# **IMPERIAL**

# $\begin{array}{c} {\bf Adding~and~Optimising}\\ {\bf Resource~Algebras~for~a~Parametric~CSE} \end{array}$

Background & Progress Report

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### Chapter 1

## Background

In this chapter we introduce some of the core ideas used throughout this document, while mentioning pre-existing work in the field. In §1.1 we introduce Separation Logic, its variations, and the idea of Separation Algebras; in §1.2 we introduce Gillian, a CSE engine; in §1.3 we mention other similar verification tools; finally, in §1.4 we evoke the legal, social, ethical and professional impacts of this work.

#### 1.1 Separation Logic

As the need to formally verify programs grew, methods needed to be built to provide a framework to do so. Hoare Logic [30] allows proving properties of programs, by describing them axiomatically. Every statement can be expressed as having a precondition – the state before execution – and a postcondition – the state after successful execution – expressed as  $\{P\} S \{Q\}$ . For assignment for instance, one could have  $\{x = x\} x := 0 \{x = 0\}$ .

While already helpful to reason about programs, an issue remains: shared mutable state. How does one describe that there is a list in memory of unknown length, that may be mutated at multiple places in the program, while ensuring properties hold at a specific point in time? Being able to describe the state and constraints upheld by global state is difficult, and past solutions scaled poorly.

Separation Logic (SL) [50, 56] is an extension of Hoare logic that permits this in a clear and scalable way. Its main addition is the separating conjunction \*: P \* Q means not only that the heap satisfies P and Q, but that it can be split into two *disjoint* parts, such that one satisfies P and the other Q. This allows us to reason compositionally about the state, by not only stating what properties are upheld, but how they may be split for further proofs. For instance, given a *list* predicate, when calling a function that mutates the list one can simply substitute the part of the state corresponding to that list with the postcondition

of the function, with the guarantee that the rest of the state is untouched, by using the frame rule:

Frame 
$$\frac{\{P\}\,S\,\{Q\}}{\{P*F\}\,S\,\{Q*F\}}$$

Because we know that S only uses (either by reading or modifying) P, any disjoint frame F can be added to the state without altering the execution of S. This is very powerful: one can prove properties of smaller parts of code (like a function, or a loop), and these properties will be able to be carried to a different context that may have a more complex state. For instance, this can be used in a Continuous Integration (CI) setting, by only analysing functions of which code is modified, while reusing past analyses of unchanged code, allowing for incremental analysis.

SL also comes equipped with the *emp* predicate, representing an empty state (i.e. P \* emp = P), and the separating implication -\*.  $P_0 -* P_1$  states that if the current state is extended with a disjoint part satisfying  $P_0$ , then  $P_1$  holds in the extended heap [56].

Further predicates can then be defined; for instance the "points to" predicate,  $a \mapsto x$ , stating that the address a stores values x. For instance,  $a \mapsto x * a \mapsto x$  does not hold, since we can't have two disjoint states both containing the address a. Similarly,  $a \mapsto x * b \mapsto y$  only holds if  $a \neq b$ .

#### 1.1.1 Incorrectness Separation Logic

Separation Logic is, by definition, over-approximate (OX):  $\{P\}$  S  $\{Q\}$  means that given a precondition P, we are guaranteed to reach a state that satisfies Q. In other words, Q may encompass more than only the states reachable from P. This can become an issue when used for real-world code, where errors that can't occur are detected and flag the code as incorrect, hindering the use of a verification tool (for instance in a CI setting).

To solve this problem, a recent innovation in the field is Incorrectness Separation Logic (ISL) [55], derived from Incorrectness Logic [51]. It is the *under-approximate* (UX) equivalent to SL, instead ensuring that the detected errors actually exist, at the cost of possibly missing some executions of the code.

Where SL uses triplets of the form  $\{\text{precondition}\}\ S$   $\{\text{postcondition}\}\$ , incorrectness separation logic uses  $[\text{presumption}]\ S$   $[\text{result}]\$ , with the result being an under-approximation of the actual result of the code. The reasoning is thus flipped, where we start from a stronger assertion at the end of the function, and then step back and broaden our assumptions, until reaching the initial presumption. This means that all paths explored this way are

guaranteed to exist, but may not encompass all possible paths. This enables us to do true bua-finding.

Another way of comparing SL to ISL is with consequence: in SL, the precondition implies the postcondition, whereas with ISL the result implies the presumption. Furthermore, ISL triplets are extended with an  $outcome\ \epsilon$ , resulting in  $[P]S[\epsilon:Q]$ , where  $\epsilon$  is the outcome of the function, for instance Ok, or an error.

While SL is over-approximate and ISL is under-approximate, we may call exact (EX) specifications that are both OX-sound and UX-sound [46]. Such triplets are then written  $\langle\!\langle P \rangle\!\rangle S \langle\!\langle o:Q \rangle\!\rangle$ , with o the outcome, and P and Q the pre and postcondition respectively.

#### 1.1.2 Separation Algebras

While the above examples of separation logic use a simple abstract heap mapping locations to values, we often need more complex state representations to verify real-world programming languages. For instance, the model used by the C language (in particular CompCert C [40], a verified C compiler) uses the notion of memory blocks and offsets: memory isn't just an assortment of different atomic cells. Furthermore, the basic "points to" predicate, while useful, has limited applications; for contexts such as concurrency, one needs to have a more precise level of sharing that goes beyond the exclusive ownership of "points to".

An example of such extension is fractional permissions [9, 10]: the "points to" predicate is extended with a permission, a fraction q in the (0; 1] range, written  $a \mapsto x$ . A permission of 1 gives read and write permission, while anything lower only gives read permission. This allows one to split permissions, for instance  $a \mapsto x$  is equivalent to  $a \mapsto x * a \mapsto x$ 

Further changes and improvements to separation logic have been made, usually with the aim of adapting a particular language feature or mechanism that couldn't be expressed otherwise. "The Next 700 Separation Logics" [52] argues that indeed too many subtly different separation logics are being created to accommodate specific challenges, which in turn require new soundness proofs that aren't compatible with each other [17]. To tackle this, research has emerged around the idea of providing one sound metatheory, that would provide the tools necessary to building more complex abstractions within that logic, rather than in parallel to it.

A first step in this quest towards a single core logic is the definition of an abstraction over the state. The main concept introduced for this is that of *Separation Algebras* [16, 22], which allow for the *construction* of complex separation algebras from simpler elements.

Because soundness is proven for these smaller elements, constructions made using them also carry this soundness, and alleviate users of complicated proofs.

Different authors used different definitions and axioms to define separation algebras. In [16] they are initially defined as a cancellative Partially Commutative Monoid (PCM)  $(\Sigma,\cdot,0)$ . This definition is however too strong: for instance, the cancellativity property implies  $\sigma\cdot\sigma'=\sigma\Rightarrow\sigma'=0$ . While this is true for some cases such as the points to predicate, this would invalidate useful constructions such as that of an agreement, where knowledge can be duplicated. For an agreement separation algebra, we have  $\sigma\cdot\sigma=\sigma$ , which dissatisfies cancellativity.

This approach is also taken in [22], where separation algebras are defined with a functional ternary relation J(x,y,z) written  $x \oplus y = z$  – the choice of using a relation rather than a partial function being due to the authors wanting to construct their models in Rocq (formerly Coq) [18], which only supports computable total functions. While the distinction is mostly syntactical and both relational and functional approaches are equivalent as long at the relation is functional, the former makes proof-work less practical when working with partial functions [54], whereas the functional approach allows one to reason equationally [34]. This paper also introduces the notion of multi-unit separation algebras, separation algebras that don't have a single unit 0, but rather enforce that every state has a unit, that can be different from other states' units. Formally, this means that instead of having  $\exists u. \forall x. x \oplus u = x$ , we have  $\forall x. \exists u_x. x \oplus u_x = x$  – this allows one to properly define the disjoint union (or sum) of separation algebras, with both sides of the sum having a distinct unit. We may note that this is a first distinction from PCMs, as a partially commutative monoid cannot have two distinct units; according to the above definition, separation algebras are rather a type of partially commutative semigroup.

In [54], further improvements are done to the axiomatisation of separation algebras. Most notably, the property of cancellativity is removed, as it is too strong and unpractical for some cases, as shown previously. Furthermore, the idea of unit, or *core*, is made explicit, with the definition of the total function  $\hat{-}$ , the core of a resource, which is its *duplicable part*.

Finally, Iris [33–35, 37] is a state of the art "higher-order concurrent separation logic". It places itself as a solution to the problem mentioned in [52], and aims to provide a sound base logic that can be reused and extended to fit all needs. It defines resource algebras (RAs), objects that are similar to separation algebras. Because Iris supports "higher-order ghost state", more sophisticated mechanisms are needed to model state, and as such the composition function  $\cdot$  needs to be total. To rule out invalid compositions, a validity function  $\overline{\mathcal{V}}$  is instead used, which returns whether a given state is valid. The core function, written |-|, is made partial, rather than total – this, they argue, makes constructing

state models from smaller components easier. This was confirmed when developing state models for Gillian, where having the core be a total function made some state models more complicated to make sound, as will be shown later.

#### 1.2 Program Verification with Gillian

While having a logic to prove programs is a great step towards verifying codebases, doing it manually is tedious and time-consuming. Tools have been developed to automate this, such as *Gillian* [26, 45, 57]. It is a *compositional symbolic execution engine*, with the added property of being *parametric on the memory model*. It allows reasoning about *correctness and incorrectness*. We may now go over exactly what this means.

#### 1.2.1 Symbolic Execution

Traditionally, software is tested by calling the code with a predetermined or randomly generated input (for instance with fuzzing), and verifying that the output is as expected. These approaches, where the code is executed directly, are in the realm of *concrete* execution, as the code is run with concrete (i.e. real, existing) values. While this method is straightforward to execute, it comes with flaws: it is limited by the imagination of the person responsible for writing the tests when written manually, or by the probability of a given input to reach a specific part of the codebase when generated. With fuzzing, methods exist to improve the odds of finding new paths [5], but it all amounts to luck nevertheless.

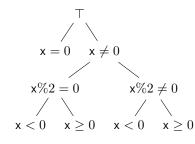
A solution to this is symbolic execution: rather than running the code with concrete values, *symbolic* values are used [4]. These values are abstract, and are restricted as an interpreter steps through the code and conditionals are encountered. Once a branch of the code terminates, a constraint solver can be used against the accumulated constraints to obtain a possible concrete value, called an *interpretation*. This method appeared in the 70s, notably with the Select [12] and EFFIGY [36] systems.

For a simple C program that checks if a given number is positive or negative and even or odd (see Figure 1.1), symbolic execution would thus branch thrice, once at each condition. This would then result in 5 different branches, each with different constraints on the program variable x: x = 0,  $x\%2 = 0 \land x < 0$ ,  $x\%2 = 0 \land x > 0$ ,  $x\%2 \neq 0 \land x < 0$  and  $x\%2 \neq 0 \land x > 0$ .

<sup>&</sup>lt;sup>1</sup>For simplicity, we omit the case where x is not an integer, which would lead to an error.

<sup>&</sup>lt;sup>2</sup>The four last constraints also contain x = 0, which is included in x > 0 and x < 0

```
if (x == 0) {
2
      print("zero");
 3
 4
    if (x % 2 == 0) {
5
6
      print("even");
7
    } else {
8
      print("odd");
9
    if (x < 0) {
10
11
      print("negative");
12
    } else {
13
      print("positive");
```



(b) Paths obtained from symbolically executing the code, with the added constraint at each node

(a) Simple branching program

Figure 1.1: Symbolic execution of a simple program

#### 1.2.2 Compositional Symbolic Execution

An issue that arises from symbolic execution is that is scales quite poorly, because of path explosion [1, 4], as each branching can potentially multiply by two the total number of branches. Furthermore, it lacks compositionality, as the code must be tested in its entirety, from main to return.

A solution to this is *compositional* symbolic execution (CSE), in which the code is tested by fragments, function by function. This allows for gradual adoption, as only parts of the code base can be tested, all while minimising the effect of path explosion.

A great tool to enable this is separation logic: one can specify the precondition and postcondition of a function, and the symbolic execution can then step through the code, by starting from a state satisfying the precondition and ensuring that once the branch terminates the postcondition is satisfied. Because of the frame rule, function specifications can then be re-used when calling functions, by symbolically executing the function specification on the corresponding state fragment. If the called function does not have a specification that can be used, it can instead be inlined, executing the code within the function, as would standard symbolic execution.

This approach was pioneered in Smallfoot [6], and was followed by later tools, such as JaVerT [28, 29], Infer [14], VeriFast [31], SpaceInvaders [59] and others – their individual contributions will be discussed in more detail in the next section.

#### 1.2.3 Parametricity

An innovation of Gillian compared to other existing tools is the fact is is parametric on the memory model (also called *state model*). Typical CSE engines usually come with a built-in notion of memory, ranging from simple heaps with memory chunks [31] to more complex state models tailored to a language, like in JaVerT [28, 29] to represent JavaScript objects.

While a fixed state model is useful for specific languages or feature sets, this can become a limitation when porting the engine to other languages. The goal behind Gillian's parametricity is to allow it to be used for virtually any language, as long as the language's state model can be encoded into a separation logic equivalent; improvements done to the engine can then be carried to all of its instantiations for free. This enables it to verify programs written in JavaScript, C or Rust, despite their significant differences in terms of state representation and level of abstraction.

To define a memory model to use in Gillian, a developer must implement it in OCaml [41]. A memory model is a partially commutative monoid, comprised of a set of states, which can be modified via *actions* and be represented in specifications via *core predicates*.

Actions serve as the interface for programs to interact with the state, for instance by storing or loading values into the state, or allocating a new block of memory and then freeing it. The language Gillian targets, GIL (Gillian's Intermediate Language), provides the base semantics of an imperative goto-language (variables, errors, functions...) which do not affect the state – actions are the only way of manipulating it.

Core predicate are predicates which have in and out parameters, and represent fragments of the state. They are the encoding of the state into separation logic, and can be used within function specifications. For example, the separation logic  $a \mapsto x$  predicate could be represented as a core predicate as  $\langle points_to \rangle(a; x)$ .

#### 1.2.4 Correctness, Incorrectness

Another key difference between Gillian and other CSE engines is that it is capable of reasoning about programs both with correctness, in SL, and incorrectness, in ISL, thus enabling OX program verification and UX true bug-finding respectively. It also supports whole-program symbolic testing, as a non-compositional engine would.

This allows one to instantiate Gillian to their preferred target language, and automatically benefit from all three modes of symbolic execution. This is in part possible thanks to the exact [46] semantics of GIL, enabling automated proofs to be done both with forwards and backwards implications, representing OX and UX respectively.

Thanks to this and the aforementioned parametricity of the memory model, Gillian is therefore a unified CSE engine, that allows scalable verification of software in a unified setting, without requiring one to adapt to two different engines, which may be prohibitively time-consuming.

#### 1.3 Existing Tools

A wide range of engines exist to verify the correctness of code, most targeted towards a specific language. This allows them to implement behaviour that is appropriate for that particular language. For instance, CBMC [38] is a bounded model checker, that allows for the verification of C programs, via whole program symbolic testing – that is, it is not compositional, and instead symbolically executes a given codebase, and verifies that assertions hold on a generated formula. While clearly useful, as shown by its success, this solution doesn't scale particularly well; larger applications need to be split modularly to be verified, and manual stubbing is required to support testing parts of the code separately.

Slightly separate from symbolic execution, KLEE [13] allows for the *concolic* execution of LLVM code – thus supporting C, C++, Rust, and any other language that can be compiled to LLVM. Concolic execution is a hybrid of symbolic and concrete execution, allowing the use of symbolic values along a concrete execution path. Similarly to CBMC, it does whole program testing, and as such doesn't scale with larger codebases. MACKE [49] is an adaptation of KLEE enabling compositional testing, proving that concolic execution can be extended to be more scalable, without losing accuracy in reported errors.

As codebases grow, there comes a need for compositional tools that can verify smaller parts of the code, enabling better scaling and integration in CI pipelines, which allow rapidly giving feedback to developers. Separation logic enables this, by allowing one to reason about a specific function and ignoring the rest of the program's state.

While Smallfoot [6] was the first tool to enable compositional symbolic execution with SL, jStar [21] was the first to allow automated verification of a real-world language, Java. It is in particular tailored towards handling design patterns that are commonly found in object-oriented programming and that may be hard to reason about with verification. It uses an intermediate representation of Java called Jimple, taken from the Java optimisation framework Soot. Similarly, VeriFast [31] is a CSE tool that is targeted at Java and a subset of C, while SpaceInvader [59] allows for the verification of C code in device drivers.

Infer [14] is a tool developed by Meta that uses separation logic and bi-abduction to find bugs in C, C++, Java and Objective-C programs, by attempting to prove correctness and extracting bugs from failures in the proof. It is the tool that pioneered bi-abduction [15], a technique enabling execution to proceed despite resources missing, via the construction of an anti-frame. Its approach, while initially stemming from separation logic, was used for finding bugs rather than proving correctness, which led to the creation of ISL [55] to provide a sound theory justifying this approach. Following this, Pulse [39] followed and

fully exploited the advances made in ISL to show that this new theory served as more than a justification to past tools, and to prove the existing of true bugs.

Also using separation logic, JaVerT [28, 29] is a CSE engine aimed at verifying JavaScript, with support for whole program testing, verification, and bi-abduction in the same fashion as Infer – it followed from Cosette [27], which already supported whole program testing.

While the above examples are targeted at specific target languages, some tools have also been designed with modularity in mind. To this end, they instead support a general intermediate language – one that isn't designed specifically for a language, like Jimple for Java – that a target language must be compiled to. This allows for greater flexibility in regards to what languages can be analysed by the engine, at the cost of requiring a compiler to the intermediate language.

For instance Viper [47] is an infrastructure capable of program verification with a permission-based separation logic, akin to what is presented in [9]. It's intermediate language is a rich object-oriented imperative typed language, that operates on a built-in heap. A range of Viper frontends exist, allowing for the verification of Scala, Java and OpenCL.

One can still take genericity further, by also abstracting the state model the engine operates on. A tool that does this is coreStar [11], the CSE engine backing jStar with all the Java-related components removed and replaced. It does not target Java, but instead a generic intermediate language. Furthermore, instead of coming with predicates predefined, it requires user-defined predicates to be provided, which are then used throughout the proof. They are defined in a language specified by coreStar, and are thus however quite limited in expressivity. Still, this is to our knowledge the first example of a tool that allows some flexibility in the underlying logic theory.

This is also what Gillian [2, 26, 45, 57] does, a CSE engine parametric on the state model. It doesn't support any one language, but instead can target any language that provides a state model implementation in OCaml and a compiler to GIL, its intermediary language. Currently Gillian has been instantiated to the C language (via CompCert-C), JavaScript, and Rust [3]. It is a generalisation of JaVerT, which only targeted JavaScript. Following Gillian, work has also been done on formalising a theoretical CSE engine, providing axioms an implementation must follow [43, 44] as well as an implementation in Rocq of the engine to prove its soundness.

In Table 1.1 we show a side-by-side comparison of the aforementioned tools.

Table 1.1: Comparison of verification tools

			Parametric	
	Target Language	Compositional	State Model	$\mathbf{Mode}$
CBMC	С	Х	Х	WPST
KLEE	LLVM	X	X	Concolic
MACKE	LLVM	✓	X	Concolic
$jStar^*$	Java	✓	×	OX
VeriFast	C, Java	✓	×	OX
$coreStar^*$	IR (coreStarIL)	✓	✓	OX
Infer	C, C++, Java, Objective-C	✓	×	OX
Cosette*	JavaScript	✓	×	WPST
$JaVerT^*$	JavaScript	✓	X	WPST, OX
Viper	IR (Viper)	✓	×	OX
Pulse	C, C++, Java, Objective-C	✓	×	UX
Gillian	IR (GIL)	✓	✓	WPST, $OX$ , $UX$

WPST = Whole Program Symbolic Testing.

\*Unmaintained

#### 1.4 Legal, Social, Ethical and Professional Requirements

There aren't particular legal, social or ethical considerations associated with compositional symbolic execution. To the contrary, improving code verification and facilitating its use and adoption may result in positive outcomes in a wide range of fields where programs are critical. This includes, among others, medical, nuclear and spatial fields, where program failures can cause life loss, catastrophic environmental degradation, or significant economic loss. For instance, the infamous Therac-25 radiation therapy machine caused the death of four patients and the injury of two others between 1985 and 1987 due to an error when handling shared mutable state [42]. This could have been detected and avoided if proper verification of the program had been done.

Verification tools have the potential to cause harm, if used on open source software by malevolent users that want to find possible exploits.<sup>3</sup> Exploits such as Heartbleed [19] or more recently the xz backdoor [20] happened on Open Source code, and an attacker could very well automate verification of these libraries privately in an effort to find new exploits. It is thus important to facilitate and democratise verification tools to counteract these actors.

<sup>&</sup>lt;sup>3</sup>Though, to our knowledge, no exploit was found this way.

### Chapter 2

# Theory

In this chapter we will cover the theoretical foundations for Resource Algebras in a CSE setting. We will first cover in §2.1 the current state of the art regarding state models, then in §2.2 what modifications are needed for a CSE to handle RAs. This is followed by a list of simple state models in §2.3, that can be used to construct more complex state models via transformers, as shown in §2.4. Finally, in §2.5 we look at the theory justifying some optimisations of the PMAP state model transformer.

#### 2.1 State of Affairs

#### 2.1.1 State Models with PCMs for CSE

Abstract separation logic historically has mostly used Partially Commutative Monoids (PCMs) [8, 16, 22, 35]. These are defined as a tuple  $(M, (\cdot): M \times M \rightharpoonup M, 0)$ , corresponding to the carrier set, composition operator, and a unique identity (or unit) element, which must satisfy the properties highlighted in Figure 2.1. PCMs are useful for representing state in SL, as they "mathematically represent the essential notions of state ownership and ownership transfer" [25], while being simple and abstract enough that they can represent a variety of concrete states (the traditional case being that of heaps).

$$\forall a, b, c. (a \cdot b) \cdot c = a \cdot (b \cdot c)$$
 (PCM-Assoc)  
$$\forall a, b. a \cdot b = b \cdot a$$
 (PCM-Comm)  
$$\forall a. a \cdot 0 = a$$
 (PCM-Identity)

Am I missing something, to highlight the fact it's partial?

Figure 2.1: Properties of Partially Commutative Monoids

When bringing this concept to a Compositional Symbolic Execution (CSE) engine, the PCM must be endowed with a set of actions  $\mathcal{A}$ , of core predicates  $\Delta$ , and functions that allow modifying it; in particular, an execute\_action function that reflects program commands that modify the state, a produce function to add an assertion to the state, a consume function to remove an assertion from the state, and a fix function to provide fixes for missing errors in bi-abduction. In [2, 43, 44], the notion of state model is thus introduced, to represent a PCM equipped with these attributes, which can be seen in Figure 2.2.

All of these together enable a CSE engine to be *parametric* on the state model; as long as the provided functions follow a set of axioms, the engine using the state model can be proved to be sound, either in over-approximate (OX) or under-approximate (UX) mode, in turn allowing for program verification, or true bug finding.

Core predicates (sometimes simply called predicates) are used to make the assertion language of the engine parametric. They take inspiration from [16], and are written  $\delta \in \Delta$ . To enable this parametricity, the assertions of the language are extended with a core predicate assertion, written  $\langle \delta \rangle (\vec{v}_i; \vec{v}_o)$ , where  $\delta$  is the core predicate,  $\vec{v}_i$  are the *in-values* (or "ins") of the predicate, and  $\vec{v}_o$  are the *out-values* (or "outs"). This distinction is used for automation purposes, to create matching plans: they tell the engine that given the values  $\vec{v}_i$ , the state can return the corresponding  $\vec{v}_o$  associated with it; this is akin to the "parameter modes" described in [48]. For instance, the traditional "points to" assertion  $a \mapsto v$ , where a is an address and v a value, can be written as  $\langle \text{points\_to} \rangle (a;v)$ : given the address a, a heap can return the value store at the location v. Core predicate also imply a satisfiability relation, denoted  $\sigma \models_{\Delta} \langle \delta \rangle (\vec{v}_i; \vec{v}_o)$  to signify a state satisfies a given predicate. Predicate satisfiability happens at the concrete level, and is later lifted to the symbolic realm by evaluating the ins and outs using a substitution  $\theta$  and variable store s; expression evaluation is denoted  $\|\hat{e}\|_{\theta,s} = e$ . Another traditional example is the emp core predicate, that is satisfied by the empty memory; it can be written  $\langle emp \rangle (;)$ .

Remark (Semantics of emp). There are two accepted and clashing semantics of the emp assertion; either it is equivalent to true and any state satisfies emp, or it represents strictly the empty memory and nothing else – the former is more appropriate for garbage collected languages while the latter for alloc/free languages. [17]

Here we make the arbitrary choice of defining it as only being satisfied by the empty memory. Because the engine is parametric on the core predicates, a user is free to chose the semantics appropriate to their use case – this is a strength of the parametric approach.

Actions are used to represent program actions that interact with the state. Typical actions include load, store, alloc or free. These actions are called with a list of arguments,

#### Elements:

- $(\Sigma, (\cdot): \Sigma \times \Sigma \longrightarrow \Sigma, 0)$ : PCM of concrete states.
- $\hat{\Sigma}$ : Set of symbolic states.
- A: Set of actions.
- $\Delta$ : Set of core predicates.

#### **Functions:**

- execute\_action: execute an action with arguments, returns a value.
- produce: add a core predicate assertion into the state, returns the new state.
- consume: removes a core predicate assertion from the state, returns the new state and the outs of the assertion.
- fix: returns the assertions needed to fix the given missing error.

#### Theoretical Properties: Is 'properties' the right word?

- \( \section : \symbolic interpretation relation. \)
- $\models_{\Delta}$ : core predicate satisfiability relation.

Figure 2.2: High-level description of the elements of a state model

return a list of results, and may modify the state, as long as the modification are sound with respect to a set of axioms, presented later.

We may now look at an example of function, to understand how the state model is used. In code, we denote a program variable x by it's name, a symbolic variable  $\hat{x}$  by #x, a core predicate  $\langle \delta \rangle(\vec{v}_i; \vec{v}_o)$  as  $\langle \delta \rangle(\vec{v}_i; \vec{v}_o)$ , and commands that use an action  $\alpha$  as  $[\alpha](\arg s)$ . The pre and postcondition are written before the function's declaration, enclosed in double square brackets  $[[\ldots]]$ . Now looking at the function  $\sec_v = 1$  which has the following specification:

$$\langle a = \hat{a} * x = \hat{x} * \langle points\_to \rangle (\hat{a}; \hat{v}) \rangle$$
 set\_value(a, x)  $\langle ret = \hat{a} * \langle points\_to \rangle (\hat{a}; \hat{x}) \rangle$ 

If the engine aims to verify the function, it will start with an empty state and produce all assertions of the pre-condition (the order being determined by the matching plan). Here, this means adding the assertions  $\mathbf{a} = \hat{a}$  and  $\mathbf{x} = \hat{x}$  to the path condition, and producing the core predicate  $\langle \text{points\_to} \rangle (\hat{a}; \hat{v})$  into the state. It will then execute the action store on the state model, with arguments  $[\hat{a}, \hat{x}]$  (note the program variables are substituted with the symbolic variable associated). In the linear heap, this simply modifies the value stored at  $\hat{a}$ , setting it to  $\hat{x}$ . The code then returns  $\mathbf{a}$ , thus assigning its value,  $\hat{a}$ , to the special variable ret which represents the value returned by the function. Finally, to ensure the state at the end of the function matches the postcondition, the assertions are consumed; first asserting ret  $= \hat{a}$  holds, and then consuming  $\langle \text{points\_to} \rangle (\hat{a}; \hat{x})$ , effectively removing

```
1  [[ a == #a * x == #x * <points_to>(#a; #v) ]]
2  [[ ret == #a * <points_to>(#a; #x) ]]
3  proc set_value(a, x) {
4    [store](a, x);
5  return a;
6  }
```

Figure 2.3: Simple function to set a value

the binding from the state. All the above executes successfully, and the function is thus verified.

The same also applies with more complex functions. For the case of function calls using function specifications, the inverse is done: when calling a function its precondition is consumed, and its postcondition is produced, effectively *framing off* the corresponding state, and then framing back on the post condition.

#### 2.1.2 Where it goes wrong

While PCMs are an obvious and straightforward choice for representing state ownership, they come with certain drawbacks. Firstly, PCMs can be overly restrictive; in particular, the presence of a unit 0 is restrictive, as it makes construction of sums impossible: for that, one needs to allow for multiple units in the object. In [22], it is shown that where a PCM fulfills  $\exists u. \forall x. x \cdot u = x$ , one can allow for more units by simply swapping the existentials:  $\forall x. \exists u. x \cdot u = x$ . This rule is further relaxed in Iris [34], where it is shown that not having a unit at all can also be useful, in particular when one wants to construct Resource Algebras (RAs) from simpler elements. Finally, and as noted in [34], they are not enough for certain applications, such as when wanting to support higher-order logic.

In [2], the concept of RA constructions from [34] is brought into a CSE setting, using PCMs. The original work however contained some errors in constructions, that created unsoundness in the verification tool. These errors stemmed, from the most part, to the use of PCMs, and the existence of the unit 0. Indeed, this made some constructions, such as sum, freeable or the partial map unsound. We will now look at examples of these.

#### Unsoundness in Freeable

The first example is with FREEABLE<sub>PCM</sub>( $\mathbb{S}$ ). This state model transformer is given a state model  $\mathbb{S}$ , and extends it with the capacity to free state, via the free action. It also adds a freed predicate, to represent freed memory.

Now consider the simple state model FREEABLE<sub>PCM</sub>(EX<sub>PCM</sub>( $\mathbb{N}$ )). Here EX<sub>PCM</sub>(X) is the exclusive state model, that represents exclusive ownership of a value of type  $X^1$ . Owner-

<sup>&</sup>lt;sup>1</sup>Note we write  $\text{Ex}_{\text{PCM}}(X)$  and  $\text{Freeable}_{\text{PCM}}(\mathbb{S})$  – the  $\text{Ex}_{\text{PCM}}$  state model accepts a set, X, whereas  $\text{Freeable}_{\text{PCM}}$  accepts a state model,  $\mathbb{S}$ .

```
[[ <ex>(; 1) ]]
    [[ ret == 0 * <freed>(;) ]]
    proc dispose() {
      [free]();
5
      return 0;
8
    [[ <ex>(; 1) ]]
    [[ ret == 0 * <freed>(;) ]]
10
    proc call_dispose() {
11
      dispose();
12
      return 0;
    }
13
```

Figure 2.4: Simple function to set a value

ship is indicated via the ex core predicate:  $\langle ex \rangle (;x)$  means the state currently has the value x – this state model will be further defined later, and is only used here for the example. For now, we define the PCM version of  $Ex_{PCM}(X)$  as:

$$\operatorname{Ex}_{\operatorname{PCM}}(X) \triangleq \operatorname{ex}(x \colon X) \mid 0$$

The notation  $\operatorname{ex}(x\colon X)$  is equivalent to  $\{\operatorname{ex}(x): \forall x\in X\}$ . We thus define  $\operatorname{Ex}_{\operatorname{PCM}}(X)$  as the set of elements of X and the 0 element. We also define composition such that  $0\cdot\sigma=\sigma\cdot 0=\sigma$  and  $\operatorname{ex}(x)\cdot\operatorname{ex}(y)$  is always undefined: since the value is owned exclusively, we can't own two values simultaneously. This satisfies the axioms of PCMs. We now consider a naive definition of FREEABLEPCM( $\mathbb S$ )'s PCM, where  $\Sigma$  is the carrier set of  $\mathbb S$ 's PCM:

$$FREEABLE_{PCM}(S) \triangleq sub(\sigma : \Sigma) \mid freed \mid 0$$

We then again define composition as  $0 \cdot \sigma = \sigma \cdot 0 = \sigma$ , and  $\operatorname{sub}(\sigma) \cdot \operatorname{sub}(\sigma') = \operatorname{sub}(\sigma \cdot \sigma')$ , which forms a valid PCM. FREEABLE<sub>PCM</sub> exposes a core predicate freed, as well as the core predicates of the underlying memory model. While we omit the full details of the state model, we note that producing freed only succeeds when in state 0 (since a state can't be both freed and something else), producing something other that freed calls the produce function of the underlying state model  $\mathbb S$ , and stores the resulting state  $\sigma$  as  $\operatorname{sub}(\sigma)$ . Similarly, consuming a predicate of  $\mathbb S$  while in state  $\operatorname{sub}(\sigma)$  calls consume on that predicate with  $\sigma$ , and then wraps the result  $\sigma'$  back into  $\operatorname{sub}(\sigma')$ .

Now consider the dispose function, as seen in Figure 2.4.<sup>2</sup> Given an owned memory value storing 1, it simply frees it. This function is called by call\_dispose, which does just that. It follows that these two functions should be successfully verified, as it is fairly obvious that they fullfil their specifications.

<sup>&</sup>lt;sup>2</sup>For these examples we only use concrete values, to make them simpler and omit substitutions that happen with symbolic values – the same results would also hold when using symbolic values, of course.

Let's now, step by step, inspect the verification of call\_dispose. First, we set up the initial state, producing  $\langle ex \rangle(;1)$  onto the empty state 0 – our state is now, according to the above definitions, sub(ex(1)). We now execute the function's body, calling the dispose function. To do this, we first need to consume its precondition: we consume  $\langle ex \rangle(;1)$  from the state sub(ex(1)). To do this, FREEABLE<sub>PCM</sub> calls the consume function of  $Ex_{PCM}(\mathbb{N})$ , with the state ex(1), which returns 0. FREEABLE<sub>PCM</sub> then wraps this state back, resulting in sub(0). We then produce the post-condition of dispose, thus producing  $\langle freed \rangle(;)$  into sub(0). This however is not valid, according to the rules outlined earlier! Indeed the state can't both be freed and something else. Of course, sub(0) represents an empty state, but that means that the consume and produce rules of FREEABLE<sub>PCM</sub> must take this into account – a simple implementation could miss it, resulting in such errors.

This difficulty is in fact caused by the existence of both 0 and  $\operatorname{sub}(0)$  as two distinct states, despite them being the same semantically: empty. To fix this, one must either make  $\operatorname{sub}(0)$  invalid, defining it as  $\operatorname{sub}(x \colon \Sigma \setminus \{0\})$ , or instead remove the definition of 0 and instead use  $\operatorname{sub}(0)$  as the unit of FREEABLE<sub>PCM</sub>. Either cases require one to be quite careful around the unit, as forgetting about it may lead to unwanted errors.

#### Unsoundness in Sum

The sum state model  $Sum(S_1, S_2)$ , also written  $S_1 + S_2$ , represents a state where either one of the two sides is owned, but not both. It allows representing states with transitions – for instance, FREEABLE(S) can also be defined as  $S + Ex(\{freed\})$ .

While this state is useful in that it allows modelling transitions from one side of the sum to another, it is in fact non-trivial to define in a way that is both OX-sound and UX-sound. In particular, we must enforce three properties: neither sides of the sum must permit the 0 element, if a state  $\sigma$  allows swapping no other smaller state  $\sigma' \preccurlyeq \sigma$  must allow swapping, and the state swapped into must be *exclusively owned*. This last property is taken from [34], and for PCMs can be defined as:

is\_exclusively\_owned\_{PCM} 
$$\sigma \triangleq \forall \sigma'. \sigma' \neq 0 \implies \neg(\sigma \# \sigma')$$

The proof that the above properties must hold can be found in appendix C. Again we note how the presence of a 0 adds requirements around these constructions, cluttering otherwise elegant definitions.

#### Unsoundness in PMap

Finally, we consider how PCMs can cause unsoundness in the partial map state model transformer, PMAP $(I, \mathbb{S})$ . It allows mappings from a domain I to a codomain state from  $\mathbb{S}$ . For this example we consider the state model construction PMAP<sub>PCM</sub> $(\mathbb{N}, Ex_{PCM}(\mathbb{N}))$ , and we define the PCM of PMAP<sub>PCM</sub> as:

$$\mathrm{PMap}_{\mathrm{PCM}}(I,\mathbb{S}) \triangleq (I \xrightarrow{fin} \Sigma) \times \mathcal{P}(I)^{?}$$

We omit the precise details of how exactly it works – it will be extensively discussed in §2.4. PMAP<sub>PCM</sub>( $I, \mathbb{S}$ ) lifts all predicates of  $\mathbb{S}$  with an additional in-value corresponding to the location  $i \in I$  to which the predicate corresponds. All action executions receive an additional argument with the location, and similarly all consume and produce calls must have the index in their ins. The behaviour of a naive implementation would then be similar to what was explained for FREEABLE<sub>PCM</sub>: calling consume or produce when the binding is not present executes it against the 0 element of the underlying PCM, and otherwise calls it on the state at the location and stores the result back at that location, overriding the previous binding. The right hand side of the product,  $\mathcal{P}(I)^2$ , represents the domain set of the partial map, in other words the set of addresses that are known to exist – any address outside of the set does not exist. The –? signifies that we extend  $\mathcal{P}(I)$  with the element  $\bot$ , in which case the domain set is not known. The core predicate  $\langle \text{domainset} \rangle (; d)$  is satisfied by the state with the domain set d. The produce function of PMAP<sub>PCM</sub> also ensures that a well-formedness constraint is upheld; in particular, given a state (h, d) with h the heap and d the domainset, if  $d \neq \bot$ , then dom $(h) \subseteq d$ .

Now, consider the following function specification, that moves a value:

$$\langle\!\langle (ex)(1;1)*\langle domainset\rangle(;\{1\})\rangle\!\rangle$$
 move()  $\langle\!\langle (ex)(2;1)*\langle domainset\rangle(;\{2\})\rangle\!\rangle$ 

Let the initial state be  $(\{1 \mapsto \text{ex}(1)\}, \{1\})$ : a heap with one cell, at address 1 – this matches the precondition of move. The engine first consumes the ex predicate from ex(1), at address 1 – this means the pointed-to state becomes 0, and the outer state becomes  $(\{1 \mapsto 0\}, \{1\})$ . The domainset predicate is then successfully consumed too, leaving  $(\{1 \mapsto 0\}, \bot)$ . Now, the post-condition is produced onto the state. First, the engine produces  $\langle \text{ex} \rangle (2; 1)$ , which adds a binding to PMAP<sub>PCM</sub> and creates the cell: our bigger state is now  $(\{1 \mapsto 0, 2 \mapsto \text{ex}(1)\}, \bot)$ . Finally,  $\langle \text{domainset} \rangle (; \{2\})$  is produced onto the state, however this doesn't succeed. Indeed, we have dom $(\{1 \mapsto 0, 2 \mapsto \text{ex}(1)\}) = \{1, 2\}$ , and of course  $\{1, 2\} \not\subseteq \{2\}$ : the domain of the heap is not a subset of the produced domain set  $\{2\}$ . This is caused because again we forgot

$$\forall a,b,c.\,(a\cdot b)\cdot c=a\cdot (b\cdot c) \qquad \qquad \text{(Iris-RA-Assoc)}$$
 
$$\forall a,b.\,a\cdot b=b\cdot a \qquad \qquad \text{(Iris-RA-Comm)}$$
 
$$\forall a.\,|a|\in M\Rightarrow |a|\cdot a=a \qquad \qquad \text{(Iris-RA-Core-ID)}$$
 
$$\forall a.\,|a|\in M\Rightarrow ||a||=|a| \qquad \qquad \text{(Iris-RA-Core-Idem)}$$
 
$$\forall a,b.\,|a|\in M\land a\preccurlyeq b\Rightarrow |b|\in M\land |a|\preccurlyeq |b| \qquad \qquad \text{(Iris-RA-Core-Mono)}$$
 
$$\forall a,b.\,\overline{\mathcal{V}}(a\cdot b)\Rightarrow \overline{\mathcal{V}}(a) \qquad \qquad \text{(Iris-RA-Valid-Op)}$$

where 
$$M^? \triangleq M \uplus \{\bot\}$$
, with  $a \cdot \bot \triangleq \bot \cdot a \triangleq a$   
 $a \preccurlyeq b \triangleq \exists c. b = a \cdot c$   
 $a \leadsto B \triangleq \forall c^? \in M^? . \overline{\mathcal{V}}(a \cdot c^?) \Rightarrow \exists b \in B. \overline{\mathcal{V}}(b \cdot c^?)$   
 $a \leadsto b \triangleq a \leadsto \{b\}$ 

A unital resource algebra is an RA M with an element  $\epsilon \in M$  such that:

$$\overline{\mathcal{V}}(\epsilon)$$
  $\forall a \in M.\epsilon \cdot a = a$   $|\epsilon| = \epsilon$ 

Figure 2.5: Definition of resource algebras in Iris

to check if the state at location 1 became 0 when consuming ex(1). We would thus need to check for this, and potentially redefine the PCM of PMAP<sub>PCM</sub> to forbid mappings to 0.

These examples hopefully showed that PCMs enforce a unit that is unwieldy, and can easily cause to unsound behaviour when forgotten. Even when taken into account, it leads to unpleasant definitions, as every state model must define a 0, and every state model transformer must then either exclude it in its construction, or explicitly take it into account.

#### 2.1.3 RAs for Iris

Iris [33–35, 37] departs from the tradition of PCMs, and introduces Resource Algebras (RAs) to model state, defined as a tuple  $(M, \overline{\mathcal{V}} \colon M \to \mathrm{iProp}, |-|\colon M \to M^?, (\cdot) \colon M \times M \to M)$ , containing the carrier set, a validity function, a partial core function and a *total* composition operator. This definition makes use of the option RA  $M^? \triangleq M \uplus \{\bot\}$  [7], that extends the set of M with an "empty" element  $\bot$ , such that  $m \cdot \bot = \bot \cdot m = \bot$ . Similarly to PCMs, a set of axioms must be satisfied, as seen in Figure 2.5.

The core function  $|-|: M \to M^?$  associates to an element m of the RA it's duplicable core. The crucial difference here, compared to PCMs, is that aside from allowing multiple units, it allows having no unit, in which case  $|m| = \bot$ . As we will see later, having no unit is a requirement for a state to be exclusively owned (i.e. it has no frame). Indeed, if the state m has a core  $|m| \neq \bot$ , then that core always constitutes a valid frame.

The validity function  $\overline{\mathcal{V}}$ :  $M \to i \text{Prop}$  is used to rule out invalid compositions, rather than relying on the partiality of the composition operator. It returns an Iris property, iProp,

which allows for step indexing and some more sophisticated logic, that it out of the scope of this project. As such, most RAs are extended with an invalid element  $\xi$ , and validity is often defined as  $\overline{\mathcal{V}}(m) \triangleq m \neq \xi$ . For instance for the Ex<sub>IRIS</sub> RA:

$$\operatorname{Ex}_{\operatorname{IRIS}}(X) \triangleq \operatorname{ex}(x \colon X) \mid \sharp$$

$$\overline{\mathcal{V}}(a) \triangleq a \neq \sharp$$

$$|\operatorname{ex}(a)| \triangleq \bot$$

$$a \cdot b \triangleq \sharp$$

The validity and core functions thus make Iris states more powerful, in that they have more flexibility in what can be expressed; while a regular PCM can be trivially converted to an RA, the opposite is not true. An example of construction that can easily be more easily constructed is the sum state model. The requirement for a transition from one side of the sum to the other can be more elegantly expressed, without having to worry about the existence of 0s [34]. Given exclusive(a)  $\triangleq \forall b. \neg \overline{\mathcal{V}}(a \cdot b)$ , we have:

EXCLUSIVE UPDATE
$$\frac{\text{exclusive(a)}}{a \leadsto b}$$

The Iris udpate  $a \rightsquigarrow b$  is a form of frame preserving update: if we update a into b, we know that any frame that was compatible with a remains compatible with b.

