

Formal Definition of the Parameterized Aspect Calculus

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TR #03-12b
October 2003
revised November 2003

Keywords: Parameterized aspect calculus, object calculus, join point model, point cut description language, aspect-oriented programming, AspectJ, advice, HyperJ, hyperslices, DemeterJ, adaptive methods
2003 CR Categories: D.3.1 [*Programming Languages*] Formal Definitions and Theory — Semantics D.3.2 [*Programming Languages*] Language Classifications — object-oriented languages D.3.3 [*Programming Languages*] Language Constructs and Features — classes and objects

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November 12, 2003

Abstract

This paper gives the formal definition of the parameterized aspect calculus, or ς_{asp} . The ς_{asp} calculus is a core calculus for the formal study of aspect-oriented programming languages. The calculus consists of a base language, taken from Abadi and Cardelli's object calculus, and point cut description language. The calculus is parameterized to accept a variety of point cut description languages, simplifying the study of a variety of aspect-oriented language features. The calculus exposes a rich join point model on the base language, granting great flexibility to point cut description languages.

1 Introduction

This paper gives the formal definition of the parameterized aspect calculus, or ς_{asp} . The ς_{asp} calculus is a core calculus for the formal study of aspect-oriented programming languages. The calculus consists of a base language, taken from Abadi and Cardelli's object calculus [1], and point cut description language. The calculus is parameterized to accept a variety of point cut description languages, simplifying the study of a variety of aspect-oriented language features. The calculus exposes a rich join point model on the base language, granting great flexibility to point cut description languages.

This paper primarily provides technical details and sample reductions. The interested reader can find a more intuitive explanation of the calculus in a separate report [2].

2 Formal Definitions

2.1 Syntax

$$\begin{array}{llllll}
x \in Vars & d \in Consts & f \in FConsts & l \in Labels & S \in \mathbf{P}(Labels) \cup Consts & pcd \in \mathcal{C} \\
\\
\text{programs} & \mathcal{P} ::= a \otimes \vec{\mathcal{A}} \\
\text{terms} & a, b, c ::= x \mid v \mid a.k \mid \\
& a.l \Leftarrow \varsigma(x)b \mid \\
& \text{proceed}_{\text{VAL}}() \mid \\
& \text{proceed}_{\text{IVK}}(a) \mid \\
& \text{proceed}_{\text{UPD}}(a, \varsigma(x)b) \mid \pi \\
\text{values} & v ::= d \mid [\bar{l}_i = \varsigma(x_i)\bar{b}_i]^{i \in I} \\
\text{selectors} & k ::= l \mid f \\
\text{proceed} & \pi ::= \Pi_{\text{VAL}}\{B, v\}() \mid \\
\text{closures} & \Pi_{\text{IVK}}\{B, S, k\}(a) \mid \\
& \Pi_{\text{UPD}}\{B, k\}(a, \varsigma(x)b) \\
\text{naked methods} & B ::= \overline{\varsigma(\vec{y})}\vec{b} \\
\text{advice} & \mathcal{A} ::= pcd \triangleright \varsigma(\vec{y})\vec{b} \\
\text{step kinds} & \rho ::= \text{VAL} \mid \text{IVK} \mid \text{UPD}
\end{array}$$

Although they are part of the term syntax, proceed closures are only generated dynamically; they may not be written in user programs.

To prevent ambiguity when evaluating programs, the sets of functional constants and labels are disjoint: $FConsts \cap Labels = \emptyset$. Similarly, $Vars \cap Consts = \emptyset$.

2.2 Meta-Syntax

$$\begin{array}{ll}
\text{reduction judgments} & \mathcal{K} \vdash_{M, \vec{\mathcal{A}}} a \rightsquigarrow v \\
\text{evaluation contexts} & \mathcal{K} ::= \epsilon \mid \kappa \cdot \mathcal{K} \\
\text{evaluation steps} & \kappa ::= \text{ib}(\bar{l}, l) \mid \text{va} \mid \text{ia} \mid \text{ua}
\end{array}$$

2.3 Well-formedness Rules

$$\frac{\text{CTX } \epsilon}{\epsilon \vdash_{M, \vec{\mathcal{A}}} \diamond} \quad \frac{\text{CTX } \kappa \quad \mathcal{K} \vdash_{M, \vec{\mathcal{A}}} \diamond}{\kappa \cdot \mathcal{K} \vdash_{M, \vec{\mathcal{A}}} \diamond}$$

2.4 Reduction Rules

2.4.1 When No Advice Matches

$$\begin{array}{c}
\text{RED VAL 0} \\
\frac{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} \diamond \quad \text{advFor}_M(\langle \text{VAL}, \mathcal{K}, \text{sig}(v), \epsilon \rangle, \vec{\mathcal{A}}) = \bullet}{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} v \rightsquigarrow v}
\end{array}$$

$$\begin{array}{c}
\text{RED SEL 0 (where } o \triangleq [\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}) \\
\frac{l_j \in [\overline{l_i}]^{i \in I} \quad \text{advFor}_M(\langle \text{IVK}, \mathcal{K}, [\overline{l_i}]^{i \in I}, l_j \rangle, \vec{\mathcal{A}}) = \bullet \quad \text{ib}([\overline{l_i}]^{i \in I}, l_j) \cdot \mathcal{K} \vdash_{M, \vec{\mathcal{A}}} b_j \{x_j \leftarrow o\} \rightsquigarrow v}{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} a.l_j \rightsquigarrow v}
\end{array}$$

$$\begin{array}{c}
\text{RED UPD 0 (where } o \triangleq [\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}) \\
\frac{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} a \rightsquigarrow o \quad l_j \in [\overline{l_i}]^{i \in I} \quad \text{advFor}_M(\langle \text{UPD}, \mathcal{K}, [\overline{l_i}]^{i \in I}, l_j \rangle, \vec{\mathcal{A}}) = \bullet}{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} a.l_j \Leftarrow \varsigma(x)b \rightsquigarrow [\overline{l_i = \varsigma(x_i)b_i}]^{i \in I \setminus \{j\}}, l_j = \varsigma(x)b]
}
\end{array}$$

$$\begin{array}{c}
\text{RED FCONST 0} \\
\frac{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} a \rightsquigarrow v' \quad \text{advFor}_M(\langle \text{IVK}, \mathcal{K}, \text{sig}(v'), f \rangle, \vec{\mathcal{A}}) = \bullet \quad \text{ib}(\text{sig}(v'), f) \cdot \mathcal{K} \vdash_{M, \vec{\mathcal{A}}} \delta(f, v') \rightsquigarrow v}{\mathcal{K} \vdash_{M, \vec{\mathcal{A}}} a.f \rightsquigarrow v}
\end{array}$$

The meaning of functional constants is given by a function $\delta: FConsts \times Values \rightarrow Values$. We leave δ underspecified but, following Crank and Felleisen, restrict it to just depend on the observable characteristics of its arguments [3]. This restriction is based on a notion of term context.

Definition 1 (Term Context) A term context is a $\varsigma_{asp}(M)$ term with a single hole generated by the recursion:

$$\begin{aligned}
\mathcal{E}[_] &::= - \mid [\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}, l = \varsigma(x)\mathcal{E}[_] \mid \\
&\quad \mathcal{E}[_].k \mid \mathcal{E}[_].l \Leftarrow \varsigma(x)c \mid \\
&\quad c.l \Leftarrow \varsigma(x)\mathcal{E}[_] \mid \text{proceed}_{\text{IVK}}(\mathcal{E}[_]) \mid \\
&\quad \text{proceed}_{\text{UPD}}(\mathcal{E}[_], \varsigma(x)c) \mid \\
&\quad \text{proceed}_{\text{UPD}}(c, \varsigma(x)\mathcal{E}[_])
\end{aligned}$$

The size of a term context is a natural n , counting the number of recursive applications of the above rule used to generate the context.

