounting of these systems is beyond the scope of this paper; for an overview of some of them, see Henglein et al. (2005). We briefly discuss a type system for regions (Tofte and Talpin 1997), in which allocation is central. Regions organize subsets of data, so that they can be deallocated together. The type system tracks each block's region, which in turn requires effects on types: for example, a function whose effect is to return a block within a given region. Our type system shares region typing's emphasis on allocation, but we differ in how we treat the names of allocated objects. First, names in our system are fine-grained: each allocated object is uniquely named. Second, names have structure and are related by that structure: the names root.1.1 and root.1.2 are not arbitrary distinct names (because they share a prefix).

Techniques for general-purpose incremental computation. Incremental algorithms, variously called online algorithms and dynamic algorithms, are expressly designed to be run repeatedly on changing inputs. Typically, the designers of incremental algorithms study each problem in isolation, and exploit domain-specific structure to craft a specific solution: search algorithms in robotics are incremental reworkings of standard search algorithms, and motion simulation algorithms can be viewed as incremental versions of standard computational geometry algorithms (Agarwal et al. 2002; Basch 1999; Alexandron et al. 2005).

Instead of crafting a novel algorithm for each problem, programming languages researche has developed general-purpose programming language abstractions, so that the language (not the program) abstracts over the incremental aspect of the desired program behavior Ramalingam and Reps (1993, 1996); Demers et al. (1981); Reps and Teitelbaum (1988); Liu and Teitelbaum (1995); Liu et al. (1998); Acar (2009); Guo and Engler (2011). These incremental languages are aware of incremental change, and through specially-designed abstractions, they help the programmer avoid thinking about changes directly. Instead, she thinks about how to apply the abstractions, which are general purpose.

Through careful language design, modern incremental abstractions elevate implementation questions, such as "how does this particular change pattern affect a particular incremental state of the system?" into simpler, more general questions, answered with special programming abstractions (e.g., via special annotations). These abstractions identify changing data and reusable subcomputations (Acar et al. 2008a; Hammer et al. 2015c, 2014, 2009; Mitschke et al. 2014), and they allow the programmer to relate the expression of incremental algorithm with the ordinary version of the algorithm that operates over

fixed, unchanging input: The gap between these two programs is witnessed by the special abstractions offered by the incremental language. Through careful algorithm and run-time system design, these abstractions admit a fast *change propagation implementation*. In particular, after an initial run of the program, as the input changes dynamically, change propagation provides a general (provably sound) approach for recomputing the affected output (Acar et al. 2006; Acar and Ley-Wild 2009; Hammer et al. 2015c). Further, IC can deliver *asymptotic* speedups (Acar et al. 2007; Hammer et al. 2007; Acar et al. 2008c,b,a; Acar 2009; Sümer et al. 2011), and has even addressed open problems (Acar et al. 2010). These IC abstractions exist in many languages (Shankar and Bodik 2007; Hammer et al. 2007; Hammer et al. 2009; Chen et al. 2014).

Functional reactive programming. Incremental computation and reactive programming (especially functional reactive programming or FRP) share common elements: both attempt to respond to outside changes and their implementations often both employ dependence graphs to model dependencies in a program that change over time (Cooper and Krishnamurthi 2006; Krishnaswami and Benton 2011; Czaplicki and Chong 2013). In a sketch of future work below, we hope to marry the feedback that is unique to FRP with the incremental data structures and algorithms that are unique to IC.

8 Conclusion and Future Work

In this report, we motivate the need for generic naming strategies in programs that use nominal memoization. We define a refinement type system that gives practical static approximations of these strategies. We prove that our type system enforces that well-typed programs that use nominal memoization always correspond with a purely functional program. Meanwhile, prior work shows that these programs can dramatically outperform non-incremental programs as well as those using traditional memoization.

Future work: Meta-level programs. The entire point of incremental computation (IC) is to update input with changes, and then propagate these changes (efficiently) into a changed output. Hence, imperative updates are fundemental to IC. To address this fact, future work should follow the direction of Hammer et al. (2014) and give explicit type-based annotations that permit such imperative behavior in explicit locations. We discuss feedback in further detail, below.

Future work: Functional reactive programming with explicit names. The effect patterns of feedback and churn, described as problems in Sec. 2, may also be viewed as desirable patterns, especially in contexts such as functional reactive programming (FRP). By definition, feedback occurs when a proram overwrites prior allocations with new data, and thus these overwrite effects violate a purely functional allocation strategy. However, we can view this allocation strategy

as corresponding to an operational view of FRP with an explicit store where controlled (annotated) feedback may occur safely.

The type system presented here suggests that we can go further than modeling FRP in isolation, and (potentially) marry the controlled feedback of FRP with nominal memoization, and thus, with general-purpose incremental computation. In particular, future work may explore explicit programming annotations for marking intended places and names where feedback occurs:

$$\frac{\Gamma \vdash e : \mathsf{LoopComp} \ A \rhd \langle \mathsf{R} \bot \mathsf{F}; W \cup \mathsf{F}; \mathsf{F} \rangle}{\Gamma \vdash \mathsf{loop} \ e \ \mathsf{over} \ \mathsf{F} : \mathsf{F} \ A \rhd \langle \mathsf{R} \bot \mathsf{F}; W \cup \mathsf{F}; \emptyset \rangle} \ \mathrm{feedback\text{-}loop}$$

The statics of this rule says that sub-expression e has type LoopComp $A \triangleright e$, which is a computation that produces the following (recursive) sum type:

$$(\mathsf{LoopComp}\ A \rhd \varepsilon) = (\mathsf{F}(A + \mathsf{Thk}[X]\ (\mathsf{LoopComp}_X\ A) \rhd \varepsilon) \rhd \varepsilon)$$

Statically, the expression e will either produce a value of type A, or a thunk that produces more thunks (and perhaps possibly a value) in the future.

The dynamics of this rule would re-run expression e until it produces its value (if ever) and in so doing, the write effects of expression e would be free to overwrite the reads of e, creating a feedback loop. In the rule above, the annotation \cdots over F explicates the over-written set of names F. To statically distinguish delayed, feedback writes from ordinary (immediate) writes, we may track three (not two) name sets in each effect e: the read set, the write set, and the set of feedback writes, here F. Operationally, the dynamic semantics treats feedback writes specially, by delaying them until the LoopComp fully completes an iteration. This proposal corresponds closely with prior work on synchronous, discrete-time FRP (Krishnaswami and Benton 2011).

Future work: Bidirectional type system. Drawing closer to an implementation, we intend to derive a bidirectional version of the type system, and prove that it corresponds to our declarative type system. A key challenge in implementing the bidirectional system would be to handle constraints over names and name sets.

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