Furthermore, Iris is a battle-tested logic,<sup>3</sup> that comes with plenty of proofs and properties making them easy to use and adapt, whereas PCMs can prove unwieldy even for simpler state models, as shown previously.

We note an interesting similarity between Iris and the PCM approach however, which is that "the global RA must be unital, which means it should have a unit element  $\epsilon$ " [34]. This is similar to what one would find in a PCM, where we have the 0 element. Any RA can be trivially extended to have a unit, via the option RA. This in particular means that while the building blocks of both approaches are differents (PCMs or RAs), the end result that is used is similar. This is good news: it means that adapting an RA construction to a CSE engine that uses PCMs should be fairly straightforward, as one can rely on the added unit as a de facto 0.

<sup>&</sup>lt;sup>3</sup>See https://iris-project.org/#publications, where more than a hundred publications are cited as using Iris, allowing for the verification of, among others, C, Go, OCaml, Rust, Scala or WASM.

A resource algebra (RA) is a triple  $(M, |-|: M \to M^?, (\cdot): M \times M \to M)$ 

$$\forall a,b,c.\,(a\cdot b)\cdot c=a\cdot(b\cdot c) \tag{RA-Assoc}$$
 
$$\forall a,b,a\cdot b=b\cdot a \tag{RA-Comm}$$
 
$$\forall a.\,|a|\in M\Rightarrow |a|\cdot a=a \tag{RA-Core-ID}$$
 
$$\forall a.\,|a|\in M\Rightarrow ||a||=|a| \tag{RA-Core-Idem}$$
 
$$\forall a,b.\,|a|\in M\land a\preccurlyeq b\Rightarrow |b|\in M\land |a|\preccurlyeq |b| \tag{RA-Core-Mono}$$
 where 
$$M^?\triangleq M\uplus\{\bot\}, \text{ with } a\cdot\bot\triangleq\bot\cdot a\triangleq a$$
 
$$a\preccurlyeq b\triangleq\exists c.\,b=a\cdot c$$
 
$$a\#b\triangleq a\cdot b \text{ is defined}$$

A unital resource algebra is a resource algebra M with an element  $\epsilon \in M$  such that:

$$\forall a \in M. \, \epsilon \cdot a = a \qquad |\epsilon| = \epsilon$$

Figure 2.6: Definition of (partial) resource algebras

#### 2.2 State Models with RAs

We have seen how PCMs are unpractical for a CSE engine, as they easily create incorrect behaviour that can be quite tricky to notice. We will now see how we may adapt a CSE to instead use RAs. We first define the notion of partial RAs, followed by the description of the changes done to the engine's layers to support partial RAs. Finally, we list the new axioms that must be respected by the functions of the state model.

#### 2.2.1 Partial RAs

A property of Iris RAs is that composition is total – to take into account invalid composition, states are usually extended with a  $\nleq$  state, such that  $\neg \overline{\mathcal{V}}(\nleq)$  (while for states  $\sigma \neq \nleq$ ,  $\overline{\mathcal{V}}(\sigma)$  holds). While this is needed in the Iris framework for higher-order ghost state and step-indexing, this doesn't come into play when one is only manipulating RAs. As such, because having to define and later verify the validity of states can be quite unwieldy, we remove it by instead making use of partiality, as was originally the case in PCMs. This translation is done by removing the invalid state  $\nleq$  when present, and instead defining compositions that lead to it as undefined. From this, we can get rid of the validity function, and define composition as being partial:  $(\cdot)$ :  $M \times M \rightharpoonup M$ . This is also in line with the core function (-) being partial.

**Remark** (Partiality and the option RA). One may note that we use the term "partial" to refer to two different signatures; the core  $|-|: M \to M$ ? is called partial but is in fact total,

with undefined mappings being equal to  $\bot$ , while the composition (·):  $M \times M \rightharpoonup M$  is an actual partial function.

While one could adapt the core to be  $M \rightharpoonup M$ , this would prove unpractical, in cases where one still wants to use |m| even when it is  $\bot$ . On the other hand, defining composition as  $M \times M \to M^?$  is simply incorrect: let  $a \cdot b$  undefined, it is untrue that  $(a \cdot b) \cdot c = c$ , while defining an undefined composition to equal  $\bot$  would render this statement true.

Partial RAs are equivalent to regular RAs, as long as  $\neg \overline{\mathcal{V}}(a) \Rightarrow \neg \overline{\mathcal{V}}(a \cdot b)$  holds; luckily for us, this is equivalent to the axiom IRIS-RA-VALID-OP. Compositions that yield  $\not$  (or any state m such that  $\neg \overline{\mathcal{V}}(m)$ ) can be made undefined, and the validity function removed, to gain partiality, and inversely to go back to the Iris definition.

An interesting property of this change is that because validity is replaced by the fact composition is defined, the validity of a composition  $\overline{\mathcal{V}}(a \cdot b)$  in a RA is exactly equivalent to the fact two elements are disjoint in a partial RA a # b. We now define the properties of partial RAs taking this change into account – see Figure 2.6. From now, the term RA will be used to refer to these partial RAs.

We may compare the definition of  $Ex_{IRIS}$ , presented earlier, with the equivalent definition using partial RAs, Ex. The main difference is the removal of  $\overline{\mathcal{V}}$  and the  $\sharp$  element, making the definition simpler.

$$\begin{aligned} \operatorname{Ex}_{\operatorname{IRIS}}(X) &\triangleq \operatorname{ex}(x \colon X) \mid \sharp \\ &\overline{\mathcal{V}}(a) \triangleq a \neq \sharp \\ &|\operatorname{ex}(a)| \triangleq \bot \\ &a \cdot b \triangleq \sharp \end{aligned} \qquad \begin{aligned} \operatorname{Ex}(X) &\triangleq \operatorname{ex}(x \colon X) \\ &|\operatorname{ex}(a)| \triangleq \bot \\ &a \cdot b \text{ is always undefined} \end{aligned}$$

#### 2.2.2 Core Engine

We're now interested into what changes need to be brought to our CSE to handle RAs. To do this, we use the axioms and definitions specified in [43]. We will also specify additional changes that exist in Gillian and that were formalised in [2], but that haven't been added to the aforementioned formalisation.

This CSE engine is split into three layers, each adding functionality to the layer below, allowing for a clear separation of concerns. These layers are, from bottom to top: the core engine, the compositional engine, and the bi-abduction engine (see Figure 2.7).

The core engine enables whole-program symbolic execution. For this, state models must firstly define the set of states the execution will happen on. While *concrete* compositional state is modelled as an RA, *symbolic* compositional state is simply a set of elements that

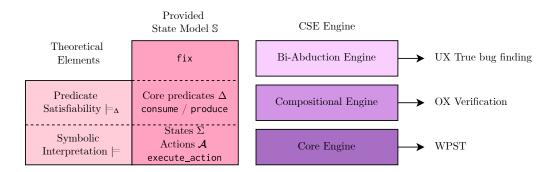


Figure 2.7: Layers of the CSE engine

does not come with composition or cores. This is because when one handles symbolic values, composition isn't a partial function anymore, as it can branch, and thus output more than one value. Take for instance the RA of trees, where nodes are defined as having an offset and size. Composing two trees made of a single node with a symbolic offset may lead to different results, depending on the interpretation of the offsets: are the two nodes adjacent? Which is on the left or the right? This cannot be modelled with an RA. Soundness of the engine is instead verified by unlifting symbolic states to concrete states via symbolic interpretation ( $\models$ ), and then using the properties of these concretes states, which do form a valid RA.

Once the set of symbolic states  $\hat{\Sigma} \ni \hat{\sigma}$  is defined, it must be further equipped with a set of actions  $\mathcal{A}$  and an execute\_action function that allows the program to modify the state. We also include the definition of the  $\models$  relation, called sat, noting that this is a purely theoretical construct that isn't implemented in the actual engine.

execute\_action: 
$$\Sigma^? \to \mathcal{A} \to \mathrm{Val} \ \mathrm{list} \to \mathcal{O}_e \times \Sigma^? \times \mathrm{Val} \ \mathrm{list}$$
 execute\_action:  $\hat{\Sigma}^? \to \mathcal{A} \to \mathrm{LVal} \ \mathrm{list} \to \mathcal{P}(\mathcal{O}_e \times \hat{\Sigma}^? \times \mathrm{LVal} \ \mathrm{list} \times \Pi)$  sat:  $\Theta \to \mathrm{Store} \to \hat{\Sigma} \to \mathcal{P}(\Sigma)$ 

We define two different versions of execute\_action: one operating on concrete states, that is deterministic, and one operating on symbolic states, that allows branching. The concrete version of the function is used in the theory, to prove the soundness of its symbolic counterpart, while the symbolic version is what is actually used by the engine to execute the analysed program.

In the concrete case, the arguments of execute\_action are, in order: the *optional* state the action is executed on, the action, and the received arguments. It returns an outcome, the new state and the returned values. It is pretty-printed as  $\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o)$ .

For its symbolic counterpart, the arguments are lifted to the symbolic realm. It also returns a set of branches, along with a path condition (PC)  $\pi$  representing the condition for the branch to exist – if the PC is not satisfiable, the branch must be cut. It is pretty-printed as  $\alpha(\hat{\sigma}, \vec{v}_i) \rightsquigarrow (o, \hat{\sigma}', \vec{v}_o, \pi)$ .

Here, the outcome is in the set of execution outcomes  $\mathcal{O}_e = \{0k, Err, Miss\}$ . Where 0k represents successful execution and Err program errors, Miss outcomes come from the compositional nature of the state. An action that leads to a Miss doesn't represent a bug in the code, but rather that additional state is needed to determine whether execution is successful or not.

The main difference with [43] is that the state may be  $\bot$ , if the action is executed on empty state; this is made explicit from using  $\hat{\Sigma}$ ? (rather than  $\hat{\Sigma}$ ). This ensures non-unital RAs are not ruled out as invalid – indeed, many useful RAs are not unital and sometimes don't have a unit at all, as is the case for instance for Ex.

We also modify the definition of a path condition. It is typically defined as a logical value LVal that must evaluate to a boolean, and which is extended via conjunction. Internally, the engine however still needs to split the conjunction into a list of terms to manipulate it. As such, to make the theory closer to what is effectively done within the engine, we make the choice of defining the path condition as a list of logical values:  $\pi \in \Pi = \text{LVal list}$ . Extension is then defined as concatening new logical values to the path condition, and strenghtening of the path condition can be trivially checked with  $\pi' \supseteq \pi$ . We further define the predicate SAT, that is true if, given the substitution  $\theta$ , store s and a path condition  $\pi$ , the conjunction of the elements of  $\pi$  resolves to true once evaluated:

$$\mathsf{SAT}_{ heta,s}(\pi) riangleq \left[igwedge_i^{|\pi|} \pi(i)
ight] = \mathsf{true}$$

The original definition of the symbolic execute\_action presented in [43] also has an "SV" argument, a set containing all existing symbolic variables. This set is used when an action creates a fresh symbolic variable, to ensure the variable is fresh. While necessary for proofs within the engine in Rocq, we omit it here, as it is only relevant for one action (alloc), and pen and paper proofs do not require the same rigidity as a Rocq proof.

Finally, the user must define the sat relation, relating concrete and symbolic states. It is pretty printed as  $\theta$ , s,  $\sigma \models \hat{\sigma}$  for  $\sigma \in \mathsf{sat} \; \theta \; s \; \hat{\sigma}$ , meaning that given a substitution  $\theta$  and a store s, the concrete state  $\sigma$  can be matched by a symbolic state  $\hat{\sigma}$ . The user must only define this relation for non- $\bot$  states; we can then lift it to the option RA, by simply adding that  $\forall \theta, s$ . sat  $\theta \; s \; \bot = \{\bot\}$ . This relieves users from needing to take  $\bot$  into account when defining  $\models$  and from proving the axiom Empty Memory.

#### 2.2.3 Compositional Engine

The compositional engine, built on top of the core engine, allows for verification of function specifications, and handles calls by specification. The state model must be extended with a set of core predicates  $\Delta$  and a pair of consume and produce functions (equivalent, respectively, to a resource assert and assume). Finally, to link core predicates to states, a core predicate satisfiability relation  $\models_{\Delta}$  (called  $\mathtt{sat}_{\Delta}$ ) must be defined in the theory, to prove soundness of consume and produce.

for 
$$M = \{ \mathsf{OX}, \mathsf{UX} \}$$

$$\mathsf{consume} \colon M \to \hat{\Sigma}^? \to \Delta \to \mathsf{LVal\ list} \to \mathcal{P}(\mathcal{O}_l \times \hat{\Sigma}^? \times \mathsf{LVal\ list} \times \Pi)$$

$$\mathsf{produce} \colon \hat{\Sigma}^? \to \Delta \to \mathsf{LVal\ list} \to \mathsf{LVal\ list} \to \mathcal{P}(\hat{\Sigma}^? \times \Pi)$$

$$\mathsf{sat}_\Delta \colon \Sigma^? \to \Delta \to \mathsf{Val\ list} \to \mathcal{P}(\mathsf{Val\ list})$$

Similarly to execute\_action, the input state can be  $\bot$ . While intuitively one may assume that the input state of consume and the output state of produce may never be  $\bot$  (as there must be a non-empty state to consume an assertion from or produce an assertion into), this would limit what core predicates can do. In particular, this means an *emp* predicate couldn't be defined, since it's production on an empty state results in an empty state.

The arguments of consume are, in order: the mode of execution, to distinguish between under-approximate and over-approximate reasoning, the state, the core predicate being consumed, the ins of the predicate. It outputs a *logical outcome*, the state with the matching predicate removed (which may result in an empty state  $\bot$ ), the outs of the predicate and the associated path condition. It is pretty-printed as  $\operatorname{consume}(m, \hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi)$ , and when the consumption is valid for both OX and UX the mode is omitted.

For produce, the arguments are the state, the core predicate being produced, the ins and the outs of the predicate, resulting in a set of new states and their associated path condition. As an example, producing  $\hat{x} \mapsto 0$  in a state  $[1 \mapsto 2]$  results in a new state  $[1 \mapsto 2, \hat{x} \mapsto 0]$  with the path condition  $[\hat{x} \neq 1]$ . If the produced predicate is incompatible with the state (e.g. producing  $1 \mapsto y$  in a state containing  $1 \mapsto x$ ), the producer vanishes. Inversely, if the assertion can be interpreted in several ways, the producer may branch. It is pretty-printed as  $\operatorname{produce}(\hat{\sigma}, \delta, \hat{v}_i, \hat{v}_o) \leadsto (\hat{\sigma}', \pi)$ .

The  $\mathsf{sat}_{\Delta}$  relation relates a possibly empty *concrete* state, core predicate and in-values to a set of out-values. It is pretty-printed as  $\sigma \models_{\Delta} \langle \delta \rangle(\vec{v}_i; \vec{v}_o)$  for  $\vec{v}_o \in \mathsf{sat}_{\Delta} \ \sigma \ \delta \ \vec{v}_i$ . For instance in the linear heap state model, we have  $[1 \mapsto 2] \models_{\Delta} \langle \mathsf{points\_to} \rangle(1; 2)$ . This definition is similar to the definition of predicates in [16, 60], where a predicate is defined



Figure 2.8: Overlap of outcomes

exactly as a set of states. We lift  $\mathsf{sat}_\Delta$  to the symbolic realm:

$$\theta, s, \sigma \models_{\Delta} \langle \delta \rangle (\vec{v}_i; \vec{v}_o) \triangleq [\![\vec{v}_i]\!]_{\theta, s} = \vec{v}_i \wedge [\![\vec{v}_o]\!]_{\theta, s} = \vec{v}_o \wedge \sigma \models_{\Delta} \langle \delta \rangle (\vec{v}_i; \vec{v}_o)$$

Here we define logical outcomes  $\mathcal{O}_l = \{\mathsf{Ok}, \mathsf{LFail}, \mathsf{Miss}\}$ . These are outcomes that happen during reasoning; in particular,  $\mathsf{LFail}$  equates to a logical failure due to an *incompatibility* between the consumed predicate and the state. For instance, consuming  $1 \mapsto 1$  from a state  $[1 \mapsto 2]$  would yield a  $\mathsf{LFail}$ , while consuming it from state  $[2 \mapsto 1]$  would yield a  $\mathsf{Miss}$ , as the state  $[1 \mapsto 1]$  could be composed with it to yield a non-miss outcome.

An addition to what [43] previously defined, taking inspiration from [2], is thus the split of what was the Abort outcome into LFail and Miss, which improves the quality of error messages and allows fixing consumptions that yield a Miss – this will be described in the next subsection.

Remark (Miss in consume). It is argued in [44] that consume shouldn't report Miss, as the missing resource might not actually be used by the function, which would then be UX unsound; we believe this preoccupation is not linked to consume but to function call semantics. If including Miss here is problematic, it should be up to the engine to correctly abort when doing function calls by specification in UX mode, rather to the consume to unnecessarily limit its capabilities.

We thus have  $\mathcal{O}_e$  for execution outcomes, and  $\mathcal{O}_l$  for reasoning outcomes – one is returned by action execution, and one for predicate consumption (see Figure 2.8). They overlap on one particular feature of the engine, which is function calls by specification: when calling a function, any outcome can happen: Ok if the function call succeeds, Err if the function call faults (as defined by its specification), Miss if additional resources are needed to satisfy the precondition of the function specification, and LFail if the state is incompatible with the function specification.

A last change done to what [43] defines is that the path condition input of consume and produce is removed, disallowing both functions from inspecting the path condition – the functions instead directly return the path condition required for the resulting branches. This brings several advantages: firstly, it simplifies the axioms for consume and produce, as there is no need to require that they strengthen the PC. Instead, the returned PC is always concatenated with the current PC, ensuring it can only be strengthened. Secondly, it simplifies the definition of consume, as whether the branch is cut or not (to satisfy OX and UX soundness) is also delegated to the engine. For instance, in UX all consumption branches that result in LFail can be dropped, as dropping branches is UX-sound – in OX however, this is not sound. A typical pattern in consume implementations is thus to evaluate the modified PC, and if an LFail is reached drop the branch depending on the mode. By omitting the PC from the arguments, this responsability is thus delegated to the engine. Finally, this makes a particular optimisation of the engine simpler to implement: incremental SMT solving.<sup>4</sup> This is a mechanism that allows storing an intermediary state in the SMT solver (a "checkpoint"), adding to the state, and later backtracking to the checkpointed state to evaluate another branch if needed. By ensuring consume and produce only return what to append to the PC, the checkpoint can easily be added before the call, and then each branch's PC added and evaluated. Allowing consume and produce to modify the PC directly would mean the PC must be parsed and filtered, to tell apart old from new terms - modified terms would make this even harder.

#### 2.2.4 Bi-Abduction Engine

To support bi-abduction in the style of Infer:Pulse [39], Miss outcomes must be fixed. This enables true UX bug finding in under-specified or unspecified functions. For this, the state model must provide a fix function, that given the details of a miss error (these details being of type LVal list and returned with the outcome) returns a list of sets of assertions that must be produced to fix the missing error.

$$fix: LVal \ list \rightarrow \mathcal{P}(Asrt) \ list$$

A list of different fixes is returned, which themselves are a set of assertions – this is because, for a given missing error, multiple fixes may be possible which causes branching. For instance, trying to load a cell that is not found in the state can lead to two possible fixes: either the cell is present, and the fix is  $\{\langle points\_to \rangle (\hat{a}; \hat{x})\}$ , either the cell is not present

 $<sup>^4\</sup>mathrm{Z}3$ , which is used in Gillian, uses "scopes" for this – see https://microsoft.github.io/z3guide/docs/logic/basiccommands/#using-scopes

anymore because it has been freed (which would then lead to a use after free error), giving the fix  $\{\langle freed \rangle (\hat{a};) \}$ .

#### 2.2.5 **Axioms**

We may now go over the axioms that must be respected by the above defined functions for the soundness of the engine. Note we only mention the axioms related specifically to the state models – axioms relating to the overall soundness of the engine or the core semantics are out of the scope of this report.

For all of the axioms we assume we have a symbolic state model  $\mathbb{S}$ , made of the concrete states RA  $\Sigma \ni \sigma$ , symbolic states set  $\hat{\Sigma} \ni \hat{\sigma}$ , actions  $\mathcal{A}$  and core predicates  $\Delta$ .

We split the axioms into two parts: those relating to the symbolic nature of state and ensuring soundness with respect to a concrete state, and those relating to the parametricity of the state.

#### Symbolicness Axioms

$$\theta, s, \sigma \models \bot \iff \sigma = \bot$$
 (Empty Memory)

The above axiom is obtained for free, by lifting the  $\models$  relation to the option RA.

$$\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o) \land \theta, s, \sigma \models \hat{\sigma} \land \llbracket \vec{v}_i \rrbracket_{\theta, s} = \vec{v}_i \implies \exists \hat{\sigma}', \vec{v}_o, \pi, \theta'.$$

$$\alpha(\hat{\sigma}, \vec{v}_i) \leadsto (o, \hat{\sigma}', \vec{v}_o, \pi) \land \mathsf{SAT}_{\theta', s}(\pi) \land \theta', s, \sigma' \models \hat{\sigma}' \land \llbracket \vec{v}_o \rrbracket_{\theta', s} = \vec{v}_o$$
(Memory Model OX Soundness)

$$\alpha(\hat{\sigma}, \vec{v}_i) \leadsto (o, \hat{\sigma}', \vec{v}_o, \pi) \land \mathsf{SAT}_{\theta', s}(\pi) \land \theta, s, \sigma' \models \hat{\sigma}' \land$$

$$[\![\vec{v}_o]\!]_{\hat{s}, \pi} \leadsto (\vec{v}_o, \pi') \land [\![\vec{v}_i]\!]_{\hat{s}, \pi'} \leadsto (\vec{v}_i, \pi'') \implies \qquad (\mathsf{Memory Model UX Soundness})$$

$$\exists \sigma. \, \alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o) \land \theta, s, \sigma \models \hat{\sigma}$$

The above two axioms are at the core of OX and UX soundness, respectively. The former ensures that if a concrete execution exists, then it must also exist in the symbolic semantics; it however doesn't ensure that more executions that those that actually exist may happen. The latter ensures that if a symbolic execution exists, then it must also exist in the concrete world; this time, no guarantees are made regarding the existence of such symbolic transition: for instance, symbolic semantics that never result in any branches are UX-sound (though they are of limited use).

Because incorrectness separation logic does execution backwards, we also prove UX soundness of the memory model backwards: if the resulting state has a concrete model,

then so must the state before. In OX, soundness is instead forwards: if the starting state is sound, so must be the resulting state.

#### Compositionality Axioms

$$\begin{split} \sigma \, \# \, \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) &= (o, \sigma', \vec{v}_o) \implies \\ \exists \sigma'', o', \vec{v}_o' \cdot \alpha(\sigma, \vec{v}_i) &= (o', \sigma'', \vec{v}_o') \wedge \\ (o' \neq \mathsf{Miss} \implies o' &= o \wedge \vec{v}_o' &= \vec{v}_o \wedge \sigma' &= \sigma'' \cdot \sigma_f) \end{split}$$
 (Frame subtraction) 
$$\alpha(\sigma, \vec{v}_i) &= (o, \sigma', \vec{v}_o) \wedge o \neq \mathsf{Miss} \wedge \sigma' \, \# \, \sigma_f \implies \\ \sigma \, \# \, \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) &= (o, \sigma' \cdot \sigma_f, \vec{v}_o) \end{split}$$
 (Frame Addition)

The above two axioms are what enable the traditional frame rule of separation logic, but bringing it to operational semantics. Frame substraction in particular is almost identical to the "Frame property" defined in [60]: if executing an action on a state from which a frame was removed yields a non-Miss, then that action did not need that frame, and the frame can be added to the result. For frame addiction, which is needed in UX, we go backwards again: if a frame can be added to the state after executing an action and the action doesn't cause a Miss, then it can also be added to the state before without interfering with the action execution.

$$\begin{array}{lll} \operatorname{consume}(\hat{\sigma},\delta,\vec{\hat{v}}_i) \leadsto (\operatorname{Ok},\hat{\sigma}_f,\vec{\hat{v}}_o,\pi) &\Longrightarrow \forall \theta,s,\sigma_f,\sigma_\delta. \\ &\theta,s,\sigma_f \models \hat{\sigma}_f \land \theta,s,\sigma_\delta \models_\Delta \langle \delta \rangle (\vec{\hat{v}}_i;\vec{\hat{v}}_o) \land \sigma_f \# \sigma_\delta \Longrightarrow & (\operatorname{Consume OX Soundness}) \\ &\exists \sigma.\, \sigma = \sigma_\delta \cdot \sigma_f \land \theta,s,\sigma \models \hat{\sigma} \land \operatorname{SAT}_{\theta,s}(\pi) \\ &(\forall o,\hat{\sigma}_f,\vec{\hat{v}}_o,\pi.\operatorname{consume}(\operatorname{OX},\hat{\sigma},\delta,\vec{\hat{v}}_i) \leadsto (o,\hat{\sigma}_f,\vec{\hat{v}}_o,\pi) \Rightarrow o_c = \operatorname{Ok}) \Longrightarrow \\ &\exists \hat{\sigma}_f',\vec{\hat{v}}_o',\pi'.\operatorname{consume}(\operatorname{OX},\hat{\sigma},\delta,\vec{v}_i,\pi) \leadsto (\operatorname{Ok},\hat{\sigma}_f',\vec{\hat{v}}_o',\pi') \\ & (\operatorname{Consume OX: No Path Drops}) \\ &\operatorname{consume}(\hat{\sigma},\delta,\vec{\hat{v}}_i) \leadsto (\operatorname{Ok},\hat{\sigma}_f,\vec{\hat{v}}_o,\pi) \Longrightarrow \forall \theta,s,\sigma. \\ &\theta,s,\sigma \models \hat{\sigma} \land \operatorname{SAT}_{\theta,s}(\pi) \Longrightarrow \exists \sigma_\delta,\sigma_f. \\ &\sigma_\delta \# \sigma_f \land \sigma = \sigma_\delta \cdot \sigma_f \land \theta,s,\sigma_\delta \models_\Delta \langle \delta \rangle (\vec{\hat{v}}_i;\vec{\hat{v}}_o) \land \theta,s,\sigma_f \models \hat{\sigma}_f \\ &(\operatorname{Consume UX Soundness}) \end{array}$$

The above three axioms show the soundness of consume for OX and UX operations. They are similar to Memory Model OX Soundness and Memory Model UX Soundness respectively. We also have an additional requirement for OX execution, stating that if there is no erroneous execution of consume then at least one successful execution exists.

$$\begin{split} \theta, s, \sigma_f &\models \hat{\sigma}_f \wedge \theta, s, \sigma_\delta \models_\Delta \langle \delta \rangle(\vec{\hat{v}}_i; \vec{v}_o) \wedge \sigma_f \,\#\, \sigma_\delta \implies \\ \exists \hat{\sigma}. \, \mathsf{produce}(\hat{\sigma}_f, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}, \pi) \wedge \mathsf{SAT}_{\theta, s}(\pi) \wedge \theta, s, (\sigma_f \cdot \sigma_\delta) \models \hat{\sigma} \end{split}$$
 (Produce: OX Soundness)

$$\begin{split} \operatorname{produce}(\hat{\sigma}_f, \delta, \vec{v}_i, \vec{v}_o) &\leadsto (\hat{\sigma}, \pi) \implies \\ \forall \theta, s, \sigma. \operatorname{SAT}_{\theta, s}(\pi) \wedge \theta, s, \sigma &\models \hat{\sigma} \implies \exists \sigma_\delta, \sigma_f. \\ \sigma_\delta \# \sigma_f \wedge \sigma &= \sigma_\delta \cdot \sigma_f \wedge \theta, s, \sigma_\delta \models_\Delta \langle \delta \rangle (\hat{v}_i; \vec{v}_o) \wedge \theta, s, \sigma_f \models \hat{\sigma}_f \\ &\qquad \qquad (\operatorname{Produce: UX Soundness}) \end{split}$$

Again, these two axioms show soundness of produce for OX and UX operation. For the former, we show that if two disjoint states exist in the concrete world and one of the two satisfies a core predicate, then producing that predicate onto the other yields their composition. For the latter, we state that if produce yields a branch that is satisfiable then there must exist a model of the result.

There are no axioms that fix needs to satisfy – as it simply results in assertions, it cannot break soundness. Either it produces no fixes or only incompatible fixes and we vanish, which is UX sound, or it finds fixes that work, in which case we already know the execution exists in the concrete world and we remain UX sound.

#### 2.3 State Models

Now that the CSE engine is defined and the properties of state models are axiomatised, we focus on state model that can be later used to build up complex state. We first look at the base case, Ex, followed by the state model of knowledge or agreement, Ag. Finally, we look at an in-between, FRAC, that allows exclusive ownership with a form of sharing.

#### 2.3.1 Exclusive

The exclusive state model Ex(X) presented earlier is the simplest form of state model. It asserts ownership of an element of the set X, provides one core predicate ex, and two actions to modify the state, load and store. The ex predicate has no ins, and one out: the value stored.

$$\operatorname{Ex}(X) \triangleq \operatorname{ex}(x \colon X)$$
 
$$|\operatorname{ex}(x)| \triangleq \bot$$
 
$$\operatorname{ex}(x_1) \cdot \operatorname{ex}(x_2) \text{ is always undefined}$$

```
EXSTOREOK STOREOK EXSTOREMISS STORE(ex(\hat{x}),[\hat{x}']) \leadsto (Ok, ex(\hat{x}'),[],[]) EXCONSOK EXPROD CONSUME(ex(\hat{x}), ex,[]) \leadsto (Ok, \bot,[\hat{x}],[]) produce(\bot, ex,[],[\hat{x}]) \leadsto (ex(\hat{x}),[])  ExFIX \\ fix [] = [\{\exists \hat{x}. \langle ex \rangle (; \hat{x})\}]
```

Figure 2.9: Some rules for Ex

```
CEXSTOREOK EXSTOREOK Store(ex(x), [x']) = (0k, ex(x'), []) EXSTOREMISS Store(\pm, [x']) = (Miss, \pm, []) EXSTOREMISS Store(\pm, [x']) = (Miss, \pm, [])
```

Figure 2.10: Side by side comparison of concrete and symbolic action rules for Ex – concrete rules are prefixed with C.

The first notation,  $\operatorname{Ex}(X)$  is the state model instantiation from the set X to the  $\operatorname{Ex}$  resource algebra. The notation  $\operatorname{ex}(x\colon X)$  stands for  $\{\operatorname{ex}(x): \forall x\in X\}$  – here " $\operatorname{ex}(x)$ " refers to the particular element of the  $\operatorname{Ex}(X)$  RA where the value is  $x\in X$ .

See Figure 2.9 for an extract of the rules for Ex – the full definition can be found in section A.1. We also show side by side the rules for the concrete and symbolic actions in Figure 2.10: in particular we note that they are strikingly similar, the latter simply being lifted from the former.

The only difference for Ex between RAs and PCMs is the absence of a 0, which is instead  $\perp$ ; in fact  $\text{Ex}(X)^?$  is strictly equivalent to  $\text{Ex}_{\text{PCM}}(X)$ .

To provide at least one full of example of state model, we also include the definition of the symbolic interpretation and core predicate satisfiability relations – both are trivial, simply checking for equality:

#### 2.3.2 Agreement

The agreement state model AG(X) is the counterpart to the exclusively owned state model – it allows sharing, and is duplicable. It defines one predicate, ag, and a single action, load. Again, ag has no ins and one out, the value. Mutating state is not sound with a duplicable resource, as that doesn't satisfy Frame Addition or Frame Addition: a frame

```
\begin{array}{lll} \operatorname{AgConsOk} & \operatorname{AgConsOk} \\ \operatorname{load}(\operatorname{ag}(\hat{x}),[]) \leadsto (\operatorname{Ok},\operatorname{ag}(\hat{x}),[\hat{x}],[]) & \operatorname{consume}(\operatorname{ag}(\hat{x}),\operatorname{ag},[]) \leadsto (\operatorname{Ok},\operatorname{ag}(\hat{x}),[\hat{x}],[]) \\ \operatorname{AgProdBot} & \operatorname{AgProdEQ} \\ \operatorname{produce}(\bot,\operatorname{ag},[],[\hat{x}]) \leadsto (\operatorname{ag}(\hat{x}),[]) & \operatorname{produce}(\operatorname{ag}(\hat{x}),\operatorname{ag},[],[\hat{x}']) \leadsto (\operatorname{ag}(\hat{x}),[\hat{x}=\hat{x}']) \end{array}
```

Figure 2.11: Some rules for AG

that is compatible with the original value is not compatible with the modified value, as they are not equal anymore.

$$\begin{split} \operatorname{AG}(X) &\triangleq \operatorname{ag}(x \colon X) \\ |\operatorname{ag}(x)| &\triangleq \operatorname{ag}(x) \\ \operatorname{ag}(x) \cdot \operatorname{ag}(x') &\triangleq \begin{cases} \operatorname{ag}(x) & \text{if } x = x' \\ \text{undefined} & \text{otherwise} \end{cases} \end{split}$$

See Figure 2.11 for an extract of its rules. Its duplicative nature can be seen in the rules AgConsOk and AgProdEq: consuming the predicate doesn't turn it into  $\bot$  but leaves it untouched, while producing into an already present state succeeds as long as the predicate's out and the state are equal, as seen in the path condition:  $[\hat{x} = \hat{x}']$ .

An observation that can be made from this is that for consumption to be sound, the resulting state can never be anything less than the input's core. Otherwise function call by specification is not frame preserving, I think? (At least with Ag.) But this isn't defined as an axiom or anything, so I'm not sure what to do with it. It doesn't follow from any axiom, and I haven't proved it for any state model (though I'm sure it holds). Is it OK to just leave it here as it is still an insight, without building upon it?

$$\operatorname{consume}(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (\operatorname{Ok}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \land \theta, s, \sigma \models \hat{\sigma} \land \theta, s, \sigma' \models \hat{\sigma}' \implies |\sigma| \preccurlyeq \sigma'$$

Unline Ex, AG is not cancellative: in other words,  $a \cdot b = a \cdot c \not\Rightarrow b = c$ , since for instance  $ag(x) \cdot \bot = ag(x) \cdot ag(x)$ , but  $\bot \neq ag(x)$ . Cancellativity is a property that was initially thought to be required for separation algebras [16, 22], but that has since been dropped [34, 54].

The agreement state model is thus useful to model immutable, shareable state – for instance, it is used in the JavaScript instantiation (discussed in §3.2) to represent the address of an object's metadata, as this information is immutable. This was first brought to a CSE engine in JaVerT [28, 29], though when that was developed the notion of an "agreement state model" didn't exist and as such it wasn't identified as one.

```
\begin{aligned} & \operatorname{FracConsAll} \\ & \operatorname{consume}(\operatorname{frac}(\hat{x},\hat{q}),\operatorname{frac},[\hat{q}']) \leadsto (\operatorname{Ok},\bot,[\hat{x}],[\hat{q}=\hat{q}']) \\ & \operatorname{FracConsSome} \\ & \operatorname{consume}(\operatorname{frac}(\hat{x},\hat{q}),\operatorname{frac},[\hat{q}']) \leadsto (\operatorname{Ok},\operatorname{frac}(\hat{x},\hat{q}-\hat{q}'),[\hat{x}],[0<\hat{q}'<\hat{q}]) \\ & \operatorname{FracConsMiss} \\ & \operatorname{consume}(\operatorname{frac}(\hat{x},\hat{q}),\operatorname{frac},[\hat{q}']) \leadsto (\operatorname{Miss},\operatorname{frac}(\hat{x},\hat{q}),[\hat{q}'-\hat{q}],[\hat{q}<\hat{q}'\leq 1]) \\ & \operatorname{FracConsFail} \\ & \operatorname{consume}(\operatorname{frac}(\hat{x},\hat{q}),\operatorname{frac},[\hat{q}']) \leadsto (\operatorname{LFail},\operatorname{frac}(\hat{x},\hat{q}),[],[\hat{q}'\leq 0 \lor 1<\hat{q}']) \end{aligned}
```

Figure 2.12: consume rules for FRAC

#### 2.3.3 Fraction

The fractional state model FRAC(X) allows for exclusive ownership while allowing sharing, via fractional permissions. This idea originates from [9, 10], where the "points to" assertion is equipped with a fraction q: a state satisfying  $a \mapsto x$  has full ownership (read and write) of x, and can be split into  $a \mapsto x * a \mapsto x$ , creating two states with read-only permissions. The FRAC memory model is a generalisation of this at the value level. It defines one predicate, frac, that has one in, the fraction q, and one out, the value x. Just like Ex, it provides a load and a store action – the only difference being that a fraction of 1 is required to modify the value. Because the max fraction is 1, we know that when it is 1 no frame can have ownership of the value, and it can thus safely be modified.

$$\begin{aligned} \operatorname{FRAC}(X) &\triangleq \operatorname{frac}(x \colon X, q \colon (0; 1]) \\ |\operatorname{frac}(x, q)| &\triangleq \bot \\ \operatorname{frac}(x, q) \cdot \operatorname{frac}(x', q') &\triangleq \begin{cases} \operatorname{frac}(x, q + q') & \text{if } x = x' \land q + q' \leq 1 \\ \text{undefined} & \text{otherwise} \end{cases} \end{aligned}$$

We put the fraction q in the in values of the fractore predicate to allow consume to subtract exactly the permission required from the state – if both the permission and the value were out-values, then consume would always evaluate to  $\bot$ , cancelling the main advantage of this state model. The rules for consume can be seen in Figure 2.12. We note here that in the symbolic state model, the fraction part is also lifted to the symbolic realm, allowing for a symbolic permission.

Because the outcome of actions depends on the amount of permission owned, FRAC is also a good example to show how one may adapt a concrete state model into a symbolic state model, by moving the relevant concrete conditions into the path condition. The rules for the concrete and symbolic store are shown in Figure 2.13. Note how, between

$$\begin{aligned} & \text{CFracStoreOk} \\ & q = 1 \\ & \text{store}(\text{frac}(x,q),[x']) = (\text{Ok},\text{frac}(x',q),[]) \end{aligned}$$
 
$$\begin{aligned} & \text{FracStoreOk} \\ & \text{store}(\text{frac}(\hat{x},\hat{q}),[\hat{x}']) \leadsto (\text{Ok},\text{frac}(\hat{x}',\hat{q}),[],[\hat{q}=1]) \end{aligned}$$
 
$$\begin{aligned} & \text{CFracStorePerm} \\ & q < 1 \\ & \overline{\text{store}(\text{frac}(x,q),[x']) = (\text{Miss},\text{frac}(x,q),[1-q])} \end{aligned}$$
 
$$\begin{aligned} & \text{FracStorePerm} \\ & \text{store}(\text{frac}(\hat{x},\hat{q}),[\hat{x}']) \leadsto (\text{Miss},\text{frac}(\hat{x},\hat{q}),[1-\hat{q}],[\hat{q}<1]) \end{aligned}$$

Figure 2.13: store rules for FRAC

CFRACSTOREPERM and FRACSTOREPERM, the condition q < 1 is replaced by  $\hat{q} < 1$ , and is returned in the path condition, rather than being checked for in the rule itself, since the value of  $\hat{q}$  depends on the current PC.

#### 2.4 State Model Transformers

While the above state models can prove useful, they are limited in use, only allowing the storage of one value. One would likely want to model state that is more complex than this, storing a range of values in a variety of ways. Iris introduces the idea of RA constructions [34] – this idea has then been adapted to Gillian in [2], with PCMs. Here we revisit these constructions, porting them to RAs. We call them *state model transformers*,<sup>5</sup> or state transformers.

<sup>&</sup>lt;sup>5</sup>Not to be confused with "state monad transformers", which allow adding state to a monad.

#### 2.4.1 Sum

The sum state model Sum( $\mathbb{S}_1, \mathbb{S}_2$ ), also written  $\mathbb{S}_1 + \mathbb{S}_2$ , allows representing states that can be in either the states of  $\mathbb{S}_1$  or  $\mathbb{S}_2$ . This state model doesn't introduce any core predicates or actions, and instead simply lifts those of the two underlying state models. Let  $\mathcal{A}_1$  and  $\Delta_1$  the actions and core predicates of  $\mathbb{S}_1$ , and  $\mathcal{A}_2$  and  $\Delta_2$  those of  $\mathbb{S}_2$ , the actions of the sum are defined as  $\mathcal{A} = \{\mathsf{L}\alpha : \alpha \in \mathcal{A}_1\} \uplus \{\mathsf{R}\alpha : \alpha \in \mathcal{A}_2\}$ , and the core predicates as  $\Delta = \{\mathsf{L}\delta : \delta \in \Delta_1\} \uplus \{\mathsf{R}\delta : \delta \in \Delta_2\}$ . Actions and predicates are simply tagged with what side of the sum they originate from.

Let  $\Sigma_1$  and  $\Sigma_2$  be the carrier sets of the RAs of  $\mathbb{S}_1$  and  $\mathbb{S}_2$  respectively; the RA of  $Sum(\mathbb{S}_1, \mathbb{S}_2)$  is:

$$\operatorname{Sum}(\mathbb{S}_{1}, \mathbb{S}_{2}) \triangleq \mathbb{S}_{1} + \mathbb{S}_{2} \triangleq l(\sigma \colon \Sigma_{1}) \mid r(\sigma \colon \Sigma_{2})$$

$$l(\sigma) \cdot l(\sigma') \triangleq l(\sigma \cdot \sigma')$$

$$r(\sigma) \cdot r(\sigma') \triangleq r(\sigma \cdot \sigma')$$

$$|l(\sigma)| \triangleq \begin{cases} \bot & \text{if } |\sigma| = \bot \\ l(|\sigma|) & \text{otherwise} \end{cases}$$

$$|r(\sigma)| \triangleq \begin{cases} \bot & \text{if } |\sigma| = \bot \\ r(|\sigma|) & \text{otherwise} \end{cases}$$

In §2.1.2 we showed how Sumpcm must disallow the unit on each side to be sound. In RAs however, this is not an issue anymore, as the underlying state models can simply not define a unit. We may also redefine is\_exclusively\_owned to take into account that the unit is not present in the carrier set, giving a much simpler definition:

is\_exclusively\_owned 
$$\sigma \triangleq \nexists \sigma' \in \Sigma$$
.  $\sigma \# \sigma'$ 

The rules of the Sum state model are otherwise straightforward, lifting execute\_action, consume and produce to the sum from the underlying state models. The only minor difference is when a Miss occurs – because fixes are only generated from a LVal list, Sum has no way of knowing which side raised the Miss, and can thus not call the fix function of the correct side. To avoid this, when the outcome of execute\_action or consume is a Miss we concatenate an identifier to the values to allow discriminating between the two. An extract of these rules is shown in Figure 2.14.