Application of a term context to a term a is modeled by non-capturing substitution, treating the hole, $-$, as a variable:

$$\mathcal{E}[a] = \mathcal{E}[_]\{- \leftarrow a\}$$

Let v be a value that is not a basic constant; that is, $v = [\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}$. Then $\delta(f, v) = a$ implies that there exists a term context $\mathcal{E}[_]$ such that for all values v' , $v' \notin Consts$, $\delta(f, v') = \mathcal{E}[v']$.

2.4.2 When Some Advice Matches

$$\text{RED VAL 1} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} \diamond \quad \text{advFor}_M(\langle \text{VAL}, \mathcal{K}, \text{sig}(v), \epsilon \rangle, \bar{A}) = \varsigma()b + B \quad \text{close}_{\text{VAL}}(b, \{B, v\}) = b' \quad \text{va} \cdot \mathcal{K} \vdash_{M,\bar{A}} b' \rightsquigarrow v'}{\mathcal{K} \vdash_{M,\bar{A}} v \rightsquigarrow v'}$$

$$\text{RED SEL 1 (where } o \triangleq [\overline{l_i = \varsigma(x_i)b_i}^{i \in I}]) \text{)} \\ \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow o \quad l_j \in \overline{l_i}^{i \in I} \quad \text{advFor}_M(\langle \text{IVK}, \mathcal{K}, \overline{l_i}^{i \in I}, l_j \rangle, \bar{A}) = \varsigma(y)b + B \\ \text{close}_{\text{IVK}}(b, \{(B + \varsigma(x_j)b_j), \overline{l_i}^{i \in I}, l_j\}) = b' \quad \text{ia} \cdot \mathcal{K} \vdash_{M,\bar{A}} b' \{y \leftarrow o\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} a.l_j \rightsquigarrow v}$$

$$\text{RED FCONST 1} \\ \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow v' \quad \text{advFor}_M(\langle \text{IVK}, \mathcal{K}, \text{sig}(v'), f \rangle, \bar{A}) = \varsigma(y)b + B \\ \text{close}_{\text{IVK}}(b, \{B, \text{sig}(v'), f\}) = b' \quad \text{ia} \cdot \mathcal{K} \vdash_{M,\bar{A}} b' \{y \leftarrow v'\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} a.f \rightsquigarrow v}$$

$$\text{RED UPD 1 (where } o \triangleq [\overline{l_i = \varsigma(x_i)b_i}^{i \in I}]) \text{)} \\ \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow o \quad \text{advFor}_M(\langle \text{UPD}, \mathcal{K}, \overline{l_i}^{i \in I}, l_j \rangle, \bar{A}) = \varsigma(\text{targ}, \text{rval})b' + B \\ \text{close}_{\text{UPD}}(b', \{B, l_j\}) = b'' \quad \text{ua} \cdot \mathcal{K} \vdash_{M,\bar{A}} b'' \{rval \leftarrow b \{x \leftarrow \text{targ}\}\}_{\text{targ}} \{\text{targ} \leftarrow o\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} a.l_j \Leftarrow \varsigma(x)b \rightsquigarrow v}$$

2.4.3 Proceeding from Advice

$$\text{RED VPRCD 0} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} \diamond}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{VAL}} \{\bullet, v\}() \rightsquigarrow v}$$

$$\text{RED VPRCD 1} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} \diamond \quad \text{close}_{\text{VAL}}(b, \{B, v\}) = b' \quad \text{va} \cdot \mathcal{K} \vdash_{M,\bar{A}} b' \rightsquigarrow v'}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{VAL}} \{(\varsigma(b + B), v)\}() \rightsquigarrow v'}$$

$$\text{RED SPRCD 0} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow o \quad \text{ib}(\bar{l}, l) \cdot \mathcal{K} \vdash_{M,\bar{A}} b \{y \leftarrow o\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{IVK}} \{\varsigma(y)b, \bar{l}, l\}(a) \rightsquigarrow v}$$

$$\text{RED SPRCD 1} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow o \quad B \neq \bullet \quad \text{close}_{\text{IVK}}(b, \{B, \bar{l}, l\}) = b' \quad \text{ia} \cdot \mathcal{K} \vdash_{M,\bar{A}} b' \{y \leftarrow o\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{IVK}} \{(\varsigma(y)b + B), \bar{l}, l\}(a) \rightsquigarrow v}$$

$$\text{RED FPRCD 0} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow v' \quad \text{ib}(S, f) \cdot \mathcal{K} \vdash_{M,\bar{A}} \delta(f, v') \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{IVK}} \{\bullet, S, f\}(a) \rightsquigarrow v}$$

$$\text{RED FPRCD 1} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow v' \quad \text{close}_{\text{IVK}}(b, \{B, S, f\}) = b' \quad \text{ia} \cdot \mathcal{K} \vdash_{M,\bar{A}} b' \{y \leftarrow v'\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{IVK}} \{(\varsigma(y)b + B), S, f\}(a) \rightsquigarrow v}$$

$$\text{RED UPRCD 0} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow [\overline{l_i = \varsigma(x_i)b_i}^{i \in I}] \quad l_j \in \overline{l_i}^{i \in I}}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{UPD}} \{\bullet, l_j\}(a, \varsigma(x)b) \rightsquigarrow [\overline{l_i = \varsigma(x_i)b_i}^{i \in I \setminus j}, l_j = \varsigma(x)b]}$$

$$\text{RED UPRCD 1} \quad \frac{\mathcal{K} \vdash_{M,\bar{A}} a \rightsquigarrow o \quad \text{close}_{\text{UPD}}(b', \{B, l_j\}) = b'' \quad \text{ua} \cdot \mathcal{K} \vdash_{M,\bar{A}} b'' \{rval \leftarrow b \{x \leftarrow \text{targ}\}\}_{\text{targ}} \{\text{targ} \leftarrow o\} \rightsquigarrow v}{\mathcal{K} \vdash_{M,\bar{A}} \Pi_{\text{UPD}} \{(\varsigma(\text{targ}, \text{rval})b' + B), l_j\}(a, \varsigma(x)b) \rightsquigarrow v}$$

2.5 Helper Functions

2.5.1 Advice Lookup

$$advFor_M(jp, \bullet) = \bullet$$

$$advFor_M(jp, (pcd \triangleright \varsigma(\vec{y})b) + \vec{\mathcal{A}}) = \\ match(pcd \triangleright \varsigma(\vec{y})b, jp) + advFor_M(jp, \vec{\mathcal{A}})$$

2.5.2 Value Signature

$$sig(v) = \begin{cases} \overline{l_i}^{i \in I} & \text{if } v = [\overline{l_i = \varsigma(x_i)b_i}^{i \in I}] \\ v & \text{otherwise} \end{cases}$$