This example is interesting in that it also shows the constraint imposed by the fact  $\perp \notin \hat{\Sigma}$ : it is often needed to define auxiliary *unwrap* and *wrap* functions, that get the state

$$\begin{aligned} \text{Given } wrap_l(x) &= \begin{cases} \bot & \text{if } x = \bot \\ l(x) & \text{otherwise} \end{cases} \text{ and } unwrap_l(x_l) = \begin{cases} \bot & \text{if } x = \bot \\ x & \text{if } x_l = l(x) \\ \text{undefined otherwise} \end{cases} \\ \frac{SUMLACTION}{\hat{\sigma} = unwrap_l(\hat{\sigma}_l)} & \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi) & \hat{\sigma}_l' = wrap_l(\hat{\sigma}') & o \neq \texttt{Miss} \\ \hline \bot \alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_l', \vec{\hat{v}}_o, \pi) \end{cases} \\ \frac{SUMLACTIONMISS}{\hat{\sigma} = unwrap_l(\hat{\sigma}_l)} & \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (\texttt{Miss}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) & \hat{\sigma}_l' = wrap_l(\hat{\sigma}') \\ \hline \bot \alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (\texttt{Miss}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) & \hat{\sigma}_l' = wrap_l(\hat{\sigma}') \\ \hline \bot \alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (\texttt{Miss}, \hat{\sigma}_l', ``l' :: \vec{\hat{v}}_o, \pi) \end{cases} & \frac{SUMLFIX}{\texttt{fix} ``l' :: \vec{\hat{v}}_i = a} \\ \hline \end{bmatrix}$$

Figure 2.14: Rules for handling Miss in Sum

out a value if it is not  $\bot$ , and returns  $\bot$  otherwise. In other words,  $unwrap: \hat{\Sigma}_1^? \to \hat{\Sigma}_2^?$  and  $wrap: \hat{\Sigma}_2^? \to \hat{\Sigma}_1^?$ . While these are straightforward to define and use, they can clutter rules a bit, in favour of not needing to handle  $\bot$  and non- $\bot$  cases separately.

#### 2.4.2 Product

The product state model PRODUCT( $\mathbb{S}_1, \mathbb{S}_2$ ), also written  $\mathbb{S}_1 \times \mathbb{S}_2$ , allows representing pairs of states, where each side belongs to a specific state model. For instance, one could chose to represent the FRAC(X) state model as  $\mathrm{Ex}(X) \times \mathrm{FRAC}_{\mathbb{Q}}$ , where FRAC $_{\mathbb{Q}}$  is an exclusively owned rational in (0; 1], and with composition defined as addition. This is, in fact, the way Iris defines FRAC [32] – however, this approach makes using the state model less practical, as one would need to define a separate predicate for each side.

Iris also defines the product RA, by simply lifting pointwise the operations on each side of the product. We call this initial definition PRODUCT<sub>0</sub>, defining it as:

$$\begin{aligned} \operatorname{PRODUCT}_0(\mathbb{S}_1, \mathbb{S}_2) &\triangleq \Sigma_1 \times \Sigma_2 \\ (\sigma_l, \sigma_r) \cdot (\sigma_l', \sigma_r') &\triangleq (\sigma_l \cdot \sigma_l', \sigma_r \cdot \sigma_r') \\ |(\sigma_l, \sigma_r)| &\triangleq \begin{cases} (|\sigma_l|, |\sigma_r|) & \text{if } |\sigma_l| \neq \bot \land |\sigma_r| \neq \bot \\ \bot & \text{otherwise} \end{cases} \end{aligned}$$

This definition, while straightforward, has a problem: indeed, while its sets of states is  $\Sigma_1 \times \Sigma_2$ , we still have that  $\bot \notin \Sigma$ . If an empty state ( $\bot$ ) produces a core predicate for one of its sides, what happens to the other side? Indeed, while one side becomes defined, such that  $\sigma_l \in \Sigma_1$ , ( $\sigma_l, \bot$ ) is not a valid state, since  $\bot \notin \Sigma_2$ . This is a problem, since in our CSE engine all states are initialised as  $\bot$  and can only be built upon via actions or predicate production, one predicate at a time. It seems here that the consume-produce interface

Given 
$$wrap(x,y) = \begin{cases} \bot & \text{if } x = \bot \land y = \bot \\ (x,y) & \text{otherwise} \end{cases}$$
 and  $unwrap(s) = \begin{cases} (\bot,\bot) & \text{if } s = \bot \\ (x,y) & \text{otherwise} \end{cases}$ 

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = unwrap(\hat{\sigma})}{L\alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \rightsquigarrow (o, \hat{\sigma}'_l, \vec{\hat{v}}_o, \pi)} \qquad \hat{\sigma}' = wrap(\hat{\sigma}'_l, \hat{\sigma}_r) \qquad o \neq \texttt{Miss}}{L\alpha(\hat{\sigma}, \vec{\hat{v}}_i) \rightsquigarrow (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi)}$$

Figure 2.15: Example of action rules in Product

of our engine, which allows creating state bit by bit, can thus be unadapted for certain RAs. Of course one can define additional predicates for a specific product instantiation that combines predicates of underlying state models, however this goes against the core idea of state model transformers to minimise the amount of effort needed to construct new state models. Finally, we also note that this problem is not encountered in PRODUCT<sub>PCM</sub>, since there the unit of the PCM can be defined as  $(0_1, 0_2)$  and work straightforwardly.

We define an alternative product RA that is more suited to this engine:

$$\begin{aligned} \text{Product}(\mathbb{S}_{1}, \mathbb{S}_{2}) &\triangleq \mathbb{S}_{1} \times \mathbb{S}_{2} \triangleq (\Sigma_{1}^{?} \times \Sigma_{2}^{?}) \setminus \{(\bot, \bot)\} \\ &(\sigma_{l}, \sigma_{r}) \cdot (\sigma'_{l}, \sigma'_{r}) \triangleq (\sigma_{l} \cdot \sigma'_{l}, \sigma_{r} \cdot \sigma'_{r}) \\ &|(\sigma_{l}, \sigma_{r})| \triangleq \begin{cases} \bot & \text{if } |\sigma_{l}| = \bot \wedge |\sigma_{r}| = \bot \\ (|\sigma_{l}|, |\sigma_{r}|) & \text{otherwise} \end{cases} \end{aligned}$$

We take advantage of the option state model -? and use that instead, thus allowing either sides of the product to be  $\bot$ . Crucially however, we exclude the  $(\bot, \bot)$  state and ensure in the wrap helper that it never occurs (see Figure 2.15). This is important, because it means exclusivity is carried: if for  $(\sigma_l, \sigma_r)$  we have is\_exclusively\_owned  $\sigma_l \land$  is\_exclusively\_owned  $\sigma_r$ , then it also follows that is\_exclusively\_owned  $(\sigma_l, \sigma_r)$ . If we hadn't excluded  $(\bot, \bot)$ , then it would always be the case that  $(\sigma_l, \sigma_r) \cdot (\bot, \bot)$  holds, meaning that the state could never be exclusively owned. This is unpractical, as it would mean that disposing of a product is not frame preserving.

### 2.4.3 Freeable

The FREEABLE(S) state model transformer allows extending a state model with a free action, that allows freeing a part of memory. The freed memory can then not be accessed, and attempting to use it raises a use-after-free error. The notion of marking freed memory as freed, rather than simply discarding it, has come from Incorrectness Separation Logic [55], as indeed simply discarding the memory would break Frame Addition.

Figure 2.16: Symbolic action rule of Freeable

FREEABLE is similar to the ONESHOT RA of Iris, with the key difference being that the freed predicate used to mark a resource as freed is not duplicable. Iris defines  $ONESHOT(X) \triangleq Ex(\{1\}) + AG(X)$  [34], which is not UX-sound; in order for the switch of the side of a sum to be UX-sound, the resulting state (here, freed) must be exclusively owned, which AG never is. This is not a problem for Iris, that is only concerned with OX soundness, but because our goal is to make our state models both OX and UX sound for additional flexibility, we need to make this change.

To ensure frame preservation, the underlying state model must provide an additional function is\_exclusively\_owned:  $\Sigma \to \mathbb{B}$ , which equates to the property "exclusive" presented before. Because actions in the symbolic realm need to be sound with respect to the concrete realm, we then lift its definition to is\_exclusively\_owned:  $\hat{\Sigma} \to \text{LVal}$ , making it return a symbolic value that can be appended to the path condition.

While not needed for the core and compositional engines, the bi-abductive engine requires that misses may be fixed; FREEABLE( $\mathbb S$ ) however cannot access directly the core predicates of  $\mathbb S$  to provide fixes when freeing an empty state. As such, the state model must also provide a fix\_owned:  $\hat{\Sigma}^? \to \text{LVal list}$  function that returns possible fixes to make the given state exclusively owned. This also implies that for any state  $\sigma$  of  $\mathbb S$ , if it is not exclusively owned then there exists a bigger state  $\sigma'$  such that  $\sigma \preccurlyeq \sigma' \land \text{is\_exclusively\_owned} \sigma'$ . This is not the case, for instance, for AG, meaning a construction such as FREEABLE(AG(X)) is not sound.

Similarly to Iris, we define the FREEABLE( $\mathbb{S}$ ) state model by construction: FREEABLE( $\mathbb{S}$ )  $\triangleq$   $\mathbb{S} + \mathrm{Ex}(\{\mathrm{freed}\})$ , which allows all associated rules to be kept. For clarity, we rename the core predicate R ex (defined by  $\mathrm{Ex}(\{\mathrm{freed}\})$ ) as freed. We also extend its actions with the free action.

Because FREEABLE is constructed via other state model transformers, we only need to describe the rules for free, which are shown in Figure 2.16 – the rest of the construction is already defined. This shows how simpler state models can be extended while alleviating the user from the burden of proving the soundness of the base construction.

Conveniently, use-after-free errors are already handled by the sum construction, thanks to the SumLactionIncompat rule which forbids actions from one side of the sum to be executed on the other side.

### 2.4.4 Partial Map

The partial map state model transformer PMAP $(I,\mathbb{S})$  is perhaps one of the most important state model transformers, as it allows having multiple state instances at different addresses and accessing them. It is constructed from a domain,  $I \subseteq Val$ , and a state model for the codomain. For instance, the traditional separation logic heap can be modelled as  $PMAP(\mathbb{N}, Ex(Val))$ . The C and the JavaScript memory models can also both be modelled using a PMAP, despite them being radically different languages.

It defines one action, alloc, and one core predicate, domainset. It also lifts all actions and predicates of the underlying model, adding to them an index argument or in-value. For instance, while for Ex(X) one has the core predicate  $\langle ex \rangle(x)$ , for PMAP(Ex(X)) one would instead have predicates of the form  $\langle ex \rangle(i,x)$ , which corresponds to the "points to" assertion  $i \mapsto x$ . We define it's RA as:

$$\begin{split} \operatorname{PMAP}(I,\mathbb{S}) &\triangleq (I \xrightarrow{\operatorname{fin}} \mathbb{S}.\Sigma) \times \mathcal{P}(I)^? \\ (h,d) \cdot (h',d') &\triangleq (h'',d'') \\ \text{where } h'' &\triangleq \lambda i. \begin{cases} h(i) \cdot h'(i) & \text{if } i \in \operatorname{dom}(h) \cap \operatorname{dom}(h') \\ h(i) & \text{if } i \in \operatorname{dom}(h) \setminus \operatorname{dom}(h') \\ h'(i) & \text{if } i \in \operatorname{dom}(h') \setminus \operatorname{dom}(h) \\ \text{undefined} & \text{otherwise} \end{cases} \\ \text{and } d'' &\triangleq \begin{cases} d & \text{if } d' = \bot \\ d' & \text{if } d = \bot \\ \text{undefined} & \text{otherwise} \end{cases} \\ \text{and } d'' &= \bot \vee \operatorname{dom}(h'') \subseteq d'' \\ |(h,d)| &\triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(h') = \emptyset \\ (h',\bot) & \text{otherwise} \end{cases} \\ \text{where } h' &\triangleq \lambda i. \begin{cases} |h(i)| & \text{if } i \in \operatorname{dom}(h) \wedge |h(i)| \neq \bot \\ \text{undefined} & \text{otherwise} \end{cases} \end{split}$$

An element of PMAP is of the form (h, d), where h are the mappings from locations to states, and d is the *domain set*, the set of indices that are known to exist. It is defined such that  $dom(h) \subseteq d$  always holds. The domain set can also be  $\bot$ , in which case there is no knowledge on what addresses exist or don't. The aim of the domain set is to allow separating invalid accesses from misses: if  $d \ne \bot$ , then any access to  $i \notin d$  leads to an Erroutcome, however if  $i \in d$  and the binding does not exist in the heap, then a Miss is raised.

As a consequence of this,  $PMAP(I, \mathbb{S})$  has an additional requirement for  $\mathbb{S}$ : any action execution on  $\bot$  must lead to a Miss.

$$\alpha(\perp, \vec{v_i}) \leadsto (o, \sigma', \vec{v_o}, \pi) \implies o = \text{Miss}$$

This is needed for frame preservation; if this wasn't the case, one could run into a case where the action of a cell in the heap succeeds on an empty heap, but an empty domain set would be compatible with the state, despite its composition with the state leading to a different outcome (an Err).

This constraint also comes from the fact that PMAP can't automatically yield a Miss when executing actions on missing cells, and must instead execute the action on  $\bot$ . Indeed, because PMAP $(I, \mathbb{S})$  can accept any underlying state model, it does not "know" what values are needed to fix the missing cell. Instead, it needs  $\mathbb{S}$  to raise a Miss that it can then lift with an index.

As mentioned before, PMAP lifts actions and core predicates, adding to them an index. Getting the state associated to an index is not simple, and will in fact be the target of several optimisations in §2.5. For now we consider the simplest form of symbolic matching: either the index is present in the heap, or it isn't – if it isn't and the domain set is not  $\bot$ , we also know that it must be in the domain set. This creates three rules one must consider. For instance, given a state ( $[\hat{a} \mapsto \hat{x}, \hat{b} \mapsto \hat{y}], \bot$ ), when an action is executed with index  $\hat{c}$ , three branches are created: either  $\hat{a} = \hat{c}$ , or  $\hat{b} = \hat{c}$ , or  $\hat{c}$  is not part of the current heap, giving us  $\hat{c} \notin \{\hat{a}, \hat{b}\}$ .

Because most execute\_action, consume and produce rules of PMAP require retrieving states at a specific index, we introduce an abstraction over this, with the get and set functions. The former allows getting the state at an index with branching; its signature is get:  $PMAP(I, \mathbb{S})^? \to I \to \mathcal{P}(I \times \Sigma^? \times \Pi)$ , as it returns the index to modify exactly the returned state, the state itself, and a path condition corresponding to the branch. The set function has signature set:  $PMAP(I, \mathbb{S})^? \to I \to \Sigma^? \to PMAP(I, \mathbb{S})^?$ . Unlike for get, we don't allow branching when setting: because branching already happened when getting, and get returns the index to use to modify the state, we can use this new index and

$$\begin{split} & \underbrace{ \begin{array}{ll} \text{PMAPGETMATCH} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) & \hat{i}' \in \text{dom}(\hat{h}) & \hat{\sigma}_i = \hat{h}(\hat{i}') \\ & \text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}']) \\ \\ & \underbrace{ \begin{array}{ll} \text{PMAPGETADD} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) & \hat{i} \notin \text{dom}(\hat{h}) & \hat{d} \neq \bot \\ \text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}) \land \hat{i} \in \hat{d}]) \\ \\ \\ & \underbrace{ \begin{array}{ll} \text{PMAPGETBOTDOMAIN} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) & \hat{i} \notin \text{dom}(\hat{h}) & \hat{d} = \bot \\ \text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h})]) \\ \\ \hline \end{array} } \end{split} }$$

Figure 2.17: Rules for get for PMAP

$$\begin{split} \frac{\text{PMAPALLOC}}{\hat{d} \neq \bot \qquad \hat{i} = \text{fresh } I \qquad \hat{\sigma}_i = \text{instantiate}(\hat{\vec{v}}_i) \qquad \hat{h}' = \hat{h}[\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{d}' = \hat{d} \uplus \{\hat{i}\} \\ & \text{alloc}((\hat{h}, \hat{d}), \hat{\vec{v}}_i) \leadsto (\text{Ok}, (\hat{h}', \hat{d}'), [\hat{i}], [\hat{i} = \hat{i}]) \\ & \frac{\text{PMAPALLOCMISS}}{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{d} = \bot} \\ & \frac{1}{\text{alloc}(\hat{\sigma}, \hat{\vec{v}}_i) \leadsto (\text{Miss}, \hat{\sigma}, [\text{'domainset'}], [])} \end{split}$$

Figure 2.18: Rules for alloc for PMAP

be guaranteed that it is sound. For instance for the above example, the three executions are  $\operatorname{\mathsf{get}}(\hat{\sigma},\hat{c}) \leadsto (\hat{a},\hat{\sigma}_a,[\hat{c}=\hat{a}]), \operatorname{\mathsf{get}}(\hat{\sigma},\hat{c}) \leadsto (\hat{b},\hat{\sigma}_b,[\hat{c}=\hat{b}])$  and  $\operatorname{\mathsf{get}}(\hat{\sigma},\hat{c}) \leadsto (\hat{c},\bot,[\hat{c}\notin\{\hat{a},\hat{b}\}])$ . We present the rules for  $\operatorname{\mathsf{get}}$  in Figure 2.17.

Another unseemingly complex aspect of PMAP is allocation, for which the rules are presented in Figure 2.18. To allow instantiating, the wrapped state model must provide an instantiate function defined as instantiate: Val  $\rightarrow \Sigma$ , such that given the arguments given to alloc, it returns a newly instantiated state. An initial definition of PMAP had alloc always succeed; for instance, a  $\bot$  state could alloc, resulting in  $([\hat{i} \mapsto \sigma_i], \bot)$ , with  $\sigma_i$  the instantiated state. This however does not satisfy Frame subtraction: the frame  $(\emptyset, \{\})$ , made of an empty heap and the empty domain set, is compatible with the state before the allocation, but is not compatible with the state after, since the domain of the heap  $\{\hat{i}\}$  is not a subset of  $\{\}$ . We must thus enforce ownership of the domain set when allocating. This is odd: in allocation-based languages like C, one can always allocate a new cell, without needing to "own" anything, as the allocator is global. With this definition of alloc however, one can't allocate without it, which can be problematic: for instance in a multi-threaded setting, two threads cannot own the domain set, since it is exclusively owned (in order to allow modification).

This is a weakness of this PMAP definition, as we require a more restrictive behaviour than what is usually admitted with allocation. This also has the unfortunate side effect that whenever a function needs to allocate, it must specify a domainset core predicate in its pre and postcondition, which can become quite verbose and unpractical. Attempts have been made to preserve the domain set while allowing for more flexible allocation semantics, for instance by having the domain set be a duplicable resource that allows modification, by defining composition as set union. However these relaxations often had as a consequence that out of bounds accesses couldn't reliably be detected anymore, as indeed more heap cells with a larger domain set could always be composed to yield a successful outcome. In fact, allocation seems to be a topic that is in essence hard to tackle with separation logic; [56, 60] already mentioned some of these difficulties regarding freshness of the new index.

A solution to the above problem is to simply get rid of the domain set, and define PMAP only as the mappings from indices to states. This works, however one loses the ability to detect out of bounds accesses entirely, as additional cells can always be composed with the state to yield a non-Miss outcome. For our purposes however, we keep the definition using a domain set, as it provides better automation and is adapted to our needs.

#### 2.4.5 Partial Map Variations

The behaviour of PMAP is not always adapted for the state one wants to model. Here we describe in a shorter form some state models that function similarly to PMAP.

#### Dynamic Partial Map

The dynamic partial map, DYNPMAP $(I,\mathbb{S})$ , allows dynamically allocating missing cells when they're accessed. It is used, for instance, when modelling objects in JavaScript. In JavaScript, accessing a property that hasn't been set simply yields a default value, undefined, and any property of an object can be set directly without needing to allocate it first. This can be easily achieved by modifying PMAP, such that any out of bounds accesses have the effect of instantiating the state at that location, storing it, and executing the action against it. In other words, instead of using the domain set to distinguish between Err and Miss, we use it to distinguish between  $\mathbb{C}$  and  $\mathbb{C}$  and  $\mathbb{C}$  between  $\mathbb{C}$  and  $\mathbb{C}$  between  $\mathbb{C}$  between  $\mathbb{C}$  and  $\mathbb{C}$  between  $\mathbb{C}$  by  $\mathbb{C}$  between  $\mathbb{C}$  betwe

We also need to remove the alloc action, since allocation has no reason to exist; the cell can be modified directly instead.

#### List

The list state model, LIST( $\mathbb{S}$ ), allows storing a list of states up to a bound. Instead of having a domain set, we define it as having a bound n such that all indices i are in [0; n(. It is useful when wanting to represent continuous blocks of memory of known size. We use it

for the state model of WISL, which uses a simple block-offset memory model: each memory access requires an address of the block, and the offset within the block to read from.

Because the bound can also be  $\perp$ , it serves the exact same purpose as the domain set in PMAP: distinguishing out of bounds from misses.

#### General Map

Due to the strong similarity between PMAP and LIST, it is tempting to define a more general sort of map state model, that allows replicating both behaviours. The general map GMAP allows this; constructed as  $GMAP(I, \mathbb{S}, \mathbb{S}_D)$ , it receives the same domain set and codomain state model as before, but it also receives a discriminator state model  $\mathbb{S}_D$ , which represents the state which is capable of discriminating out of bounds accesses from missing accesses.

The discriminator must come equipped with an is\_within:  $\Sigma_D \to I \to \mathbb{B}$  function that given an index returns whether or not the state is within bounds. Here, the input discriminator state is never  $\bot$ : if the discriminator is not known then a Miss must be issued, since composing a state with a non- $\bot$  discriminator can change the outcome. is\_within must also be lifted to the symbolic realm, resulting in is\_within:  $\hat{\Sigma}_D \to I \to \mathcal{P}(LVal)$ .

Defining PMAP and LIST with a GMAP would require using the discriminator state models  $\text{Ex}(\mathcal{P}(I))$  and  $\text{Ex}(\mathbb{N})$  respectively, with the following is\_within definitions:

is\_within
$$_{\mathrm{PMAP}} \ d \ i \triangleq i \in d$$
 
$$\mathrm{is\_within}_{\mathrm{LIST}} \ n \ i \triangleq 0 \leq i \land i < n$$

For the above two examples it is also needed to exclude the store actions defined by the discriminator (via Ex), as modifying the domain set of a partial map or the bound of a list is not frame preserving. In general, it is unsound to modify the discriminator directly without also modifying the map: "increasing" it breaks Frame Addition (as states compatible with the new state aren't compatible with the old state), while "decreasing" it breaks Frame subtraction (for the inverse reason). Here increasing and decreasing refer to modifying the discriminator such that more or less indices are considered within it.

## 2.5 Optimising Partial Maps

Map-like state models are both widely used and of varied applications, as seen before. They are also quite complex, as they can lead to many branches, even when only a small number of the branches is actually feasible. Consider the PMAP state ( $[\hat{a}_1 \mapsto x_1, ..., \hat{a}_n \mapsto x_n], \bot$ ).

$$\begin{split} & \underbrace{\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma})}{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma})} \quad \hat{i} \in \text{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i})}_{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{\sigma}_i, [])} \\ & \underbrace{\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma})}{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma})} \quad \hat{i} \notin \text{dom}(\hat{h}) \qquad \hat{i}' \in \text{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i}')}_{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])} \end{split}}$$

Figure 2.19: Modified rules for syntactic matching

Executing a lookup on it will yield n+1 branches: one for each possible match, and one for the case where the index is different from all others. Such a state also needs to carry the fact that all addresses are distinct; for n addresses, this is  $\frac{n(n+1)}{2}$  inequalities. While at the theory level this isn't an issue, as we only worry about the soundness of the actions when the path condition is satisfiable, this can have a significant performance impact on the implementation.

Optimisations exist to reduce the number of branches generated, and while they have been implemented successfully in Gillian, there currently exists no theoretical justification to them proving their soundness. Here, we *theoretically* define how to optimise PMAP-like state models, how to prove the soundness of these optimisations, and explore three different optimisations that already exist in Gillian.

#### 2.5.1 Syntactic Checking

The simplest optimisation done to lookups is *syntactic* checking: before attempting to branch on every index on the map, we check if the symbolic index is already present in the map, in which case no branching is needed.

This only requires modifying the matching rule for the get function, adding  $\hat{i} \notin \text{dom}(\hat{h})$  to the precondition, and adding a rule that directly gets the relevant state; both rules can be seen in Figure 2.19.

This optimisation is what justifies returning an index for every action in PMAP. As an example, take the state model PMAP(Ex(Val)) in the state ( $[\hat{a}_1 \mapsto x_a, ..., \hat{a}_n \mapsto x_n]$ ). When executing load( $\hat{b}$ ) with a fresh symbolic variable  $\hat{b}$ , we branch n+1 times, and return the value stored there along with the used index, for instance returning  $[\hat{a}_3, \hat{x}_3]$ . The program is then free to use  $\hat{a}_3$  instead of  $\hat{b}$  for further accesses to that cell, and thanks to syntactic checking no branching will be needed. Without syntactic checking, every access invariably causes branching, even if the index is already present.

An observation is that thanks to our get and set abstractions over accessing states at a given index, we can restrict ourselves to only modify these two definitions and carry over the rest of the rules of PMAP. Because these optimisations are purely due to the symbolicness of the heap indices, the concrete state model is untouched and we can prove soundness with respect to the concrete PMAP. Furthermore, we define symbolic get *axioms*, which further ease proving soundness of the optimisations: when proving soundness of PMAP, we only use these axioms, meaning that any optimisation that satisfies the axioms is also automatically sound.

$$\theta, s, \sigma \models \hat{\sigma} \land \mathsf{get}(\sigma, i) = \sigma_i \land \llbracket \hat{i} \rrbracket_{\theta, s} = i \implies \exists \hat{\sigma}_i, \hat{i}', \pi.$$

$$\mathsf{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land \llbracket \hat{i}' \rrbracket_{\theta, s} = i \land \mathsf{SAT}_{\theta, s}(\pi)$$
(Get OX Soundness)

$$\gcd(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \llbracket \hat{i} \rrbracket_{\theta, s} = \llbracket \hat{i}' \rrbracket_{\theta, s} = i \land \mathsf{SAT}_{\theta, s}(\pi) \implies \\ \forall \sigma. \, \theta, s, \sigma \models \hat{\sigma} \implies \exists \sigma_i. \, \mathsf{get}(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$$
 (Get UX Soundness)

They are quite simple: for the OX case, any get call that exists in the concrete world must exist in the symbolic world. For the UX soundness, if a symbolic get returns a satisfiable branch then there is also a matching state  $\sigma_i$  in the concrete world.

We also define two additional axioms that we do not prove but that must hold for all get and set pairs:

$$\begin{split} \operatorname{get}(\hat{\sigma}, \hat{i}) &\leadsto (\hat{i}', \hat{\sigma}_i) \wedge \operatorname{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}' \implies \\ (\exists \sigma, \sigma_i'. \, \theta, s, \sigma \models \hat{\sigma} \wedge \theta, s, \sigma_i' \models \hat{\sigma}_i') &\Longrightarrow \end{split} \qquad \text{(Get-Set Forwards Soundness)} \\ \exists \sigma'. \, \theta, s, \sigma' \models \hat{\sigma}' \end{split}$$

$$\begin{split} & \mathsf{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i) \land \mathsf{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}' \implies \\ & (\exists \sigma', \sigma_i. \, \theta, s, \sigma' \models \hat{\sigma}' \land \theta, s, \sigma_i \models \hat{\sigma}_i) \implies \\ & \exists \sigma. \, \theta, s, \sigma \models \hat{\sigma} \end{split} \tag{Get-Set Backwards Soundness}$$

These are quite straightforward: the former states that if a state exists in the concrete world and is modified with a substate that also exists in the concrete world, then the result must exist too. The latter states the reverse: if a state after set exists in the concrete world, and we know that the subtate that used to be at the modified location also exists in the concrete world, then the original state also must exist.

These two axioms simply follow from the fact that  $\hat{i}'$  represents the location to directly modify the returned state, and that **set** only modifies the state at the given index, the rest being untouched. The axioms are used when proving OX and UX soundness respectively, as they each use either forwards or backwards reasoning.

## 2.5.2 Split PMap

The split partial map, PMAP<sub>SPLIT</sub>, goes further to reduce branches: it split the heap into two, one side being the concrete part, and the other the symbolic. A concrete value is a value that contains no symbolic variables, such that it's evaluation is unaffected by substitutions:  $\exists e. \forall \theta, s. \ [ \hat{e} ] _{\theta,s} = e.$  We introduce the predicate is\_concrete that is true if a value is concrete. The symbolicness of a binding in the heap is determined from the value (the substate) and the key; a concrete key pointing to a symbolic substate, or a symbolic key pointing to a concrete substate, both go in the symbolic part of the heap. Similarly to values, a state is concrete if its interpretation is unaffected by substitutions:  $\exists \sigma. \forall \theta, s. \theta, s, \sigma \models \hat{\sigma}.$  PMAP<sub>SPLIT</sub>(I,S) thus requires the state model to implement an is\_concrete<sub>S</sub>:  $\hat{\Sigma} \to \mathbb{B}$  function, which determines whether the entire state is concrete. For instance for the Ex(X) state model, we have is\_concrete<sub>EX</sub> ex( $\hat{x}$ )  $\triangleq$  is\_concrete  $\hat{x}$ .

This has two advantages; the first minor advantage is that when looking for syntactic matches in the heaps, if the address is symbolic then we only must check in the symbolic part of the heap, as symbolic keys are not permitted in the concrete part.<sup>6</sup> The second major advantage of this technique is related to an implementation detail of some CSE engines: substitution.

Substitution is the process of replacing logical values in a state with a new value.<sup>7</sup> For instance when learning  $\hat{a}_1 = \hat{a}_2$ , the engine may decide to substitute all occurrences of  $\hat{a}_2$  with  $\hat{a}_1$  and remove the equality from the PC if it is sound to do so (if  $\hat{a}_2$  never occurs again). Because this process applies to the entire state, it can lead to unfortunate consequences, such as needing to merge two state fragments that live at the same location. Reusing the above example, if the state contained  $[\hat{a}_1 \mapsto \hat{\sigma}_1, \hat{a}_2 \mapsto \hat{\sigma}_2]$ , then the substituted state would be  $[\hat{a}_1 \mapsto \hat{\sigma}_1 \cdot \hat{\sigma}_2]$ . This is problematic, as composition is not defined for symbolic state; symbolic states are an arbitrary set, unlike concrete states which form an RA. Current implementations usually mimic the composition of the concrete counterpart, branching when relevant; this is however not justified in the theory presented in this project, as it is caused by an implementation quirk rather than a theoretical necessity. In fact, what causes this is also a "defect" from the implementation, as it would mean the engine didn't assert that two different locations are in fact different. This phenomenon exists in Gillian under the name substitutions [45], and in Viper under the name consolidation [24].

Despite the fact it is only needed due to implementation details, we argue that optimisations are also something that is only needed because of an implementation, which justifies

<sup>&</sup>lt;sup>6</sup>This doesn't hold for concrete keys, as it can be that the concrete key maps to a symbolic state, in which case it goes in the symbolic part of the heap.

<sup>&</sup>lt;sup>7</sup>Here we define substitution in states; this is different to substitution in expressions – also known as  $\alpha$ -conversion – which cannot be avoided. State substitution usually uses expression substitution.

$$\begin{split} & \text{SPLITPMAPSETSOMECON} \\ & (\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{\sigma}_i \neq \bot \quad \text{is\_concrete}_{\Sigma} \; \hat{\sigma}_i \\ & \text{is\_concrete} \; \hat{i} \quad \hat{h}'_c = \hat{h}_c [\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{h}'_s = \hat{h}_s [\hat{i} \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}'_c, \hat{h}'_s, \hat{d}) \\ & \text{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}' \\ \\ & \text{SPLITPMAPSETSOMESYM} \\ & (\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{s}_i \neq \bot \qquad \neg (\text{is\_concrete}_{\Sigma} \; \hat{\sigma}_i \lor \text{is\_concrete} \; \hat{i}) \\ & \qquad \qquad \hat{h}'_c = \hat{h}_c [\hat{i} \not\leftarrow] \qquad \hat{h}'_s = \hat{h}_s [\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{\sigma}' = wrap(\hat{h}'_c, \hat{h}'_s, \hat{d}) \\ & \qquad \qquad \text{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}' \\ \\ & \qquad \qquad \text{SPLITPMAPSETNONE} \\ & \qquad \qquad (\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \\ & \qquad \qquad \hat{\sigma}_i = \bot \qquad \hat{h}'_c = \hat{h}_c [\hat{i} \not\leftarrow] \qquad \hat{h}'_s = \hat{h}_s [\hat{i} \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}'_c, \hat{h}'_s, \hat{d}) \\ & \qquad \qquad \text{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}' \\ \\ & \qquad \qquad \text{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}' \\ \end{split}$$

Figure 2.20: Rules for set in  $PMap_{Split}$ 

attempting to prove soundness of techniques that are caused by theoretically unsound solutions.

The second advantage of PMAP<sub>SPLIT</sub> relating to substitutions is that these can be entirely skipped for the concrete part of the heap, since we already know that it doesn't contain any symbolic variables. Substitutions are computationally expensive, as they require traversing the entire state and attempting substitutions on all values; for large states, this can take a significant amount of time.

In Figure 2.20 we present the rules relating to set for PMAP<sub>Split</sub>. In particular, when this requires modifying the binding for one side of the heap, we must delete it from the other, as it can be that the state was concrete and became symbolic, or inversely.

We also note that this optimisation, while at times performant, also comes with a cost: checking if a state is concrete also requires traversing the tree (although it doesn't require modifying it), which comes at a cost for larger trees. This could be further optimised, for instance by caching whether larger states are concrete, and invalidating the cache on modification. However, we have not found this cost to be significant enough to warrant this modification. This improvement is thus highly dependent on what code is verified; if there is a majority of symbolic values, then the speed gain is minimal, and it could even be cancelled by the cost of needing to check for concreteness of substates.

## 2.5.3 Abstract Location PMap

The last, and by far most agressive optimisation we present is the *abstract location* optimisation. Rather than straightforwardly stating how the optimisation works, we will first take a detour, explaining what abtract locations are.

#### **Abstract Locations?**

First, we define locations. A location  $loc(x) \in Loc \subseteq Val$  is an uninterpreted value, representing a location in memory. It only has a name x, such that two locations with different names are always different.

Abstract locations (or ALocs) are, as their name implies, locations lifted to the symbolic realm. An abstract location  $aloc(x) \in ALoc \subseteq LVal$  also is characterized by its name, and evaluates to a location:  $\forall a. \exists b. \, [aloc(a)]_{\theta,s} = loc(b)$ . Unlike for locations, two syntactically distinct ALocs can still be equal once concretised. Whether they are equal is decided by the engine according to an additional flag of the path condition: the *matching mode*. We denote it  $m \in \{MATCH, NO\_MATCH\}$ . When matching is enabled, abstract locations behave similarly to symbolic variables: their equality is decided by the path condition, such that for an empty path condition and  $a \neq b$ , both aloc(a) = aloc(b) and  $aloc(a) \neq aloc(b)$  are satisfiable. When matching is disabled however, equality of ALocs is only decided by their name:  $a = b \Leftrightarrow aloc(a) = aloc(b)$ . Matching is always enabled for consume, and always disabled for produce and execute\_action.

To be able to use abstract locations, the engine must be able to get the ALoc associated to a value; to this end, we introduce the function  $to\_aloc$ : LVal  $\rightarrow$  Str?. Given a logical value, it attempts to find a location or abstract location that it is equal to, and returns its name, if it finds it – otherwise, it returns  $\bot$ . It attempts to guess the ALoc by traversing the value's AST, and by looking in the PC for an equality. The crucial point of this is that this is a best effort function, that can fail to find an ALoc the value is associated to. When  $to\_aloc$   $e = \bot$ , the engine is then free to use fresh\_aloc to generate a fresh new abstract location, and add e = aloc(a) to the PC, where a is the name of the fresh ALoc. This mechanism is a form of semantic hash consing, where the engine looks for any value semantically equal to the expression, and looks for an ALoc associated to it. This approach is sound, as the generated location is fresh, such that even if the value was already bound to a location, it can always be found later that the new location is equal to the old one.

#### **Abstract Location Optimisation**

We now present the optimisation of PMAP using abstract locations:  $PMAP_{ALoc}(S)$ . Firstly, and unlike the syntactic optimisation or  $PMAP_{SPLIT}$ , this optimisation is only applicable to PMAP(Loc, S), since abstract locations can only evaluate to locations. The optimisation consists in retrieving the abstract location associated with the given index, and then looking for matches. The strength (and as will be shown later, weakness) of this optimisation is that there is no branching when matching is disabled, as indeed the SMT solver is not needed

$$\begin{aligned} & \text{with } \hat{i} \in ? \ \hat{d} \triangleq \begin{cases} \text{true} & \text{if } \hat{d} = \bot \\ \hat{i} \in \hat{d} & \text{otherwise} \end{cases} \\ & \underbrace{ \text{ALocPMapGetNoMatchNotFound} }_{ \begin{array}{c} (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a \neq \bot \quad a \notin \text{dom}(\hat{h}) \\ & \text{get}(\hat{\sigma}, \hat{i}, \text{NO\_MATCH}) \leadsto (\text{aloc}(a), \bot, [\hat{i} \in ? \ \hat{d}]) \end{cases} \\ & \underbrace{ \begin{array}{c} \text{ALocPMapGetMatchNotFound} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a \neq \bot \quad a \notin \text{dom}(\hat{h}) \\ & \text{get}(\hat{\sigma}, \hat{i}, \text{MATCH}) \leadsto (\text{aloc}(a), \bot, [\hat{i} \notin \{\text{aloc}(a') : a' \in \text{dom}(\hat{h})\} \land \hat{i} \in ? \ \hat{d}]) \end{cases} }_{ \begin{array}{c} \text{ALocPMapGetNoMatchNew} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a = \bot \quad a' = \text{fresh\_aloc } () \\ & \text{get}(\hat{\sigma}, \hat{i}, \text{NO\_MATCH}) \leadsto (\text{aloc}(a'), \bot, [\hat{i} = \text{aloc}(a') \land \hat{i} \in ? \ \hat{d}]) \end{cases} \\ & \underbrace{ \begin{array}{c} \text{ALocPMapGetMatchNew} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a = \bot \quad a' = \text{fresh\_aloc } () \\ & \text{get}(\hat{\sigma}, \hat{i}, \text{MATCH}) \leadsto (\text{aloc}(a'), \bot, [\hat{i} = \text{aloc}(a') \land \hat{i} \notin \{\text{aloc}(a'') : a'' \in \text{dom}(\hat{h})\} \land \hat{i} \in ? \ \hat{d}]) \end{cases} } \\ & \underbrace{ \begin{array}{c} \text{ALocPMapGetMatchNew} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a = \bot \quad a' = \text{fresh\_aloc } () \\ \text{get}(\hat{\sigma}, \hat{i}, \text{MATCH}) \leadsto (\text{aloc}(a'), \bot, [\hat{i} = \text{aloc}(a') \land \hat{i} \notin \{\text{aloc}(a'') : a'' \in \text{dom}(\hat{h})\} \land \hat{i} \in ? \ \hat{d}]) \end{cases} } \\ & \underbrace{ \begin{array}{c} \text{ALocPMapGetMatchNew} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a = \bot \quad a' = \text{fresh\_aloc } () \\ \text{get}(\hat{\sigma}, \hat{i}, \text{MATCH}) \leadsto (\text{aloc}(a'), \bot, [\hat{i} = \text{aloc}(a') \land \hat{i} \notin \{\text{aloc}(a'') : a'' \in \text{dom}(\hat{h})\} \land \hat{i} \in ? \ \hat{d}]) } \end{cases} } \\ & \underbrace{ \begin{array}{c} \text{ALocPMapGetMatchNew} \\ (\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \quad a = \text{to\_aloc } \hat{i} \quad a = \bot \quad a' = \text{fresh\_aloc } () \\ \text{get}(\hat{\sigma}, \hat{i}, \text{MATCH}) \leadsto (\text{aloc}(a'), \bot, [\hat{i} = \text{aloc}(a') \land \hat{i} \notin \{\text{aloc}(a'') : a'' \in \text{dom}(\hat{h})\} \land \hat{i} \in ? \ \hat{d}] } \end{cases} } } \\ \end{aligned}}$$

Figure 2.21: Extract of rules for get in PMAPALOC

to match the given abstract location with the locations in the heap: syntactic equality is sufficient. Its set of symbolic elements is very similar to PMAP, but with strings as keys (the name of the abstract location):

$$\mathrm{PMap}_{\mathrm{ALoc}}(\mathbb{S}) \triangleq \mathrm{Str} \xrightarrow{\mathit{fin}} \mathbb{S}.\Sigma \times \mathcal{P}(\mathrm{LVal})^?$$

To take advantage of matching, the definition of get is modified, adding an argument  $m \in \{\text{MATCH}, \text{NO\_MATCH}\}$ . Two rules, in their matching and non-matching form, are shown in Figure 2.21. The only difference between the two is the inclusion in the path condition of  $\hat{i} \notin \{\text{aloc}(a'') : a'' \in \text{dom}(\hat{h})\}$ , such that the branch is only valid if we're *sure* that the address is not already in the heap. This cannot be included when matching is disabled, because it is too strong a condition: when not matching we already do not attempt to match the abstract location to any of the other indices, so if it results that we skipped a correct branch then the action would vanish (result in no branches), which is not OX sound.

This optimisation is particularly powerful in the case where a value that is not known to be equal to any already present location is accessed: for a heap of size n, PMAP would branch n+1 times, whereas PMAP<sub>ALOC</sub> does not branch (if not matching), from the rule ALOCPMAPNOMATCHNEW. However, this is also too strong an optimisation. Consider the state model PMAP(Loc, PMAP(Str, Ex(Val))): the state maps locations to objects, which are string-value mappings<sup>8</sup>. It exposes predicates of the form  $\langle ex \rangle (a, n; x)$ , where a is the address in the heap, n is the name of the attribute, and x its value. Now consider

 $<sup>^8{\</sup>rm This}$  is very similar to the JavaScript state model.

Figure 2.22: Symbolic state satisfiability for PMAP<sub>ALOC</sub>

the following precondition:

$$\langle ex \rangle (\hat{a}_1, 'val'; \hat{x}) * \langle ex \rangle (\hat{a}_2, 'len'; \hat{n})$$

When producing this precondition with PMAP, we get two branches: one where  $\hat{a}_1 = \hat{a}_2$  and both properties live in the same object, and one where  $\hat{a}_1 \neq \hat{a}_2$ . However with PMAPALOC this only results in one branch: both properties are empty and the path condition is empty. In the current definition of symbolic interpretation for PMAPALOC (shown in Figure 2.22), this is not OX sound. Substitutions play a role here, because if at a later point in execution the engine finds  $\hat{a}_1 = \hat{a}_2$  then it may do a substitution and merge the two states, however this is not guaranteed. A solution to this would be to redefine satisfiability of PMAPALOC, to take into account the merging of states. Informally, this would mean that a symbolic state models a concrete state if for all concrete addresses, the composition of all symbolic substates at matching symbolic addresses is modelled by the concrete substate. For example, the symbolic state  $[\hat{a}_1 \mapsto [\text{`val'} \mapsto \hat{x}], \hat{a}_2 \mapsto [\text{`len'} \mapsto \hat{n}]]$  would then satisfy a concrete state  $[a \mapsto [\text{`val'} \mapsto x, \text{'len'} \mapsto n]]^9$  for an empty path condition, since we could compose the two substates together (assuming again that composition is defined for symbolic states, which it isn't in the current formalism).

In fact, this unsoundness is also visible when attempting to prove Get OX Soundness, and another similar issue occurs in the proof for Get UX Soundness. In both cases, the issue only occurs when the matching mode is NO\_MATCH, confirming this is the source of the problems.

#### In Defense of Abstract Locations

As shown above, there is currently an unsoundness in optimisations using abstract locations, due to the *matching mode*, and in particular when it is disabled. This occurs because the engine eagerly assumes difference, when it shouldn't always. We argue here that this behaviour, while currently unsound, can be made sound, and has a practical justification.

 $<sup>^9\</sup>mathrm{We}$  take some liberties when representing the states, for the sake of readability, by omitting the domain sets.

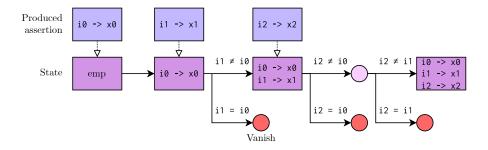


Figure 2.23: Branching when producing cells

Consider a simple linear heap, PMAP(Loc, Ex(Val)), with the core predicate  $\langle \exp \rangle(\hat{i}; \hat{x})$ , that we denote  $\hat{i} \mapsto \hat{x}$ . In a heap  $\hat{i}_0 \mapsto \hat{x}_0 * \dots * \hat{i}_n \mapsto \hat{x}_n$ , one's intuition is of course that all locations are different, and indeed both PMAP(Loc, Ex(Val)) and PMAP<sub>ALoc</sub>(Ex(Val)) will create the same state when producing these assertions. In the optimised case, because each index will be new when producing the associated core predicate, PMAP<sub>ALoc</sub> will always generate a new abstract location to associate to it, and consider that it is different from all other locations. This is done *implicitly* – that is, the path condition is not extended to take this into account. The base version, on the other side, will need to branch for each new assertion. Because these assertions do not already explicitly specify that all addresses are distinct  $(\hat{i}_0 \neq \hat{i}_1 \wedge \hat{i}_0 \neq \hat{i}_2 \wedge \dots)$ , the branch  $\hat{i}_0 = \hat{i}_1$  is satisfiable; however of course producing  $\hat{i}_1 \mapsto \hat{x}_1$  into a state  $[\hat{i}_0 \mapsto \hat{x}_0]$  when  $\hat{i}_0 = \hat{i}_1$  will vanish. This then happens for each address combination; for n addresses, this results in  $\frac{n(n+1)}{2}$  inequalities. This makes the PC grow quadratically, slowing down queries sent to the SMT solver, as well as branching and vanishing right after just as many times (see Figure 2.23).

One could of course specify, along with the assertions, that all addresses are distinct – this is however quite effortful, and detracts from the already complex task of formulating function specifications.

Furthermore, one can't design a general map state model that makes this assumption — while for a simple case like PMAP(Loc, Ex(Val)) it is always true that different cells must have different addresses (since the cell is exclusively owned), this is not true for more complex state models. For instance take PMAP(Loc, PMAP(Str, Val)), where the map contains objects, which are string-value pairs. When describing two objects with non-overlapping keys, if the addresses are not explicitly said to be different then both the branches where they're distinct and where they're separate should be considered. While these two branches should work the same, it results that disabling this abstract location optimisation when producing (i.e. enabling matching on produce) results in several functions not verifying with Gillian, in WISL, C and JS. It thus seems that this lack of completeness is at least in part caused by developer intuition, where one "knows" when writing the specification that

the locations should be distinct, without formalising it, and instead relying on this unsound behaviour. This problem seems partly caused by recursive user predicates, where addresses are "discovered" as they are unfolder; these locations are, before the unfolding, not named, and as such cannot be stated to be distinct from the others.

This difference between intuition and actual application to a verification setting seems to be caused by "Barendregt Variable Convention" [58], where variables named differently are informally assumed to be different, which can become an issue for an engine that does not properly take it into account. A solution to this would be to better formalise the principles governing abstract locations, for instance using *Nominal Logic* [53, 58]. It provides a sound framework for reasoning about syntactic equivalence of terms within recursive structures, such as recursive user defined predicates.

## Chapter 3

# Implementation

We now present the implementation of the state models and state model transformers presented. In §3.1, we present the library which exposes these state models, some relevant implementation details, and notable differences with the theory. In §3.2, we then present the instantiations of Gillian for three different target languages using this library.

## 3.1 State Model Library

All of the presented state models have been implemented in OCaml 5.2 [41] for Gillian [26, 45, 57], a CSE engine that is parametric on the state model. We first present the general interface for state models, and how it connects to the Gillian engine. We then look more in depth into the implementation of two state models: Ex for its canonical simplicity, and PMAP for its complexity. Finally, we list some additional "utility" state models that were omitted in the previous chapter but that are frequently used when instantiating state models.

#### 3.1.1 State Model Interface

State models are built as standalone *modules*, while state model transformers are *functors*, that accept one or more state model modules, and return a new state model module. Gillian already provides an interface for state models, MonadicSMemory, whose name originates from its extensive use of the Delayed monad: a *symbolic execution monad* that allows branching and SMT queries in a sugared syntax that is easy to read and use, as seen in Figure 3.1. The signature of the monad, in OCaml syntax, is 'a Delayed.t = Pc.t -> ('a \* Pc.t) list: given a PC, we return a list of branches, containing the new value and an updated PC. We make extensive use of *let bindings*: 1 let\* v = d in e is the bind operator, passing the

<sup>&</sup>lt;sup>1</sup>See https://ocaml.org/manual/5.2/bindingops.html

```
1 let produce pred s args =
2 match (pred, args) with
3 | SubPred pred, idx :: args →
4 let* ss = validate_index s idx in
1et+ ss' = S.produce pred ss args in
6 update_entry s idx ss'
```

Figure 3.1: Example of usage of the Delayed monad

value v contained in the delayed element d to the expression e, which must result in another delayed value. let+ is the map operator, which modifies the value without branching. For the code shown, we branch when validating the index, and then branch again when calling S.produce; we then update the entry with ss', without branching.

The interface provided by Gillian is not satisfactory: it is quite complex, requiring the implementation of 26 different functions whose purpose is at times unclear.<sup>2</sup> We thus create a simpler interface, MyMonadicSMemory, reducing the number of required functions and type definitions, and simplifying the definition of some functions (removing arguments when unneeded). Gillian also passes actions and core predicates as strings; while this is manageable for monolithic state models, this is unpractical. We instead require state models to define action and pred types, along with their to\_string and from\_string parsers, ensuring decoupling between parsing and execution. See Table 3.1 for a side-by-side comparison of the interfaces. Because Gillian still requires the state model to satisfy MonadicSMemory, we provide a Make functor converting from our interface to Gillian's.

Most of the removals are possible because some default function can be used for all state models, under the assumption they behave a particular way; for instance, we remove the copy function, because we assume all state models only work on immutable code. Most of the additions are due to the modular nature of our state models, or to enforce a stronger typing of actions and predicates.

A significant difference with the theory presented is that the execute\_action, consume and produce functions do not take as input states of type t option (to imitate  $\hat{\Sigma}^{?}$ ), but instead t. While both options are possible, using the option monad proved to at times be somewhat unweildy (for the same reason it is unpractical to use wrap and unwrap in the rules) – we instead rely on the is\_empty function when necessary. We also require a compose function, despite it not being justified in the theory, because of substitutions.

One of the most significant improvements of our interface is the re-work of fixes in the engine, to behave precisely like the definition of fix in the theory. Gillian uses an intermediary fix\_t type, that is obtained from an error with get\_fixes, and must then be applied with apply\_fix: t -> fix\_ t -> (t, err\_t) result Delayed.t. Aside from being

<sup>&</sup>lt;sup>2</sup>This is due to the fact Gillian has been in development for several years, and as it grew some unused functions were never cleaned up.

Table 3.1: Difference between Gillian's and our state model interface

Gillian interface	Our interface	Outcome
apply_fix		Removed
clean_up		Removed
clear		Removed
сору		Removed
<pre>get_failing_constraint</pre>		Removed
get_init_data		Removed
get_print_info		Removed
is_overlapping_asrt		Removed
mem_constraints		Removed
pp_by_need		Removed
pp_c_fix		Removed
split_further		Removed
sure_is_nonempty		Removed
alocs	alocs	
assertions	assertions	Typed the core predicates
can_fix	can_fix	
consume	consume	Typed the core predicate
execute_action	execute_action	Typed the action
get_fixes	get_fixes	Simplified, use assertions
<pre>get_recovery_tactic</pre>	<pre>get_recovery_tactic</pre>	Simplified
init	empty	Renamed
lvars	lvars	
рр	рр	
pp_err	pp_err	
produce	produce	Typed core predicate
substitution_in_place	substitution_in_place	
	action_from_str	Added
	action_to_str	Added
	assertions_others	Added
	compose	Added
	instantiate	Added
	is_concrete	Added
	is_empty	Added
	is_exclusively_owned	Added
	pred_from_str	Added
	pred_to_str	Added

an unpractical system to use, generating a fix and then applying it separately, it *introduces* unsoundness into the bi-abductive engine. Indeed, applying a fix modifies the state directly; there is no requirement for this fix to be describable in terms of assertions. This means that applying the same fix on the state and the anti-frame may give different results, that do not result in the appropriate change for the function's generated pre-condition.

For instance in [2], an example implementation of PMAP is given, where executing an action on an empty cell creates a Miss, called MissingBinding. A naive implementation of PMAP for Gillian could resolve this fix by simply adding a binding that points to the empty substate in the heap at the corresponding index. This however is not sound, as it *cannot be* expressed as an assertion: it manipulates the representation of state, but no assertion exists that creates an empty binding, since PMAP doesn't define core predicates for bindings – it only lifts the core predicates of the wrapped state model. This would therefore result in a fix that doesn't change the anti-frame or the specification of the function, which is invalid.

We fix this by enforcing fix\_t to be of type MyAsrt.t: strongly typed assertions. The implementation of apply\_fix then produces all the fixes into the state model. This change thus not only makes state models less prone to error, but also simplifies their code.

#### 3.1.2 Exclusive

We now present the implementation of the Ex state model, step by step (skipping the uninteresting details such as parsing or auxiliary functions). Firstly, we must define the types for the state model: the state type, the errors (which include misses), the actions and core predicates. This is trivial:

```
1 type t = Expr.t option
2 type err_t = MissingState
3 type action = Load | Store
4 type pred = Ex
```

Here we define states as being an optional expression, where the None case stands for  $\bot$ . There is also one singular error, for misses. We then must define the function of the state model: execute\_action, consume and produce, following the rules outlined in appendix A.1:

```
let execute_action action s args =
 1
2
       match (action, s, args) with
       | _, None, _{-} \rightarrow DR.error MissingState
 3
       | Load, Some v, [] \rightarrow DR.ok (Some v, [ v ])
 4
       | Store, Some _, [ v' ] \rightarrow DR.ok (Some v', [])
 5
       | _ → failwith "Invalid Ex execute_action"
6
8
     let consume core_pred s ins =
9
      match (core pred. s. ins) with
10
       | Ex, Some v, [] \rightarrow DR.ok (None, [v])
11
       \mid Ex, None, \_ \rightarrow DR.error MissingState
       | \_ \rightarrow failwith "Invalid Ex consume"
12
13
14
     let produce core_pred s insouts =
      match (core_pred, s, insouts) with
15
       | Ex, None, [ v ] \rightarrow Delayed.return (Some v)
16
       | Ex, Some _, _ 
ightarrow Delayed.vanish ()
```

```
| \_ \rightarrow failwith "Invalid Ex produce"
```

Again, this is straightforward; a simple pattern matching suffices. DR.ok stands for Delayed.return (Ok -), returning the input in a Result monad, wrapped in a Delayed monad. The outcome is thus Ok - for Err, LFail and Miss outcomes, we use DR.error; the state model then provides a can\_fix function that given an error returns whether it is fixable, signifying it's a Miss. Experiments were done to create an Outcome monad, with variants Ok | Miss | Err | LFail, as a replacement for the Result monad with Ok | Err. While this allowed getting rid of the can\_fix function, it was unpractical to use, as it was purely "aesthetic"; the engine still needed to parse everything back to a Result. Furthermomre, this made lifting errors quite tedious: instead of simply defining a variant SubError of S.err\_t for the error type and lifting the errors of the underlying state model through it, one had to handle all three error cases, creating a lot of boilerplate code for little to no added value.

Finally here we mention some of the additional functions that the state model must implement:

```
let empty () = None
    let is_empty = Option.is_none
 3
    let is_exclusively_owned = Option.is_some
    let instantiate = function
      | [] \rightarrow (Some (Expr.Lit Undefined), [])
      | [v] \rightarrow (Some v, [])
      | _ → failwith "Invalid Ex instantiation"
8
    let assertions = function
      | None \rightarrow []
      | Some v \rightarrow [(Ex, [], [v])]
10
11
    let can_fix MissingState = true
    let get_fixes MissingState =
12
      [ [ MyAsrt.CorePred (Ex, [], [ LVar (Generators.fresh_svar ()) ]) ] ]
13
```

These functions are, in order: the constructor for the empty  $(\bot)$  state, a function to tell if the state is  $\bot$ , the is\_exclusively\_owned presented before, the instantiate function required by PMAP, a function to return all the core predicates associated with a state, the function to tell apart Err and LFail from Miss, and the fix function presented in chapter 2. For fix, we note that specifying the existential for the fix  $\exists \hat{x}. \langle ex \rangle(; \hat{x})$  is not needed, as all unbound logical variables in Gillian are existentially quantified by default.

This is – skipping the irrelevant functions – all of the code for a simple state model in Gillian, using our interface MyMonadicSMemory. The entire code is 77 lines long, with the biggest function taking 9 lines. This shows that for simple state models, our interface allows for just as simple implementations.

#### 3.1.3 Partial Map

We now present some details of the  $PMAP(I, \mathbb{S})$  implementation. It is a functor, receiving a PMapIndex module and a MyMonadicSMemory module, corresponding to the I and  $\mathbb{S}$  arguments of the constructor. Both PMAP and DYNPMAP are implemented as one state model;

what enables the "dynamic" behaviour is the value mode: Static | Dynamic that is contained in the input index. The only difference in behaviour this causes is enabling the alloc action in static mode, and catching any NotAllocated errors and replacing them with an implicit allocation in dynamic mode. Aside from this, PMapIndex has an is\_valid\_index, which returns the "actual" index associated with an input index. This is where the abstract location optimisations plays a role: for the LocationIndex module, it is:

```
module LocationIndex : PMapIndex = struct
 1
2
      let mode = Static
3
4
      let make_fresh () =
5
        let loc = ALoc.alloc () in
6
        Expr.loc_from_loc_name loc
7
8
      let is_valid_index = function
        | (Expr.Lit (Loc _) | Expr.ALoc _) as l \rightarrow Delayed.return (Some l)
9
10
         \mid e when not (Expr.is_concrete e) \rightarrow (
             let* loc = Delayed.resolve_loc e in
11
            match loc with
12
13
             | Some 1 \rightarrow Delayed.return (Some (Expr.loc_from_loc_name 1))
14
            \mid None \rightarrow
                let loc' = make_fresh () in
15
                Delayed.return ~learned:[ (loc' #== e) ] (Some loc'))
16
         | _ 
ightarrow Delayed.return None
17
18
    end
```

We see here how Delayed.resolve\_loc, named to\_aloc in §2.5, may not find a result, in which case we generate a new fresh abstract location to assign to the expression, by extending the PC: "~learned: [ (loc' #== e) ]" appends the expression loc' = e to it, with #== signifying the symbolic equality.

This is only one part of the implementation for what was called get in the theory. The rest of it is two functions:<sup>3</sup>

```
let validate_index (h, d) idx =
2
      let* idx' = I.is_valid_index idx in
      match idx' with
3
      | None → DR.error (InvalidIndexValue idx)
4
      | Some idx' \rightarrow (
6
          let* match_val = ExpMap.sym_find_opt idx' h in
          match (match_val, d) with
          | Some (idx'', v), _ \rightarrow DR.ok (idx'', v)
8
          | None, None \rightarrow DR.ok (idx', S.empty ())
10
          | None, Some d 
ightarrow
              if%sat Formula.SetMem (idx', d) then DR.ok (idx', S.empty ())
11
12
              else DR.error (NotAllocated idx'))
```

This is the function in PMAP responsible for retrieving the substate associated to an index; it calls ExpMap.sym\_find\_opt, which symbolically tries to match an index against all entries of the map, and if that finds no result it either returns an empty substate  $\bot$  with S. empty (), or raises an error if a bound exists and  $\hat{i}' \notin \hat{d}$ . Here if%sat is sugared syntax to allow branching according to a symbolic formula, using Delayed. If both the formula and its negation can be true then the value it returns evaluates to two branches, extended with the formula or its negation in the PC.

```
1 module ExpMap = struct
```

<sup>&</sup>lt;sup>3</sup>We omit some optimisations for clarity.

```
2
       include Map.Make (Expr)
3
4
       let sym_find_opt k m =
5
         match find opt k m with
6
         | Some v \rightarrow Delayed.return (Some (k, v))
7
8
             let rec find match = function
9
               | [] \rightarrow Delayed.return None
               (k', v) :: t1 \rightarrow if%sat k' #== k then Delayed.return (Some (k', v))
10
11
                   else find_match tl
13
14
             find_match (bindings m)
15
     end
```

This last function is the symbolic matching function, which checks if  $SAT_{\theta,s}(\hat{i}=\hat{i}')$  for every index  $\hat{i}'$  in the map, and creates a branch for every possible match. It also does *syntactic matching*, as seen at the start of the function: it first attempts to find the matching key directly, using Map.find\_opt, and only if that fails does it proceed to match with every key.

The Delayed monad and the associated syntactic sugar, initially described in [2], allow us to write branching-heavy code with ease, abstracting the details of the PC away from us. We finally take a brief look at the implementation of produce, to show how a state model transformer accesses the wrapped state model via the interface we defined:

```
let produce pred s args =
       match (pred, args) with
3
       | SubPred pred, idx :: args \rightarrow
4
           let*? idx, ss = validate_index s idx in
5
           let+ ss' = S.produce pred ss args in
           update_entry s idx idx ss'
6
7
       | DomainSet, [ d' ] \rightarrow (
8
           match s with
9
           \mid _, Some \_ \rightarrow Delayed.vanish ()
10
           \mid h, None \rightarrow Delayed.return (h, Some d'))
       \mid _ \rightarrow failwith "Invalid PMap produce"
```

Here again, and even for a transformer as complex as  $PMAP(I, \mathbb{S})$ , the code is short and minimal. For the case of a core predicate from  $\mathbb{S}$  (SubPred pred), we get the substate associated with the index (let\*? gets the value associated with a Result and vanishes if it's an error), we produce the predicate onto the substate, and then update the entry. For the domain set, we simply add it to the state, vanishing if a domain set is already present as it is exclusively owned.

The entire PMAP code is 343 lines of code – a sizeable amount, but still rather short given its complexity.

#### 3.1.4 Helper State Models

We now define some additional state model transformers that are often used when one wants to instantiate a state model with the aforementioned library. These transformers focus on applying small, customisable modifications to state models, allowing for an even greater re-use. In particular these utility state model transformers are useful when instantiating our state model with the goal to mimic a pre-existing state model.

ACTIONADD( $\mathbb{S}$ , A) is a transformer that equips the given state model with additional actions defined by A, without needing to define any predicates, overriding functions, or re-implementing repetitive functions.

FILTER( $\mathbb{S}, \mathcal{A}, \Delta$ ) is a transformer that allows filtering the actions and predicates exposed by a state model, for instance to limit the functionality of a state model if one of its provided actions shouldn't exist in the semantics of the language.

Mapper( $\mathbb{S}, F_{\mathcal{A}}, F_{\Delta}$ ) allows renaming predicates or actions.

INJECTOR( $\mathbb{S}, I$ ) adds hooks into consume, produce and execute\_action, allowing the developer to apply transformations to the inputs or outputs of these functions. For instance, this makes re-ordering the arguments of actions trivial, or allows treating an out-value as an in-value that makes the branch vanish if it doesn't correspond.

## 3.2 Instantiations

We now look at how this library of state models can be used to instantiate Gillian. In particular we focus on instantiating Gillian to target languages (TLs) that have already been instantiated for it: WISL, JavaScript and C.

To instantiate a language to Gillian, the developer must provide its state model, and a compiler from the TL to GIL. Because this project focuses on constructing state models, it is thus natural that we picked languages for which the compiler already exists.

We note that all forms of PMAP(Loc, S) implicitly make use of the abstract location optimisation described in §2.5. This is because all of Gillian's instantiations also make use of it, and the goal of our instantiations is to have *parity* with the reference implementation, rather than to have additional completeness. Indeed, while removing the unsoundness caused by ALocs is straightforward in our library, it causes a majority of tests to fail, making any comparison with Gillian impossible.

## 3.2.1 WISL

WISL (WhIle language for Separation Logic) is an imperative toy language used to teach students the principles of software verification. It uses a simple block-offset memory:

```
PMAP(Loc, FREEABLE(LIST(Ex(Val))))
```

Unsurprisingly for a simple state model like this, its instantiation is also trivial when brought to code:

```
1 module ExclusiveNull = struct
2 include Exclusive
```

```
4
       let instantiate = function
5
         | [] \rightarrow (Some (Expr.Lit Null), [])
6
         | [v] \rightarrow (Some v, [])
         | _ \rightarrow failwith "ExclusiveNull: instantiate: too many arguments"
7
8
9
10
     module BaseMemory = PMap (LocationIndex) (Freeable (MList (ExclusiveNull)))
11
     module WISLSubst : NameMap = struct
12
13
       let action_substitutions =
14
15
           ("alloc", "alloc");
           ("dispose", "free");
("setcell", "store");
("getcell", "load");
16
17
18
19
20
21
       let pred_substitutions =
         [ ("cell", "points_to"); ("freed", "freed"); ("bound", "length") ]
22
23
24
25
    module WISLMonadicSMemory = Mapper (WISLSubst) (BaseMemory)
```

This is <u>all</u> the code needed for the instantiation of WISL to match exactly Gillian's original monolithic instantiation.

A consequence of our modular approach to state models is functor-heavy code, with many intermediary modules to allow for further constructions. This can be a bit unweigly if not organised properly, which explains why here we take some intermediary steps, instead of writing down one large and confusing functor application.<sup>4</sup>

We now describe, module by module, how the instantiation of WISL is done. We first override the Ex state model, to modify its default value on instantiation, as WISL requires instantiation to null and the base implementation of Ex initialises it to undefined instead. We then create the intermediary module BaseMemory, equivalent to PMAP(Loc, FREEABLE(LIST(EX))), which corresponds to the bulk of the state model. We finish by applying the substitutions needed to match the target instantiation, via the Mapper module, and voilà! Thanks to our library, what used to be about 800 lines of code is now 25 lines. We point out that our instantiation is also easier to verify the soundness of, as all modules are independent and don't require understanding how the other state model transformers work.

## 3.2.2 JavaScript

Gillian originated from JaVerT [28, 29], which eventually became Gillian-JS, a JavaScript instantiation for Gillian, which targets ECMAScript 5.1 [23]. Its state model is:

$$PMAP(Loc, DYNPMAP(Str, Ex(Val)) \times Ag(Loc))$$

<sup>&</sup>lt;sup>4</sup>It is often not possible to write the whole state model as a single functor application anyways, due to constraints in OCaml abount naming modules – see https://github.com/ocaml/ocaml/issues/6917#issuecomment-626586129.

 $<sup>^5\</sup>mathrm{The}$  List state model transformer is called MList to avoid shadowing OCaml's List module.

Here DynPMap(Str, Ex(Val)) represents objects, and AG(Loc) is the location of the object's metadata, which itself is also an object. An object's metadata is immutable, and thus shareable, explaining the choice of an agreement.

Its implementation with our library is more involved than for WISL; indeed, Gillian-JS doesn't quite line up with the actions and predicates we defined for our state models. These differences mostly stem from different design, meaning most of the work required to fix them is juggling values to fit the right shape.

For instance in Gillian-JS the domainset predicate has one in-value, the domain set, and no out. In our definitions however, the same predicate has no in, and one out, the domain set. To this end, we override consume for domainset, and vanish when the given in doesn't match the out.

```
module MoveInToOut (S : States.MyMonadicSMemory.S) :
      States.MyMonadicSMemory.S with type t = S.t = struct
2
3
      include S
4
5
      let consume pred s ins =
6
        match (pred_to_str pred, ins) with
7
         | "domainset", [ out ] \rightarrow (
            let** s', outs = S.consume pred s [] in
8
9
            match outs with
10
            | [ out' ] 
ightarrow
                if%sat out #== out' then Delayed_result.ok (s', [])
11
                else Delayed.vanish ()
12
13
            | \_ \rightarrow Delayed_result.ok (s', outs))
14
         \mid _ \rightarrow consume pred s ins
15
    end
```

Another difference is that for the DYNPMAP, when accessing a field that didn't exist, Gillian-JS instantiates it to Nono, a special value to represent values that are not actually there. When using the get\_domainset action or consuming the domainset predicate, all keys mapping to a Nono value must thus be removed:

```
module PatchDomainsetObject (S : sig
2
      include MvMonadicSMemory
3
      val is_not_nono : t \rightarrow Expr.t \rightarrow bool
    end) =
5
      Injector
6
         (struct
7
          include DummyInject (S)
8
9
          let post_execute_action a (s, args, rets) =
10
            match (a, rets) with
11
             | "get_domainset", [ Expr.EList dom ] →
12
                 let dom = List.filter (S.is_not_nono s) dom in
                Delayed.return (s, args, [ Expr.EList dom ])
13
14
             \mid _ \rightarrow Delayed.return (s, args, rets)
15
16
          let post_consume p (s, outs) =
17
            match (p, outs) with
18
             | "domainset", [ Expr.ESet dom ] 
ightarrow
19
                let dom = List.filter (S.is_not_nono s) dom in
20
                Delayed.return (s, [ Expr.ESet dom ])
21
              \_ \rightarrow \text{Delayed.return} (s, outs)
22
         end)
23
         (S)
```

This is done via an injection, that receives a state model enhanced with the is\_not\_nono function; we do this instead of hardcoding the lookup to ease the process of instantiating

the JavaScript stack with different optimisations of PMAP that may not have the same state type. An example implementation is then:

```
1 module ObjectBase = struct
2 include PMap (StringIndex) (Exclusive)
3
4 let is_not_nono (h, _) idx =
5 match ExpMap.find_opt idx h with
6 | Some (Some (Expr.Lit Nono)) → false
7 | _ → true
8 end
```

The entire instantiation to JavaScript is 253 lines of code, including all versions of the state model with different optimisations enabled. Most of these lines are used on functor and module definitions, to build up the state one modification at a time. This type of code is hard to read and understand, but this complexity is not caused by the target state model as much as the need for matching the Gillian-JS interface; if the compiler from JavaScript to GIL was reworked to be more aligned with our state model definitions, the instantiation of JavaScript would likely be much simpler and succinct.

#### 3.2.3 C

Gillian-C [57] targets a subset of C, via the CompCert-C verified compiler [40]. Due to the intricacies of the C memory model, its construction uses custom components, not part of the shared library. Its state model is:

$$PMap(Loc, Freeable(BlockTree)) \times CGEnv$$

The first custom state model used is the BLOCKTREE component, which represents chunk of memorys as a binary tree, where each node has a range, a value, and zero or two children. Discussing the behaviour of this state model is out of the scope of this report, and is described in more detail in [2].

The second custom state model is CGENV, the C global environment. Gillian-C handles function pointers and global variables by constructing an "initial data" when parsing the input code, and then passing this data to the state model on instantiation. While this behaviour was hard-coded in Gillian-C, we generalise it, considering simply that the state is a product of the memory and the global environment.

For the implementation of BlockTree and CGEnv, the code from Gillian-C was reused and adapted where necessary to our interface. A particular improvement that was done to it is that in Gillian-C the notion of Freeable and of BlockTree were coupled, meaning any action executed on the BlockTree had to first check if the memory had been freed before proceeding. Decoupling this behaviour was straightforward, and allowed for simpler code in BlockTree.

The bulk of the new code needed for the C instantiation lies in implementing a new action, that was previously hardcoded in the Gillian-C state model: move. This action allows copying some memory from a source into a target location, merging the BLOCKTREES when appropriate. It is quite inelegant to implement, as it cannot rely on the abstraction provided by state models, since it needs to "cross layers" between the state models: it operates on two different indices of the PMAP, but applies a merge operation that only exists for BLOCKTREE. It seems that for such crossing actions, our modular state model approach is not sufficient; one instead needs to implement these actions for a specific instantiation, making the proof work much more intricate.

Omitting this addition done via ActionAdd, the rest of the state model is straightforward to instantiate:

```
module BaseMemory = PMap (LocationIndex) (Freeable (BlockTree))
2
3
    module Wrap (S : C_PMapType) =
4
      Product
5
        (struct
          let id1 = "mem_"
6
7
          let id2 = "genv_"
8
        end)
9
        (ExtendMemory (S))
10
        (CGEnv)
11
    module CMonadicSMemory = Wrap (BaseMemory)
```

Here ExtendMemory is the functor responsible for adding the move action to the state model. We make use of the Wrap functor to avoid code duplication when building the state model with different optimisations of PMAP. The module defined on lines 5-8 specifies the suffixes used to distinguish actions and core predicates from the left and right side of the product.

The instantiation of C with our library is a good example of mixing modular, general use state models with specialised state models; one can still make use of this library while implementing a perhaps more performant or more complex state model for some of the construction.

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# Appendix A

# State Model Definitions

We list here the different state models developed. In particular, we list their components in the following order: the definition of the Resource Algebra, along with its core |-| and composition  $(\cdot)$  rules, the list of actions  $\mathcal{A}$  and core predicates  $\Delta$  it has, the rules for symbolic interpretation  $\models$ , the rules for predicate satisfiability  $\models_{\Delta}$ , the rules for execute\_action on concrete states, the rules for execute\_action on symbolic state, the rules for consume, the rules for produce, and finally the rules for fix.

### A.1 Exclusive

### A.1.1 Resource Algebra

$$\operatorname{Ex}(X) \triangleq \operatorname{ex}(x \colon X)$$
 $|\operatorname{ex}(x)| \triangleq \bot$ 

 $\operatorname{ex}(x_1) \cdot \operatorname{ex}(x_2)$  is always undefined

It defines two actions,  $A = \{load, store\}$  and a predicate  $\Delta = \{ex\}$ .

### A.1.2 Satisfiability

EXPREDSAT EXSYMINTERPRETATION 
$$\frac{\sigma = \operatorname{ex}(x)}{\sigma \models_{\Delta} \langle \operatorname{ex} \rangle([]; [x])} \frac{[\![\hat{x}]\!]_{\theta,s} = x}{\theta, s, \operatorname{ex}(x) \models \operatorname{ex}(\hat{x})}$$

### A.1.3 Compositional Concrete Rules

CEXLOADOK CEXLOADMISS 
$$\begin{aligned} \log(\operatorname{ex}(x),[]) &= (\operatorname{Ok},\operatorname{ex}(x),[x]) & \log(\bot,[]) &= (\operatorname{Miss},\bot,[]) \end{aligned}$$
 CEXSTOREOK CEXSTOREMISS 
$$\begin{aligned} \operatorname{Store}(\operatorname{ex}(x),[x']) &= (\operatorname{Ok},\operatorname{ex}(x'),[]) & \operatorname{store}(\bot,[x']) &= (\operatorname{Miss},\bot,[]) \end{aligned}$$

### A.1.4 Compositional Symbolic Rules

$$\begin{split} &\operatorname{ExLoadMiss} \\ &\operatorname{load}(\operatorname{ex}(\hat{x}), []) \leadsto (\operatorname{Ok}, \operatorname{ex}(\hat{x}), [\hat{x}], []) \\ &\operatorname{ExStoreOk} \\ &\operatorname{ExStoreMiss} \\ &\operatorname{store}(\operatorname{ex}(\hat{x}), [\hat{x}']) \leadsto (\operatorname{Ok}, \operatorname{ex}(\hat{x}'), [], []) \\ &\operatorname{ExConsOk} \\ &\operatorname{ExConsMiss} \\ &\operatorname{consume}(\operatorname{ex}(\hat{x}), \operatorname{ex}, []) \leadsto (\operatorname{Ok}, \bot, [\hat{x}], []) \\ &\operatorname{ExConsMiss} \\ &\operatorname{consume}(\bot, \operatorname{ex}, []) \leadsto (\operatorname{Ok}, \bot, [\hat{x}], []) \\ &\operatorname{ExPROD} \\ &\operatorname{ExFix} \\ &\operatorname{produce}(\bot, \operatorname{ex}, [], [\hat{x}]) \leadsto (\operatorname{ex}(\hat{x}), []) \\ &\operatorname{fix} \ [] = [\{\exists \hat{x} : \langle \operatorname{ex} \rangle (; \hat{x})\}] \end{split}$$

## A.2 Agreement

### A.2.1 Resource Algebra

$$\operatorname{AG}(X) \triangleq \operatorname{ag}(x \colon X)$$
$$|\operatorname{ag}(x)| \triangleq \operatorname{ag}(x)$$
$$\operatorname{ag}(x) \cdot \operatorname{ag}(x') \triangleq \begin{cases} \operatorname{ag}(x) & \text{if } x = x' \\ \text{undefined} & \text{otherwise} \end{cases}$$

It defines  $A = \{load\}$ , and one predicate,  $\Delta = \{ag\}$ .

### A.2.2 Satisfiability

$$\begin{array}{ccc} \operatorname{AgPredSat} & \operatorname{AgSymInterpretation} \\ & \sigma = \operatorname{ag}(x) & & & \|\hat{x}\|_{\theta,s} = x \\ & \sigma \models_{\Delta} \langle \operatorname{ag} \rangle([];[x]) & & \theta, s, \operatorname{ag}(x) \models \operatorname{ag}(\hat{x}) \end{array}$$

### A.2.3 Compositional Concrete Rules

$$\mathsf{load}(\mathsf{ag}(x),[]) = (\mathsf{0k},\mathsf{ag}(x),[x]) \qquad \qquad \mathsf{load}(\bot,[]) = (\mathsf{Miss},\bot,[])$$

### A.2.4 Compositional Symbolic Rules

$$\begin{array}{lll} \operatorname{AGLoadMiss} \\ \operatorname{load}(\operatorname{ag}(\hat{x}), []) \leadsto (\operatorname{Ok}, \operatorname{ag}(\hat{x}), [\hat{x}], []) & \operatorname{load}(\bot, []) \leadsto (\operatorname{Miss}, \bot, [], []) \\ \\ \operatorname{AGConsOk} & \operatorname{AgConsMiss} \\ \operatorname{consume}(\operatorname{ag}(\hat{x}), \operatorname{ag}, []) \leadsto (\operatorname{Ok}, \operatorname{ag}(\hat{x}), [\hat{x}], []) & \operatorname{consume}(\bot, \operatorname{ag}, []) \leadsto (\operatorname{Miss}, \bot, [][]) \\ \\ \operatorname{AGPRODBOT} & \operatorname{AGPRODEQ} \\ \operatorname{produce}(\bot, \operatorname{ag}, [], [\hat{x}]) \leadsto (\operatorname{ag}(\hat{x}), []) & \operatorname{produce}(\operatorname{ag}(\hat{x}), \operatorname{ag}, [], [\hat{x}']) \leadsto (\operatorname{ag}(\hat{x}), [\hat{x} = \hat{x}']) \\ \\ \operatorname{AGFIX} & \\ \operatorname{fix} \ [] = [\{\exists \hat{x}. \, \langle \operatorname{ag} \rangle (; \hat{x})\}] \end{array}$$

### A.3 Fractional

### A.3.1 Resource Algebra

$$\begin{aligned} \operatorname{FRAC}(X) &\triangleq \operatorname{frac}(x \colon X, q \colon (0; 1]) \\ |\operatorname{frac}(x, q)| &\triangleq \bot \\ \operatorname{frac}(x, q) \cdot \operatorname{frac}(x', q') &\triangleq \begin{cases} \operatorname{frac}(x, q + q') & \text{if } x = x' \land q + q' \leq 1 \\ \text{undefined} & \text{otherwise} \end{cases} \end{aligned}$$

We define the actions  $A = \{load, store\}$  and the core predicate  $\Delta = \{frac\}$ .

### A.3.2 Satisfiability

### A.3.3 Compositional Concrete Rules

$$\begin{array}{ll} \text{CFRacLoadOK} & \text{CFRacLoadMiss} \\ & \text{load}(\text{frac}(x,q),[]) = (\text{Ok}, \text{frac}(x,q),[x],[]) & \text{load}(\bot,[]) = (\text{Miss},\bot,[1]) \\ & & & \\ & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & & \\ & & &$$

### A.3.4 Compositional Symbolic Rules

store(frac(x, q), [x']) = (Miss, frac(x, q), [1 - q])

$$\begin{aligned} & \text{FracLoadMiss} \\ & \text{load}(\text{frac}(\hat{x}, \hat{q}), []) \leadsto (\text{Ok}, \text{frac}(\hat{x}, \hat{q}), [\hat{x}], []) & \text{load}(\bot, []) \leadsto (\text{Miss}, \bot, [1], []) \\ & \text{FracStoreOk} \\ & \text{store}(\text{frac}(\hat{x}, \hat{q}), [\hat{x}']) \leadsto (\text{Ok}, \text{frac}(\hat{x}', \hat{q}), [], [\hat{q} = 1]) \\ & \text{FracStorePerm} \\ & \text{store}(\text{frac}(\hat{x}, \hat{q}), [\hat{x}']) \leadsto (\text{Miss}, \text{frac}(\hat{x}, \hat{q}), [1 - \hat{q}], [\hat{q} < 1]) \\ & \text{FracStoreMiss} & \text{FracConsAll} \\ & \text{store}(\bot, [\hat{x}']) \leadsto (\text{Miss}, \bot, [1], []) & \text{consume}(\text{frac}(\hat{x}, \hat{q}), \text{frac}, [\hat{q}']) \leadsto (\text{Ok}, \bot, [\hat{x}], [\hat{q} = \hat{q}']) \\ & \text{FracConsSome} \\ & \text{consume}(\text{frac}(\hat{x}, \hat{q}), \text{frac}, [\hat{q}']) \leadsto (\text{Ok}, \text{frac}(\hat{x}, \hat{q} - \hat{q}'), [\hat{x}], [0 < \hat{q}' < \hat{q}]) \\ & \text{FracConsMiss} \\ & \text{consume}(\text{frac}(\hat{x}, \hat{q}), \text{frac}, [\hat{q}']) \leadsto (\text{Miss}, \text{frac}(\hat{x}, \hat{q}), [\hat{q}' - \hat{q}], [\hat{q} < \hat{q}' \leq 1]) \\ & \text{FracProDBot} \\ & \text{produce}(\bot, \text{frac}, [\hat{q}], [\hat{x}]) \leadsto (\text{frac}(\hat{x}, \hat{q}), [0 < \hat{q} \leq 1]) \\ & \text{FracProDEQ} \\ & \text{produce}(\text{frac}(\hat{x}, \hat{q}), \text{frac}, [\hat{q}'], [\hat{x}']) \leadsto (\text{frac}(\hat{x}, \hat{q} + \hat{q}'), [\hat{x} = \hat{x}' \land 0 < \hat{q}' \land \hat{q} + \hat{q}' \leq 1]) \\ & \text{FracFix} \end{aligned}$$

 $fix [\hat{q}] = [\{\exists \hat{x}. \langle frac \rangle (\hat{q}; \hat{x})\}]$ 

### A.4 Sum

### A.4.1 Resource Algebra

Let  $\Sigma_1$  and  $\Sigma_2$  be the carrier sets of the RAs of  $\mathbb{S}_1$  and  $\mathbb{S}_2$  respectively.

$$\begin{split} \operatorname{Sum}(\mathbb{S}_1,\mathbb{S}_2) &\triangleq \mathbb{S}_1 + \mathbb{S}_2 \triangleq l(\sigma \colon \Sigma_1) \mid r(\sigma \colon \Sigma_2) \\ & l(\sigma) \cdot l(\sigma') \triangleq l(\sigma \cdot \sigma') \\ & r(\sigma) \cdot r(\sigma') \triangleq r(\sigma \cdot \sigma') \\ & |l(\sigma)| \triangleq \begin{cases} \bot & \text{if } |\sigma| = \bot \\ l(|\sigma|) & \text{otherwise} \end{cases} \\ & |r(\sigma)| \triangleq \begin{cases} \bot & \text{if } |\sigma| = \bot \\ r(|\sigma|) & \text{otherwise} \end{cases} \end{split}$$

Given the sum  $\mathbb{S}_1 + \mathbb{S}_2$  such that the actions and predicates of  $\mathbb{S}_1$  and  $\mathbb{S}_2$  are  $\mathcal{A}_1$  and  $\Delta_1$ , and  $\mathcal{A}_2$  and  $\Delta_2$  respectively, the actions of the sum are defined as  $\mathcal{A} = \{\mathsf{L}\alpha : \alpha \in \mathcal{A}_1\} \uplus \{\mathsf{R}\alpha : \alpha \in \mathcal{A}_2\}$ , and the core predicates as  $\Delta = \{\mathsf{L}\delta : \delta \in \Delta_1\} \uplus \{\mathsf{R}\delta : \delta \in \Delta_2\}$ .