2.5.3 Transforming Proceed Terms to Proceed Closures

$$close_\rho(x, t) = x \quad close_\rho(d, t) = d \quad close_\rho([\overline{l_i = \varsigma(x_i)b_i}^{i \in I}], t) = [\overline{l_i = \varsigma(x_i)close_\rho(b_i, t)}^{i \in I}]$$

$$close_\rho(a.k, t) = close_\rho(a, t).k \quad close_\rho(a.l \Leftarrow \varsigma(x)b, t) = close_\rho(a, t).l \Leftarrow \varsigma(x)close_\rho(b, t)$$

$$close_{VAL}(\mathsf{proceed}_{VAL}(), \{B, v\}) = \Pi_{VAL}\{B, v\}() \quad close_{IVK}(\mathsf{proceed}_{IVK}(a), \{B, S, k\}) = \\ \Pi_{IVK}\{B, S, k\}(close_{IVK}(a, \{B, S, k\}))$$

$$close_{UPD}(\mathsf{proceed}_{UPD}(a, \varsigma(x)b), \{B, k\}) = \\ \Pi_{UPD}\{B, k\}(close_{UPD}(a, \{B, k\}), \varsigma(x)close_{UPD}(b, \{B, k\}))$$

$$close_\rho(\mathsf{proceed}_{VAL}(), \{B, v\}) = \mathsf{proceed}_{VAL}() \text{ for } \rho \neq VAL$$

$$close_\rho(\mathsf{proceed}_{IVK}(a), \{B, S, k\}) = \mathsf{proceed}_{IVK}(close_\rho(a, \{B, S, k\})) \text{ for } \rho \neq IVK$$

$$close_\rho(\mathsf{proceed}_{UPD}(a, \varsigma(x)b), \{B, k\}) = \mathsf{proceed}_{UPD}(close_\rho(a, \{B, k\}), \varsigma(x)close_\rho(b, \{B, k\})) \text{ for } \rho \neq UPD$$

$$close_\rho(\Pi_{VAL}\{B, v\}(), t) = \Pi_{VAL}\{B, v\}() \quad close_\rho(\Pi_{IVK}\{B, S, k\}(a), t) = \Pi_{IVK}\{B, S, k\}(close_\rho(a, t))$$

$$close_\rho(\Pi_{UPD}\{B, k\}(a, \varsigma(x)b), t) = \Pi_{UPD}\{B, k\}(close_\rho(a, t), \varsigma(x)close_\rho(b, t))$$

2.6 Variable Scoping and Substitution

Substitutions are performed sequentially, left-to-right, not simultaneously. For nested substitutions, the inner-most substitution is performed first.

2.6.1 Variable Scoping

$$\begin{aligned}
FV(\varsigma(y)b) &\triangleq FV(b) \setminus \{y\} \\
FV(x) &\triangleq x \\
FV([\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}) &\triangleq \bigcup_{i \in I} FV(\varsigma(x_i)b_i) \\
FV(a.l) &\triangleq FV(a) \\
FV(a.l \Leftarrow \varsigma(y)b) &\triangleq FV(a) \cup FV(\varsigma(y)b) \\
FV(\text{proceed}_{\text{VAL}}()) &\triangleq \emptyset \\
FV(\text{proceed}_{\text{IVK}}(a)) &\triangleq FV(a) \\
FV(\text{proceed}_{\text{UPD}}(a, \varsigma(y)b)) &\triangleq FV(a) \cup FV(\varsigma(y)b) \\
FV(\Pi_{\text{VAL}}\{B, v\}()) &\triangleq \emptyset \\
FV(\Pi_{\text{IVK}}\{B, S, k\}(a)) &\triangleq FV(a) \\
FV(\Pi_{\text{UPD}}\{B, k\}(a, \varsigma(y)b)) &\triangleq FV(a) \cup FV(\varsigma(y)b)
\end{aligned}$$

2.6.2 Capture-Avoiding Substitution

$$\begin{aligned}
(\varsigma(y)b)\{x \leftarrow c\} &\triangleq \varsigma(y')(b\{y \leftarrow y'\}\{x \leftarrow c\}) \quad \text{where } y' \notin FV(\varsigma(y)b) \cup FV(c) \cup \{x\} \\
x\{x \leftarrow c\} &\triangleq c \\
y\{x \leftarrow c\} &\triangleq y \quad \text{if } x \neq y \\
[\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}\{x \leftarrow c\} &\triangleq [\overline{l_i = (\varsigma(x_i)b_i)}\{x \leftarrow c\}]^{i \in I} \\
(a.l)\{x \leftarrow c\} &\triangleq (a\{x \leftarrow c\}).l \\
(a.l \Leftarrow \varsigma(y)b)\{x \leftarrow c\} &\triangleq (a\{x \leftarrow c\}).l \Leftarrow ((\varsigma(y)b)\{x \leftarrow c\}) \\
(\text{proceed}_{\text{VAL}}())\{x \leftarrow c\} &\triangleq \text{proceed}_{\text{VAL}}() \\
(\text{proceed}_{\text{IVK}}(a))\{x \leftarrow c\} &\triangleq \text{proceed}_{\text{IVK}}(a\{x \leftarrow c\}) \\
(\text{proceed}_{\text{UPD}}(a, \varsigma(y)b))\{x \leftarrow c\} &\triangleq \text{proceed}_{\text{UPD}}((a\{x \leftarrow c\}), ((\varsigma(y)b)\{x \leftarrow c\})) \\
(\Pi_{\text{VAL}}\{B, v\}())\{x \leftarrow c\} &\triangleq \Pi_{\text{VAL}}\{B, v\}() \\
(\Pi_{\text{IVK}}\{B, S, k\}(a))\{x \leftarrow c\} &\triangleq \Pi_{\text{IVK}}\{B, S, k\}(a\{x \leftarrow c\}) \\
(\Pi_{\text{UPD}}\{B, k\}(a, \varsigma(y)b))\{x \leftarrow c\} &\triangleq \Pi_{\text{UPD}}\{B, k\}((a\{x \leftarrow c\}), ((\varsigma(y)b)\{x \leftarrow c\}))
\end{aligned}$$

2.6.3 Capturing Substitution

$$\begin{aligned}
(\varsigma(z)b)\{x \leftarrow c\}_z &\triangleq \varsigma(z)(\{x \leftarrow c\}_z) \\
(\varsigma(y)b)\{x \leftarrow c\}_z &\triangleq \varsigma(y')(b\{y \leftarrow y'\}\{x \leftarrow c\}_z) \quad \text{if } y \neq z, \text{ where } y' \notin FV(\varsigma(y)b) \cup FV(c) \cup \{x\} \\
x\{x \leftarrow c\}_z &\triangleq c \\
y\{x \leftarrow c\}_z &\triangleq y \quad \text{if } x \neq y \\
[\overline{l_i = \varsigma(x_i)b_i}]^{i \in I}\{x \leftarrow c\}_z &\triangleq [\overline{l_i = (\varsigma(x_i)b_i)}\{x \leftarrow c\}]_z^{i \in I} \\
(a.l)\{x \leftarrow c\}_z &\triangleq (a\{x \leftarrow c\}_z).l \\
(a.l \Leftarrow \varsigma(y)b)\{x \leftarrow c\}_z &\triangleq (a\{x \leftarrow c\}_z).l \Leftarrow ((\varsigma(y)b)\{x \leftarrow c\}_z) \\
(\text{proceed}_{\text{VAL}}())\{x \leftarrow c\}_z &\triangleq \text{proceed}_{\text{VAL}}() \\
(\text{proceed}_{\text{IVK}}(a))\{x \leftarrow c\}_z &\triangleq \text{proceed}_{\text{IVK}}(a\{x \leftarrow c\}_z) \\
(\text{proceed}_{\text{UPD}}(a, \varsigma(y)b))\{x \leftarrow c\}_z &\triangleq \text{proceed}_{\text{UPD}}((a\{x \leftarrow c\}_z), ((\varsigma(y)b)\{x \leftarrow c\}_z)) \\
(\Pi_{\text{VAL}}\{B, v\}())\{x \leftarrow c\}_z &\triangleq \Pi_{\text{VAL}}\{B, v\}() \\
(\Pi_{\text{IVK}}\{B, S, k\}(a))\{x \leftarrow c\}_z &\triangleq \Pi_{\text{IVK}}\{B, S, k\}(a\{x \leftarrow c\}_z) \\
(\Pi_{\text{UPD}}\{B, k\}(a, \varsigma(y)b))\{x \leftarrow c\}_z &\triangleq \Pi_{\text{UPD}}\{B, k\}((a\{x \leftarrow c\}_z), ((\varsigma(y)b)\{x \leftarrow c\}_z))
\end{aligned}$$

3 Sample Point Cut Description Languages

3.1 Natural Selection

Let $M_s = \langle \mathcal{C}_s, match_s \rangle$, where $\mathcal{C}_s ::= [\bar{l}].l$ and:

$$match_s([\bar{l}].l \triangleright \varsigma(\vec{y})b, \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} \langle \varsigma(\vec{y})b \rangle & \text{if } (\rho = \text{IVK}) \wedge (S = \bar{l}) \wedge (k = l) \\ \bullet & \text{otherwise} \end{cases}$$

3.2 General Matching

The general point cut description language, M_G , is defined in the following subsections.