### A.4.2 Satisfiability

SUMPREDSATL

$$\frac{\sigma_{l} = l(\sigma) \qquad \sigma \models_{\Delta} \langle \delta \rangle(\vec{v}_{i}; \vec{v}_{o})}{\sigma_{l} \models_{\Delta} \langle \mathsf{L} \delta \rangle(\vec{v}_{i}; \vec{v}_{o})} \qquad \frac{\sigma_{r} = r(\sigma) \qquad \sigma \models_{\Delta} \langle \delta \rangle(\vec{v}_{i}; \vec{v}_{o})}{\sigma_{r} \models_{\Delta} \langle \mathsf{R} \delta \rangle(\vec{v}_{i}; \vec{v}_{o})}$$

$$\frac{\mathsf{SUMPREDSATLBOT}}{\mathsf{L} \models_{\Delta} \langle \delta \rangle(\vec{v}_{i}; \vec{v}_{o})} \qquad \qquad \frac{\mathsf{L} \models_{\Delta} \langle \delta \rangle(\vec{v}_{i}; \vec{v}_{o})}{\mathsf{L} \models_{\Delta} \langle \mathsf{L} \delta \rangle(\vec{v}_{i}; \vec{v}_{o})} \qquad \qquad \frac{\mathsf{L} \models_{\Delta} \langle \delta \rangle(\vec{v}_{i}; \vec{v}_{o})}{\mathsf{L} \models_{\Delta} \langle \mathsf{R} \delta \rangle(\vec{v}_{i}; \vec{v}_{o})}$$

$$\frac{\mathsf{SUMSymInterpretationR}}{\mathsf{L} \mathsf{R} \mathsf{SUMSymInterpretationR}} \qquad \qquad \frac{\sigma_{l} = l(\sigma) \qquad \hat{\sigma}_{l} = l(\hat{\sigma}) \qquad \theta, s, \sigma \models \hat{\sigma}}{\theta, s, \sigma_{l} \models \hat{\sigma}_{l}} \qquad \qquad \theta, s, \sigma_{r} \models \hat{\sigma}_{r}$$

### A.4.3 Compositional Concrete Rules

CSUMLACTION

$$\frac{\sigma = unwrap_l(\sigma_l) \qquad \alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o) \qquad \sigma'_l = wrap_l(\sigma') \qquad o \neq \texttt{Miss}}{\mathsf{L}\alpha(\sigma_l, \vec{v}_i) = (o, \sigma'_l, \vec{v}_o)}$$

CSUMLACTIONMISS

$$\frac{\sigma = \mathit{unwrap}_l(\sigma_l) \quad \alpha(\sigma, \vec{v}_i) = (\mathtt{Miss}, \sigma', \vec{v}_o) \quad \sigma'_l = \mathit{wrap}_l(\sigma')}{\mathtt{L}\alpha(\sigma_l, \vec{v}_i) = (\mathtt{Miss}, \sigma'_l, \text{`1'} :: \vec{v}_o)}$$

CSUMRACTION

$$\frac{\sigma = \mathit{unwrap}_r(\sigma_r) \qquad \alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o) \qquad \sigma'_r = \mathit{wrap}_r(\sigma') \qquad o \neq \mathsf{Miss}}{\mathsf{R}\alpha(\sigma_r, \vec{v}_i) = (o, \sigma'_r, \vec{v}_o)}$$

CSUMRACTIONMISS

$$\frac{\sigma = \mathit{unwrap}_r(\sigma_r) \qquad \alpha(\sigma, \vec{v}_i) = (\mathsf{Miss}, \sigma', \vec{v}_o) \qquad \sigma'_r = \mathit{wrap}_r(\sigma')}{\mathsf{R}\alpha(\sigma_r, \vec{v}_i) = (\mathsf{Miss}, \sigma'_r, `r' :: \vec{v}_o)}$$

CSUMLACTIONINCOMPAT

CSumR.ActionIncompat

$$\mathsf{L}\alpha(r(\sigma),\vec{v}_i) = (\mathsf{Err},r(\sigma),[]) \qquad \qquad \mathsf{R}\alpha(l(\sigma),\vec{v}_i) = (\mathsf{Err},l(\sigma),[])$$

#### A.4.4 Compositional Symbolic Rules

SUMLACTION

$$\frac{\hat{\sigma} = \mathit{unwrap}_l(\hat{\sigma}_l) \qquad \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}_l' = \mathit{wrap}_l(\hat{\sigma}') \qquad o \neq \mathsf{Miss}}{\mathsf{L}\alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_l', \vec{\hat{v}}_o, \pi)}$$

SUMLACTIONMISS

$$\frac{\hat{\sigma} = \mathit{unwrap}_l(\hat{\sigma}_l) \qquad \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}_l' = \mathit{wrap}_l(\hat{\sigma}')}{\mathsf{L}\alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}_l', \text{`1'} :: \vec{\hat{v}}_o, \pi)}$$

SUMRACTION

$$\frac{\hat{\sigma} = \mathit{unwrap}_r(\hat{\sigma}_r) \qquad \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}'_r = \mathit{wrap}_r(\hat{\sigma}') \qquad o \neq \mathsf{Miss}}{\mathsf{R}\alpha(\hat{\sigma}_r, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}'_r, \vec{\hat{v}}_o, \pi)}$$

SumractionMiss

$$\frac{\hat{\sigma} = \mathit{unwrap}_r(\hat{\sigma}_r) \qquad \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}'_r = \mathit{wrap}_r(\hat{\sigma}')}{\mathsf{R}\alpha(\hat{\sigma}_r, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}'_r, \text{`r'} :: \vec{\hat{v}}_o, \pi)}$$

SUMLACTIONINCOMPAT

SUMRACTIONINCOMPAT

$$\mathsf{L}\alpha(r(\hat{\sigma}),\vec{\hat{v}_i}) \leadsto (\mathsf{Err},r(\hat{\sigma}),[],[]) \\ \qquad \mathsf{R}\alpha(l(\hat{\sigma}),\vec{\hat{v}_i}) \leadsto (\mathsf{Err},l(\hat{\sigma}),[],[]) \\$$

SumLCons

$$\frac{\hat{\sigma} = \mathit{unwrap}_l(\hat{\sigma}_l) \qquad \mathsf{consume}(\hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}_l' = \mathit{wrap}_l(\hat{\sigma}') \qquad o \neq \mathsf{Miss}}{\mathsf{consume}(\hat{\sigma}_l, \mathsf{L}\delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_l', \vec{\hat{v}}_o, \pi)}$$

SumLConsMiss

$$\frac{\hat{\sigma} = \mathit{unwrap}_l(\hat{\sigma}_l) \quad \mathsf{consume}(\hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \quad \hat{\sigma}_l' = \mathit{wrap}_l(\hat{\sigma}')}{\mathsf{consume}(\hat{\sigma}_l, \mathsf{L}\delta, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}_l', \text{`1'} :: \vec{\hat{v}}_o, \pi)}$$

SumRCons

$$\frac{\hat{\sigma} = \mathit{unwrap}_r(\hat{\sigma}_r) \qquad \mathsf{consume}(\hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}'_r = \mathit{wrap}_r(\hat{\sigma}') \qquad o \neq \mathsf{Miss}}{\mathsf{consume}(\hat{\sigma}_r, \mathsf{R}\delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}'_r, \vec{\hat{v}}_o, \pi)}$$

SumRConsMiss

$$\frac{\hat{\sigma} = \mathit{unwrap}_r(\hat{\sigma}_r) \qquad \mathsf{consume}(\hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}'_r = \mathit{wrap}_r(\hat{\sigma}')}{\mathsf{consume}(\hat{\sigma}_r, \mathsf{R}\delta, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}'_r, `r' :: \vec{\hat{v}}_o, \pi)}$$

SUMLCONSINCOMPAT

SumRConsIncompat

$$\mathsf{consume}(r(\hat{\sigma}), \mathsf{L}\delta, \vec{\hat{v}_i}) \leadsto (\mathsf{LFail}, r(\hat{\sigma}), [], []) \\ \qquad \mathsf{consume}(l(\hat{\sigma}), \mathsf{R}\delta, \vec{\hat{v}_i}) \leadsto (\mathsf{LFail}, l(\hat{\sigma}), [], []) \\$$

SumLProd

$$\frac{\hat{\sigma} = unwrap_l(\hat{\sigma}_l) \qquad \mathsf{produce}(\hat{\sigma}, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}', \pi) \qquad \hat{\sigma}'_l = wrap_l(\hat{\sigma}')}{\mathsf{produce}(\hat{\sigma}_l, \mathsf{L}\delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}'_l, \pi)}$$

SumRProd

$$\frac{\hat{\sigma} = unwrap_r(\hat{\sigma}_r) \qquad \text{produce}(\hat{\sigma}, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}', \pi) \qquad \hat{\sigma}'_r = wrap_r(\hat{\sigma}')}{\text{produce}(\hat{\sigma}_r, R\delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}'_r, \pi)}$$

$$\begin{array}{ccc} \text{SumLFix} & & \text{SumRFix} \\ & \underline{\mathbb{S}_1.\text{fix } \vec{v}_i = a} & & \underline{\mathbb{S}_2.\text{fix } \vec{v}_i = a} \\ & & & \\ \hline \text{fix '1'} :: \vec{v}_i = a & & & \\ \hline \end{array}$$

### A.5 Product

### A.5.1 Resource Algebra

Let  $\Sigma_1$  and  $\Sigma_2$  be the carrier sets of the RAs of  $\mathbb{S}_1$  and  $\mathbb{S}_2$  respectively.

$$\begin{aligned} \text{Product}(\mathbb{S}_1, \mathbb{S}_2) &\triangleq \mathbb{S}_1 \times \mathbb{S}_2 \triangleq \Sigma_1^? \times \Sigma_2^? \setminus \{(\bot, \bot)\} \\ &(\sigma_l, \sigma_r) \cdot (\sigma_l', \sigma_r') \triangleq (\sigma_l \cdot \sigma_l', \sigma_r \cdot \sigma_r') \\ &|(\sigma_l, \sigma_r)| \triangleq \begin{cases} \bot & \text{if } |\sigma_l| = \bot \wedge |\sigma_r| = \bot \\ (|\sigma_l|, |\sigma_r|) & \text{otherwise} \end{cases} \end{aligned}$$

Given the product  $\mathbb{S}_1 \times \mathbb{S}_2$  such that the actions and predicates of  $\mathbb{S}_1$  and  $\mathbb{S}_2$  are  $\mathcal{A}_1$  and  $\Delta_1$ , and  $\Delta_2$  respectively, the actions of the product are defined as  $\mathcal{A} = \{\mathsf{L}\alpha : \alpha \in \mathcal{A}_1\} \uplus \{\mathsf{R}\alpha : \alpha \in \mathcal{A}_2\}$ , and the core predicates as  $\Delta = \{\mathsf{L}\delta : \delta \in \Delta_1\} \uplus \{\mathsf{R}\delta : \delta \in \Delta_2\}$ .

#### A.5.2 Satisfiability

$$\begin{array}{lll} \text{PRODUCTPREDSATL} & \text{PRODUCTPREDSATR} \\ & \sigma_l \models_\Delta \langle \delta \rangle(\vec{v}_i; \vec{v}_o) & \sigma_r \models_\Delta \langle \delta \rangle(\vec{v}_i; \vec{v}_o) \\ & (\sigma_l, \sigma_r) \models_\Delta \langle \mathsf{L} \delta \rangle(\vec{v}_i; \vec{v}_o) & (\sigma_l, \sigma_r) \models_\Delta \langle \mathsf{R} \delta \rangle(\vec{v}_i; \vec{v}_o) \\ & \text{PRODUCTPREDSATLBOT} & \text{PRODUCTPREDSATRBOT} \\ & \bot \models_\Delta \langle \delta \rangle(\vec{v}_i; \vec{v}_o) & \bot \models_\Delta \langle \mathsf{R} \delta \rangle(\vec{v}_i; \vec{v}_o) \\ & \bot \models_\Delta \langle \mathsf{L} \delta \rangle(\vec{v}_i; \vec{v}_o) & \bot \models_\Delta \langle \mathsf{R} \delta \rangle(\vec{v}_i; \vec{v}_o) \end{array}$$

PRODUCTSYMINTERPRETATION

$$\frac{\theta, s, \sigma_l \models \hat{\sigma}_l \qquad \theta, s, \sigma_r \models \hat{\sigma}_r}{\theta, s, (\sigma_l, \sigma_r) \models (\hat{\sigma}_l, \hat{\sigma}_r)}$$

### A.5.3 Compositional Concrete Rules

Given 
$$wrap(x,y) = \begin{cases} \bot & \text{if } x = \bot \land y = \bot \\ (x,y) & \text{otherwise} \end{cases}$$
 and  $unwrap(s) = \begin{cases} (\bot,\bot) & \text{if } s = \bot \\ (x,y) & \text{otherwise} \end{cases}$ 

CPRODUCTLACTION

$$\frac{(\sigma_l, \sigma_r) = unwrap(\sigma) \qquad \alpha(\sigma_l, \vec{v}_i) = (o, \sigma'_l, \vec{v}_o) \qquad \sigma' = wrap(\sigma'_l, \sigma_r) \qquad o \neq \texttt{Miss}}{\mathsf{L}\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o)}$$

CPRODUCTLACTIONMISS

$$\frac{(\sigma_l,\sigma_r) = \mathit{unwrap}(\sigma) \qquad \alpha(\sigma_l,\vec{v_i}) = (\mathtt{Miss},\sigma_l',\vec{v_o}) \qquad \sigma' = \mathit{wrap}(\sigma_l',\sigma_r)}{\mathtt{L}\alpha(\sigma,\vec{v_i}) = (\mathtt{Miss},\sigma',\text{`1'}::\vec{v_o})}$$

CPRODUCTRACTION

$$\frac{(\sigma_l, \sigma_r) = unwrap(\sigma) \qquad \alpha(\sigma_r, \vec{v}_i) = (o, \sigma'_r, \vec{v}_o) \qquad \sigma' = wrap(\sigma_l, \sigma'_r) \qquad o \neq \texttt{Miss}}{\mathsf{R}\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o)}$$

**CPRODUCTRACTIONMISS** 

$$\frac{(\sigma_l, \sigma_r) = \mathit{unwrap}(\sigma) \qquad \alpha(\sigma_r, \vec{v}_i) = (\mathtt{Miss}, \sigma'_r, \vec{v}_o) \qquad \sigma' = \mathit{wrap}(\sigma_l, \sigma'_r)}{\mathtt{R}\alpha(\sigma, \vec{v}_i) = (\mathtt{Miss}, \sigma', \text{`r'} :: \vec{v}_o)}$$

### A.5.4 Compositional Symbolic Rules

PRODUCTLACTION

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = unwrap(\hat{\sigma}) \qquad \alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}'_l, \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}' = wrap(\hat{\sigma}'_l, \hat{\sigma}_r) \qquad o \neq \texttt{Miss}}{\mathsf{L}\alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi)}$$

PRODUCTLACTIONMISS

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = \mathit{unwrap}(\hat{\sigma}) \qquad \alpha(\hat{\sigma}_l, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}'_l, \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}' = \mathit{wrap}(\hat{\sigma}'_l, \hat{\sigma}_r)}{\mathsf{L}\alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \text{`l'} :: \vec{\hat{v}}_o, \pi)}$$

PRODUCTRACTION

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = unwrap(\hat{\sigma}) \qquad \alpha(\hat{\sigma}_r, \vec{v}_i) \leadsto (o, \hat{\sigma}_r', \vec{v}_o, \pi) \qquad \hat{\sigma}' = wrap(\hat{\sigma}_l, \hat{\sigma}_r') \qquad o \neq \texttt{Miss}}{\mathsf{R}\alpha(\hat{\sigma}, \vec{v}_i) \leadsto (o, \hat{\sigma}', \vec{v}_o, \pi)}$$

PRODUCTRACTIONMISS

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = \mathit{unwrap}(\hat{\sigma}) \qquad \alpha(\hat{\sigma}_r, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}_r', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}' = \mathit{wrap}(\hat{\sigma}_l, \hat{\sigma}_r')}{\mathsf{R}\alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \text{`r'} :: \vec{\hat{v}}_o, \pi)}$$

PRODUCTLCONS

$$\begin{split} (\hat{\sigma}_l, \hat{\sigma}_r) &= unwrap(\hat{\sigma}) \\ \frac{\mathrm{consume}(\hat{\sigma}_l, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_l', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}' = wrap(\hat{\sigma}_l', \hat{\sigma}_r) \qquad o \neq \mathrm{Miss}}{\mathrm{consume}(\hat{\sigma}, \mathrm{L}\delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi)} \end{split}$$

PRODUCTLCONSMISS

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = unwrap(\hat{\sigma}) \qquad \text{consume}(\hat{\sigma}_l, \delta, \vec{\hat{v}}_i) \leadsto (\text{Miss}, \hat{\sigma}_l', \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}' = wrap(\hat{\sigma}_l', \hat{\sigma}_r)}{\text{consume}(\hat{\sigma}, \mathsf{L}\delta, \vec{\hat{v}}_i) \leadsto (\text{Miss}, \hat{\sigma}', \text{`1'} :: \vec{\hat{v}}_o, \pi)}$$

PRODUCTRCONS

$$\begin{split} &(\hat{\sigma}_l,\hat{\sigma}_r) = unwrap(\hat{\sigma})\\ &\underbrace{\mathrm{consume}(\hat{\sigma}_r,\delta,\vec{\hat{v}}_i) \leadsto (o,\hat{\sigma}'_r,\vec{\hat{v}}_o,\pi) \qquad \hat{\sigma}' = wrap(\hat{\sigma}_l,\hat{\sigma}'_r) \qquad o \neq \mathrm{Miss}}_{\mathrm{consume}(\hat{\sigma},\mathrm{R}\delta,\vec{\hat{v}}_i) \leadsto (o,\hat{\sigma}',\vec{\hat{v}}_o,\pi)} \end{split}$$

PRODUCTRCONSMISS

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = \mathit{unwrap}(\hat{\sigma}) \qquad \mathsf{consume}(\hat{\sigma}_r, \delta, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}'_r, \vec{\hat{v}}_o, \pi) \qquad \hat{\sigma}' = \mathit{wrap}(\hat{\sigma}_l, \hat{\sigma}'_r)}{\mathsf{consume}(\hat{\sigma}, \mathsf{R}\delta, \vec{\hat{v}}_i) \leadsto (\mathsf{Miss}, \hat{\sigma}', \text{`r'} :: \vec{\hat{v}}_o, \pi)}$$

$$\mathsf{consume}(\hat{\sigma}, \mathsf{R}\delta, ec{\hat{v}_i}) \leadsto (\mathsf{Miss}, \hat{\sigma}', \mathsf{`r'} :: ec{\hat{v}}_o, \pi)$$

PRODUCTLPROD

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = \mathit{unwrap}(\hat{\sigma}) \qquad \mathsf{produce}(\hat{\sigma}_l, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}'_l, \pi) \qquad \hat{\sigma}' = \mathit{wrap}(\hat{\sigma}'_l, \hat{\sigma}_r)}{\mathsf{produce}(\hat{\sigma}, \mathsf{L}\delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}', \pi)}$$

PRODUCTRPROD

$$\frac{(\hat{\sigma}_l, \hat{\sigma}_r) = \mathit{unwrap}(\hat{\sigma}) \qquad \mathsf{produce}(\hat{\sigma}_r, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}'_r, \pi) \qquad \hat{\sigma}' = \mathit{wrap}(\hat{\sigma}_l, \hat{\sigma}'_r)}{\mathsf{produce}(\hat{\sigma}, \mathsf{R}\delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}', \pi)}$$

$$\begin{array}{ll} \text{ProductLFix} & \text{ProductRFix} \\ \underline{\mathbb{S}_1.\text{fix } \vec{v}_i = a} & \underline{\mathbb{S}_1.\text{fix } \vec{v}_i = a} \\ \text{fix '1'} :: \vec{v}_i = a & \text{fix 'r'} :: \vec{v}_i = a \end{array}$$

#### Freeable **A.6**

#### A.6.1 Resource Algebra

The RA for Freeable(S) is defined as a construction: Freeable(S)  $\triangleq$  S + Ex({freed}), which allows all associated rules to be kept. For clarity, we rename the core predicate R ex (defined by Ex({freed})) as freed. We also extend its actions with the free action.

#### A.6.2 Satisfiability

The symbolic interpretation and core predicate satisfiability are carried by construction via the sum.

#### A.6.3Compositional Concrete Rules

We only define the additional action free, the rest being carried by construction via the sum. The wrapped concrete state model  $\mathbb S$  must provide an is\_exclusively\_owned:  $\Sigma \to \mathbb B$ function such that:

is\_exclusively\_owned 
$$\sigma \iff \nexists .\sigma' . \sigma \# \sigma'$$

It also must provide a fix\_owned:  $\Sigma \to \mathrm{Val}\ \mathrm{list}\ \mathrm{giving}\ \mathrm{values}\ \mathrm{to}\ \mathrm{fix}\ \mathrm{the}\ \mathrm{Miss}.$ 

CFREEABLEACTIONFREE

CFREEABLEACTIONFREEERR

is\_exclusively\_owned  $\sigma$ 

 $\neg$ is\_exclusively\_owned  $\sigma$ 

$$free(l(\sigma), []) = (Ok, r(ex(freed)), [])$$

$$\mathsf{free}(l(\sigma),[]) = (\mathsf{Miss},l(\sigma),\mathsf{fix\_owned}\;x)$$

CFREEABLEACTIONFREEMISS

CFREEABLEACTIONDOUBLEFREE

$$free(\bot,[]) = (Miss, \bot, fix\_owned \bot)$$

$$free(r(ex(freed)), []) \rightsquigarrow (Err, r(ex(freed)), [])$$

### A.6.4 Compositional Symbolic Rules

Similarly, we carry the symbolic rules for execute\_action, consume, produce and fix by construction via the sum. The wrapped symbolic state model  $\mathbb{S}$  must provide an is\_exclusively\_owned:  $\hat{\Sigma} \to \text{LVal}$  function such that:

is\_exclusively\_owned 
$$\hat{\sigma} = \text{true} \iff (\forall \theta, s, \sigma. \theta, s, \sigma \models \hat{\sigma} \implies \nexists.\sigma'. \sigma \# \sigma')$$

It also must provide a fix\_owned:  $\hat{\Sigma}^? \to LVal$  list giving values to fix the Miss.

FREEABLEACTIONFREE

$$free(l(\hat{\sigma}), []) \leadsto (Ok, r(ex(freed)), [], [is\_exclusively\_owned \hat{\sigma}])$$

FREEABLEACTIONFREEERR

$$\mathsf{free}(l(\hat{\sigma}),[]) \leadsto (\mathsf{Miss},l(\hat{\sigma}),\mathsf{fix\_owned}\ \hat{\sigma},[\neg \mathsf{is\_exclusively\_owned}\ \hat{\sigma}])$$

FREEABLEACTIONFREEMISS

FreeableActionDoubleFree

$$\mathsf{free}(\bot,[]) \leadsto (\mathsf{Miss},\bot,\mathsf{fix\_owned}\ \bot,[]) \qquad \mathsf{free}(r(\mathsf{ex}(\mathsf{freed})),[]) \leadsto (\mathsf{Err},r(\mathsf{ex}(\mathsf{freed})),[],[]) \\$$

# A.7 Partial Map

### A.7.1 Resource Algebra

$$\begin{split} \operatorname{PMAP}(I,\mathbb{S}) &\triangleq (I \xrightarrow{fin} \mathbb{S}.\Sigma) \times \mathcal{P}(I)^? \\ (h,d) \cdot (h',d') &\triangleq (h'',d'') \\ \text{where } h'' &\triangleq \lambda i. \begin{cases} h(i) \cdot h'(i) & \text{if } i \in \operatorname{dom}(h) \cap \operatorname{dom}(h') \\ h(i) & \text{if } i \in \operatorname{dom}(h) \setminus \operatorname{dom}(h') \\ h'(i) & \text{if } i \in \operatorname{dom}(h') \setminus \operatorname{dom}(h) \\ \text{undefined} & \text{otherwise} \\ \end{cases} \end{split}$$

$$\operatorname{and} d'' \triangleq \begin{cases} d & \text{if } d' = \bot \\ d' & \text{if } d = \bot \\ \text{undefined otherwise} \end{cases}$$

$$\operatorname{and} d'' = \bot \vee \operatorname{dom}(h'') \subseteq d''$$

$$|(h,d)| \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(h') = \emptyset \\ (h',\bot) & \text{otherwise} \end{cases}$$

$$\operatorname{where} h' \triangleq \lambda i. \begin{cases} |h(i)| & \text{if } i \in \operatorname{dom}(h) \wedge |h(i)| \neq \bot \\ \text{undefined otherwise} \end{cases}$$

For a wrapped state model  $\mathbb S$  such that its actions and predicates are  $\mathcal A_{\mathbb S}$  and  $\Delta_{\mathbb S}$ , we define the actions of  $\mathrm{PMAP}(I,\mathbb S)$  as  $\mathcal A=\mathcal A_{\mathbb S}\uplus\{\mathtt{alloc}\}$ , and the predicates  $\Delta=\Delta_{\mathbb S}\uplus\{\mathtt{domainset}\}$ .

### A.7.2 Satisfiability

$$\begin{array}{ll} \operatorname{PMAPPREDSAT} & \operatorname{PMAPPREDSATBOT} \\ \sigma_i \models_\Delta \langle \delta \rangle(\vec{v}_i; \vec{v}_o) & \frac{\bot \models_\Delta \langle \delta \rangle(\vec{v}_i; \vec{v}_o)}{\bot \models_\Delta \langle \delta \rangle(i :: \vec{v}_i; \vec{v}_o)} & \frac{(\emptyset, d) \models_\Delta \langle \operatorname{domainset} \rangle(; d)}{\bot \models_\Delta \langle \delta \rangle(i :: \vec{v}_i; \vec{v}_o)} \\ \\ \operatorname{PMAPSYMINTERPRETATION} \\ \forall \hat{i} \in \operatorname{dom}(\hat{h}). \ \llbracket \hat{i} \rrbracket_{\theta,s} = i \wedge i \in \operatorname{dom}(h) \wedge \theta, s, h(i) \models \hat{h}(\hat{i}) \\ \\ \underline{ \llbracket \operatorname{dom}(\hat{h}) \rrbracket_{\theta,s} = \operatorname{dom}(h) } & \llbracket \hat{d} \rrbracket_{\theta,s} = d \\ \\ \theta, s, (h, d) \models (\hat{h}, \hat{d}) \\ \end{array}$$

### A.7.3 Compositional Concrete Rules

We define the helper functions get and set as:

$$\begin{split} & \mathsf{get} \colon \mathrm{PMap}(I,\mathbb{S}) \to I \to \mathbb{S}.\Sigma \\ & \mathsf{set} \colon \mathrm{PMap}(I,\mathbb{S}) \to I \to \mathbb{S}.\Sigma \to \mathrm{PMap}(I,\mathbb{S}) \end{split}$$

We pretty-print get and set as  $get(\sigma, i) = \sigma_i$  and  $set(\sigma, i, \sigma_i) = \sigma'$ .

Given 
$$wrap(h, d) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(h) = \emptyset \land d = \bot \\ (h, d) & \text{otherwise} \end{cases}$$

$$unwrap(\sigma) \triangleq \begin{cases} ([], \bot) & \text{if } \sigma = \bot \\ (h, d) & \text{if } \sigma = (h, d) \end{cases}$$

$$\begin{array}{c} \text{CPMapGetMatch} \\ \underline{(h,d) = unwrap(\sigma)} \quad i \in \text{dom}(h) \quad \sigma_i = h(i) \\ \\ \text{get}(\sigma,i) = \sigma_i \\ \\ \hline \\ \text{CPMapGetAdd} \\ \underline{(h,d) = unwrap(\sigma)} \quad i \notin \text{dom}(h) \quad d \neq \bot \quad i \in d \\ \\ \text{get}(\sigma,i) = \bot \\ \\ \hline \\ \text{CPMapGetBotDomain} \\ \underline{(h,d) = unwrap(\sigma)} \quad i \notin \text{dom}(h) \quad d = \bot \\ \\ \text{get}(\sigma,i) = \bot \\ \\ \hline \\ \text{CPMapSetSome} \\ \underline{(h,d) = unwrap(\sigma)} \quad \sigma_i \neq \bot \quad h' = h[i \leftarrow \sigma_i] \quad \sigma' = wrap(h',d) \\ \\ \text{set}(\sigma,i,\sigma_i) = \sigma' \\ \\ \hline \\ \text{CPMapSetNone} \\ \underline{(h,d) = unwrap(\sigma)} \quad \sigma_i = \bot \quad h' = h[i \not\leftarrow] \quad \sigma' = wrap(h',d) \\ \\ \text{set}(\sigma,i,\sigma_i) = \sigma' \\ \\ \hline \\ \text{The action rules are then:} \\ \hline \\ \text{CPMapAction} \\ \underline{\text{get}(\sigma,i) = \sigma_i \quad \alpha(\sigma_i,\vec{v}_i) = (\sigma,\sigma'_i,\vec{v}_\sigma) \quad \text{set}(\sigma,i,\sigma'_i) = \sigma'} \\ \\ \alpha(\sigma,i::\vec{v}_i) = (\sigma,\sigma',i::\vec{v}_\sigma) \\ \hline \\ \text{CPMapActionOutOebounds} \\ \underline{d \neq \bot \quad i \notin d} \\ \hline \\ \alpha((h,d),i::\vec{v}_i) = (\text{Err},(h,d),[]) \\ \hline \\ \text{CPMapAlloc} \\ \underline{d \neq \bot \quad i = \text{fresh} \quad \sigma_i = \text{instantiate}(\vec{v}_i) \quad h' = h[i \leftarrow \sigma_i] \quad d' = d \uplus \{i\} \\ \hline \\ \text{alloc}((h,d),\vec{v}_i) = (\text{Ok},(h',d'),[i]) \\ \hline \\ \text{CPMapAllocMiss} \\ \hline \end{array}$$

### A.7.4 Compositional Symbolic Rules

We now redefine get and set, lifting them to the symbolic realm (note set is unchanged, as it does not perform any sort of matching). get now returns a state and an index, that may be different from the input index – it corresponds to the actual index of the state in the map.

 $(h,d) = unwrap(\sigma)$   $d = \bot$ 

 $alloc(\sigma, \vec{v}_i) = (Miss, \sigma, ['domainset'])$ 

get: 
$$PMAP(I, \mathbb{S}) \to I \to \mathcal{P}(I \times \mathbb{S}.\hat{\Sigma} \times \Pi)$$
  
set:  $PMAP(I, \mathbb{S}) \to I \to \mathbb{S}.\hat{\Sigma} \to PMAP(I, \mathbb{S})$ 

We pretty-print get and set as  $\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi)$  and  $\text{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'$ .

$$\begin{aligned} \text{Given } wrap(\hat{h}, \hat{d}) &\triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(\hat{h}) = \emptyset \wedge \hat{d} = \bot \\ (\hat{h}, \hat{d}) & \text{otherwise} \end{cases} \\ unwrap(\hat{\sigma}) &\triangleq \begin{cases} ([], \bot) & \text{if } \hat{\sigma} = \bot \\ (\hat{h}, \hat{d}) & \text{if } \hat{\sigma} = (\hat{h}, \hat{d}) \end{cases} \end{aligned}$$

РМАРСЕТМАТСН

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i}' \in \text{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i}')}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])}$$

PMapGetAdd

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \notin \text{dom}(\hat{h}) \qquad \hat{d} \neq \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}) \land \hat{i} \in \hat{d}])}$$

PMAPGETBOTDOMAIN

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \notin \text{dom}(\hat{h}) \qquad \hat{d} = \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h})])}$$

PMAPSETSOME

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{\sigma}_i \neq \bot \qquad \hat{h}' = \hat{h}[\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\text{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

PMapSetNone

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{\sigma}_i = \bot \qquad \hat{h}' = \hat{h}[\hat{i} \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\operatorname{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

And now for the relevant rules:

$$\text{Given } \textit{lift\_if\_miss}(o, \hat{i}, \vec{\hat{v}_i}) \triangleq \begin{cases} \hat{i} :: \vec{\hat{v}_i} & \text{if } o = \texttt{Miss} \\ \vec{\hat{v}_i} & \text{otherwise} \end{cases}$$

PMAPACTION

$$\underbrace{ \frac{ \operatorname{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \qquad \alpha(\hat{\sigma}_i, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}'_i, \vec{\hat{v}}_o, \pi') \qquad \operatorname{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}'_i) = \hat{\sigma}'}_{\alpha(\hat{\sigma}, \hat{i} :: \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \hat{i}' :: \vec{\hat{v}}_o, \pi :: \pi') }$$

**PMAPACTIONOUTOFBOUNDS** 

$$\frac{\hat{d} \neq \bot}{\alpha((\hat{h}, \hat{d}), \hat{i} :: \vec{\hat{v_i}}) \leadsto (\mathsf{Err}, (\hat{h}, \hat{d}), [], [\hat{i} \notin \hat{d}])}$$

PMapAlloc

$$\frac{\hat{d} \neq \bot \qquad \hat{i} = \mathsf{fresh} \qquad \hat{\sigma}_i = \mathsf{instantiate}(\vec{v}_i) \qquad \hat{h}' = \hat{h}[\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{d}' = \hat{d} \uplus \{\hat{i}\}}{\mathsf{alloc}((\hat{h}, \hat{d}), \vec{v}_i) \leadsto (\mathsf{Ok}, (\hat{h}', \hat{d}'), [\hat{i}], [\hat{i} = \hat{i}])}$$

PMAPALLOCMISS

$$(\hat{h}, \hat{d}) = unwrap(\hat{\sigma})$$
  $\hat{d} = \bot$ 

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{d} = \bot}{\text{alloc}(\hat{\sigma},\vec{\hat{v}_i}) \leadsto (\texttt{Miss},\hat{\sigma},[\text{'domainset'}],[])}$$

**PMapCons** 

$$\operatorname{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi)$$

$$\frac{\mathrm{consume}(\hat{\sigma}_i, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_i', \vec{\hat{v}}_o, \pi') \qquad \mathrm{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}' \qquad \vec{\hat{v}}_o' = \mathit{lift\_if\_miss}(o, \hat{i}', \vec{\hat{v}}_o)}{\mathrm{consume}(\hat{\sigma}, \delta, \hat{i} :: \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o', \pi :: \pi')}$$

$$\mathsf{consume}(\hat{\sigma}, \delta, \hat{i} :: \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o', \pi :: \pi')$$

PMAPCONSINCOMPAT

$$\hat{d} \neq \bot$$

$$consume((\hat{h}, \hat{d}), \delta, \hat{i} :: \vec{\hat{v}_i}) \leadsto (LFail, (\hat{h}, \hat{d}), [], [\hat{i} \notin \hat{d}])$$

PMapConsDomainSet

$$\hat{d} \neq \bot$$

$$\frac{\hat{d} \neq \bot}{\mathsf{consume}((\hat{h}, \hat{d}), \mathsf{domainset}, []) \leadsto (\mathsf{Ok}, (\hat{h}, \bot), [\hat{d}], [])}$$

PMapConsDomainSetMiss

ODMAINSETMISS
$$(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{d} = \bot$$

$$\mathsf{consume}(\hat{\sigma}, \mathsf{domainset}, []) \leadsto (\mathsf{Miss}, \hat{\sigma}, [\text{`domainset'}], [])$$

PMapProd

$$\frac{\text{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i}',\hat{\sigma}_i,\pi) \qquad \text{produce}(\hat{\sigma}_i,\delta,\vec{\hat{v}}_i,\vec{\hat{v}}_o) \leadsto (\hat{\sigma}_i',\pi') \qquad \text{set}(\hat{\sigma},\hat{i}',\hat{\sigma}_i') = \hat{\sigma}'}{\text{produce}(\hat{\sigma},\delta,\hat{i}::\vec{\hat{v}}_i,\vec{\hat{v}}_o) \leadsto (\hat{\sigma}',\pi::\pi')}$$

PMapProdDomainSet

$$(\hat{\alpha}) = u n u v a n (\hat{\alpha})$$

$$\frac{(\hat{h},\bot) = \textit{unwrap}(\hat{\sigma})}{\mathsf{produce}(\hat{\sigma},\mathsf{domainset},[],[\hat{d}]) \leadsto ((\hat{h},\hat{d}),[\mathsf{dom}(\hat{h}) \subseteq \hat{d}])} \qquad \underbrace{\begin{array}{c} \mathsf{S.fix} \ \hat{v}_i = a \\ & a' = \textit{lift}(a,\hat{i}) \end{array}}_{}$$

PMAPFIXDOMAINSET

fix ['domainset'] = 
$$\exists \hat{d}$$
.  $\langle domainset \rangle (; \hat{d})$ 

Note in the above we define  $lift(a,\hat{i})$  as a function that traverses an assertion a and lifts all core predicate assertions by adding the value  $\hat{i}$  at the start of its in-values, such that  $lift(\langle \delta \rangle(\vec{v}_i;\vec{v}_o),\hat{i}) = \langle \delta \rangle(\hat{i}::\vec{v}_i;\vec{v}_o).$ 

## A.8 Syntactic PMap

For the variation of PMAP that has syntactic checking, we only need to modify the rules for the symbolic version of get.

$$\begin{split} & \text{SyntacticPMapGetMatch} \\ & \underline{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \in \text{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i})} \\ & \text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{\sigma}_i, []) \end{split}$$

SYNTACTICPMAPGETBRANCH

$$\frac{(\hat{h}, \hat{d}) = \mathit{unwrap}(\hat{\sigma}) \qquad \hat{i} \notin \mathrm{dom}(\hat{h}) \qquad \hat{i}' \in \mathrm{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i}')}{\mathrm{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])}$$

SYNTACTICPMAPGETADD

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \not\in \text{dom}(\hat{h}) \qquad \hat{d} \neq \bot}{\text{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i},\bot,[\hat{i} \not\in \text{dom}(\hat{h}) \land \hat{i} \in \hat{d}])}$$

SYNTACTICPMAPGETBOTDOMAIN

$$\frac{(\hat{h},\hat{d}) = \mathit{unwrap}(\hat{\sigma}) \qquad \hat{i} \not\in \mathrm{dom}(\hat{h}) \qquad \hat{d} = \bot}{\mathtt{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i},\bot,[\hat{i} \not\in \mathrm{dom}(\hat{h})])}$$

# A.9 Split PMap

#### A.9.1 Symbolic States

Because we re-use the concrete RA from PMAP, we only need to define the set of symbolic states, which does not come equipped with composition or core.

$$\mathrm{PMap}_{\mathrm{Split}}(I,\mathbb{S}) \triangleq I \xrightarrow{\mathit{fin}} \mathbb{S}.\Sigma \times I \xrightarrow{\mathit{fin}} \mathbb{S}.\Sigma \times \mathcal{P}(I)^?$$

### A.9.2 Satisfiability

Predicate satisfiability is defined with regards to the concrete compositional states, so we do not need to redefine it. We however need to redefine symbolic interpretation, as follows.

We denote  $\hat{h}_c$  and  $\hat{h}_s$  the concrete and symbolic parts of the heap respectively.

SPLITPMAPSYMINTERPRETATION

$$\forall \hat{i} \in \operatorname{dom}(\hat{h}_c). \ (\nexists \hat{i}' \in \hat{h}_s. \llbracket \hat{i}' \rrbracket_{\theta,s} = \llbracket \hat{i} \rrbracket_{\theta,s}) \wedge \llbracket \hat{i} \rrbracket_{\theta,s} = i \wedge i \in \operatorname{dom}(h) \wedge \theta, s, h(i) \models \hat{h}_c(\hat{i}) \\ \forall \hat{i} \in \operatorname{dom}(\hat{h}_s). \ (\nexists \hat{i}' \in \hat{h}_c. \llbracket \hat{i}' \rrbracket_{\theta,s} = \llbracket \hat{i} \rrbracket_{\theta,s}) \wedge \llbracket \hat{i} \rrbracket_{\theta,s} = i \wedge i \in \operatorname{dom}(h) \wedge \theta, s, h(i) \models \hat{h}_s(\hat{i}) \\ \underline{\llbracket \operatorname{dom}(\hat{h}_c) \uplus \operatorname{dom}(\hat{h}_s) \rrbracket_{\theta,s} = \operatorname{dom}(h)} \qquad \llbracket \hat{d} \rrbracket_{\theta,s} = d \\ \theta, s, (h, d) \models (\hat{h}_c, \hat{h}_s, \hat{d})$$

#### A.9.3get and set rules

Given 
$$wrap(\hat{h}_c, \hat{h}_s, \hat{d}) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(\hat{h}_c) = \emptyset \wedge \operatorname{dom}(\hat{h}_s) = \emptyset \wedge \hat{d} = \emptyset \\ (\hat{h}_c, \hat{h}_s, \hat{d}) & \text{otherwise} \end{cases}$$

$$unwrap(\hat{\sigma}) \triangleq \begin{cases} ([], [], \emptyset) & \text{if } \hat{\sigma} = \bot \\ (\hat{h}_c, \hat{h}_s, \hat{d}) & \text{if } \hat{\sigma} = (\hat{h}_c, \hat{h}_s, \hat{d}) \end{cases}$$

SPLITPMAPGETMATCHCON

$$\frac{(\hat{h}_c,\hat{h}_s,\hat{d}) = unwrap(\hat{\sigma}) \qquad \text{is\_concrete } \hat{i} \qquad \hat{i} \in \text{dom}(\hat{h}_c) \qquad \hat{\sigma}_i = \hat{h}_c(\hat{i})}{}$$

$$\mathtt{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i},\hat{\sigma}_i,[])$$

SPLITPMAPGETMATCHSYM

$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \in \text{dom}(\hat{h}_s) \qquad \hat{\sigma}_i = \hat{h}_s(i)}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{\sigma}_i, \lceil \mid)}$$

SPLITPMAPGETBRANCH

$$\begin{split} (\hat{h}_c, \hat{h}_s, \hat{d}) &= \mathit{unwrap}(\hat{\sigma}) \\ \frac{\hat{h}_{\mathit{all}} = \hat{h}_c \cup \hat{h}_s \qquad \hat{i} \notin \mathit{dom}(\hat{h}_{\mathit{all}}) \qquad \hat{i}' \in \mathit{dom}(\hat{h}_{\mathit{all}}) \qquad \hat{\sigma}_i = \hat{h}_{\mathit{all}}(\hat{i}')}{\mathsf{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])} \end{split}$$

SplitPMapGetAdd

$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \qquad \hat{i} \notin \text{dom}(\hat{h}_{all}) \qquad \hat{d} \neq \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}_{all}) \land \hat{i} \in \hat{d}])}$$

SPLITPMAPGETBOTDOMAIN

SplitPMapGetBotDomain 
$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \qquad \hat{i} \notin \text{dom}(\hat{h}_{all}) \qquad \hat{d} = \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}_{all})])}$$

SPLITPMAPSETSOMECON

$$\begin{split} \hat{(h_c,\hat{h}_s,\hat{d})} &= unwrap(\hat{\sigma}) & \hat{\sigma}_i \neq \bot \quad \text{is\_concrete}_{\Sigma} \; \hat{\sigma}_i \\ \\ \underline{\text{is\_concrete} \; \hat{i} \quad \; \hat{h}'_c = \hat{h}_c[\hat{i} \leftarrow \hat{\sigma}_i] \quad \; \; \hat{h}'_s = \hat{h}_s[\hat{i} \not\leftarrow] \quad \; \; \hat{\sigma}' = wrap(\hat{h}'_c,\hat{h}'_s,\hat{d})} \\ \\ \underline{\text{set}(\hat{\sigma},\hat{i},\hat{\sigma}_i) = \hat{\sigma}'} \end{split}$$

SPLITPMAPSETSOMESYM

$$\begin{split} (\hat{h}_c, \hat{h}_s, \hat{d}) &= unwrap(\hat{\sigma}) & \qquad \hat{s}_i \neq \bot \qquad \neg (\texttt{is\_concrete}_\Sigma \ \hat{\sigma}_i \lor \texttt{is\_concrete} \ \hat{i}) \\ & \qquad \qquad \hat{h}'_c = \hat{h}_c[\hat{i} \not\leftarrow] & \qquad \hat{h}'_s = \hat{h}_s[\hat{i} \leftarrow \hat{\sigma}_i] & \qquad \hat{\sigma}' = wrap(\hat{h}'_c, \hat{h}'_s, \hat{d}) \\ & \qquad \qquad \texttt{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}' \end{split}$$

SplitPMapSetNone

$$\begin{split} \hat{(h_c,\hat{h}_s,\hat{d})} &= unwrap(\hat{\sigma}) \\ \frac{\hat{\sigma}_i = \bot \qquad \hat{h}_c' = \hat{h}_c[\hat{i} \not\leftarrow] \qquad \hat{h}_s' = \hat{h}_s[\hat{i} \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}_c',\hat{h}_s',\hat{d})}{\text{set}(\hat{\sigma},\hat{i},\hat{\sigma}_i) = \hat{\sigma}'} \end{split}$$

## A.10 Abstract Locations PMap

### A.10.1 Symbolic States

$$PMAP_{ALoc}(S) \triangleq Str \xrightarrow{fin} S.\Sigma \times \mathcal{P}(LVal)^?$$

### A.10.2 Satisfiability

ALocPMapSymInterpretation

$$\forall a \in \operatorname{dom}(\hat{h}). \, [\![\operatorname{aloc}(a)]\!]_{\theta,s} = i \wedge i \in \operatorname{dom}(h) \wedge \theta, s, h(i) \models \hat{h}(a)$$
$$\underbrace{ [\![\{\operatorname{aloc}(a) : a \in \operatorname{dom}(\hat{h})\}]\!]_{\theta,s} = \operatorname{dom}(h) }_{\theta,s,(h,d) \models (\hat{h},\hat{d})}$$

### A.10.3 get and set rules

Here we use the matching mode  $m \in \{MATCH, NO\_MATCH\}$ .