3.2.1 Syntax of M_G

$$\begin{array}{ll} \text{descriptions} & pcd ::= \text{VAL} \mid \text{IVK} \mid \text{UPD} \mid \\ & k = k \mid S = S \mid K \in r \mid \\ & \neg pcd \mid pcd \wedge pcd \mid pcd \vee pcd \\ \text{context} & r ::= \epsilon \mid \text{ib}(M, m) \mid \text{va} \mid \text{ia} \mid \text{ua} \mid \\ \text{expr.} & \cdot \mid r + r \mid rr \mid r^* \\ \text{signatures} & M ::= d \mid \bar{l} \mid \cdot \\ \text{messages} & m ::= f \mid l \mid \cdot \end{array}$$

3.2.2 Semantics of M_G

Join Point Matching

$$match_G(pcd \triangleright \varsigma(\vec{y})b, jp) = \begin{cases} \langle \varsigma(\vec{y})b \rangle & \text{if } matches(pcd, jp) \\ \bullet & \text{otherwise} \end{cases}$$

$$matches(\text{VAL}, \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} T & \text{if } \rho = \text{VAL} \\ F & \text{otherwise} \end{cases}$$

$$matches(\text{IVK}, \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} T & \text{if } \rho = \text{IVK} \\ F & \text{otherwise} \end{cases}$$

$$matches(\text{UPD}, \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} T & \text{if } \rho = \text{UPD} \\ F & \text{otherwise} \end{cases}$$

$$matches(k = k', \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} T & \text{if } k = k' \\ F & \text{otherwise} \end{cases}$$

$$matches(S = S', \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} T & \text{if } S = S' \\ F & \text{otherwise} \end{cases}$$

$$matches(K \in r, \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} T & \text{if } K \in r \\ F & \text{otherwise} \end{cases}$$

$$matches(pcd_1 \wedge pcd_2, \langle \rho, \mathcal{K}, S, k \rangle) = matches(pcd_1, \langle \rho, \mathcal{K}, S, k \rangle) \wedge matches(pcd_2, \langle \rho, \mathcal{K}, S, k \rangle)$$

$$matches(pcd_1 \vee pcd_2, \langle \rho, \mathcal{K}, S, k \rangle) = matches(pcd_1, \langle \rho, \mathcal{K}, S, k \rangle) \vee matches(pcd_2, \langle \rho, \mathcal{K}, S, k \rangle)$$

$$matches(\neg pcd, \langle \rho, \mathcal{K}, S, k \rangle) = \neg matches(pcd, \langle \rho, \mathcal{K}, S, k \rangle)$$

Context Pattern Matching

$$\begin{array}{cccccc}
\overline{\epsilon \in \epsilon} & \overline{\text{ib}(S, k) \in \text{ib}(S, k)} & \overline{\text{ib}(S, k) \in \text{ib}(\cdot, k)} & \overline{\text{ib}(S, k) \in \text{ib}(S, \cdot)} & \overline{\text{ib}(S, k) \in \text{ib}(\cdot, \cdot)} & \overline{\text{va} \in \text{va}} \\
\\
\overline{\text{ia} \in \text{ia}} & \overline{\text{ua} \in \text{ua}} & \overline{\text{ib}(S, k) \in \cdot} & \overline{\text{va} \in \cdot} & \overline{\text{ia} \in \cdot} & \overline{\text{ua} \in \cdot} & \frac{\mathcal{K} \in r_1 \vee \mathcal{K} \in r_2}{\mathcal{K} \in r_1 + r_2} \\
\\
\frac{\mathcal{K}_1 \in r_1 \quad \mathcal{K}_2 \in r_2}{\mathcal{K}_1 \cdot \mathcal{K}_2 \in r_1 r_2} & & \frac{}{\epsilon \in r^*} & & \frac{\mathcal{K}_1 \in r \quad \mathcal{K}_2 \in r^*}{\mathcal{K}_1 \cdot \mathcal{K}_2 \in r^*} &
\end{array}$$

3.3 Basic Reflection

We can use ς_{asp} to add some reflective capabilities to the object calculus. For example, we can construct a point cut description language and advice that allows a base program term to query an object for the existence of a particular label by selection on a special functional constant: `a.hasLabel_m`.

We construct a point cut description language, $M_B = \langle \mathcal{C}_B, \text{match}_B \rangle$, that binds advice to selection on these special query labels:

$$\begin{aligned}
\mathcal{C}_B ::= & \text{found} \mid \text{notFound} \\
\text{match}_B(\text{found} \triangleright \varsigma(x)b, \langle \rho, \mathcal{K}, L, k \rangle) = & \begin{cases} \langle \varsigma(x)b \rangle & \text{if } \rho = \text{IVK} \wedge k = \text{hasLabel_}l \wedge l \in L \\ \bullet & \text{otherwise} \end{cases} \\
\text{match}_B(\text{notFound} \triangleright \varsigma(x)b, \langle \rho, \mathcal{K}, L, k \rangle) = & \begin{cases} \langle \varsigma(x)b \rangle & \text{if } \rho = \text{IVK} \wedge k = \text{hasLabel_}l \wedge l \notin L \\ \bullet & \text{otherwise} \end{cases}
\end{aligned}$$

To use this reflective mechanism, a program in $\varsigma_{asp}(M_B)$ must include two pieces of advice:

$$\begin{aligned}
\text{found} \triangleright \varsigma(s) \text{ true} \\
\text{notFound} \triangleright \varsigma(s) \text{ false}
\end{aligned}$$

where we assume basic constants for the boolean values.

With this advice, and the given definition of match_B , a term `a.hasLabel_m` will reduce to true if `a` reduces to an object containing the label `m`. Otherwise the term will reduce to false.

3.4 Quantified Advice and Adaptive Methods

To model adaptive methods [4] we establish a convention that labels for fields begin with “f_” and define a point cut description language, $M_R = \langle \mathcal{C}_R, \text{match}_R \rangle$, that extends M_G with a mechanism to quantify over the fields of an object.

All point cut descriptions in \mathcal{C}_G are valid in \mathcal{C}_R . Additionally the suffix “ $\forall l \in \text{fieldsOf}(S)$ ” may be added to any of \mathcal{C}_G ’s point cut descriptions. This suffix causes match_R to create a sequence of advice from a single matching piece of advice. The generated sequence has one element for each field in the target object of the join point.