Given 
$$wrap(\hat{h}, \hat{d}) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(\hat{h}) = \emptyset \land \hat{d} = \bot \\ (\hat{h}, \hat{d}) & \text{otherwise} \end{cases}$$
 
$$unwrap(\hat{\sigma}) \triangleq \begin{cases} ([], \bot) & \text{if } \hat{\sigma} = \bot \\ (\hat{h}, \hat{d}) & \text{if } \hat{\sigma} = (\hat{h}, \hat{d}) \end{cases}$$
 
$$\hat{i} \in \hat{d} \triangleq \begin{cases} \text{true} & \text{if } \hat{d} = \bot \\ \hat{i} \in \hat{d} & \text{otherwise} \end{cases}$$

ALOCPMAPGETMATCH

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \; \hat{i} \qquad a \neq \bot \qquad a \in \mathsf{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(a)}{\mathsf{get}(\hat{\sigma}. \hat{i}, m) \leadsto (\mathsf{aloc}(a), \hat{\sigma}_i, \Vert)}$$

ALOCPMAPGETNOMATCHNOTFOUND

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \ \hat{i} \qquad a \neq \bot \qquad a \notin \mathsf{dom}(\hat{h})}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{NO\_MATCH}) \leadsto (\mathsf{aloc}(a),\bot,[\hat{i} \in ^? \hat{d}])}$$

ALocPMapGetNoMatchNew

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \text{to\_aloc } \hat{i} \qquad a = \bot \qquad a' = \text{fresh\_aloc ()}}{\text{get}(\hat{\sigma},\hat{i},\text{NO\_MATCH}) \leadsto (\text{aloc}(a'),\bot,[\hat{i} = \text{aloc}(a') \land \hat{i} \in \hat{d}])}$$

ALocPMapGetMatchNotFound

$$\frac{(\hat{h},\hat{d}) = \mathit{unwrap}(\hat{\sigma}) \quad a = \mathsf{to\_aloc} \; \hat{i} \quad a \neq \bot \quad a \notin \mathsf{dom}(\hat{h})}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{MATCH}) \leadsto (\mathsf{aloc}(a),\bot,[\hat{i} \notin \{\mathsf{aloc}(a'): a' \in \mathsf{dom}(\hat{h})\} \land \hat{i} \in \hat{}^? \; \hat{d}])}$$

ALocPMapGetMatchNew

$$\frac{(\hat{h},\hat{d}) = \mathit{unwrap}(\hat{\sigma}) \quad a = \mathsf{to\_aloc} \; \hat{i} \quad a = \bot \quad a' = \mathsf{fresh\_aloc} \; ()}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{MATCH}) \leadsto (\mathsf{aloc}(a'),\bot,[\hat{i} = \mathsf{aloc}(a') \land \hat{i} \notin \{\mathsf{aloc}(a'') : a'' \in \mathsf{dom}(\hat{h})\} \land \hat{i} \in \hat{d}])}$$

ALOCPMAPMATCHING

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma})}{a = \mathsf{to\_aloc} \; \hat{i} \qquad a \neq \bot \qquad a \notin \mathrm{dom}(\hat{h}) \qquad a' \in \mathrm{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(a')}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{MATCH}) \leadsto (\mathrm{aloc}(a'),\hat{\sigma}_i,[\mathrm{aloc}(a) = \mathrm{aloc}(a')])}$$

ALOCPMAPMATCHINGBOT

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \ \hat{i} \qquad a = \bot \qquad a' \in \mathrm{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(a')}{\mathsf{get}(\hat{\sigma}, \hat{i}, \mathsf{MATCH}) \leadsto (\mathrm{aloc}(a'), \hat{\sigma}_i, [\hat{i} = \mathrm{aloc}(a')])}$$

ALOCPMAPSETSOME

$$\frac{a = \mathsf{to\_aloc} \; \hat{i} \qquad a \neq \bot \qquad \hat{\sigma}_i \neq \bot \qquad \hat{h}' = \hat{h}[a \leftarrow \hat{\sigma}_i] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\mathsf{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

ALOCPMAPSETNONE

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma})}{a = \mathsf{to\_aloc}\; \hat{i} \qquad a \neq \bot \qquad \hat{\sigma}_i = \bot \qquad \hat{h}' = \hat{h}[a \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}',\hat{d})}{\mathsf{set}(\hat{\sigma},\hat{i},\hat{\sigma}_i) = \hat{\sigma}'}$$

# Appendix B

# Proofs of soundness

### B.1 Exclusive

### B.1.1 Resource Algebra

$$\operatorname{Ex}(X) \triangleq \operatorname{ex}(x:X)$$

$$|\operatorname{ex}(x)| \triangleq \bot$$

$$\operatorname{ex}(x_1) \cdot \operatorname{ex}(x_2) \text{ is always undefined}$$

We define the actions of the state model as  $\mathcal{A} = \{\text{load}, \text{store}\}$ , and the predicates  $\Delta = \{\text{ex}\}$ . We define predicate satisfiability and symbolic interpretation as:

EXPREDSAT EXSYMINTERPRETATION 
$$\frac{\sigma = \operatorname{ex}(x)}{\sigma \models_{\Delta} \langle \operatorname{ex} \rangle([]; [x])} \frac{\llbracket \hat{x} \rrbracket_{\theta,s} = x}{\theta, s, \operatorname{ex}(x) \models \operatorname{ex}(\hat{x})}$$

### **B.1.2** Compositional Concrete Rules

CEXLOADOK CEXLOADMISS 
$$\begin{aligned} \log(\operatorname{ex}(x),[]) &= (\operatorname{Ok},\operatorname{ex}(x),[x]) & \log(\bot,[]) \leadsto (\operatorname{Miss},\bot,[]) \end{aligned}$$
 CEXSTOREOK CEXSTOREMISS 
$$\begin{aligned} \operatorname{Store}(\operatorname{ex}(x),[x']) \leadsto (\operatorname{Ok},\operatorname{ex}(x'),[]) & \operatorname{store}(\bot,[x']) \leadsto (\operatorname{Miss},\bot,[]) \end{aligned}$$

### **B.1.3** Compositional Symbolic Rules

$$\begin{split} &\operatorname{ExLoadMiss} \\ &\operatorname{load}(\operatorname{ex}(\hat{x}),[]) \leadsto (\operatorname{Ok},\operatorname{ex}(\hat{x}),[\hat{x}],[]) \\ &\operatorname{ExStoreOk} \\ &\operatorname{ExStoreMiss} \\ &\operatorname{store}(\operatorname{ex}(\hat{x}),[\hat{x}']) \leadsto (\operatorname{Ok},\operatorname{ex}(\hat{x}'),[],[]) \\ &\operatorname{ExConsOk} \\ &\operatorname{ExConsMiss} \\ &\operatorname{consume}(\operatorname{ex}(\hat{x}),\operatorname{ex},[]) \leadsto (\operatorname{Ok},\bot,[\hat{x}],[]) \\ &\operatorname{ExConsume}(\bot,\operatorname{ex},[]) \leadsto (\operatorname{Miss},\bot,[],[]) \\ &\operatorname{ExProd} \\ &\operatorname{ExFix} \\ &\operatorname{produce}(\bot,\operatorname{ex},[],[\hat{x}]) \leadsto (\operatorname{ex}(\hat{x}),[]) \\ &\operatorname{fix}\ [] = [\{\exists \hat{x}.\langle\operatorname{ex}\rangle(;\hat{x})\}] \end{split}$$

#### **B.1.4** Soundness Proofs

Proof.

#### Proposition: OX Soudness

Assume

(H1) 
$$\theta, s, \sigma \models \hat{\sigma} \land \alpha(\sigma, \vec{v_i}) = (o, \sigma', \vec{v_o}) \land [\![\hat{v_i}]\!]_{\theta, s} = \vec{v_i}$$

$$\textbf{(H2)} \ \forall o', \hat{\sigma}', \vec{\hat{v}}_o', \pi'. \ \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (o', \hat{\sigma}', \vec{\hat{v}}_o', \pi') \Rightarrow o' \in \{\texttt{Ok}, \texttt{Err}\}$$

To prove

(G1) 
$$\exists \hat{\sigma}', \hat{v}_o, \pi, \theta'. \hat{\alpha}(\hat{\sigma}, \hat{v}_i) \rightsquigarrow (o, \hat{\sigma}', \hat{v}_o, \pi) \land \theta', s, \sigma' \models \hat{\sigma}' \land \mathsf{SAT}_{\theta', s}(\pi) \land \llbracket \hat{v}_o \rrbracket_{\theta', s} = \vec{v}_o$$

The proof is analogous for both actions, so we only consider the case where  $\alpha = \mathsf{load}$ . Given (H1) and the definition of  $\models$ , there are two further cases:  $\sigma = \mathsf{ex}(x)$  and  $\hat{\sigma} = \mathsf{ex}(\hat{x})$ , or  $\sigma = \bot$  and  $\hat{\sigma} = \bot$ . Again, both cases are analogous, so we only consider  $\sigma = \mathsf{ex}(x)$ ,  $\hat{\sigma} = \mathsf{ex}(\hat{x})$ .

- **(H3)** From our hypothesis,  $\sigma = ex(x)$  and  $\hat{\sigma} = ex(\hat{x})$
- (H4) From the definition of concrete actions CExLoadok, we get  $\vec{v}_i = [], o = 0k, \sigma' = \sigma,$   $\vec{v}_o = [x].$
- **(H5)** From the definition of  $\models$  we also have  $[\![\hat{x}]\!]_{\theta,s} = x$
- **(H6)** We can pick  $\vec{v}_o$ ,  $\pi$  and  $\theta'$  such that  $\vec{v}_o = [\hat{x}]$ ,  $\pi = []$  and  $\theta' = \theta$ .
- **(H7)** From (H6), we get  $[\hat{v}_o]_{\theta,s} = \vec{v}_o$ , as well as  $SAT_{\theta,s}(\pi)$
- **(H8)** From EXLOADOK and (H6), we have  $load(\hat{\sigma}, \vec{\hat{v}}_i) \rightsquigarrow (Ok, \hat{\sigma}', \vec{\hat{v}}_o, \pi)$  with  $\hat{\sigma}' = \hat{\sigma}$ .

**(H9)** Finally, from (H4) and (H8), we have  $\sigma' = \sigma$  and  $\hat{\sigma}' = \hat{\sigma}$ , thus from (H1) it follows that  $\theta, s, \sigma' \models \hat{\sigma}'$ .

Combining (H7), (H8) and (H9), we get our goal (G1).

#### Proposition: UX Soudness

Assume

(H1) 
$$\hat{\alpha}(\hat{\sigma}, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \vec{\hat{v}}_o, \pi) \land \mathsf{SAT}_{\theta, s}(\pi) \land \theta, s, \sigma' \models \hat{\sigma}' \land \\ \|\vec{\hat{v}}_o\|_{\hat{s}, \pi} \leadsto (\vec{v}_o, \pi') \land \|\vec{\hat{v}}_i\|_{\hat{s}, \pi'} \leadsto (\vec{v}_i, \pi'')$$

To prove

(G1) 
$$\exists \sigma. \theta, s, \sigma \models \hat{\sigma} \land \alpha(\sigma, \vec{v_i}) = (o, \sigma', \vec{v_o})$$

We again only consider the case where  $\alpha = \text{load}$  and  $\sigma' = \text{ex}(x)$ ,  $\hat{\sigma}' = \text{ex}(\hat{x})$  – the other three cases are analogous.

- (H2) From ExLoadon, we get  $\hat{\sigma} = \hat{\sigma}' = \exp(\hat{x})$ ,  $\hat{v}_i = []$ , o = 0k,  $\hat{v}_o = [\hat{x}]$ ,  $\pi = []$ , with  $s = [x \mapsto \hat{x}]$
- **(H3)** From (H2) and (H1) we have  $\vec{v}_i = []$  and  $\vec{v}_o = [x]$
- **(H4)** We can pick  $\sigma = \sigma' = \text{ex}(x)$ , which from (H1) and (H2) give us  $\theta, s, \sigma \models \hat{\sigma}$
- **(H5)** From CEXLOADOK, (H3) and (H4), we have load( $\sigma, \vec{v}_i$ ) = (0k,  $\sigma', \vec{v}_o$ ).

Combinding (H4) and (H5) gives our goal (G1).

#### Proposition: Frame subtraction is satisfied

Assume

**(H1)** 
$$\sigma \# \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma', \vec{v}_o)$$

To prove

- **(H2)** load and store are always defined for respectively 0 and 1 arguments, so from (H1) we know  $\exists \sigma'', o', \vec{v}'_o$ .  $\alpha(\sigma, \vec{v}_i) = (o', \sigma'', \vec{v}'_o)$ .
- **(H3)** Assume  $o' \neq \text{Miss}$
- **(H4)** If  $\sigma = \bot$ , the rules say o' = Miss, contradicting (H3). Thus  $\sigma = \text{ex}(x)$ .
- **(H5)** From (H1), we know  $\sigma \cdot \sigma_f$  is defined, so it must be that  $\sigma_f = \bot$  as  $ex(x) \cdot ex(y)$  is undefined.

From (H5) and composition rules we know  $\sigma \cdot \sigma_f = \operatorname{ex}(x) \cdot \bot = \operatorname{ex}(x) = \sigma$ . This gives our goal (G1).

#### Proposition: Frame addition is satisfied

#### Assume

- **(H1)**  $\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o)$
- **(H2)**  $\sigma_f \# \sigma'$
- (H3)  $o \neq \text{Miss}$

### To prove

- **(G1)**  $\sigma \# \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma' \cdot \sigma_f, \vec{v}_o)$
- **(H4)** From (H3) and the action rules, we know  $\sigma' = ex(x)$ .
- **(H5)** From composition, (H4) and (H2), we get  $\sigma_f = \bot$ .

From (H5) we get  $\sigma' \cdot \sigma_f = \sigma' \cdot \bot = \sigma'$ , and from (H1) our goal (G1) follows.

#### Proposition: Consume soundness

#### Assume

- (H1) consume $(\hat{\sigma}, \delta, \vec{v_i}) \rightsquigarrow (Ok, \hat{\sigma}_f, \vec{v_o}, \pi)$
- **(H2)**  $\theta, s, \sigma \models \hat{\sigma} \land \mathsf{SAT}_{\theta, s}(\pi)$

#### To prove

- (G1)  $\exists \sigma_{\delta}, \sigma_{f}, \sigma_{\delta} \# \sigma_{f} \land \sigma = \sigma_{\delta} \cdot \sigma_{f} \land \theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{v}_{i}; \vec{v}_{o}) \land \theta, s, \sigma_{f} \models \hat{\sigma}_{f}$
- (H3) There is only one consume rules yielding 0k, giving us  $\delta = \exp(\hat{x})$ ,  $\hat{\sigma}_f = \perp$ ,  $\vec{\hat{v}}_i = []$ ,  $\vec{\hat{v}}_o = [\hat{x}]$ ,  $\pi = []$
- **(H4)** From the definition of  $\models$ , we must have  $\sigma = \operatorname{ex}(x)$  and  $\sigma_f = \bot$  such that  $\theta, s, \sigma \models \hat{\sigma}$  and  $\theta, s, \sigma_f \models \hat{\sigma}_f$ .
- **(H5)** From (H4), we have  $[\hat{x}]_{\theta,s} = x$ , such that  $[\vec{v}_i]_{\theta,s} = []$  and  $[\vec{v}_o]_{\theta,s} = [x]$ .
- **(H6)** From the definition of composition, we must have  $\sigma_{\delta} = \operatorname{ex}(x)$  such that  $\sigma = \sigma_{\delta} \cdot \sigma_f$ , which also implies  $\sigma_{\delta} \# \sigma_f$
- **(H7)** From (H6), (H5) and the definition of  $\models_{\Delta}$ , we have  $\theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o)$ .

Combining (H4), (H6) and (H7) gives our goal (G1).

### Proposition: Consume completeness

#### Assume

- **(H1)** consume $(\hat{\sigma}, \delta, \vec{\hat{v}_i}) \leadsto (\mathsf{Ok}, \hat{\sigma}_f, \vec{\hat{v}_o}, \pi)$
- **(H2)**  $\theta, s, \sigma_f \models \hat{\sigma}_f \land \theta, s, \sigma_\delta \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o) \land \sigma_f \# \sigma_\delta$

#### To prove

- (G1)  $\exists \sigma. \sigma = \sigma_{\delta} \cdot \sigma_f \wedge \theta, s, \sigma \models \hat{\sigma} \wedge \mathsf{SAT}_{\theta,s}(\pi)$
- (H3) There is only one consume rules yielding Ok, giving us  $\delta = \exp(\hat{x})$ ,  $\hat{\sigma}_f = \perp$ ,  $\hat{v}_i = [], \hat{v}_o = [\hat{x}], \pi = []$
- **(H4)** From (H3) and the definition of  $\models$ , if  $\theta, s, \sigma_f \models \hat{\sigma}_f$  we have  $\sigma_f = \bot$ .
- **(H5)** From (H3) and the definition of  $\models_{\Delta}$ , if  $\theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o)$ , we have  $\sigma_{\delta} = \operatorname{ex}(x)$ , with  $\vec{v}_i = []$  and  $\vec{v}_o = [x]$  such that  $[\![\vec{\hat{v}}_i]\!]_{\theta,s} = \vec{v}_i$  and  $[\![\vec{\hat{v}}_o]\!]_{\theta,s} = \vec{v}_o$ .
- **(H6)** From (H4), (H5) and the rule for composition, we have  $\sigma_f \# \sigma_\delta$ .
- **(H7)** We can pick  $\sigma = \sigma_{\delta} = \exp(x)$ .
- **(H8)** From the definition of composition, (H4) and (H5), we have  $\sigma = \sigma_{\delta} \cdot \sigma_{f}$ .
- **(H9)** Given (H3) and (H7), we have  $\theta, s, \sigma \models \hat{\sigma}$ .
- (H10) From (H3),  $\pi = []$  thus  $SAT_{\theta,s}(\pi)$ . Together with (H8) and (H9), this gives our goal (G1).

### Proposition: Consume: OX sound

#### Assume

**(H1)** 
$$\forall o, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi. \operatorname{consume}(\operatorname{OX}, \hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi) \Rightarrow o_c = \operatorname{Ok}$$

#### To prove

$$\textbf{(G1)} \ \exists \hat{\sigma}_f', \vec{\hat{v}}_o', \pi'. \, \text{consume}(\texttt{OX}, \hat{\sigma}, \delta, \vec{v}_i, \pi) \leadsto (\texttt{Ok}, \hat{\sigma}_f', \vec{\hat{v}}_o', \pi')$$

From the consume rules, we never vanish, so if all consumptions are Ok we must have  $\hat{\sigma} = \text{ex}(\hat{x}), \ \delta = \text{ex}, \ \vec{v}_i = [], \ \hat{\sigma}_f' = \bot, \ \vec{v}_o' = [\hat{x}], \ \pi' = [].$  Our goal (G1) follows.

#### **Proposition: Produce soundness**

#### Assume

- **(H1)** produce $(\hat{\sigma}_f, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \rightsquigarrow (\hat{\sigma}, \pi)$
- **(H2)** SAT<sub> $\theta,s$ </sub> $(\pi) \wedge \theta, s, \sigma \models \hat{\sigma}$

To prove

- (G1)  $\exists \sigma_{\delta}, \sigma_{f}. \sigma_{\delta} \# \sigma_{f} \land \sigma = \sigma_{\delta} \cdot \sigma_{f} \land \theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\hat{v}_{i}; \hat{v}_{o}) \land \theta, s, \sigma_{f} \models \hat{\sigma}_{f}$
- **(H3)** From the produce rule, we have  $\hat{\sigma}_f = \bot$ ,  $\delta = \exp$ ,  $\vec{v}_i = []$ ,  $\vec{v}_o = [\hat{x}]$ ,  $\hat{\sigma} = \exp(\hat{x})$  and  $\pi = []$ .
- **(H4)** From (H2) and the definition of  $\models$  we have  $\sigma = \operatorname{ex}(x)$  and  $[\![\hat{x}]\!]_{\theta,s} = x$
- **(H5)** Given (H4), we have  $\sigma_{\delta} = \operatorname{ex}(x)$  and  $\sigma_{f} = \bot$ , such that  $\sigma_{\delta} \# \sigma_{f}$  and  $\sigma = \sigma_{\delta} \cdot \sigma_{f}$ .
- **(H6)** From (H4), we have  $[\![\vec{\hat{v}}_i]\!]_{\theta,s} = [\![\!]$  and  $[\![\vec{\hat{v}}_o]\!]_{\theta,s} = [\![\!]\!]$ , thus from (H5) and the definition of  $\models_{\Delta}$  we have  $\theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o)$ .
- (H7) From (H3) and (H5) we have  $\sigma_f = \bot$  and  $\hat{\sigma}_f = \bot$ , thus from the definition of  $\models$ , we have  $\theta, s, \sigma_f \models \hat{\sigma}_f$ . This with (H6) gives our goal (G1).

#### Proposition: Produce completeness

Assume

**(H1)** 
$$\theta, s, \sigma_f \models \hat{\sigma}_f \land \theta, s, \sigma_\delta \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{v}_o) \land \sigma_f \# \sigma_\delta$$

To prove

- (G1)  $\exists \hat{\sigma}. \operatorname{produce}(\hat{\sigma}_f, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}, \pi) \land \operatorname{SAT}_{\theta,s}(\pi) \land \theta, s, (\sigma_f \cdot \sigma_\delta) \models \hat{\sigma}$
- (H2) From (H9),  $\sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\hat{v}_i; \hat{v}_o)$ , thus from the definition of  $\models_{\Delta}$  we have  $\sigma_{\delta} = \operatorname{ex}(x)$ ,  $\delta = \operatorname{ex}, \ \vec{v}_i = [], \ \vec{v}_o = [\hat{x}] \text{ with } [\![\hat{x}]\!]_{\theta,s} = x.$
- **(H3)** From (H9),  $\sigma_f \# \sigma_\delta$ , thus from (H2) and the rules of composition, we have  $\sigma_f = \bot$ .
- **(H4)** From the rules of  $\models$  and (H3), we have  $\hat{\sigma}_f = \bot$ .
- **(H5)** We can pick  $\hat{\sigma} = \exp(\hat{x})$ .
- **(H6)** From (H4), (H2), (H5) and the rules of produce, we have  $\operatorname{produce}(\sigma_f, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \rightsquigarrow (\hat{\sigma}, \pi)$ , with  $\pi = []$ .
- **(H7)** From (H6),  $SAT_{\theta,s}(\pi)$ .
- **(H8)** From (H3) and (H2),  $\sigma_f = \bot$  and  $\sigma_\delta = \operatorname{ex}(x)$ , thus from the rules of composition,  $\sigma_f \cdot \sigma_\delta = \operatorname{ex}(x)$ .

**(H9)** From (H2),  $[\![\hat{x}]\!]_{\theta,s} = x$ , thus from (H5) and (H8) we have  $\theta, s, (\sigma_f \cdot \sigma_\delta) \models \hat{\sigma}$ , which combined with (H6) and (H7) gives our goal (G1).

## B.2 Partial Map

### B.2.1 Resource Algebra

$$\begin{split} \operatorname{PMAP}(I,\mathbb{S}) &\triangleq I \xrightarrow{fin} \mathbb{S}.\Sigma \times \mathcal{P}(I)^? \\ (h,d) \cdot (h',d') &\triangleq (h'',d'') \\ \text{where } h'' &\triangleq \lambda i. \begin{cases} h(i) \cdot h'(i) & \text{if } i \in \operatorname{dom}(h) \cap \operatorname{dom}(h') \\ h(i) & \text{if } i \in \operatorname{dom}(h) \backslash \operatorname{dom}(h') \\ h'(i) & \text{if } i \in \operatorname{dom}(h') \backslash \operatorname{dom}(h) \\ \text{undefined} & \text{otherwise} \end{cases} \\ &= \begin{cases} d & \text{if } d' = \bot \\ d' & \text{if } d = \bot \\ \text{undefined} & \text{otherwise} \end{cases} \\ &= \operatorname{and} d'' = \bot \vee \operatorname{dom}(h'') \subseteq d'' \\ |(h,d)| &\triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(h') = \emptyset \\ (h',\bot) & \text{otherwise} \end{cases} \\ &\text{where } h' \triangleq \lambda i. \begin{cases} |h(i)| & \text{if } i \in \operatorname{dom}(h) \wedge |h(i)| \neq \bot \\ \text{undefined} & \text{otherwise} \end{cases} \end{split}$$

For a wrapped state model  $\mathbb{S} = \{\Sigma_{\mathbb{S}}, \mathcal{A}_{\mathbb{S}}, \Delta_{\mathbb{S}}\}$ , we define the actions of PMAP $(I, \mathbb{S})$  as  $\mathcal{A} = \mathcal{A}_{\mathbb{S}} \uplus \{\text{alloc}\}$ , and the predicates  $\Delta = \Delta_{\mathbb{S}} \uplus \{\text{domainset}\}$ .

We define predicate satisfiability as:

We define symbolic interpretation as:

PMAPSYMINTERPRETATION 
$$\forall \hat{i} \in \text{dom}(\hat{h}). \, [\![\hat{i}]\!]_{\theta,s} = i \wedge i \in \text{dom}(h) \wedge \theta, s, h(i) \models \hat{h}(\hat{i})$$
 
$$\underline{ [\![\text{dom}(\hat{h})]\!]_{\theta,s} = \text{dom}(h) } \quad [\![\hat{d}]\!]_{\theta,s} = d$$
 
$$\theta, s, (h,d) \models (\hat{h},\hat{d})$$

### **B.2.2** Compositional Concrete Rules

We define the helper functions get and set as:

$$\begin{split} & \mathsf{get}: \mathsf{PMAP}(I,\mathbb{S}) \to I \to \mathbb{S}.\Sigma \\ & \mathsf{set}: \mathsf{PMAP}(I,\mathbb{S}) \to I \to \mathbb{S}.\Sigma \to \mathsf{PMAP}(I,\mathbb{S}) \end{split}$$

We pretty-print get and set as  $get(\sigma, i) = \sigma_i$  and  $set(\sigma, i, \sigma_i) = \sigma'$ .

Given 
$$wrap(h, d) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(h) = \emptyset \land d = \bot \\ (h, d) & \text{otherwise} \end{cases}$$

$$unwrap(\sigma) \triangleq \begin{cases} ([], \bot) & \text{if } \sigma = \bot \\ (h, d) & \text{if } \sigma = (h, d) \end{cases}$$

$${\operatorname{CPMapGetMatch}}$$

$$\frac{(h,d) = \mathit{unwrap}(\sigma) \qquad i \in \mathrm{dom}(h) \qquad \sigma_i = h(i)}{\mathtt{get}(\sigma,i) = \sigma_i}$$

CPMapGetAdd

$$\frac{(h,d) = unwrap(\sigma) \qquad i \notin \text{dom}(h) \qquad d \neq \bot \qquad i \in d}{\text{get}(\sigma,i) = \bot}$$

CPMAPGETBOTDOMAIN

$$\frac{(h,d) = \mathit{unwrap}(\sigma) \qquad i \notin \mathrm{dom}(h) \qquad d = \bot}{\mathrm{get}(\sigma,i) = \bot}$$

CPMAPSETSOME

$$\frac{(h,d) = unwrap(\sigma) \qquad \sigma_i \neq \bot \qquad h' = h[i \leftarrow \sigma_i] \qquad \sigma' = wrap(h',d)}{\operatorname{set}(\sigma,i,\sigma_i) = \sigma'}$$

CPMapSetNone

$$\frac{(h,d) = unwrap(\sigma) \qquad \sigma_i = \bot \qquad h' = h[i \neq] \qquad \sigma' = wrap(h',d)}{\operatorname{set}(\sigma,i,\sigma_i) = \sigma'}$$

The action rules are then:

$$\frac{\text{get}(\sigma,i) = \sigma_i \qquad \alpha(\sigma_i,\vec{v}_i) = (o,\sigma_i',\vec{v}_o) \qquad \text{set}(\sigma,i,\sigma_i') = \sigma'}{\alpha(\sigma,i::\vec{v}_i) = (o,\sigma',i::\vec{v}_o)}$$

$$\frac{\alpha(\sigma,i::\vec{v}_i) = (o,\sigma',i::\vec{v}_o)}{\text{CPMapActionOutOfBounds}}$$

$$\frac{d \neq \bot \qquad i \notin d}{\alpha((h,d),i::\vec{v}_i) = (\text{Err},(h,d),[])}$$

$$\frac{d \neq \bot \qquad i = \text{fresh} \qquad \sigma_i = \text{instantiate}(\vec{v}_i) \qquad h' = h[i \leftarrow \sigma_i] \qquad d' = d \uplus \{i\}$$

$$\frac{1}{\text{alloc}((h,d),\vec{v}_i) = (\text{Ok},(h',d'),[i])}$$

$$\frac{(\text{CPMapAllocMiss})}{(h,d) = unwrap(\sigma)} \qquad d = \bot$$

### **B.2.3** Compositional Symbolic Rules

We now re-define get and set, lifting them to the symbolic realm (note set is unchanged, as it does not perform any sort of matching). get now returns a state and an index, that may be different from the input index – it corresponds to the actual index of the state in the map. For example, for a map  $[1 \mapsto \hat{x}]$ , calling get with index  $\hat{y}$  may return index 1, along with the state  $\hat{x}$  and the path condition  $[\hat{y} = 1]$ .

 $alloc(\sigma, \vec{v_i}) = (Miss, \sigma, ['domainset'])$ 

$$\begin{split} & \mathsf{get}: \mathsf{PMAP}(I,\mathbb{S}) \to I \to \mathcal{P}(I \times \mathbb{S}.\hat{\Sigma} \times \Pi) \\ & \mathsf{set}: \mathsf{PMAP}(I,\mathbb{S}) \to I \to \mathbb{S}.\hat{\Sigma} \to \mathsf{PMAP}(I,\mathbb{S}) \end{split}$$

We pretty-print get and set as  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi)$  and  $set(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'$ .

Given 
$$wrap(\hat{h}, \hat{d}) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(\hat{h}) = \emptyset \land \hat{d} = \bot \\ (\hat{h}, \hat{d}) & \text{otherwise} \end{cases}$$

$$unwrap(\hat{\sigma}) \triangleq \begin{cases} ([], \bot) & \text{if } \hat{\sigma} = \bot \\ (\hat{h}, \hat{d}) & \text{if } \hat{\sigma} = (\hat{h}, \hat{d}) \end{cases}$$

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i}' \in \text{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i}')}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])}$$

PMapGetAdd

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \notin \text{dom}(\hat{h}) \qquad \hat{d} \neq \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}) \land \hat{i} \in \hat{d}])}$$

PMAPGETBOTDOMAIN

$$\frac{(\hat{h}, \hat{d}) = \mathit{unwrap}(\hat{\sigma}) \qquad \hat{i} \not\in \mathrm{dom}(\hat{h}) \qquad \hat{d} = \bot}{\mathrm{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \not\in \mathrm{dom}(\hat{h})])}$$

PMapSetSome

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{\sigma}_i \neq \bot \qquad \hat{h}' = \hat{h}[\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\operatorname{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

PMapSetNone

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{\sigma}_i = \bot \qquad \hat{h}' = \hat{h}[\hat{i} \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\operatorname{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

$$\text{Given } \textit{lift\_if\_miss}(o, \hat{i}, \vec{\hat{v}_i}) \triangleq \begin{cases} \hat{i} :: \vec{\hat{v}_i} & \text{if } o = \texttt{Miss} \\ \vec{\hat{v}_i} & \text{otherwise} \end{cases}$$

PMAPACTION

$$\frac{ \gcd(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \qquad \alpha(\hat{\sigma}_i, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_i', \vec{\hat{v}}_o, \pi') \qquad \gcd(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}'}{\alpha(\hat{\sigma}, \hat{i} :: \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}', \hat{i}' :: \vec{\hat{v}}_o, \pi :: \pi')}$$

PMAPACTIONOUTOFBOUNDS

$$\frac{\hat{d} \neq \bot}{\alpha((\hat{h}, \hat{d}), \hat{i} :: \vec{\hat{v_i}}) \leadsto (\mathsf{Err}, (\hat{h}, \hat{d}), [], [\hat{i} \notin \hat{d}])}$$

PMAPALLOC

$$\frac{\hat{d} \neq \bot \qquad \hat{i} = \text{fresh} \qquad \hat{\sigma}_i = \text{instantiate}(\vec{\hat{v}}_i) \qquad \hat{h}' = \hat{h}[\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{d}' = \hat{d} \uplus \{\hat{i}\}}{\text{alloc}((\hat{h}, \hat{d}), \vec{\hat{v}}_i) \leadsto (\text{Ok}, (\hat{h}', \hat{d}'), [\hat{i}], [\hat{i} = \hat{i}])}$$

PMAPALLOCMISS

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{d} = \bot}{\text{alloc}(\hat{\sigma},\vec{\hat{v}_i}) \leadsto (\text{Miss},\hat{\sigma},[\text{'domainset'}],[])}$$

**PMapCons** 

$$\gcd(\hat{\sigma},\hat{i})\leadsto(\hat{i}',\hat{\sigma}_i,\pi)$$
 
$$\gcd(\hat{\sigma}_i,\delta,\vec{\hat{v}}_i)\leadsto(o,\hat{\sigma}_i',\vec{\hat{v}}_o,\pi') \qquad \gcd(\hat{\sigma},\hat{i}',\hat{\sigma}_i')=\hat{\sigma}' \qquad \vec{\hat{v}}_o'=\mathit{lift\_if\_miss}(o,\hat{i}',\vec{\hat{v}}_o)$$
 
$$\gcd(\hat{\sigma},\delta,\hat{i}::\vec{\hat{v}}_i)\leadsto(o,\hat{\sigma}',\vec{\hat{v}}_o',\pi::\pi')$$

PMAPCONSINCOMPAT

$$\begin{split} \hat{d} \neq \bot \\ & \text{consume}((\hat{h}, \hat{d}), \delta, \hat{i} :: \vec{\hat{v}}_i) \leadsto (\text{LFail}, (\hat{h}, \hat{d}), [], [\hat{i} \notin \hat{d}]) \\ & \text{PMapConsDomainSet} \\ & \qquad \qquad \hat{d} \neq \bot \\ & \text{consume}((\hat{h}, \hat{d}), \text{domainset}, []) \leadsto (\text{Ok}, (\hat{h}, \bot), [\hat{d}], []) \end{split}$$

PMapConsDomainSetMiss

$$(\hat{h}, \hat{d}) = unwrap(\hat{\sigma})$$
  $\hat{d} = \bot$ 

 $\frac{(\hat{h},\hat{d}) = \textit{unwrap}(\hat{\sigma}) \qquad \hat{d} = \bot}{\mathsf{consume}(\hat{\sigma},\mathsf{domainset},[]) \leadsto (\mathsf{Miss},\hat{\sigma},[\text{'domainset'}],[])}$ 

PMapProd

$$\frac{\text{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i}',\hat{\sigma}_i,\pi) \qquad \text{produce}(\hat{\sigma}_i,\delta,\vec{\hat{v}}_i,\vec{\hat{v}}_o) \leadsto (\hat{\sigma}_i',\pi') \qquad \text{set}(\hat{\sigma},\hat{i}',\hat{\sigma}_i') = \hat{\sigma}'}{\text{produce}(\hat{\sigma},\delta,\hat{i}::\vec{\hat{v}}_i,\vec{\hat{v}}_o) \leadsto (\hat{\sigma}',\pi::\pi')}$$

$$\begin{split} & & \text{PMapProdDomainSet} \\ & & & & (\hat{h}, \bot) = \textit{unwrap}(\hat{\sigma}) \\ & & & & & & & & \\ & & & & & \\ & & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \\$$

**PMapFixDomainSet** 

fix ['domainset'] = 
$$\exists \hat{d}$$
.  $\langle domainset \rangle (; \hat{d})$ 

Note in the above we define  $lift(a, \hat{i})$  as a function that traverses an assertion a and lifts all core predicate assertions by adding the value  $\hat{i}$  at the start of its in-values, such that  $lift(\langle \delta \rangle(\hat{v}_i; \hat{v}_o), \hat{i}) = \langle \delta \rangle(\hat{i} :: \hat{v}_i; \hat{v}_o).$ 

#### B.2.4 Soundness Proofs

To facilitate the soundness proof for PMAP, and due to its extensive use of the get and set helper methods to modify the elements in the heap, we first define and prove axioms about these auxiliary functions.

We introduce a pair of axioms for get, similar to OX and UX soundness of actions, to ensure that the symbolic function retrieves the right state if it exists in the concrete counterpart of the state.

$$\theta, s, \sigma \models \hat{\sigma} \land \text{get}(\sigma, i) = \sigma_i \land [\hat{i}]_{\theta, s} = i \implies \exists \hat{\sigma}_i, \hat{i}', \pi.$$

$$\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land [\hat{i}']_{\theta, s} = i \land \text{SAT}_{\theta, s}(\pi)$$
(Get OX Soundness)

$$\gcd(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land [\![\hat{i}]\!]_{\theta,s} = [\![\hat{i}']\!]_{\theta,s} = i \land \mathsf{SAT}_{\theta,s}(\pi) \implies \\ \forall \sigma. \, \theta, s, \sigma \models \hat{\sigma} \implies \exists \sigma_i. \gcd(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$$
 (Get UX Soundness)

Proof.

#### Proposition: Get OX Soundness

Assume

**(H1)** 
$$\theta, s, \sigma \models \hat{\sigma} \land \text{get}(\sigma, i) = \sigma_i \land [\hat{i}]_{\theta, s} = i$$

To prove

(G1) 
$$\exists \hat{\sigma}_i, \hat{i}', \pi. \operatorname{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land [[\hat{i}']]_{\theta,s} = i \land \operatorname{SAT}_{\theta,s}(\pi)$$

We proceed by proving the property holds for all rules of the concrete get, resulting in three cases.

#### Case CPMAPGETMATCH:

- **(H2)** Assume  $(h, d) = unwrap(\sigma) \land i \in dom(h) \land \sigma_i = h(i)$ .
- **(H3)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \hat{d})$  such that  $\exists \hat{i}'. \hat{i}' \in \text{dom}(\hat{h})$ ,  $[\hat{i}']_{\theta,s} = i$  and  $\theta, s, \sigma_i \models \hat{h}(\hat{i}')$ .
- **(H4)** We can then apply PMAPGETMATCH, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{h}(\hat{i}'), [\hat{i}' = \hat{i}])$ .
- **(H5)** From (H3) and (H1) it follows that  $SAT_{\theta,s}([\hat{i}'=\hat{i}])$ , which completes our goal (G1).

### Case CPMAPGETADD:

- **(H6)** Assume  $(h, d) = unwrap(\sigma) \land i \notin dom(h) \land d \neq \bot \land i \in d$ .
- **(H7)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \hat{d})$  such that  $\hat{i} \in \hat{d}$  and  $\hat{i} \notin \text{dom}(\hat{h})$ .
- **(H8)** We can then apply PMAPGETADD, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \perp, |\hat{i} \notin dom(\hat{h}) \land \hat{i} \notin \hat{d}|)$ .
- **(H9)** From (H8) and (H7) it follows that  $SAT_{\theta,s}([\hat{i} \notin dom(\hat{h}) \land \hat{i} \notin \hat{d}])$ , which completes our goal (G1).

#### Case CPMAPGETBOTDOMAIN:

**(H10)** Assume  $(h, d) = unwrap(\sigma) \land i \notin dom(h) \land d = \bot$ .

- **(H11)** From (H10), (H1) and CPMAPGETBOTDOMAIN, we have  $\sigma_i = \bot$ .
- **(H12)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \perp)$  such that  $\hat{i} \notin \text{dom}(\hat{h})$ .
- **(H13)** We can then apply PMAPGETBOTDOMAIN, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \perp, [\hat{i} \notin dom(\hat{h})])$ .
- **(H14)** From (H13) and (H12) it follows that  $SAT_{\theta,s}([\hat{i} \notin dom(\hat{h})])$ , which completes our goal (G1).

#### Proposition: Get UX Soundness

#### Assume

$$\textbf{(H1)} \ \gcd(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \llbracket \hat{i} \rrbracket_{\theta, s} = \llbracket \hat{i}' \rrbracket_{\theta, s} = i \land \mathsf{SAT}_{\theta, s}(\pi)$$

**(H2)** 
$$\theta, s, \sigma \models \hat{\sigma}$$

#### To prove

**(G1)** 
$$\exists \sigma_i. get(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$$

We proceed by proving the property holds for all rules of the symbolic get, resulting in three cases.

#### Case PMAPGETMATCH:

- **(H3)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{i}' \in dom(\hat{h}) \wedge \hat{\sigma}_i = \hat{h}(\hat{i}')$
- **(H4)** From (H3), (H1) and PMAPGETMATCH, we have  $\pi = [\hat{i} = \hat{i}']$
- **(H5)** From (H3), we have  $\hat{\sigma} \neq \bot$ , thus from (H2) we have  $\sigma = (h, d)$
- **(H6)** From (H2), (H3) and the definition of  $\models$ , we have  $i \in h$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$
- (H7) From (H6) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

#### Case PMAPGETADD:

- **(H8)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{i} \notin dom(\hat{h}) \wedge \hat{d} \neq \bot$
- **(H9)** From (H8), (H1) and PMAPGETADD, we have  $\hat{i}' = \hat{i}$ ,  $\hat{\sigma}_i = \bot$  and  $\pi = [\hat{i} \notin \text{dom}(\hat{h}) \land \hat{i} \in \hat{d}]$
- **(H10)** From (H8) we have  $\hat{d} \neq \bot$ , thus  $\hat{\sigma} = (\hat{h}, \hat{d})$  and from (H2)  $\sigma = (h, d)$  such that  $[\![\hat{d}]\!]_{\theta,s} = d$ .
- **(H11)** From (H1)  $SAT_{\theta,s}(\pi)$ , thus from (H9) and (H2),  $i \notin dom(h) \land i \in d$

- (H12) From (H11) and (H10) we can apply CPMAPGETADD, thus  $get(\sigma, i) = \bot$ .
- **(H13)** From Empty Memory,  $\theta, s, \perp \models \perp$ , which along with (H12) completes our goal (G1).

#### Case PMAPGETBOTDOMAIN:

- **(H14)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{i} \notin dom(\hat{h}) \wedge \hat{d} = \bot$
- (H15) From (H14), (H1) and PMAPGETBOTDOMAIN, we have  $\hat{i}' = \hat{i}$ ,  $\hat{\sigma}_i = \bot$  and  $\pi = [\hat{i} \notin \text{dom}(\hat{h})]$
- **(H16)** From (H14) we have  $\hat{d} = \bot$ , so given  $(h, d) = unwrap(\sigma)$  and (H2), we have  $d = \bot$ .
- **(H17)** From (H1)  $SAT_{\theta,s}(\pi)$ , thus from (H15) and (H2),  $i \notin dom(h)$ .
- (H18) From (H17) and (H16) we can apply CPMAPGETBOTDOMAIN, thus  $get(\sigma, i) = \bot$ .
- **(H19)** From Empty Memory,  $\theta, s, \perp \models \perp$ , which along with (H18) completes our goal (G1).

We can now proceed with the standard proofs.

Proof.

#### Proposition: OX Soudness

#### Assume

**(H1)** 
$$\theta, s, \sigma \models \hat{\sigma} \land \alpha(\sigma, \vec{v_i}) = (o, \sigma', \vec{v_o}) \land [[\hat{v_i}]]_{\theta, s} = \vec{v_i}$$

**(H2)** 
$$\forall o', \hat{\sigma}', \vec{\hat{v}}'_o, \pi'. \alpha(\hat{\sigma}, \vec{\hat{v}}_i) \rightsquigarrow (o', \hat{\sigma}', \vec{\hat{v}}'_o, \pi') \Rightarrow o' \in \{0k, Err\}$$

(H3) The actions on S are OX sound.

#### To prove

(G1) 
$$\exists \hat{\sigma}', \hat{v}_o, \pi, \theta'. \hat{\alpha}(\hat{\sigma}, \hat{v}_i) \rightsquigarrow (o, \hat{\sigma}', \hat{v}_o, \pi) \land \theta', s, \sigma' \models \hat{\sigma}' \land \mathsf{SAT}_{\theta', s}(\pi) \land \llbracket \hat{v}_o \rrbracket_{\theta', s} = \vec{v}_o$$

There are two action cases to consider:  $\alpha \in A_{\mathbb{S}}$  and  $\alpha = \mathsf{alloc}$ .