For a point cut description without the quantifier suffix, match_R is identical to match_G . For a point cut description $\text{pcd}_G \cdot \forall l \in \text{fieldsOf}(S)$, match_R is defined as follows:

$$\text{match}_R(\text{pcd}_G \cdot \forall l \in \text{fieldsOf}(S) \triangleright \varsigma(\vec{y})b, \langle \rho, \mathcal{K}, S, k \rangle) = \begin{cases} \bullet & \text{if } \text{match}_G(\text{pcd}_G \triangleright \varsigma(y)b, \langle \rho, \mathcal{K}, S, k \rangle) = \bullet \\ \langle \varsigma(\vec{y})b_1, \dots, \varsigma(\vec{y})b_m \rangle & \text{otherwise} \end{cases}$$

where $\{l_1, \dots, l_m\}$ is the set of field labels in S and each b_i is formed from b by first finding all occurrences of l as a selection or update label, and then replacing them with l_i . This relabeling step is formalized as

follows:

$$\begin{aligned}
\text{relabel}(x, l, l_i) &= x & \text{relabel}(d, l, l_i) &= d & \text{relabel}([\overline{l_j = \varsigma(x_j)b_j}]^{j \in J}, l, l_i) &= [\overline{l_j = \varsigma(x_j)(\text{relabel}(b_j, l, l_i))}]^{j \in J} \\
\text{relabel}(a.f, l, l_i) &= (\text{relabel}(a, l, l_i)).f & \text{relabel}(a.l', l, l_i) &= \begin{cases} (\text{relabel}(a, l, l_i)).l_i & \text{if } l' = l \\ (\text{relabel}(a, l, l_i)).l' & \text{otherwise} \end{cases} \\
\text{relabel}(a.l' \Leftarrow \varsigma(x)b, l, l_i) &= \begin{cases} (\text{relabel}(a, l, l_i)).l_i \Leftarrow \varsigma(x)\text{relabel}(b, l, l_i) & \text{if } l' = l \\ (\text{relabel}(a, l, l_i)).l' \Leftarrow \varsigma(x)\text{relabel}(b, l, l_i) & \text{otherwise} \end{cases} \\
\text{relabel}(\text{proceed}_{\text{VAL}}, l, l_i) &= \text{proceed}_{\text{VAL}} & \text{relabel}(\text{proceed}_{\text{IVK}}(a), l, l_i) &= \text{proceed}_{\text{IVK}}(\text{relabel}(a, l, l_i)) \\
\text{relabel}(\text{proceed}_{\text{UPD}}(a, \varsigma(x)b), l, l_i) &= \text{proceed}_{\text{UPD}}(\text{relabel}(a, l, l_i), \varsigma(x)\text{relabel}(b, l, l_i))
\end{aligned}$$

With M_R we modeling of traversals using update advice for walking the object graph. Update advice has two parameters; we use one to track the root of the object graph to be traversed and the other to hold a visitor object for accumulating results.

Suppose we have a traversal described by “to Point”, where Point is an object with the set of labels $\{\text{f_n, pos}\}$. We can model this traversal in M_R with the following advice, where traverse is a special functional constant and traversing is a special label, Visitor $\triangleq \{\text{f_obj, f_acc, visit}\}$, and Point $\triangleq \{\text{f_n, pos}\}$:

```

// initiates the traversal
IVK ∧ S = Visitor ∧ k = traverse ▷
     $\varsigma(v)$  proceedIVK((v.f_obj).traversing  $\Leftarrow$   $\varsigma(y)v$ )
// recurses to fields for non-points
UPD ∧  $\neg(S = \text{Point})$  ∧ k = traversing ·  $\forall \text{field} \in \text{fieldsOf}(S)$  ▷
     $\varsigma(t,r)$  proceedUPD(t,  $\varsigma(y)((r.f_obj.field).traversing \Leftarrow$ 
         $\varsigma(q)(r.f_obj \Leftarrow \varsigma(z)r.f_obj.field).f_obj \Leftarrow \varsigma(q)t)$ )
// prevents proceeding to actual update of traversing label, which would stick
UPD ∧  $\neg(S = \text{Point})$  ∧ k = traversing ▷
     $\varsigma(t,r)$  r
// selects visit method for points
UPD ∧ S = Point ∧ k = traversing ▷
     $\varsigma(t,r)$  r.visit
// extracts result from visitor object
IVK ∧ S = Visitor ∧ k = traverse ▷  $\varsigma(v)$  v.f.acc

```

This traversal is used in a program as follows:

```

[ f_obj= $\varsigma(y)o$ , // starting object, y not free in o
  f_acc= $\varsigma(y)a$ , // result accumulator
  visit= $\varsigma(y)b$  // updates accumulator based on object
].traverse

```

4 Sample Reductions

The sample reductions given in this section were automatically generated and typeset using an interpreter for $sasp$, implemented in Java. The interpreter is open-source and is available from <http://www.cs.iastate.edu/~ccclifton/sasp>. The reductions are presented in the Abadi and Cardelli style, where a hooked arrow indicates all the premises leading to a given judgment [1, §7].

4.1 Base Language Reductions

We first present some reductions without advice as examples of calculation in the base language.

Reducing an object value:

$$A \triangleq \bullet$$

$$\begin{array}{c} \left. \begin{array}{l} \epsilon \vdash \diamond \\ advFor(\langle VAL, \epsilon, \{\}, \epsilon \rangle, A) = \bullet \end{array} \right\} \\ \epsilon \vdash [] \rightsquigarrow [] \end{array} \quad \text{RED VAL 0}$$

Reducing a method selection:

$$A \triangleq \bullet$$

$$\begin{array}{c} \left. \begin{array}{l} \epsilon \vdash \diamond \\ advFor(\langle VAL, \epsilon, \{\}, \epsilon \rangle, A) = \bullet \end{array} \right\} \\ \epsilon \vdash [l = \varsigma(x)] \rightsquigarrow [l = \varsigma(x)] \\ l \in \{\} \\ advFor(\langle IVK, \epsilon, \{\}, l \rangle, A) = \bullet \\ \left. \begin{array}{l} ib(\{\}, l) \vdash \diamond \\ advFor(\langle VAL, ib(\{\}, l), \{\}, \epsilon \rangle, A) = \bullet \end{array} \right\} \\ ib(\{\}, l) \vdash [] \rightsquigarrow [] \\ \epsilon \vdash [l = \varsigma(x)].l \rightsquigarrow [] \end{array} \quad \begin{array}{l} \text{RED VAL 0} \\ \text{RED SEL 0} \end{array}$$

Reducing a method update:

$$A \triangleq \bullet$$

$$\begin{array}{c} \left. \begin{array}{l} \epsilon \vdash \diamond \\ advFor(\langle VAL, \epsilon, \{\}, \epsilon \rangle, A) = \bullet \end{array} \right\} \\ \epsilon \vdash [l = \varsigma(x)x] \rightsquigarrow [l = \varsigma(x)x] \\ advFor(\langle UPD, \epsilon, \{\}, l \rangle, A) = \bullet \\ l \in \{\} \\ \epsilon \vdash ([l = \varsigma(x)x].l \Leftarrow \varsigma(y)) \rightsquigarrow [l = \varsigma(y)] \end{array} \quad \begin{array}{l} \text{RED VAL 0} \\ \text{RED UPD 0} \end{array}$$

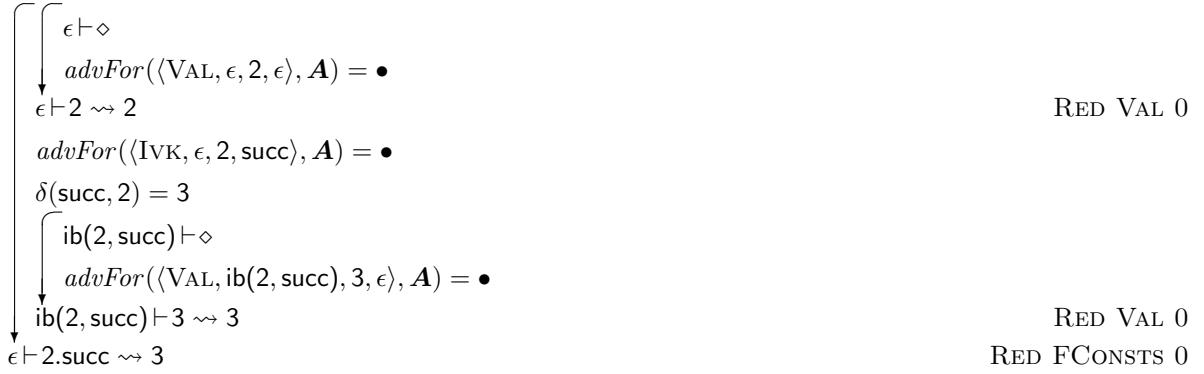
Reducing a basic constant:

$$A \triangleq \bullet$$

$$\begin{array}{c} \left. \begin{array}{l} \epsilon \vdash \diamond \\ advFor(\langle VAL, \epsilon, 2, \epsilon \rangle, A) = \bullet \end{array} \right\} \\ \epsilon \vdash 2 \rightsquigarrow 2 \end{array} \quad \text{RED VAL 0}$$