### Case $\alpha \in A_{\mathbb{S}}$ :

(H4) If the action gets executed successfully (it is not an out of bounds error), from CPMA-PACTION we have  $get(\sigma, i) = \sigma_i$ ,  $\alpha(\sigma_i, \vec{v}'_i) = (o, \sigma'_i, \vec{v}'_o)$  and  $set(\sigma, i, \sigma'_i) = \sigma'$ , with  $\vec{v}_i = i :: \vec{v}'_i$  and  $\vec{v}_o = i :: \vec{v}'_o$ 

- **(H5)** From (H1) and (H4), we know we have  $\vec{v}_i = \hat{i} :: \vec{v}_i'$  such that  $[\hat{i}]_{\theta,s} = i$  and  $[\vec{v}_i']_{\theta,s} = \vec{v}_i$ .
- **(H6)** From (H1), (H4) and (H5), we can apply Get OX Soundness, giving  $\exists \hat{\sigma}_i, \hat{i}', \pi. \operatorname{get}(\hat{\sigma}, \hat{i}) \rightsquigarrow (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land \|\hat{i}'\|_{\theta,s} = i \land \operatorname{SAT}_{\theta,s}(\pi)$
- (H7) From (H4), (H6), (H2), (H5) and (H3), we can apply Memory Model OX Soundness, giving us  $\exists \vec{v}_o', \hat{\sigma}_i', \pi', \theta' . \alpha(\hat{\sigma}_i, \vec{v}_i') \leadsto (o, \hat{\sigma}_i', \vec{v}_o', \pi') \land \theta', s, \sigma_i' \models \hat{\sigma}_i' \land \mathsf{SAT}_{\theta', s}(\pi') \land \llbracket \hat{v}_o' \rrbracket_{\theta', s} = \vec{v}_o'.$
- **(H8)** From (H1) and (H7), we have  $\theta, s, \sigma \models \hat{\sigma}$  and  $\theta', s, \sigma'_i \models \hat{\sigma}'_i$ . From the definition of set, it follows that only modifying the state at  $\hat{i}'$  preserves symbolic interpretation, thus given  $\operatorname{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}'_i) = \hat{\sigma}'$ , we have  $\theta', s, \sigma' \models \hat{\sigma}'$ .
- (H9) By applying PMAPACTION and from (H7), (H6), (H8) our goal (G1) follows.
- **(H10)** If this is an out of bounds access, we have  $\sigma = \sigma' = (h, d)$  with  $d \neq \bot$  and  $\vec{v_i} = i :: \vec{v_i'}$ , with  $i \notin d$ , as well as  $\vec{\hat{v_i}} = \hat{i} :: \vec{\hat{v}_i'}$
- **(H11)** From (H1) and the definition of  $\models$ , we have  $\hat{\sigma} = (\hat{h}, \hat{d})$  such that  $\pi = [\hat{i} \notin \hat{d}]$  and  $SAT_{\theta,s}(\pi)$ .
- **(H12)** From the rules of action execution and (H11), we get  $\alpha(\hat{\sigma}, \vec{v}_i) \rightsquigarrow (\text{Err}, \hat{\sigma}, [], \pi)$ . The rest of our goal (G1) follows.

#### Case $\alpha = \text{alloc}$ :

Both cases (when  $d \neq \bot$  and  $d = \bot$ ) are trivially lifted from the concrete to the symbolic realm, with  $\pi = [i = i]$  and  $\pi = []$  respectively. It follows that  $SAT_{\theta,s}(\pi)$  in both cases, giving our goal (G1).

For the successful allocation case, we note there is an additional requirement for soundness of state instantiation to be able to prove the resulting states  $\hat{\sigma}'$  and  $\sigma'$  are compatible.

$$\|\vec{\hat{v}}_i\|_{\theta,s} = \vec{v}_i \implies (\forall \sigma, \hat{\sigma}. \sigma = \text{instantiate } \vec{v}_i \land \hat{\sigma} = \text{instantiate } \vec{\hat{v}}_i \implies \theta, s, \sigma \models \hat{\sigma})$$

### Proposition: UX Soundness

Assume

$$\textbf{(H1)} \quad \begin{array}{ll} \hat{\alpha}(\hat{\sigma},\vec{\hat{v}_i}) \leadsto (o,\hat{\sigma}',\vec{\hat{v}_o},\pi) \land \mathsf{SAT}_{\theta,s}(\pi) \land \theta, s,\sigma' \models \hat{\sigma}' \land \\ & [\![\vec{\hat{v}_o}]\!]_{\hat{s},\pi} \leadsto (\vec{v_o},\pi') \land [\![\vec{\hat{v}_i}]\!]_{\hat{s},\pi'} \leadsto (\vec{v_i},\pi'') \end{array}$$

(H2) The actions on S are UX sound.

To prove

(G1)  $\exists \sigma. \theta, s, \sigma \models \hat{\sigma} \land \alpha(\sigma, \vec{v_i}) = (o, \sigma', \vec{v_o})$ 

Case  $\alpha \in A_{\mathbb{S}}$ :

- (H3) If the action gets executed successfully (it is not an out of bounds error), from PMA-PACTION we have  $\operatorname{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi'), \alpha(\hat{\sigma}_i, \vec{\tilde{v}}_i') \leadsto (o, \hat{\sigma}_i', \vec{\tilde{v}}_o', \pi'')$  and  $\operatorname{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}'$ , with  $\vec{\tilde{v}}_i = \hat{i} :: \vec{\tilde{v}}_i', \vec{\tilde{v}}_o = \hat{i}' :: \vec{\tilde{v}}_o'$  and  $\pi = \pi' :: \pi''$
- **(H4)** From (H2), (H3) and (H1), we get  $\exists \sigma_i.\theta, s, \sigma_i \models \hat{\sigma}_i \land \alpha(\sigma_i, \vec{v}_i') = (p, \sigma', \vec{v}_o')$
- **(H5)** From the definition of  $\models$ , it follows that if  $\theta, s, \sigma' \models \hat{\sigma}'$  where  $\mathsf{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}'_i) = \hat{\sigma}'$ , then if  $\theta, s, \sigma_i \models \hat{\sigma}_i$  there exists an  $\sigma$  and i such that  $\mathsf{set}(\sigma, i, \sigma'_i) = \sigma'$ , giving us  $\theta, s, \sigma \models \hat{\sigma}$
- **(H6)** From (H3), (H1) and (H5) we can use Get UX Soundness, giving us  $\exists \sigma_i$ . get $(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$
- (H7) From (H6), (H4) and (H5) we can apply CPMAPACTION, giving us  $\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o)$ , which together with (H5) gives our goal (G1).
- (H8) If the action fails as it is out of bounds, from PMAPACTIONOUTOFBOUNDS we have  $\vec{v}_i = \hat{i} :: \vec{v}_i'$  and  $\pi = [\hat{i} \notin \hat{d}]$ .
- **(H9)** From (H1) and the definition of  $\models$  we thus have  $\sigma = (h, d)$ , with  $\vec{v}_i = i :: \vec{v}_i', d \neq \bot$  and  $i \notin d$
- (H10) From (H9), we can apply CPMAPACTIONOUTOFBOUNDS, which along with the fact the state is unmodified gives our goal (G1).

#### ${\it Case} \ \alpha = {\it alloc:}$

Again, both cases (when the domains set d is owned and when it is  $\bot$ ) are directly lifted from the concrete cases; it follows that if  $\theta, s, \sigma' \models \hat{\sigma}'$  then  $\theta, s, \sigma \models \hat{\sigma}$  where  $\sigma$  is the state without the added binding.

#### Proposition: Frame subtraction is satisfied

Assume

- **(H1)**  $\sigma \# \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma', \vec{v}_o)$
- **(H2)** The actions on  $\mathbb{S}$  satisfy frame subtraction.
- **(H3)**  $\forall \alpha, \vec{v_i}, \vec{v_o}. \alpha(\bot, \vec{v_i}) = (\text{Miss}, \sigma', \vec{v_o})$

To prove

$$\begin{aligned} \textbf{(G1)} \quad & \exists \sigma'', o', \vec{v}_o'. \, \alpha(\sigma, \vec{v}_i) = (o', \sigma'', \vec{v}_o') \, \wedge \\ & (o' \neq \texttt{Miss} \implies o' = o \wedge \vec{v}_o' = \vec{v}_o \wedge \sigma' = \sigma'' \cdot \sigma_f) \end{aligned}$$

Case  $\alpha \in A_{\mathbb{S}}$ :

- **(H4)** If the action causes an out of bounds error, from CPMAPACTIONOUTOFBOUNDS we have  $\sigma \cdot \sigma_f = (h, d)$  and  $\vec{v}_i = i :: \vec{v}_i'$  such that  $i \notin d$ .
- **(H5)** From composition, the domain set d is either part of  $\sigma$  or  $\sigma_f$ . If it is part of  $\sigma$  such that  $\sigma = (h_a, d)$  and  $\sigma_f = (h_b, \bot)$ , then we still have  $i \notin d$ , resulting in the same error and outcome, giving our goal (G1).
- (H6) If the domain is in  $\sigma_f$  such that  $\sigma = (h_a, \bot)$  and  $\sigma_f = (h_b, d)$ , then the action is executed on the underlying state model. Because we have  $\text{dom}(h) \subseteq d$ , from (H4) we have  $i \notin \text{dom}(h_a)$ , so from the rules of get and CPMAPACTION the action is executed on  $\bot$ . From (H3), we have o' = Miss, satisfying (G1).
- **(H7)** If the action is not out of bounds, we have  $\alpha(\sigma_i, \vec{v}_i') = (o, \sigma_i', \vec{v}_o')$  with  $\vec{v}_i = i :: \vec{v}_i'$  and  $\vec{v}_o = i :: \vec{v}_o'$ .
- (H8) If  $\sigma_f$  only includes the domain set and indices different than i, then the action is executed on the same  $\sigma_i$  and gives the same outcomes it can then be composed with the result, giving our goal (G1).
- **(H9)** Let  $\sigma_f$  such that  $\sigma_f = (h_b, d_b)$  and  $i \in \text{dom}(h_b)$ . We denote  $\sigma_{b,i} = h_b(i)$ . For  $\sigma = (h_a, d_a)$ , we also have  $\sigma_{a,i} = \bot$  if  $i \notin \text{dom}(h_a)$ , and  $\sigma_{a,i} = h_a(i)$  otherwise, such that  $\sigma_{a,i} \# \sigma_{b,i}$ .
- (H10) With (H7) and (H9), from (H2), we can apply Frame subtraction to the action on the state at index i, which allows re-applying CPMAPACTION and completes our goal (G1).

 ${\it Case} \ \alpha = {\it alloc:}$ 

- (H11) alloc is always defined, so we know  $\exists \sigma'', o', \vec{v}_o. \alpha(\sigma, \vec{v}_i) = (o', \sigma'', \vec{v}_o').$
- **(H12)** Assume  $o' \neq \text{Miss}$ .
- **(H13)** From (H12) and the rules of alloc,  $\sigma = (h, d)$  with  $d \neq \bot$ .
- **(H14)** From (H13) and (H1), if  $\sigma \# \sigma_f$ , then  $\sigma_f = (h_f, \bot)$ .

(H15) From the definition of alloc, the heap is not modified to the exception of the added entry, so from the definition of composition we have  $\sigma' = \sigma'' \cdot \sigma_f$ . The returns value  $\vec{v_o}$  are the same (a fresh value), and the outcome is 0k in both cases, giving our goal (G1).

# Proposition: Frame addition is satisfied

Assume

**(H1)** 
$$\alpha(\sigma, \vec{v_i}) = (o, \sigma', \vec{v_o})$$

- **(H2)**  $\sigma_f \# \sigma'$
- (H3)  $o \neq \text{Miss}$
- (H4) The actions on S satisfy frame addition.

**(H5)** 
$$\forall \alpha, \vec{v_i}, \vec{v_o}, \alpha(\perp, \vec{v_i}) = (\text{Miss}, \sigma', \vec{v_o})$$

To prove

**(G1)** 
$$\sigma \# \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma' \cdot \sigma_f, \vec{v}_o)$$

Case  $\alpha \in A_{\mathbb{S}}$ :

- **(H6)** From the rules of concrete actions we have  $\vec{v}_i = i :: \vec{v}_i'$  and  $\vec{v}_o = i :: \vec{v}_o'$ .
- **(H7)** If the execution of the action is an out of bounds error, by CPMAPACTIONOUTOF-BOUNDS we have  $\sigma = (h, d)$  such that  $d \neq \bot$  and  $i \notin d$ .
- **(H8)** From composition rules, we know  $\sigma_f = (h_f, \perp)$ , meaning that we still get  $\alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (\text{Err}, \sigma' \cdot \sigma_f, [])$  as we still have  $i \notin d$ . This satisfies our goal (G1).
- (H9) We now look at the case where we do not have an out of bounds error. Assume we have  $\sigma = (h, d)$  such that  $i \notin \text{dom}(h)$ . From the rules for get, we know we executed the action on  $\bot$ . From (H5), we thus have o = Miss, which is a contradiction with (H3). Thus we know  $i \in \text{dom}(h)$ .
- (H10) From (H9) and the rules for get, we know we have  $\sigma_i = h(i)$  and  $\alpha(\sigma_i, \vec{v}_i') = (o, \sigma_i', \vec{v}_o')$ .
- **(H11)** If we have  $\sigma_f = (h_f, d_f)$  such that  $i \notin \text{dom}(h_f)$ , then it follows from (H2) that  $\sigma \# \sigma_f$  as the action does not modify any other cell. The action will also have the same outcome, completing our goal (G1).

- **(H12)** If we have  $\sigma_f = (h_f, d_f)$  such that  $i \in \text{dom}(h_f)$ , it follows from (H2) that  $\sigma'_i \# h_f(i)$ . We also know the outcome is not Miss from the rules of action execution and (H3).
- (H13) From (H12), (H10) and (H4) we can apply Frame Addition and get  $\sigma_i \# h_f(i)$  and  $\alpha(\sigma_i \cdot h_f(i), \vec{v}_i') = (o, \sigma_i' \cdot h_f(i), \vec{v}_o')$ .
- **(H14)** From (H13) and CPMAPACTION, we get  $\sigma \# \sigma_f$  and  $\alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma' \cdot \sigma_f, \vec{v}_o)$ , giving our goal (G1).

Case  $\alpha = \text{alloc}$ :

- **(H15)** From (H1), (H3) and the rules for alloc, we have  $\sigma' = (h', d')$  such that  $d' \neq \bot$ .
- **(H16)** From (H15), (H2) and the rules of composition it follows that  $\sigma_f = (h_f, \perp)$ .
- (H17) Because alloc does not modify any state aside from adding a cell, and because  $\sigma_f$  has no domain set, it follows from (H2) that  $\sigma \# \sigma_f$  and that the action has the saame outcome, thus  $\alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma' \cdot \sigma_f, \vec{v}_o)$ , giving our goal (G1).

# Proposition: Consume soundness

Assume

- (H1) consume $(\hat{\sigma}, \delta, \vec{\hat{v}}_i) \rightsquigarrow (Ok, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi)$
- **(H2)**  $\theta, s, \sigma \models \hat{\sigma} \land SAT_{\theta,s}(\pi)$
- **(H3)** consume on  $\mathbb{S}$  is sound.

To prove

(G1) 
$$\exists \sigma_{\delta}, \sigma_{f}, \sigma_{\delta} \# \sigma_{f} \land \sigma = \sigma_{\delta} \cdot \sigma_{f} \land \theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{v}_{i}; \vec{v}_{o}) \land \theta, s, \sigma_{f} \models \hat{\sigma}_{f}$$

Case  $\delta \in \Delta_{\mathbb{S}}$ :

- **(H4)** From (H1) and PMAPCONS, given  $\vec{v}_i = \hat{i} :: \vec{v}_i'$  and  $\pi = \pi' :: \pi''$  we have  $get(\hat{\sigma}, \hat{i}) \rightsquigarrow (\hat{i}', \hat{\sigma}_i, \pi')$ ,  $consume(\hat{\sigma}_i, \delta, \vec{v}_i') \rightsquigarrow (o, \hat{\sigma}_i', \vec{v}_o, \pi'')$  and  $set(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}_f$ .
- **(H5)** From (H4) and (H2), we have  $\vec{\hat{v}}_i = \hat{i} :: \vec{\hat{v}}_i' \wedge [[\hat{i}]]_{\theta,s} = i$
- **(H6)** From (H2) and Get UX Soundness, we know  $\exists \sigma_i. get(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$ .
- **(H7)** From (H2) and (H4), we have  $SAT_{\theta,s}(\pi'')$ .
- **(H8)** From (H6), (H7) and (H3) we can apply **??**, giving us  $\exists \sigma_{\delta,i}, \sigma_{f,i}. \sigma_{\delta,i} \# \sigma'_i \wedge \sigma_i = \sigma_{\delta,i} \cdot \sigma'_i \wedge \theta, s, \sigma_{\delta,i} \models_{\Delta} \langle \delta \rangle (\vec{v}'_i; \vec{v}_o) \wedge \theta, s, \sigma_{f,i} \models \hat{\sigma}_{f,i}.$

- **(H9)** We can then pick  $\sigma_{\delta} = ([i \mapsto \sigma_{\delta,i}], \bot)$ , and from (H8) and the definition of composition we know  $\sigma_{\delta} \# \sigma$  and (H4) and the definition of set,  $\sigma_f = \sigma \cdot \sigma_{\delta}$ .
- (H10) From the definition of  $\models_{\Delta}$  and (H8) we then have  $\theta, s, \sigma_{\delta} \models \langle \delta \rangle(\vec{v}_i; \vec{v}_o)$ , which with (H9) gives our goal (G1).

#### ${m Case} \ \delta = {\sf domainset:}$

- **(H11)** From (H1) and the rules for consume, we have  $\hat{\sigma} = (\hat{h}, \hat{d})$  and  $\hat{\sigma}_f = (\hat{h}, \perp)$  such that  $\hat{d} \neq \perp$ , as well as  $\vec{\hat{v}}_i = []$  and  $\vec{\hat{v}}_o = [\hat{d}]$
- **(H12)** From the definition of  $\models$  and (H11) we have  $\sigma = (h, d)$  and  $\sigma_f = (h, \bot)$  such that  $d \neq \bot$  and  $[\hat{d}]_{\theta,s} = d$ .
- **(H13)** We can pick  $\sigma_{\delta} = ([], d)$  it follows that  $\sigma_f \# \sigma_{\delta}$  and from the composition rules,  $\sigma = \sigma_f \cdot \sigma_{\delta}$ .
- (H14) From the definition of  $\models_{\Delta}$  and (H12) it follows that  $\theta, s, \sigma_{\delta} \models_{\Delta} \langle \mathsf{domainset} \rangle(\hat{v}_i; \hat{v}_o)$ . Along with (H13), this gives our goal (G1).

# Proposition: Consume completeness

### Assume

- (H1) consume $(\hat{\sigma}, \delta, \vec{\hat{v}_i}) \leadsto (\text{Ok}, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi)$
- **(H2)**  $\theta, s, \sigma_f \models \hat{\sigma}_f \land \theta, s, \sigma_\delta \models_\Delta \langle \delta \rangle (\hat{v}_i; \hat{v}_o) \land \sigma_f \# \sigma_\delta$
- **(H3)** consume on  $\mathbb{S}$  is complete.

# To prove

(G1) 
$$\exists \sigma. \sigma = \sigma_{\delta} \cdot \sigma_{f} \wedge \theta, s, \sigma \models \hat{\sigma} \wedge \mathsf{SAT}_{\theta,s}(\pi)$$

### Case $\delta \in \Delta_{\mathbb{S}}$ :

- **(H4)** From (H1) and PMAPCONS, given  $\vec{v}_i = \hat{i} :: \vec{v}_i'$  and  $\pi = \pi' :: \pi''$  we have  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi')$ ,  $consume(\hat{\sigma}_i, \delta, \vec{v}_i') \leadsto (o, \hat{\sigma}_i', \vec{v}_o, \pi'')$  and  $set(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}_f$ .
- **(H5)** There are two cases here,  $\sigma_{\delta} = \bot$  and  $\sigma_{\delta} \neq \bot$ . If  $\sigma_{\delta} = \bot$ , we trivially get  $\sigma = \bot \cdot \sigma_f$  and the rest of our goal (G1).
- **(H6)** Assume now  $\sigma_{\delta} \neq \bot$ . From the definition of  $\models_{\Delta}$  and (H2), we have  $\exists i, \sigma_{\delta,i}$ .  $[\hat{i}]_{\theta,s} = i \land \sigma_{\delta} = ([i \mapsto \sigma_{\delta,i}], \bot) \land \sigma_{\delta,i} \models_{\Delta} \langle \delta \rangle (\hat{\vec{v}}'_i; \hat{\vec{v}}_o)$ .

- **(H7)** From (H2), (H4) and the definitions of  $\models$  and composition we have  $\exists \sigma'_i. \theta, s, \sigma'_i \models \hat{\sigma}'_i \land \sigma'_i \# \sigma_{\delta,i}$ .
- **(H8)** From (H7), (H6) and (H3), we can apply ??, giving us  $\exists \sigma_i. \sigma_i = \sigma_{\delta,i} \cdot \sigma'_i \wedge \theta, s, \sigma_i \models \hat{\sigma}_i \wedge SAT_{\theta,s}(\pi'')$ .
- **(H9)** From the definition of  $\models$ , because **consume** doesn't affect any other cell, given  $\theta, s, \sigma_i \models \hat{\sigma}_i$  and  $\theta, s, \sigma_f \models \hat{\sigma}_f$  and  $\mathsf{set}(\hat{\sigma}, \hat{i}', \hat{\sigma}_i') = \hat{\sigma}_f$ , we obtain  $\theta, s, \sigma \models \hat{\sigma}$ .
- **(H10)** From (H6), we have  $\sigma_{\delta}$  of the form  $\sigma_{\delta} = (h_{\delta}, \perp)$  where  $i \in \text{dom}(h_{\delta})$ . From composition rules, it follows that if  $\sigma = \sigma_{\delta} \cdot \sigma_f$  we have  $\sigma = (h, d)$  where  $i \in \text{dom}(h)$  too.
- (H11) From (H10) and (H9), it follows that we can apply CPMAPGETMATCH such that  $get(\sigma, i) = \sigma_i$ . We can then use Get OX Soundness, giving us, from (H4) that  $SAT_{\theta,s}(\pi')$ .
- (H12) It follows from (H11) and (H8) that  $SAT_{\theta,s}(\pi)$ . This along with (H9) gives our goal (G1).

 ${m Case} \ \delta = {\sf domainset:}$ 

- **(H13)** From (H1) and the rules for consume, we have  $\hat{\sigma} = (\hat{h}, \hat{d})$  and  $\hat{\sigma}_f = (\hat{h}, \perp)$  such that  $\hat{d} \neq \perp$ , as well as  $\vec{\hat{v}}_i = []$ ,  $\vec{\hat{v}}_o = [\hat{d}]$  and  $\pi = []$
- **(H14)** From the definition of  $\models_{\Delta}$  and (H2) we have  $\sigma_{\delta} = (\emptyset, d)$  such that  $[\![\hat{d}]\!]_{\theta,s} = d$ .
- **(H15)** It follows from (H13), (H2) and the definition of  $\models$  that  $\sigma_f$  is of the form  $\sigma_f = (h, \bot)$ .
- **(H16)** From (H14), (H15) and the definition of  $\models$  it follows that for  $\sigma = \sigma_{\delta} \cdot \sigma_{f}$ , we have  $\theta, s, \sigma \models \hat{\sigma}$ . This with the fact  $SAT_{\theta,s}([])$  gives our goal (G1).

# Proposition: Consume: OX sound

Assume

$$\textbf{(H1)} \ \forall o, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi. \, \text{consume}(\texttt{OX}, \hat{\sigma}, \delta, \vec{\hat{v}}_i) \leadsto (o, \hat{\sigma}_f, \vec{\hat{v}}_o, \pi) \Rightarrow o_c = \texttt{Ok}$$

(H2) consume on S are OX sound.

To prove

$$\textbf{(G1)} \ \exists \hat{\sigma}_f', \vec{\hat{v}}_o', \pi'. \ \mathsf{consume}(\mathtt{OX}, \hat{\sigma}, \delta, \vec{v}_i, \pi) \leadsto (\mathtt{Ok}, \hat{\sigma}_f', \vec{\hat{v}}_o', \pi')$$

# Case $\delta \in \Delta_{\mathbb{S}}$ :

From PMAPPROD we know consume for  $\delta \in \Delta_{\mathbb{S}}$  only returns 0k if the underlying's consume returns 0k. It follows that if that is the case, then by (H2) there exists an execution of consume that succeeds for  $\mathbb{S}$ , that is lifted by PMAP, giving us our goal (G1).

#### $Case \ \delta = domainset:$

From the consume rules, we never vanish, so if all consumptions are 0k we must have  $\hat{\sigma} = (\hat{h}, \hat{d})$  with  $\hat{d} \neq \bot$ ,  $\vec{v}_i = []$ ,  $\hat{\sigma}'_f = (\hat{h}, \bot)$ ,  $\vec{v}'_o = [\hat{d}]$ ,  $\pi' = []$ , which gives our goal (G1).

#### Proposition: Produce soundness

#### Assume

- **(H1)** produce $(\hat{\sigma}_f, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto (\hat{\sigma}, \pi)$
- **(H2)** SAT<sub> $\theta,s$ </sub> $(\pi) \wedge \theta, s, \sigma \models \hat{\sigma}$
- (H3) produce on  $\mathbb{S}$  is sound.

# To prove

(G1) 
$$\exists \sigma_{\delta}, \sigma_{f}, \sigma_{\delta} \# \sigma_{f} \land \sigma = \sigma_{\delta} \cdot \sigma_{f} \land \theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{v}_{i}; \vec{v}_{o}) \land \theta, s, \sigma_{f} \models \hat{\sigma}_{f}$$

# Case $\delta \in \Delta_{\mathbb{S}}$ :

- **(H4)** Given (H1), from PMAPPROD we have  $\operatorname{get}(\hat{\sigma}_f, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_{f,i}, \pi') \land \operatorname{produce}(\hat{\sigma}_{f,i}, \delta, \vec{v}'_i, \vec{v}_o) \leadsto (\hat{\sigma}_i, \pi'') \land \operatorname{set}(\hat{\sigma}_f, \hat{i}', \hat{\sigma}_i) = \hat{\sigma}, \text{ given } \vec{v}_i = \hat{i} :: \vec{v}'_i \text{ and } \pi = \pi' :: \pi''.$
- **(H5)** From (H2) and  $\pi = \pi' :: \pi''$ , it follows that  $SAT_{\theta,s}(\pi')$  and  $SAT_{\theta,s}(\pi'')$ .
- **(H6)** From the definition of set, (H4) and (H2), we have  $\exists \sigma_i. \theta, s, \sigma_i \models \hat{\sigma}_i.$
- (H7) From (H4), (H6), (H5) and (H3), we can apply ??, giving us  $\exists \sigma_{\delta,i}, \sigma_{f,i}. \sigma_{\delta,i} \# \sigma_{f,i} \land \sigma_{i} = \sigma_{\delta,i} \cdot \sigma_{f,i} \land \theta, s, \sigma_{\delta,i} \models_{\Delta} \langle \delta \rangle (\vec{v}'_{i}; \vec{v}_{o}) \land \theta, s, \sigma_{f,i} \models \hat{\sigma}_{f,i}.$
- **(H8)** Given  $[\hat{i}']_{\theta,s} = i$ , we can then have  $\sigma_{\delta} = ([i \mapsto \sigma_{\delta,i}], \bot)$ , or  $\sigma_{\delta} = \bot$  if  $\sigma_{\delta,i} = \bot$ . By the definition of  $\models_{\Delta}$ ,  $\theta, s, \sigma_{\delta} \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o)$  in both cases.
- (H9) Given set only modifies the state at a specific location leaving the rest untouched, from (H2) and (H7) we have  $\theta, s, \sigma \models \hat{\sigma}$  and  $\theta, s, \sigma_{f,i} \models \hat{\sigma}_{f,i}$ , thus it follows that  $\exists \sigma_f. \theta, s, \sigma_f \models \sigma_f.$
- (H10) From (H7) and by the rules of composition, we know that if the two states  $\sigma_{\delta,i}$  and  $\sigma_{f,i}$  at index i are disjoint then the entire state  $\sigma_f$  and the singleton  $\sigma_{\delta}$  are disjoint. Thus  $\sigma_{\delta} \# \sigma_f \wedge \sigma = \sigma_{\delta} \cdot \sigma_f$ , giving our goal (G1).

 ${m Case} \ \delta = {\sf domainset:}$ 

- **(H11)** Given (H1), from PMAPPRODDOMAINSET we have  $\hat{\sigma}_f = (\hat{h}, \perp)$ ,  $\hat{\sigma} = (\hat{h}, \hat{d})$  and  $\pi = [\text{dom}(\hat{h}) \subseteq \hat{d}]$ , with  $\vec{\hat{v}}_i = []$  and  $\vec{\hat{v}}_o = [\hat{d}]$ .
- **(H12)** We can pick  $\sigma_{\delta}$  such that  $\sigma_{\delta} = (\emptyset, d)$  with  $[\![\hat{d}]\!]_{\theta,s} = d$ , which from the definition of  $\models_{\Delta}$  gives  $\theta, s, \sigma_{\delta} \models_{\Delta} \langle \mathsf{domainset} \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o)$ .
- **(H13)** We can pick  $\sigma_f$  such that  $\sigma_f = (h, \perp)$  and  $[\![\hat{h}]\!]_{\theta,s} = h$ . From the definition of  $\models$ , we have  $\theta, s, \sigma_f \models \hat{\sigma}_f$ .
- (H14) From (H12), (H13) and composition rules we have  $\sigma_f \# \sigma_\delta$  and  $\sigma = \sigma_f \cdot \sigma_\delta$ . This completes our goal (G1).

#### Proposition: Produce completeness

Assume

**(H1)** 
$$\theta, s, \sigma_f \models \hat{\sigma}_f \land \theta, s, \sigma_\delta \models_{\Delta} \langle \delta \rangle (\vec{\hat{v}}_i; \vec{\hat{v}}_o) \land \sigma_f \# \sigma_\delta$$

(H2) produce on S is complete.

To prove

(G1) 
$$\exists \hat{\sigma}. \operatorname{produce}(\hat{\sigma}_f, \delta, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \rightsquigarrow (\hat{\sigma}, \pi) \land \operatorname{SAT}_{\theta, s}(\pi) \land \theta, s, (\sigma_f \cdot \sigma_\delta) \models \hat{\sigma}$$

Case  $\delta \in \Delta_{\mathbb{S}}$ :

- **(H3)** From (H1) and the definition of  $\models_{\Delta}$ , we have  $\exists \hat{i}, i, \sigma_{\delta,i}. \vec{v}_i = \hat{i} :: \vec{v}'_i \wedge [\![\hat{i}]\!]_{\theta,s} = i \wedge \theta, s, \sigma_{\delta,i} \models_{\Delta} \langle \delta \rangle (\vec{v}'_i; \vec{v}_o) \wedge \sigma_{\delta} = ([i \mapsto \sigma_{\delta,i}], \bot).$
- (H4) Given the fact  $\sigma_f \# \sigma_\delta$  and  $\sigma_\delta \models_\Delta \langle \delta \rangle (\hat{i} :: \vec{v}_i'; \vec{v}_o)$ , it means the index i is compatible with  $\sigma_f$ , so the state located at i is obtainable via get. We thus have  $\exists \sigma_{f,i} . \operatorname{get}(\sigma_f, i) = \sigma_{f,i}$ .
- (H5) From (H4), (H3) and (H1), we can use Get OX Soundness, giving us  $\exists \hat{\sigma}_{f,i}, \hat{i}', \pi'$ .  $\mathsf{get}(\hat{\sigma}_f, \hat{i}) \rightsquigarrow (\hat{i}', \hat{\sigma}_{f,i}, \pi) \land \theta, s, \sigma_{f,i} \models \hat{\sigma}_{f,i} \land [\hat{i}']_{\theta,s} = i \land \mathsf{SAT}_{\theta,s}(\pi')$ .
- (**H6**) From (H1), because  $\sigma_{f,i}$  and  $\sigma_{\delta,i}$  are at the same index and  $\sigma_f \# \sigma_{\delta}$ , from the rules for composition we have  $\sigma_{f,i} \# \sigma_{\delta,i}$ .
- (H7) From (H3), (H5) and (H6) we can apply ?? for  $\mathbb{S}$ , giving us  $\exists \hat{\sigma}_i$ . produce  $(\hat{\sigma}_{f,i}, \delta, \vec{v}_i', \vec{v}_o) \leadsto (\hat{\sigma}_i, \pi'') \land SAT_{\theta,s}(\pi'') \land \theta, s, (\sigma_{f,i} \cdot \sigma_{\delta,i}) \models \hat{\sigma}_i$ .
- **(H8)** We may define  $\hat{\sigma}$  such that  $set(\hat{\sigma}_f, \hat{i}', \hat{\sigma}_i) = \hat{\sigma}$ .

- **(H9)** We have  $\theta$ , s,  $\sigma_f \models \hat{\sigma}_f$  and  $\theta$ , s,  $(\sigma_{f,i} \cdot \sigma_{\delta,i}) \models \hat{\sigma}_i$  from (H1) and (H7). Furthermore, we know  $\sigma_{\delta} = ([i \mapsto \sigma_{\delta,i}], \bot)$  from (H3). From (H8) and the rules for set we know no state is lost and only the state at i is extended by  $\sigma_{\delta,i}$ , it thus follows that  $\theta$ , s,  $(\sigma_f \cdot \sigma_{\delta}) \models \hat{\sigma}$ .
- **(H10)** From and (H5), (H7) and this we can apply PMAPPROD, giving us  $produce(\hat{\sigma}_f, \delta, \vec{v}_i, \vec{v}_o) \rightsquigarrow (\hat{\sigma}, \pi' :: \pi'')$ .
- (H11) From (H5) and (H7), we have  $SAT_{\theta,s}(\pi' :: \pi'')$ . This, along with (H9) and (H10), gives our goal (G1).

#### ${m Case} \ \delta = {\sf domainset:}$

- **(H12)** From (H1) and the definition of  $\models_{\Delta}$  we have  $\sigma_{\delta} = (\emptyset, d)$  such that  $\vec{\hat{v}}_i = [], \ \vec{\hat{v}}_o = [\hat{d}]$  and  $[\![\hat{d}]\!]_{\theta,s} = d$ .
- **(H13)** From (H12) and (H1), by the definition of composition, we must have  $\hat{\sigma}_f = (\hat{h}, \perp)$ ,  $\sigma_f = (h, \perp)$  and  $\text{dom}(h) \subseteq d$ .
- **(H14)** From (H12) and (H13) we may apply PMAPPRODDOMAINSET, resulting in  $produce(\hat{\sigma}_f, domainset, \vec{\hat{v}}_i, \vec{\hat{v}}_o) \leadsto ((\hat{h}, \hat{d}), [dom(\hat{h}) \subseteq \hat{d}]).$
- **(H15)** From (H1) and (H13) we have  $SAT_{\theta,s}([dom(\hat{h}) \subseteq \hat{d}])$ .
- **(H16)** From (H12), (H13) and composition rules we have we have  $\sigma_f \cdot \sigma_\delta = (h, d)$ . It follows from (H14) that we have  $\hat{\sigma} = (\hat{h}, \hat{d})$  such that  $\theta, s, (\sigma_f \cdot \sigma_\delta) \models \hat{\sigma}$ . This along with (H14) and (H15) gives our goal (G1).

# **B.3** Syntactic Partial Map

For the partial map with syntactic checks, we re-use the RA of PMAP, as well as the concrete and symbolic action rules, and the produce and consume rules, as we only modify the behaviour of the get helper function. From this, we can thus re-use the soundness proofs for all of the axioms; the only part that needs to be proved again are the axioms for symbolic get, with regards to the concrete version that is taken from PMAP.

# B.3.1 get rules

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \in \text{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i})}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{\sigma}_i, [])}$$

SYNTACTICPMAPGETBRANCH

$$\frac{(\hat{h}, \hat{d}) = \mathit{unwrap}(\hat{\sigma}) \qquad \hat{i} \notin \mathrm{dom}(\hat{h}) \qquad \hat{i}' \in \mathrm{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(\hat{i}')}{\mathrm{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])}$$

 ${\bf Syntactic PMap Get Add}$ 

$$\frac{(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{i} \notin \text{dom}(\hat{h}) \qquad \hat{d} \neq \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}) \land \hat{i} \in \hat{d}])}$$

SYNTACTICPMAPGETBOTDOMAIN

$$\frac{(\hat{h},\hat{d}) = \mathit{unwrap}(\hat{\sigma}) \qquad \hat{i} \not\in \mathrm{dom}(\hat{h}) \qquad \hat{d} = \bot}{\mathrm{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i},\bot,[\hat{i} \not\in \mathrm{dom}(\hat{h})])}$$

## **B.3.2** Soundness Proofs

Proof.

# Proposition: Get OX Soundness

Assume

**(H1)** 
$$\theta, s, \sigma \models \hat{\sigma} \land \text{get}(\sigma, i) = \sigma_i \land [\hat{i}]_{\theta, s} = i$$

To prove

$$\textbf{(G1)} \ \exists \hat{\sigma}_i, \hat{i}', \pi. \, \text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land [\![\hat{i}']\!]_{\theta, s} = i \land \mathsf{SAT}_{\theta, s}(\pi)$$

We proceed by proving the property holds for all rules of the concrete get, resulting in three cases.

Case CPMAPGETMATCH:

- **(H2)** Assume  $(h, d) = unwrap(\sigma) \land i \in dom(h) \land \sigma_i = h(i)$ .
- **(H3)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \hat{d})$ .
- **(H4)** Either  $\hat{i} \in \hat{h}$  or not. If  $\hat{i} \in \hat{h}$ , we apply SyntacticPMapGetMatch, giving us  $\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{h}(\hat{i}), [])$ . It also follows from the definition of  $\models$  that  $\theta, s, \sigma_i \models \hat{h}(\hat{i})$ , completing our goal (G1).

**(H5)** Otherwise  $\hat{i} \notin \hat{h}$ . From the definition of  $\models$  we know  $\exists \hat{i}'$ .  $[\hat{i}']_{\theta,s} = i \land \hat{i}' \in \hat{h} \land \theta, s, \sigma_i \models \hat{h}(\hat{i}')$ . We can apply SyntacticPMapGetBranch, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])$ . It follows that  $SAT_{\theta,s}([\hat{i} = \hat{i}'])$ , which completes our goal (G1).

Case CPMAPGETADD: The rule SYNTACTICPMAPGETADD is exactly the same as PMAPGETADD – we omit the proof.

Case CPMAPGETBOTDOMAIN: The rule SYNTACTICPMAPGETBOTDOMAIN is exactly the same as PMAPGETBOTDOMAIN – we omit the proof.

### Proposition: Get UX Soundness

#### Assume

(H1) 
$$get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land [\hat{i}]_{\theta,s} = [\hat{i}']_{\theta,s} = i \land SAT_{\theta,s}(\pi)$$

**(H2)** 
$$\theta, s, \sigma \models \hat{\sigma}$$
.

#### To prove

**(G1)** 
$$\exists \sigma_i. get(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$$

We proceed by proving the property holds for all rules of the symbolic get, resulting in four cases.

# Case SyntacticPMapGetMatch:

- **(H3)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \land \hat{i} \in dom(\hat{h}) \land \hat{\sigma}_i = \hat{h}(\hat{i})$ .
- (H4) From (H3), (H1) and SyntacticPMapGetMatch we have  $\pi = []$  and  $\hat{i}' = \hat{i}$ .
- **(H5)** From (H3), we know  $\hat{\sigma} \neq \bot$ , thus  $\sigma$  is of the form  $\sigma = (h, d)$ .
- **(H6)** From (H2), (H3) and the definition of  $\models$ , we have  $i \in h$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$ .
- (H7) From (H6) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

#### Case SYNTACTICPMAPGETBRANCH:

**(H8)** Assume 
$$(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{i}' \in dom(\hat{h}) \wedge \hat{\sigma}_i = \hat{h}(\hat{i}')$$

**(H9)** From (H8), (H1) and PMAPGETMATCH, we have 
$$\pi = [\hat{i} = \hat{i}']$$

**(H10)** From (H8), we have 
$$\hat{\sigma} \neq \bot$$
, thus from (H2) we have  $\sigma = (h, d)$ 

- **(H11)** From (H2), (H8) and the definition of  $\models$ , we have  $i \in h$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$
- (H12) From (H11) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

Case SYNTACTICPMAPGETADD: The rule SYNTACTICPMAPGETADD is exactly the same as PMAPGETADD – we omit the proof.

Case SyntacticPMapGetBotDomain: The rule SyntacticPMapGetBotDomain is exactly the same as PMapGetBotDomain – we omit the proof.  $\Box$ 

# B.4 Split Partial Map

For the split partial map, we define a new RA, PMAP<sub>SPLIT</sub>, which we only use for the symbolic representation; the concrete compositional RA is still that defined in PMAP. We re-use the definitions for action execution, consume and produce, as the only part of them that is modified is get and set. From this, we can thus re-use the soundness proofs for all of the axioms; the only part that needs to be proved again are the axioms for symbolic get, with regards to the concrete version that is taken from PMAP.