Reducing a functional constant application:

$$A \triangleq \bullet$$

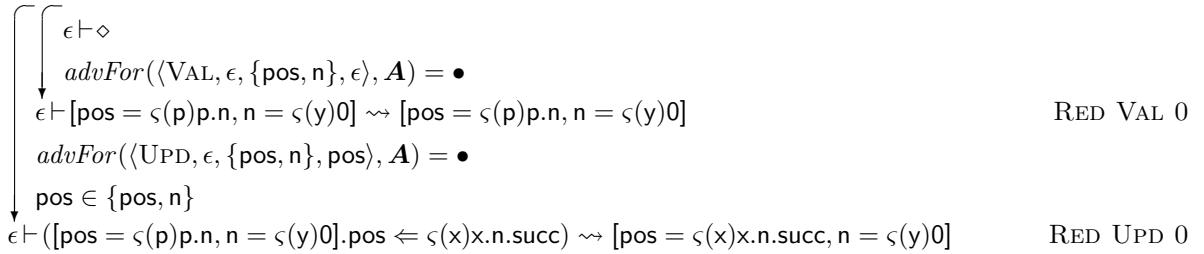


4.2 Advice on Update

This section gives sample reductions using a variant of M_S that associates advice with update operations instead of selections. Using the point cut description language lets us demonstrate the results of the substitution examples from the explanation of ς_{asp} [2, §2.1.4].

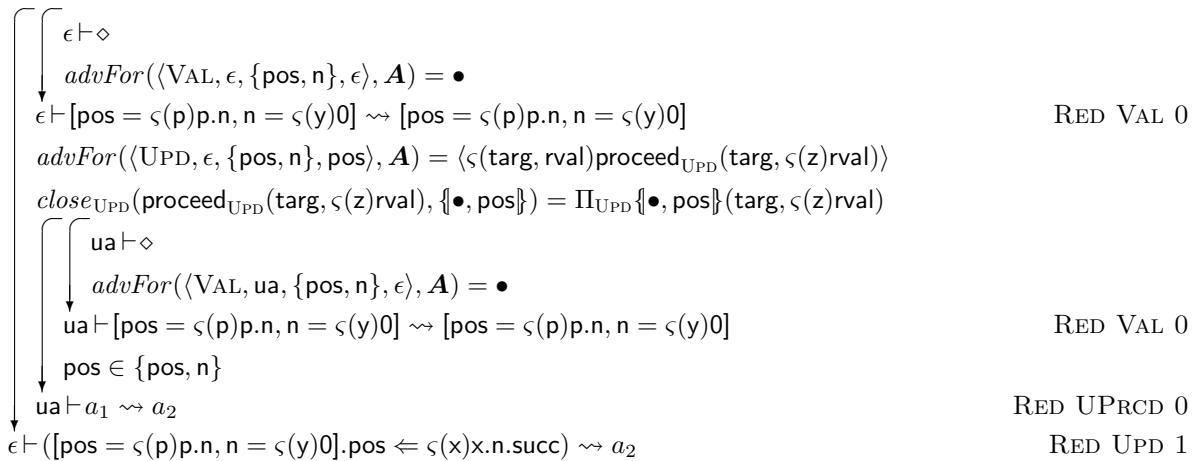
Reduction without advice:

$$A \triangleq \bullet$$



Reduction with advice that avoids capture:

$$\begin{aligned}
A &\triangleq \langle [pos, n].pos \triangleright \varsigma(targ, rval) \text{proceed}_{UPD}(targ, \varsigma(z)rval) \rangle \\
a_1 &\triangleq \Pi_{UPD}\{\bullet, pos\}([pos = \varsigma(p)p.n, n = \varsigma(y)0], \varsigma(z)[pos = \varsigma(p)p.n, n = \varsigma(y)0].n.succ) \\
a_2 &\triangleq [pos = \varsigma(z)[pos = \varsigma(p)p.n, n = \varsigma(y)0].n.succ, n = \varsigma(y)0]
\end{aligned}$$



Reduction with advice that uses capture:

$$\begin{aligned}
A &\triangleq \langle [pos, n].pos \triangleright \varsigma(targ, rval) \text{proceed}_{UPD}(targ, \varsigma(targ)rval.succ) \rangle \\
a_1 &\triangleq \Pi_{UPD}\{\bullet, pos\}([pos = \varsigma(p)p.n, n = \varsigma(y)0], \varsigma(targ)targ.n.succ.succ) \\
a_2 &\triangleq [pos = \varsigma(targ)targ.n.succ.succ, n = \varsigma(y)0]
\end{aligned}$$

$$\begin{array}{l}
\left(\begin{array}{l} \epsilon \vdash \diamond \\ \downarrow \\ \text{advFor}(\langle \text{VAL}, \epsilon, \{\text{pos}, \text{n}\}, \epsilon \rangle, A) = \bullet \\ \downarrow \\ \epsilon \vdash [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)0] \rightsquigarrow [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)0] \end{array} \right) \quad \text{RED VAL 0} \\
\text{advFor}(\langle \text{UPD}, \epsilon, \{\text{pos}, \text{n}\}, \text{pos} \rangle, A) = \langle \varsigma(\text{targ}, \text{rval}) \text{proceed}_{\text{UPD}}(\text{targ}, \varsigma(\text{targ})\text{rval}.succ) \rangle \\
\text{close}_{\text{UPD}}(\text{proceed}_{\text{UPD}}(\text{targ}, \varsigma(\text{targ})\text{rval}.succ), \{\bullet, \text{pos}\}) = \Pi_{\text{UPD}}\{\bullet, \text{pos}\}(\text{targ}, \varsigma(\text{targ})\text{rval}.succ) \\
\left(\begin{array}{l} \left(\begin{array}{l} \text{ua} \vdash \diamond \\ \downarrow \\ \text{advFor}(\langle \text{VAL}, \text{ua}, \{\text{pos}, \text{n}\}, \epsilon \rangle, A) = \bullet \\ \downarrow \\ \text{ua} \vdash [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)0] \rightsquigarrow [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)0] \end{array} \right) \quad \text{RED VAL 0} \\ \text{pos} \in \{\text{pos}, \text{n}\} \\ \text{ua} \vdash a_1 \rightsquigarrow a_2 \end{array} \right) \quad \text{RED UPRCD 0} \\
\epsilon \vdash ([\text{pos} = \varsigma(p)p.n, n = \varsigma(y)0].\text{pos} \Leftarrow \varsigma(x)x.n.succ) \rightsquigarrow a_2 \quad \text{RED UPD 1}
\end{array}$$

4.3 Before, After, and Around Advice

This section gives sample reductions using M_S to demonstrate the modeling of before, after, and around advice presented in the explanation of ς_{asp} [2, §2.1.5].