# B.4.1 Resource Algebra

While we keep the default PMAP RA for the concrete compositional states, we need to define a new set of states for the symbolic compositional states, as defined below.

$$\mathrm{PMap}_{\mathrm{Split}}(I,\mathbb{S}) \triangleq I \xrightarrow{\mathit{fin}} \mathbb{S}.\Sigma \times I \xrightarrow{\mathit{fin}} \mathbb{S}.\Sigma \times \mathcal{P}(I)^?$$

Predicate satisfiability is defined with regards to the concrete compositional states, so we re-use it from PMAP. We however need to re-define symbolic interpretation, as follows. We denote  $\hat{h}_c$  and  $\hat{h}_s$  the concrete and symbolic parts of the heap respectively.

$$\begin{split} & \text{SplitPMapSymInterpretation} \\ & \forall \hat{i} \in \text{dom}(\hat{h}_c). \, \hat{i} \notin \hat{h}_s \wedge \llbracket \hat{i} \rrbracket_{\theta,s} = i \wedge i \in \text{dom}(h) \wedge \theta, s, h(i) \models \hat{h}_c(\hat{i}) \\ & \forall \hat{i} \in \text{dom}(\hat{h}_s). \, \hat{i} \notin \hat{h}_c \wedge \llbracket \hat{i} \rrbracket_{\theta,s} = i \wedge i \in \text{dom}(h) \wedge \theta, s, h(i) \models \hat{h}_s(\hat{i}) \\ & \underline{ \llbracket \text{dom}(\hat{h}_c) \uplus \text{dom}(\hat{h}_s) \rrbracket_{\theta,s} = \text{dom}(h) } & \underline{ \llbracket \hat{d} \rrbracket_{\theta,s} = d } \\ & \underline{ \theta, s, (h, d) \models (\hat{h}_c, \hat{h}_s, \hat{d}) } \end{split}$$

# B.4.2 get and set rules

Given 
$$wrap(\hat{h}_c, \hat{h}_s, \hat{d}) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(\hat{h}_c) = \emptyset \wedge \operatorname{dom}(\hat{h}_s) = \emptyset \wedge \hat{d} = \emptyset \\ (\hat{h}_c, \hat{h}_s, \hat{d}) & \text{otherwise} \end{cases}$$

$$unwrap(\hat{\sigma}) \triangleq \begin{cases} ([], [], \emptyset) & \text{if } \hat{\sigma} = \bot \\ (\hat{h}_c, \hat{h}_s, \hat{d}) & \text{if } \hat{\sigma} = (\hat{h}_c, \hat{h}_s, \hat{d}) \end{cases}$$

SPLITPMAPGETMATCHCON

$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \text{is\_concrete } \hat{i} \qquad \hat{i} \in \text{dom}(\hat{h}_c) \qquad \hat{\sigma}_i = \hat{h}_c(\hat{i})}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{\sigma}_i, ||)}$$

 ${\bf SPLITPMAPGETMATCHSYM}$ 

$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = \mathit{unwrap}(\hat{\sigma}) \qquad \hat{i} \in \mathrm{dom}(\hat{h}_s) \qquad \hat{\sigma}_i = \hat{h}_s(i)}{\mathrm{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{\sigma}_i, [])}$$

SPLITPMAPGETBRANCH

$$\begin{split} (\hat{h}_c, \hat{h}_s, \hat{d}) &= unwrap(\hat{\sigma}) \\ \frac{\hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \qquad \hat{i} \notin \text{dom}(\hat{h}_{all}) \qquad \hat{i}' \in \text{dom}(\hat{h}_{all}) \qquad \hat{\sigma}_i = \hat{h}_{all}(\hat{i}')}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])} \end{split}$$

SPLITPMAPGETADD

$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \qquad \hat{i} \notin \text{dom}(\hat{h}_{all}) \qquad \hat{d} \neq \bot}{\text{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}_{all}) \land \hat{i} \in \hat{d}])}$$

SPLITPMAPGETBOTDOMAIN

$$\frac{(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \qquad \hat{i} \notin \text{dom}(\hat{h}_{all}) \qquad \hat{d} = \bot}{\text{get}(\hat{\sigma}, \hat{i}) \rightsquigarrow (\hat{i}, \bot, [\hat{i} \notin \text{dom}(\hat{h}_{all})])}$$

SPLITPMAPSETSOMECON

$$\begin{split} \hat{(h_c,\hat{h}_s,\hat{d})} &= unwrap(\hat{\sigma}) & \quad \hat{\sigma}_i \neq \bot \quad \text{is\_concrete}_{\Sigma} \; \hat{\sigma}_i \\ & \quad \text{is\_concrete} \; \hat{i} \quad \; \hat{h}'_c = \hat{h}_c[\hat{i} \leftarrow \hat{\sigma}_i] \quad \; \hat{h}'_s = \hat{h}_s[\hat{i} \not\leftarrow] \quad \quad \hat{\sigma}' = wrap(\hat{h}'_c,\hat{h}'_s,\hat{d}) \\ & \quad \text{set}(\hat{\sigma},\hat{i},\hat{\sigma}_i) = \hat{\sigma}' \end{split}$$

SPLITPMAPSETSOMESYM

$$(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \qquad \hat{s}_i \neq \bot \qquad \neg (\text{is\_concrete}_{\Sigma} \ \hat{\sigma}_i \lor \text{is\_concrete} \ \hat{i})$$

$$\hat{h}'_c = \hat{h}_c[\hat{i} \not\leftarrow] \qquad \hat{h}'_s = \hat{h}_s[\hat{i} \leftarrow \hat{\sigma}_i] \qquad \hat{\sigma}' = wrap(\hat{h}'_c, \hat{h}'_s, \hat{d})$$

$$\text{set}(\hat{\sigma}, \hat{k}, \hat{\sigma}_i) = \hat{\sigma}'$$

SPLITPMAPSETNONE

$$\begin{split} \hat{(h_c,\hat{h}_s,\hat{d})} &= unwrap(\hat{\sigma}) \\ \frac{\hat{\sigma}_i = \bot \qquad \hat{h}_c' = \hat{h}_c[\hat{i} \not\leftarrow] \qquad \hat{h}_s' = \hat{h}_s[\hat{i} \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}_c',\hat{h}_s',\hat{d})}{\text{set}(\hat{\sigma},\hat{i},\hat{\sigma}_i) = \hat{\sigma}'} \end{split}$$

# **B.4.3** Soundness Proofs

Proof.

Proposition: Get OX Soundness

Assume

**(H1)** 
$$\theta, s, \sigma \models \hat{\sigma} \land \text{get}(\sigma, i) = \sigma_i \land [\hat{i}]_{\theta, s} = i$$

To prove

(G1) 
$$\exists \hat{\sigma}_i, \hat{i}', \pi. \operatorname{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land \|\hat{i}'\|_{\theta,s} = i \land SAT_{\theta,s}(\pi)$$

We proceed by proving the property holds for all rules of the concrete get, resulting in three cases.

#### Case CPMAPGETMATCH:

- **(H2)** Assume  $(h, d) = unwrap(\sigma) \land i \in dom(h) \land \sigma_i = h(i)$ .
- **(H3)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}_c, \hat{h}_s, \hat{d})$ . From here we have three cases:  $\hat{i}$  is present directly in  $\hat{h}_c$ , or in  $\hat{h}_s$ , or neither.
- (H4) From (H3), assume the binding is in the concrete part of the heap:  $\hat{i} \in \hat{h}_c$ . From the definition of set, we know that if an entry is in the concrete part it must be concrete; thus is\_concrete  $\hat{i}$ . From this, we can apply SPLITPMAPGETMATCHCON, giving us  $\text{get}(\hat{\sigma},\hat{i}) \leadsto (\hat{i},\hat{h}_c(\hat{i}),[])$ . From the definition of  $\models$ , it also follows that  $\theta, s, \sigma_i \models \hat{h}_c(\hat{i})$ . This, along with the fact SAT $_{\theta,s}([])$ , gives our goal (G1).
- (H5) Assume now from (H3) that the binding is in the symbolic part of the heap,  $\hat{h}_s$ . From this, we can apply SplitPMAPGETMATCHSYM, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \hat{h}_s(\hat{i}), [])$ . From the definition of  $\models$ , it also follows that  $\theta, s, \sigma_i \models \hat{h}_s(\hat{i})$ . This again gives our goal (G1).

(H6) Finally, it may be that  $\hat{i} \notin \hat{h}_c \wedge \hat{i} \notin \hat{h}_s$ . From the definition of  $\models$ , there must however still exist a  $\hat{i}'$  such that  $[\hat{i}']_{\theta,s} = i$  and given  $\hat{h}_{all} = \hat{h}_c \cup \hat{h}_s$ ,  $\theta, s, \sigma_i \models \hat{h}_{all}(\hat{i}')$ . From this we may apply SPLITPMAPGETBRANCH, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, [\hat{i} = \hat{i}'])$ . Because we have  $[\hat{i}]_{\theta,s} = i$ , it follows that  $SAT_{\theta,s}([\hat{i} = \hat{i}'])$ , completing our goal (G1).

### Case CPMAPGETADD:

- **(H7)** Assume  $(h, d) = unwrap(\sigma) \land i \notin dom(h) \land d \neq \bot \land i \in d$ .
- **(H8)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}_c, \hat{h}_s, \hat{d})$  such that  $[\hat{i}]_{\theta,s} = i$ ,  $\hat{i} \in \hat{d}$  and  $\hat{i} \notin \text{dom}(\hat{h}_c \cup \hat{h}_s)$ .
- **(H9)** We can then apply SplitPMapGetAdd, giving us  $get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \perp, [\hat{i} \notin dom(\hat{h}_c \cup \hat{h}_s) \land \hat{i} \notin \hat{d}])$ .
- **(H10)** From (H8) it follows that  $SAT_{\theta,s}([\hat{i} \notin dom(\hat{h}) \land \hat{i} \notin \hat{d}])$ , which completes our goal (G1).

#### Case CPMAPGETBOTDOMAIN:

- **(H11)** Assume  $(h, d) = unwrap(\sigma) \land i \notin dom(h) \land d = \bot$ .
- **(H12)** From (H11), (H1) and CPMAPGETBOTDOMAIN, we have  $\sigma_i = \bot$ .
- **(H13)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}_c, \hat{h}_s, \bot)$  such that  $[\hat{i}]_{\theta,s} = i$  and  $\hat{i} \notin \text{dom}(\hat{h}_c \cup \hat{h}_s)$ .
- **(H14)** We can then apply SplitPMapGetBotDomain, giving us  $\mathsf{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}, \bot, [\hat{i} \notin \mathsf{dom}(\hat{h}_c \cup \hat{h}_s)])$ .
- (H15) From (H13) it follows that  $SAT_{\theta,s}([\hat{i} \notin dom(\hat{h}_c \cup \hat{h}_s)])$ , which completes our goal (G1).

# Proposition: Get UX Soundness

# Assume

(H1) 
$$get(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land [\hat{i}]_{\theta,s} = [\hat{i}']_{\theta,s} = i \land SAT_{\theta,s}(\pi)$$

**(H2)** 
$$\sigma.\theta, s, \sigma \models \hat{\sigma}$$

#### To prove

**(G1)** 
$$\exists \sigma_i. get(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$$

We proceed by proving the property holds for all rules of the symbolic get, resulting in five cases.

## Case SplitPMapGetMatchCon:

- **(H3)** Assume  $(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \wedge is\_concrete \hat{i} \wedge \hat{i} \in dom(\hat{h}_c) \wedge \hat{\sigma}_i = \hat{h}_c(\hat{i}).$
- (H4) From (H3), (H1) and SPLITPMAPGETMATCHCON we have  $\pi = []$  and  $\hat{i}' = \hat{i}$ .
- **(H5)** From (H3), we know  $\hat{\sigma} \neq \bot$ , thus  $\sigma$  is of the form  $\sigma = (h, d)$ .
- **(H6)** From (H2), (H3) and the definition of  $\models$ , we have  $i \in \text{dom}(h)$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$ .
- (H7) From (H6) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

Case SplitPMapGetMatchSym: This is proved analogously to the above.

#### Case SplitPMapGetBranch:

- **(H8)** Assume  $(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \wedge \hat{i} \notin dom(\hat{h}_{all}) \wedge \hat{i}' \in dom(\hat{h}_{all}) \wedge \hat{\sigma}_i = \hat{h}_{all}(\hat{i}').$
- **(H9)** From (H8), (H1) and SPLITPMAPGETBRANCH we have  $\pi = [\hat{i} = \hat{i}']$ .
- **(H10)** From (H8) we know  $\hat{\sigma} \neq \bot$ , thus  $\sigma$  is of the form  $\sigma = (h, d)$ .
- **(H11)** From (H8), (H2) and the definition of  $\models$ , we have  $i \in \text{dom}(h)$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$ .
- (H12) From (H11) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

# Case SplitPMapGetAdd:

- **(H13)** Assume  $(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \wedge \hat{i} \notin dom(\hat{h}_{all}) \wedge \hat{d} \neq \bot$ .
- (H14) From (H13), (H1) and SPLITPMAPGETADD we have  $\pi = [\hat{i} \notin \text{dom}(\hat{h}_{all}) \land \hat{i} \in \hat{d}],$   $\hat{i}' = \hat{i} \text{ and } \hat{\sigma}_i = \bot.$
- **(H15)** From (H13) we know  $\hat{\sigma} \neq \bot$ , thus  $\sigma$  is of the form  $\sigma = (h, d)$ .
- **(H16)** From (H1) and (H14), we know  $SAT_{\theta,s}(\pi)$ , such that from (H2) we have  $i \notin dom(h) \land i \in d$ .

(H17) From (H16) we can straightforwardly apply CPMAPGETADD, giving us  $get(\sigma, i) = \bot$  – together with (H14) and Empty Memory this completes our goal (G1).

Case SplitPMapGetBotDomain:

**(H18)** Assume 
$$(\hat{h}_c, \hat{h}_s, \hat{d}) = unwrap(\hat{\sigma}) \wedge \hat{h}_{all} = \hat{h}_c \cup \hat{h}_s \wedge \hat{i} \notin dom(\hat{h}_{all}) \wedge \hat{d} = \bot$$
.

- (H19) From (H18), (H1) and SPLITPMAPGETBOTDOMAIN we have  $\pi = [\hat{i} \notin \text{dom}(\hat{h}_{all})]$  and  $\hat{i} = \hat{i}$ .
- **(H20)** From (H18) we have  $\hat{d} = \bot$ , so given  $(h, d) = unwrap(\sigma)$  and (H2), we have  $d = \bot$ .
- **(H21)** From (H1)  $SAT_{\theta,s}(\pi)$ , thus from (H19) and (H2),  $i \notin dom(h)$ .
- (H22) From (H21) and (H20) we can apply CPMAPGETBOTDOMAIN, thus  $get(\sigma, i) = \bot$  together with (H19) and Empty Memory this completes our goal (G1).

# **B.5** Abstract Location Partial Map

For the abstract location partial map, we define a new set of states, PMAP<sub>ALOC</sub>, which we only use for the symbolic representation; the concrete compositional RA is still that defined in PMAP.

For the PMAP with syntactic matching and PMAP<sub>SPLIT</sub> we considered compatibility with any "regular" PMAP that has the same domain I. Here however, because PMAP<sub>ALoc</sub> enforces the domain be abstract locations (as strings), we must consider compatibility with regards to PMAP(Loc,  $\mathbb{S}$ ) only.

Concrete locations (or just locations) are values of the form loc(a), where a is the name of the location – it always holds that given two locations loc(a) and loc(b),  $loc(a) = loc(b) \iff a = b$ . They are uninterpreted values, and do not allow any operations; as such, they can be used as sorts of pointers, where they represent address in a map, to the difference that they do not allow pointer arithmetics.

Abstract locations are *symbolic* values of the form aloc(a), where a is the name of the abstract location. Unlike with concrete locations, their name does not uniquely identify them: we may have a substitution  $\theta$  such that for aloc(a) and aloc(b) with  $a \neq b$ , we have  $[aloc(a)]_{\theta,s} = [aloc(b)]_{\theta,s}$ .

We also introduce the function  $to_aloc$ , that returns the name of an abstract location associated with a symbolic value if it exists and is found, and  $\bot$  otherwise. This is a best

effort function, that may not find an abstract location that exists.

to\_aloc 
$$i = a \implies \exists a. [[i]]_{\theta,s} = [[aloc(a)]]_{\theta,s}$$

# B.5.1 Resource Algebra

$$PMAP_{ALoc}(S) \triangleq Str \xrightarrow{fin} S.\Sigma \times \mathcal{P}(LVal)^?$$

# B.5.2 get and set rules

We now present the rules for this state model; in particular, we again only need to concern ourselves with the get and set internal methods. We also extend get to receive a mode  $M = \{\text{MATCH}, \text{NO\_MATCH}\}$ ; it is MATCH in consume and produce, and NO\_MATCH during action execution. The rules for execute\_action, consume and produce are omitted as they are analogous to those in PMAP, to the exception that get receives a matching mode. To simplify the rules, we also add the notation  $\in$ ?, that checks for membership in a possibly  $\perp$  set, in which case the result is true.

Given 
$$wrap(\hat{h}, \hat{d}) \triangleq \begin{cases} \bot & \text{if } \operatorname{dom}(\hat{h}) = \emptyset \land \hat{d} = \bot \\ (\hat{h}, \hat{d}) & \text{otherwise} \end{cases}$$
 
$$unwrap(\hat{\sigma}) \triangleq \begin{cases} ([], \bot) & \text{if } \hat{\sigma} = \bot \\ (\hat{h}, \hat{d}) & \text{if } \hat{\sigma} = (\hat{h}, \hat{d}) \end{cases}$$
 
$$a \in \hat{d} \triangleq \begin{cases} \text{true} & \text{if } \hat{d} = \bot \\ a \in \hat{d} & \text{otherwise} \end{cases}$$

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \; \hat{i} \qquad a \neq \bot \qquad a \in \mathsf{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(a)}{\mathsf{get}(\hat{\sigma},\hat{i},m) \leadsto (\mathsf{aloc}(a),\hat{\sigma}_i, [])}$$

ALOCPMAPGETNOMATCHNOTFOUND

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \ \hat{i} \qquad a \neq \bot \qquad a \notin \mathsf{dom}(\hat{h})}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{NO\_MATCH}) \leadsto (\mathsf{aloc}(a),\bot,[\hat{i} \in ^? \hat{d}])}$$

ALocPMapGetNoMatchNew

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \text{to\_aloc } \hat{i} \qquad a = \bot \qquad a' = \text{fresh\_aloc ()}}{\text{get}(\hat{\sigma},\hat{i},\text{NO\_MATCH}) \leadsto (\text{aloc}(a'),\bot,[\hat{i} = \text{aloc}(a') \land \hat{i} \in \hat{}^?\hat{d}])}$$

ALOCPMAPGETMATCHNOTFOUND

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \ \hat{i} \qquad a \neq \bot \qquad a \notin \mathsf{dom}(\hat{h})}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{MATCH}) \leadsto (\mathsf{aloc}(a),\bot,[\hat{i} \notin \{\mathsf{aloc}(a'): a' \in \mathsf{dom}(\hat{h})\} \land \hat{i} \in \hat{}^? \ \hat{d}])}$$

ALocPMapGetMatchNew

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \text{to\_aloc } \hat{i} \qquad a = \bot \qquad a' = \text{fresh\_aloc ()}}{\text{get}(\hat{\sigma},\hat{i},\text{MATCH}) \leadsto (\text{aloc}(a'),\bot,[\hat{i} = \text{aloc}(a') \land \hat{i} \notin \{\text{aloc}(a''): a'' \in \text{dom}(\hat{h})\} \land \hat{i} \in \hat{d}])}$$

ALOCPMAPMATCHING

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma})}{a = \mathsf{to\_aloc} \; \hat{i} \qquad a \neq \bot \qquad a \notin \mathrm{dom}(\hat{h}) \qquad a' \in \mathrm{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(a')}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{MATCH}) \leadsto (\mathrm{aloc}(a'),\hat{\sigma}_i,[\mathrm{aloc}(a) = \mathrm{aloc}(a')])}$$

ALocPMapMatchingBot

$$\frac{(\hat{h},\hat{d}) = unwrap(\hat{\sigma}) \qquad a = \mathsf{to\_aloc} \ \hat{i} \qquad a = \bot \qquad a' \in \mathsf{dom}(\hat{h}) \qquad \hat{\sigma}_i = \hat{h}(a')}{\mathsf{get}(\hat{\sigma},\hat{i},\mathsf{MATCH}) \leadsto (\mathsf{aloc}(a'),\hat{\sigma}_i,[\hat{i} = \mathsf{aloc}(a')])}$$

ALOCPMAPSETSOME

$$\frac{a = \mathsf{to\_aloc} \; \hat{i} \qquad a \neq \bot \qquad \hat{\sigma}_i \neq \bot \qquad \hat{h}' = \hat{h}[a \leftarrow \hat{\sigma}_i] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\mathsf{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

ALOCPMAPSETNONE

$$\frac{a = \mathsf{to\_aloc}\:\hat{i} \qquad a \neq \bot \qquad \hat{\sigma}_i = \bot \qquad \hat{h}' = \hat{h}[a \not\leftarrow] \qquad \hat{\sigma}' = wrap(\hat{h}', \hat{d})}{\mathsf{set}(\hat{\sigma}, \hat{i}, \hat{\sigma}_i) = \hat{\sigma}'}$$

#### **B.5.3** Soundness Proofs

Because we extend the signature of get with a matching mode, we proceed with the proofs for get and set by doing no assumption on the value of the mode m. Interestingly, we will

see that we cannot prove OX soundness with  $m = NO\_MATCH$ , as it only holds for m = MATCH. This is central to the difference in behaviour between PMAP(Loc, S) and  $PMAP_{ALoc}(S)$ . Proof.

# Proposition: Get OX Soundness

#### Assume

**(H1)** 
$$\theta, s, \sigma \models \hat{\sigma} \land \text{get}(\sigma, i) = \sigma_i \land [\hat{i}]_{\theta, s} = i$$

### To prove

$$\textbf{(G1)} \ \exists \hat{\sigma}_i, \hat{i}', \pi. \, \mathtt{get}(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land \theta, s, \sigma_i \models \hat{\sigma}_i \land [\![\hat{i}']\!]_{\theta, s} = i \land \mathsf{SAT}_{\theta, s}(\pi)$$

We proceed by proving the property holds for all rules of the concrete get, resulting in three cases.

#### Case CPMAPGETMATCH:

- **(H2)** Assume  $(h, d) = unwrap(\sigma) \land i \in dom(h) \land \sigma_i = h(i)$ .
- **(H3)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \hat{d})$  such that  $\exists \hat{i}'. \hat{i}' = \operatorname{aloc}(a') \land a' \in \operatorname{dom}(\hat{h}), [[\hat{i}']]_{\theta,s} = i \text{ and } \theta, s, \sigma_i \models \hat{h}(a').$
- (H4) From here, multiple cases are possible: either to\_aloc  $\hat{i} = a'$ , or to\_aloc  $\hat{i} = a$  such that  $a \neq a' \wedge [aloc(a)]_{\theta,s} = [aloc(a')]_{\theta,s}$ , or to\_aloc  $\hat{i} = \bot$ . Indeed, to\_aloc is a best effort function, that may not find a match despite there being one; it can also be that  $\hat{i}$  is a fresh symbolic variable, such that in the current path condition nothing binds it to aloc(a).
- (H5) In the first case, to\_aloc  $\hat{i} = a'$  we can apply ALOCPMAPGETMATCH, giving us our goal (G1).
- (H6) In the second case, to\_aloc  $\hat{i} = a \wedge a \neq a' \wedge [aloc(a)]_{\theta,s} = [aloc(a')]_{\theta,s}$ , we may apply ALocPMapMatching, but only when m = MATCH. This gives our goal (G1). Otherwise, when  $m = \text{NO\_MATCH}$ , the only rule that suits is ALocPMapGetNo-MatchNotFound, which gives us  $\hat{\sigma}_i = \bot$ , from which it follows that  $\theta, s, \sigma_i \not\models \hat{\sigma}_i$ .
- (H7) In the last case, to\_aloc  $\hat{i} = \bot$ . Again, if m = MATCH, we can apply ALOCPMAP-MATCHINGBOT and get our goal (G1); otherwise, the only applicable rule is ALOCPMAPGET-NOMATCHNEW, which invalidates our goal, as  $\theta, s, \sigma_i \not\models \hat{\sigma}_i$ .

#### Case CPMAPGETADD:

- **(H8)** Assume  $(h, d) = unwrap(\sigma) \land i \notin dom(h) \land d \neq \bot \land i \in d$ .
- **(H9)** From (H14), (H1) and CPMAPGETADD, we have  $\sigma_i = \bot$ .
- **(H10)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \hat{d})$  such that  $\hat{i} = \text{aloc}(a') \land \hat{i} \in \hat{d}$  and  $a' \notin \text{dom}(\hat{h})$ .
- (H11) We again get three possible cases, depending on to\_aloc: either to\_aloc  $\hat{i} = a'$ , or to\_aloc  $\hat{i} = a$  such that  $a \neq a' \land [aloc(a)]_{\theta,s} = [aloc(a')]_{\theta,s}$ , or to\_aloc  $\hat{i} = \bot$ .
- (H12) In the two first cases, from the definition of  $\models$  and to\_aloc, it still holds from (H10) that  $a' \notin \hat{h}$  and  $a \notin \hat{h}$ , respectively. Depending on the mode m, we thus apply ALOCPMAPGETNOMATCHNOTFOUND or ALOCPMAPGETMATCHNOTFOUND, giving us  $\hat{\sigma}_i = \bot$  and  $\pi = [\hat{i} \in \hat{d}]$  or  $\pi = [\hat{i} \notin \{aloc(a') : a' \in dom(\hat{h})\} \land \hat{i} \in \hat{d}]$ , which from (H10) we know  $SAT_{\theta,s}(\pi)$  in both cases, giving our goal (G1).
- (H13) In the last case, we have to\_aloc  $\hat{i} = \bot$ , we thus apply ALOCPMAPGETNOMATCH-NEW or ALOCPMAPGETMATCHNEW depending on m, giving us  $a' = \text{fresh\_aloc}$  (),  $\hat{\sigma}_i = \bot$ ,  $\hat{i}' = \text{aloc}(a')$  and  $\pi = [\hat{i} = \text{aloc}(a') \land \hat{i} \in \hat{d}]$  or  $[\hat{i} = \text{aloc}(a') \land \hat{i} \notin \{\text{aloc}(a'') : a'' \in \text{dom}(\hat{h})\} \land \hat{i} \in \hat{d}]$ . From (H9) and Empty Memory, we have  $\theta, s, \sigma_i \models \hat{\sigma}_i$ . Because a' is a fresh abstract location, it's equality to one term can always be satisfied; and from (H10) we know  $\hat{i} \in \hat{d}$  holds thus  $\text{SAT}_{\theta,s}(\pi)$ , completing our goal (G1).

### Case CPMAPGETBOTDOMAIN:

- **(H14)** Assume  $(h, d) = unwrap(\sigma) \land i \notin dom(h) \land d = \bot$ .
- **(H15)** From (H14), (H1) and CPMAPGETBOTDOMAIN, we have  $\sigma_i = \bot$ .
- **(H16)** It follows from (H1) and the definition of  $\models$  that  $\hat{\sigma} = (\hat{h}, \bot)$  such that  $\hat{i} \notin \text{dom}(\hat{h})$ .
- (H17) The remainder of the proof is analogous to the above; three cases are possible depending on the result of to\_aloc  $\hat{i}$  in all three cases, we can apply one of the four [...]NOTFOUND and [...]NEW rules, such that  $\hat{\sigma}_i = \bot$ ,  $[\hat{i}']_{\theta,s} = i$  and  $SAT_{\theta,s}(\pi)$ , giving our goal (G1).

# Proposition: Get UX Soundness

# Assume

$$\textbf{(H1)} \ \gcd(\hat{\sigma}, \hat{i}) \leadsto (\hat{i}', \hat{\sigma}_i, \pi) \land [\![\hat{i}]\!]_{\theta, s} = [\![\hat{i}']\!]_{\theta, s} = i \land \mathsf{SAT}_{\theta, s}(\pi)$$

**(H2)** 
$$\theta, s, \sigma \models \hat{\sigma}$$

#### To prove

**(G1)** 
$$\exists \sigma_i. get(\sigma, i) = \sigma_i \land \theta, s, \sigma_i \models \hat{\sigma}_i$$

We proceed by proving the property holds for all rules of the symbolic get, resulting in five cases.

#### Case ALOCPMAPGETMATCH:

- **(H3)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \land a = \text{to\_aloc } \hat{i} \land a \neq \bot \land a \in \text{dom}(\hat{h}) \land \hat{\sigma}_i = \hat{h}(a)$
- **(H4)** From (H3), (H1) and ALOCPMAPGETMATCH, we have  $\pi = []$
- **(H5)** From (H3), we have  $\hat{\sigma} \neq \bot$ , thus from (H2) we have  $\sigma = (h, d)$
- **(H6)** From (H2), (H3) and the definition of  $\models$ , we have  $i \in h$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$
- (H7) From (H6) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

#### Case ALocPMapGetNoMatchNotFound:

- **(H8)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \land a = \text{to\_aloc } \hat{i} \land a \neq \bot \land a \notin \text{dom}(\hat{h}) \land m = \text{NO\_MATCH}$
- **(H9)** From (H21), (H1) and ALOCPMAPGETNOMATCHNOTFOUND, we have  $\hat{\sigma}_i = \bot$  and  $\pi = [\operatorname{aloc}(a) \in \hat{d}]$
- **(H10)** Let  $(h, d) = unwrap(\sigma)$ .
- **(H11)** From (H1)  $SAT_{\theta,s}(\pi)$ , thus from (H22), either  $d \neq \bot \land i \in d$ , or  $d = \bot$ .
- **(H12)** From this, we get two cases: either  $i \in \text{dom}(h)$ , or  $i \notin \text{dom}(h)$ .
- (H13) If  $i \notin h$ , we can apply either CPMAPGETADD or CPMAPGETBOTDOMAIN depending on d, in both cases giving us  $get(\sigma, i) = \bot$ . Together with Empty Memory, this completes our goal.
- **(H14)** However, if  $i \in \text{dom}(h)$ , we can only apply CPMAPGETMATCH, which however gives us  $\text{get}(\sigma, i) = \sigma_i$  where  $\sigma_i \neq \bot$  this however is not compatible with  $\hat{\sigma}_i$ .

#### Case ALocPMapGetNoMatchNew:

(H15) Assume  $(\hat{h},\hat{d})=unwrap(\hat{\sigma})\wedge a=$  to\_aloc  $\hat{i}\wedge a=\perp\wedge a'=$  fresh\_aloc ()  $\wedge$  m= NO\_MATCH

- (H16) From (H26), (H1) and ALOCPMAPGETNOMATCHNEW, we have  $\hat{i}' = \text{aloc}(a')$ ,  $\hat{\sigma}_i = \bot$  and  $\pi = [\hat{i} = \text{aloc}(a') \land \hat{i} \in \hat{d}]$
- **(H17)** Let  $(h, d) = unwrap(\sigma)$ .
- **(H18)** From here, we get two cases: either  $i \in \text{dom}(h)$ , or  $i \notin \text{dom}(h)$  as indeed it may be the case that  $[\hat{i}]_{\theta,s} = i$ , but that due to the nature of to\_aloc,  $a \notin \hat{h}$ .
- (H19) If  $i \notin h$ , we can apply either CPMAPGETADD or CPMAPGETBOTDOMAIN depending on d, in either cases giving us  $get(\sigma, i) = \bot$ . Together with Empty Memory, this completes our goal.
- **(H20)** However, if  $i \in \text{dom}(h)$ , we can only apply CPMAPGETMATCH, which however gives us  $\text{get}(\sigma, i) = \sigma_i$  where  $\sigma_i \neq \bot$  this however is not compatible with  $\hat{\sigma}_i$ .

### Case ALocPMapGetMatchNotFound:

- **(H21)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \land a = \text{to\_aloc } \hat{i} \land a \neq \bot \land a \notin \text{dom}(\hat{h}) \land m = \text{MATCH}$
- **(H22)** From (H21), (H1) and ALOCPMAPGETNOMATCHNOTFOUND, we have  $\hat{\sigma}_i = \bot$  and  $\pi = [\hat{i} \notin \{aloc(a') : a' \in dom(\hat{h})\} \land \hat{i} \in \hat{d}]$
- **(H23)** Let  $(h, d) = unwrap(\sigma)$ .
- **(H24)** From (H1)  $SAT_{\theta,s}(\pi)$ , thus from (H22), either  $d \neq \bot \land i \in d$ , or  $d = \bot$ . We also have, crucially, that  $i \notin dom(h)$ .
- (H25) We can thus apply either CPMAPGETADD or CPMAPGETBOTDOMAIN depending on d, in both cases giving us  $get(\sigma, i) = \bot$ . Together with Empty Memory, this completes our goal.

## Case ALocPMapGetMatchNew:

- (H26) Assume  $(\hat{h},\hat{d})=unwrap(\hat{\sigma})\wedge a=$  to\_aloc  $\hat{i}\wedge a=\perp\wedge a'=$  fresh\_aloc ()  $\wedge m=$  MATCH
- **(H27)** Let  $(h, d) = unwrap(\sigma)$ .
- (H28) From (H26), (H1) and ALOCPMAPGETNOMATCHNEW, we have  $\hat{i}' = \operatorname{aloc}(a')$ ,  $\hat{\sigma}_i = \bot$  and  $\pi = [\hat{i} = \operatorname{aloc}(a') \land \hat{i} \notin \{\operatorname{aloc}(a'') : a'' \in \operatorname{dom}(\hat{h})\} \land \hat{i} \in \hat{d}]$
- **(H29)** From (H1) SAT<sub> $\theta$ ,s</sub>( $\pi$ ), thus from (H28), either  $d \neq \bot \land i \in d$ , or  $d = \bot$ . We also have, crucially, that  $[aloc(a')]_{\theta,s} = i$ , and thus  $i \notin dom(h)$ .

(H30) From (H29), we can apply either CPMAPGETADD or CPMAPGETBOTDOMAIN depending on d, in either cases giving us  $get(\sigma, i) = \bot$ . Together with Empty Memory, this completes our goal.

# Case ALOCPMAPMATCHING:

- (H31) Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \land a = to\_aloc \ \hat{i} \land a \neq \bot \land a \notin dom(\hat{h}) \land a' \in dom(\hat{h}) \land \hat{\sigma}_i = \hat{h}(a')$
- (H32) From (H31), (H1) and ALOCPMAPMATCHING, we have  $\hat{i}' = \text{aloc}(a')$ ,  $\hat{\sigma}_i = \hat{h}(a')$ ,  $\pi = [\text{aloc}(a) = \text{aloc}(a')]$  and m = MATCH.
- **(H33)** From (H2), (H31) and the definition of  $\models$ , we have  $i \in h$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$
- (H34) From (H33) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

#### Case ALocPMapMatchingBot:

- **(H35)** Assume  $(\hat{h}, \hat{d}) = unwrap(\hat{\sigma}) \land a = to\_aloc \ \hat{i} \land a = \bot \land a' \in dom(\hat{h}) \land \hat{\sigma}_i = \hat{h}(a')$
- **(H36)** From (H35), (H1) and ALOCPMAPMATCHINGBOT, we have  $\hat{i}' = \text{aloc}(a')$ ,  $\hat{\sigma}_i = \hat{h}(a')$ ,  $\pi = [\hat{i} = \text{aloc}(a')]$  and m = MATCH.
- **(H37)** From (H2), (H35) and the definition of  $\models$ , we have  $i \in h$  such that  $\sigma_i = h(i)$  and  $\theta, s, \sigma_i \models \hat{\sigma}_i$
- (H38) From (H37) we can apply CPMAPGETMATCH, thus  $get(\sigma, i) = \sigma_i$ . This completes our goal (G1).

Appendix C

**Proof for Sum Soundness** 

We first prove that a transition in a sum is not sound when no additional properties are

specified. We then show the proof for the properties that must be held by the elements

of a sum to allow for OX and UX-sound transitions from one side of the sum to the other

when using PCMs. This shows that sound constructions using PCMs are non-trivial, even

for simple cases.

C.1 Proof of Unsoundness

The sum state model transformer,  $\mathbb{S}_1 + \mathbb{S}_2$  with  $\Sigma_1$  and  $\Sigma_2$  the carrier sets of each side's

RA, is unsound for actions that allow flipping the sum from one side to the other, if the

actions doesn't have additional requirements to preserve soundness. We define the PCM of

 $Sum_{PCM}$  PCM as:

 $Sum_{PCM}(S_1, S_2) \triangleq S_1 + S_2 \triangleq l(\sigma : \Sigma_1) \mid r(\sigma : \Sigma_2) \mid 0$ 

 $\sigma \cdot 0 = 0 \cdot \sigma = \sigma$ 

 $l(\sigma) \cdot l(\sigma') = l(\sigma \cdot \sigma')$ 

 $r(\sigma) \cdot r(\sigma') = r(\sigma \cdot \sigma')$ 

undefined otherwise

We also remind the frame substraction property, defined as:

Proof.

Proposition: Sum actions that swap sides are not frame preserving

Assume

**(H1)**  $\mathbb{S}_1$  is a sound state model with a PCM  $(\Sigma_1, \cdot, 0_1)$ .

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- **(H2)**  $\mathbb{S}_2$  is a sound state model with a PCM  $(\Sigma_2, \cdot, 0_2)$ .
- (H3) There is an action swap that may flip the side of a sum from one side to the other:

$$\exists \sigma, \sigma', \vec{v_i}, \vec{v_o}$$
. swap $(l(\sigma), \vec{v_i}) = (\mathsf{Ok}, r(\sigma'), \vec{v_o})$ 

We want to prove Frame subtraction does not hold with action swap.

- **(H4)** From (H3) we have  $\exists \sigma, \sigma', \vec{v_i}, \vec{v_o}$ . swap $(l(\sigma), \vec{v_i}) = (\mathsf{0k}, r(\sigma'), \vec{v_o})$
- **(H5)** From the definition of composition,  $l(\sigma) \cdot l(0_1) = l(\sigma)$
- **(H6)** From (H5) and (H4),  $swap(l(\sigma) \cdot l(0_1), \vec{v}_i) = (Ok, r(\sigma'), \vec{v}_o)$

To satisfy frame substraction, as the outcome in (H4) is 0k, we would need  $r(\sigma') = r(\sigma') \cdot l(0_1)$ , which is not the case, as per the definition of composition,  $r(\sigma') \cdot l(0_1)$  is undefined. Frame substraction thus does not hold for swap.

# C.2 Sound Sum

To define a sound version of sum that supports flipping the side, we thus need to remove the 0 element of both sides from the allowed states. This ensure the states  $l(0_1)$  and  $r(0_2)$ aren't allowed, avoiding the problem seen with frame substraction.

$$\mathbb{S}_1 + \mathbb{S}_2 \triangleq l(\sigma \colon \Sigma_1 \setminus \{0_1\}) \mid r(\sigma \colon \Sigma_2 \setminus \{0_2\}) \mid 0$$

Additionally, to allow for a frame-preserving swap action, another requirement must be met: the state must be *exclusively owned* – no other non-empty state can be composed with it. This thus requires for a is\_exclusively\_owned function to be provided, defined as:<sup>1</sup>

is\_exclusively\_owned 
$$\sigma \triangleq \forall \sigma'. \sigma' \neq 0 \implies \neg(\sigma \# \sigma')$$

Proof.

### **Proposition:**

If a valid frame-preserving sum actions swaps sides, then that sum cannot permit empty elements on either sides and the swapped states must be exclusively owned

Assuming

 $<sup>^1</sup>$ This definition is for PCMs – the formulation of is\_exclusively\_owned for RAs is simpler: is\_exclusively\_owned  $\sigma \triangleq \#\sigma'$ .  $\sigma \# \sigma'$ 

- **(H1)**  $\mathbb{S}_1$  is a sound state model with a PCM  $(\Sigma_1, \cdot, 0_1)$ .
- **(H2)**  $\mathbb{S}_2$  is a sound state model with a PCM  $(\Sigma_2, \cdot, 0_2)$ .
- **(H3)** A sum state model is defined as  $\mathbb{S}_1 + \mathbb{S}_2 \triangleq l(\sigma \colon \Sigma_1') \mid r(\sigma \colon \Sigma_2') \mid 0$ , such that  $\Sigma_1' \subseteq \Sigma_1$  and  $\Sigma_2' \subseteq \Sigma_2$ . We re-use the definition of composition shown before.
- (H4) There is an action swap that may flip the side of a sum from the left to then right-hand side:  $\exists \sigma, \sigma', \vec{v}_i, \vec{v}_o$ . swap $(l(\sigma), \vec{v}_i) = (\mathsf{0k}, r(\sigma'), \vec{v}_o)$

This swap action is frame preserving, thus satisfying Frame subtraction and Frame Addition:

$$\sigma \# \sigma_f \wedge \alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma', \vec{v}_o) \implies$$

$$\exists \sigma'', o', \vec{v}_o' \cdot \alpha(\sigma, \vec{v}_i) = (o', \sigma'', \vec{v}_o') \wedge$$

$$(o' \neq \mathsf{Miss} \implies o' = o \wedge \vec{v}_o' = \vec{v}_o \wedge \sigma' = \sigma'' \cdot \sigma_f)$$

(H6) 
$$\alpha(\sigma, \vec{v}_i) = (o, \sigma', \vec{v}_o) \land o \neq \text{Miss} \land \sigma' \# \sigma_f \implies \sigma \# \sigma_f \land \alpha(\sigma \cdot \sigma_f, \vec{v}_i) = (o, \sigma' \cdot \sigma_f, \vec{v}_o)$$

We aim to prove:

- (G1) The zero elements of the left side of the sum is not permitted:  $0_1 \notin \Sigma_1'$
- (G2) The zero elements of the right side of the sum is not permitted:  $0_2 \notin \Sigma_2'$
- (G3) If swap flips the side of the sum, any swap of smaller states will result in a Miss:  $\operatorname{swap}(l(\sigma), \vec{v}_i) = (\operatorname{Ok}, r(\sigma'), \vec{v}_o) \wedge \sigma_a \preccurlyeq \sigma \wedge \operatorname{swap}(l(\sigma_a), \vec{v}_i) = (o', \sigma'', \vec{v}_o')$   $\implies o' = \operatorname{Miss}$
- (G4) If swap flips the side of the sum, the new state must be exclusively owned:  $swap(l(\sigma), \vec{v}_i) = (0k, r(\sigma'), \vec{v}_o) \implies is\_exclusively\_owned \ \sigma'$

We first prove (G1).

- **(H7)** Assume  $0_1 \in \Sigma'_1$ .
- **(H8)** From composition, (H7) and (H4),  $swap(l(\sigma) \cdot l(0_1), \vec{v}_i) = (0k, r(\sigma'), \vec{v}_o)$

From (H5) and (H8),

**(H9)** 
$$\exists \sigma'', o', \vec{v}'_o$$
. swap $(l(\sigma), \vec{v}_i) = (o', \sigma'', \vec{v}'_o)$ 

**(H10)** 
$$o' \neq \text{Miss} \implies o' = o \land \vec{v}'_o = \vec{v}_o \land r(\sigma') = \sigma'' \cdot l(0_1)$$

(H11) Because there can only be one action execution for a given state, action and arguments, we know from (H8) that o' = 0k,  $\sigma'' = r(\sigma')$ ,  $\vec{v}'_o = \vec{v}_o$ .

However, from composition,  $r(\sigma') \cdot l(0_1)$  is undefined, thus despite o' = 0k,  $r(\sigma') \neq r(\sigma') \cdot l(0_1)$  – our hypothesis (H7) is thus wrong, giving us our goal  $0_1 \notin \Sigma'_1$  (G1). We now prove (G2).

- **(H12)** Assume  $0_2 \in \Sigma'_2$ .
- **(H13)** Per (H12) and composition,  $r(\sigma') \# r(0_2)$ , and  $r(\sigma') \cdot r(0_2) = r(\sigma')$

Thus per (H4), (H13) and (H6), we would have  $r(\sigma') \# r(0_2)$ . This is however not the case from the definition of composition. Our hypothesis (H12) is thus wrong, giving us our goal (G2).

We may now prove our third goal (G3).

**(H14)** Let 
$$\operatorname{swap}(l(\sigma), \vec{v_i}) = (\operatorname{Ok}, r(\sigma'), \vec{v_o}) \wedge \sigma_a \leq \sigma \wedge \operatorname{swap}(l(\sigma_a), \vec{v_i}) = (\sigma', \sigma'', \vec{v_o}).$$

- **(H15)** Assume  $o' \neq \text{Miss}$ .
- **(H16)** From the definition of  $\leq$  and (H14), we have  $\exists \sigma_b. \sigma_a \cdot \sigma_b = \sigma$ , which also implies  $l(\sigma_a) \cdot l(\sigma_b) = l(\sigma)$ .
- (H17) From (H5), (H14) and (H15), we have  $r(\sigma') = \sigma'' \cdot l(\sigma_b)$ , and o' = 0k. From composition this is however not possible, as no composition from l(-) yields r(-). Our assumption (H15) is thus false, thus o' = Miss, giving our goal (G3).

We may finally prove our fourth and last goal (G4).

- **(H18)** Let swap $(l(\sigma), \vec{v_i}) = (0k, r(\sigma'), \vec{v_o}).$
- (H19) Assume  $\neg$ is\_exclusively\_owned  $\sigma'$ .
- **(H20)** From (H19),  $\exists \sigma'' . \sigma'' \neq 0_2 \land \sigma'' \# \sigma'$ , thus from composition we also have  $r(\sigma'') \# r(\sigma')$ .
- (H21) From (H6), we must have  $swap(l(\sigma) \cdot r(\sigma''), \vec{v_i}) = (0k, r(\sigma') \cdot r(\sigma''), \vec{v_o})$ . This is however not possible, as per composition,  $l(\sigma) \cdot r(\sigma'')$  is undefined. Thus our hypothesis (H19) is false, giving us our goal (G4).