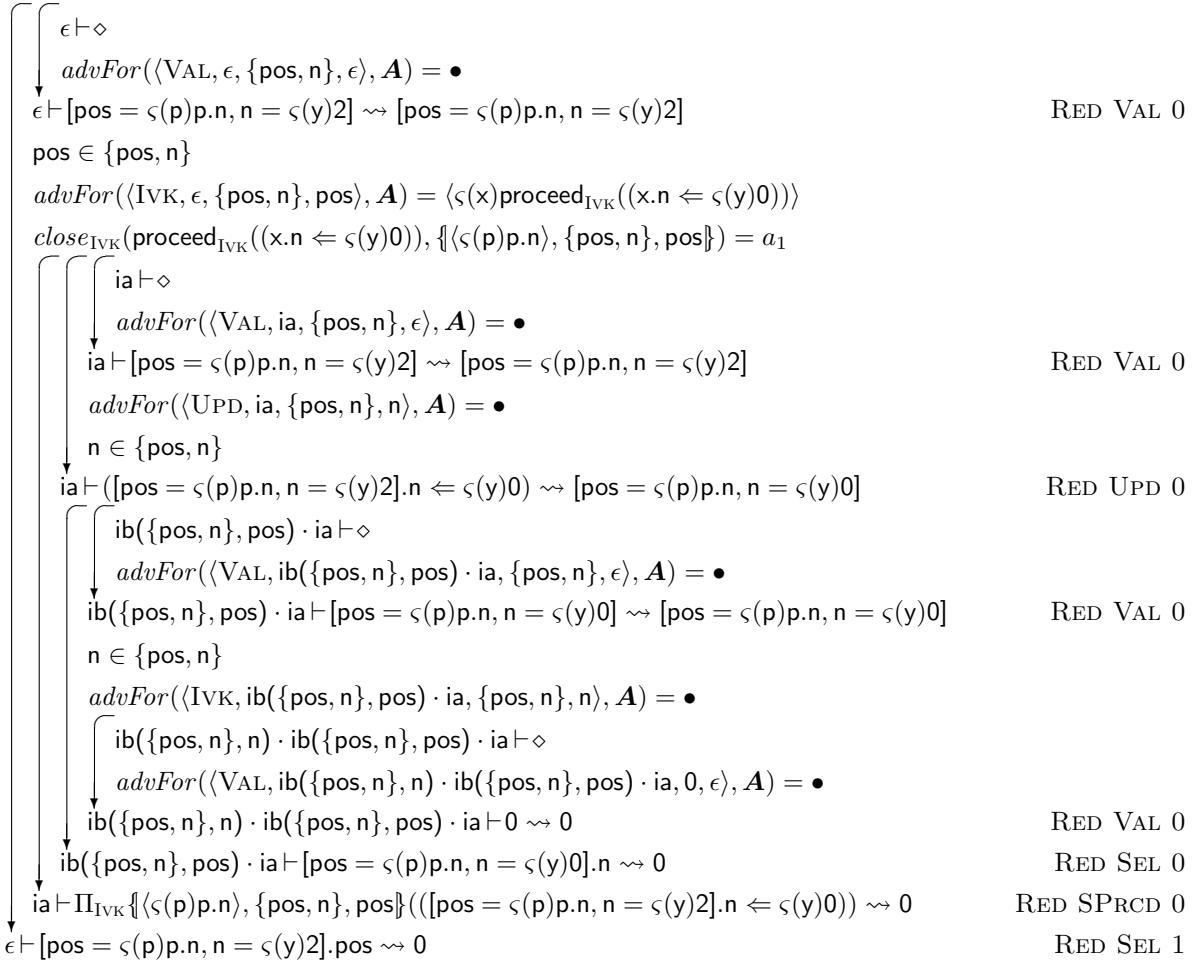
Without advice:

$$A \triangleq \bullet$$

$$\begin{array}{l}
\left(\begin{array}{l} \epsilon \vdash \diamond \\ \downarrow \\ \text{advFor}(\langle \text{VAL}, \epsilon, \{\text{pos}, \text{n}\}, \epsilon \rangle, A) = \bullet \\ \downarrow \\ \epsilon \vdash [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)2] \rightsquigarrow [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)2] \end{array} \right) \quad \text{RED VAL 0} \\
\text{pos} \in \{\text{pos}, \text{n}\} \\
\text{advFor}(\langle \text{IVK}, \epsilon, \{\text{pos}, \text{n}\}, \text{pos} \rangle, A) = \bullet \\
\left(\begin{array}{l} \left(\begin{array}{l} \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}) \vdash \diamond \\ \downarrow \\ \text{advFor}(\langle \text{VAL}, \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}), \{\text{pos}, \text{n}\}, \epsilon \rangle, A) = \bullet \\ \downarrow \\ \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}) \vdash [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)2] \rightsquigarrow [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)2] \end{array} \right) \quad \text{RED VAL 0} \\ \text{n} \in \{\text{pos}, \text{n}\} \\ \text{advFor}(\langle \text{IVK}, \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}), \{\text{pos}, \text{n}\}, \text{n} \rangle, A) = \bullet \\ \left(\begin{array}{l} \left(\begin{array}{l} \text{ib}(\{\text{pos}, \text{n}\}, \text{n}) \cdot \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}) \vdash \diamond \\ \downarrow \\ \text{advFor}(\langle \text{VAL}, \text{ib}(\{\text{pos}, \text{n}\}, \text{n}) \cdot \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}), 2, \epsilon \rangle, A) = \bullet \\ \downarrow \\ \text{ib}(\{\text{pos}, \text{n}\}, \text{n}) \cdot \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}) \vdash 2 \rightsquigarrow 2 \end{array} \right) \quad \text{RED VAL 0} \\ \text{ib}(\{\text{pos}, \text{n}\}, \text{pos}) \vdash [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)2].\text{n} \rightsquigarrow 2 \quad \text{RED SEL 0} \\ \epsilon \vdash [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)2].\text{pos} \rightsquigarrow 2 \quad \text{RED SEL 0} \end{array} \right)
\end{array}$$

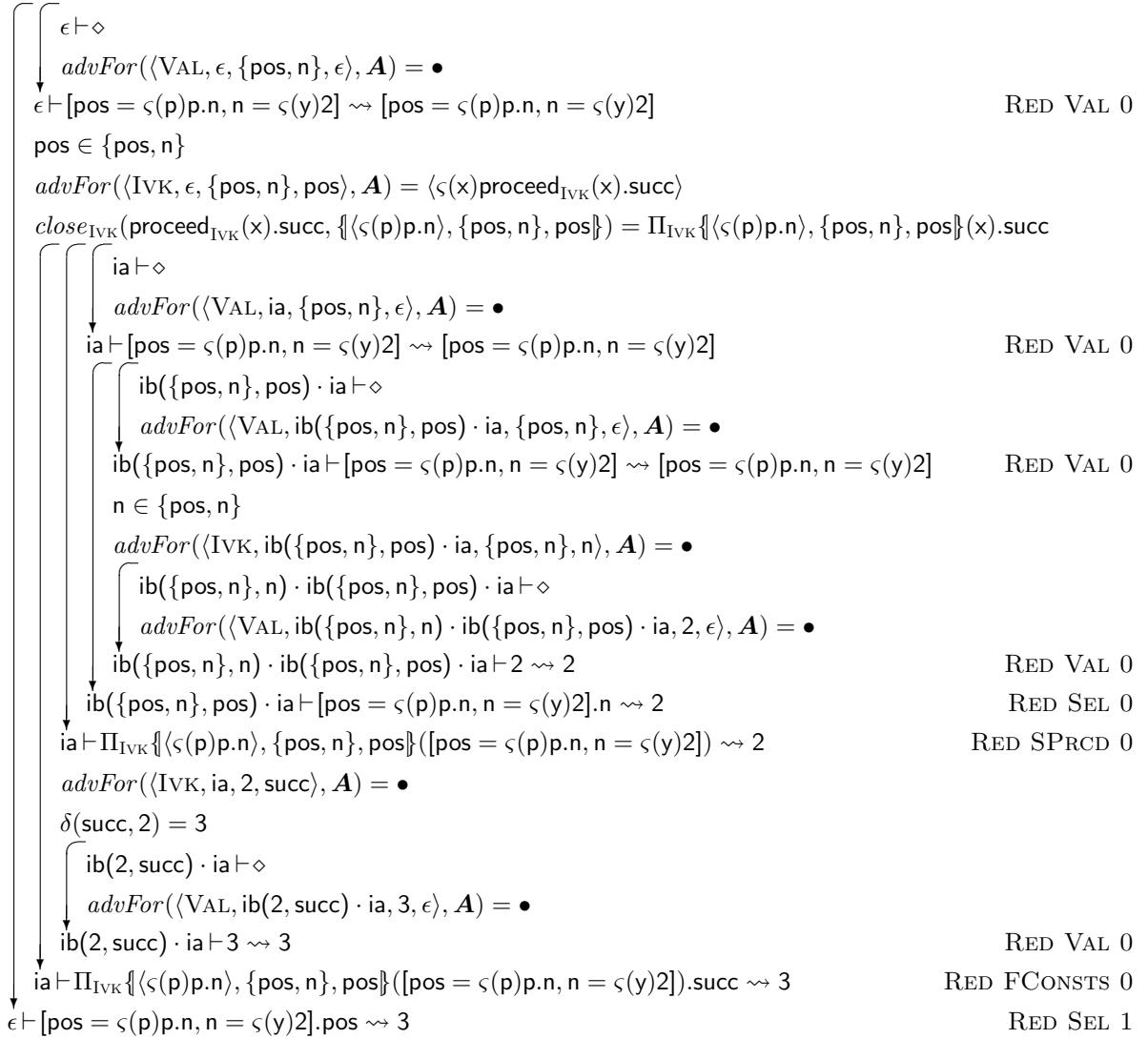
With before advice:

$$\begin{aligned}
A &\triangleq \langle [\text{pos}, \text{n}].\text{pos} \triangleright \varsigma(x)\text{proceed}_{\text{IVK}}((x.n \Leftarrow \varsigma(y)0)) \rangle \\
a_1 &\triangleq \Pi_{\text{IVK}}\{\langle \varsigma(p)p.n \rangle, \{\text{pos}, \text{n}\}, \text{pos}\}((x.n \Leftarrow \varsigma(y)0))
\end{aligned}$$



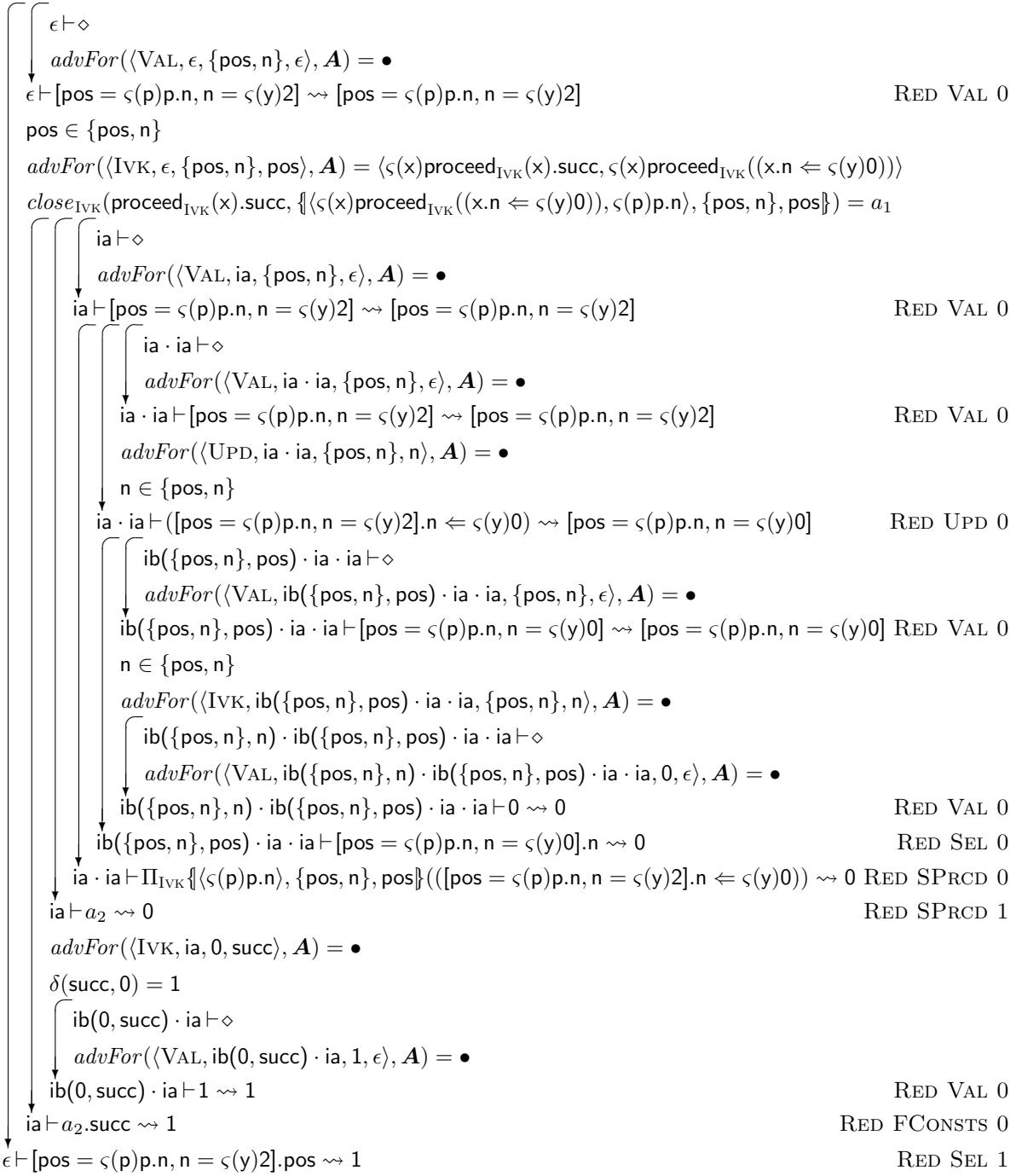
With after advice:

$$A \triangleq \langle [pos, n].pos \triangleright \varsigma(x)proceed_{I_{VK}}(x).succ \rangle$$



With around advice:

$$\begin{aligned}
 A &\triangleq \langle [pos, n].pos \triangleright \varsigma(x)proceed_{I_{VK}}(x).succ, [pos, n].pos \triangleright \varsigma(x)proceed_{I_{VK}}((x.n \Leftarrow \varsigma(y)0)) \rangle \\
 a_1 &\triangleq \Pi_{I_{VK}}\{\langle \varsigma(x)proceed_{I_{VK}}((x.n \Leftarrow \varsigma(y)0)), \varsigma(p)p.n \rangle, \{pos, n\}, pos\}(x).succ \\
 a_2 &\triangleq \Pi_{I_{VK}}\{\langle \varsigma(x)proceed_{I_{VK}}((x.n \Leftarrow \varsigma(y)0)), \varsigma(p)p.n \rangle, \{pos, n\}, pos\}([pos = \varsigma(p)p.n, n = \varsigma(y)2])
 \end{aligned}$$



4.4 Open Classes

This section gives sample reductions using M_G to model open classes, as described in §3.1.2 of the explanatory report [2].

Without advice:

$$A \triangleq \bullet$$

$\epsilon \vdash \diamond$	
\downarrow	
$advFor(\langle VAL, \epsilon, \{pos, n\}, \epsilon \rangle, A) = \bullet$	
$\epsilon \vdash [pos = \varsigma(p)p.n, n = \varsigma(y)0] \rightsquigarrow [pos = \varsigma(p)p.n, n = \varsigma(y)0]$	RED VAL 0
$advFor(\langle UPD, \epsilon, \{pos, n\}, n \rangle, A) = \bullet$	
$n \in \{pos, n\}$	
$\epsilon \vdash ([pos = \varsigma(p)p.n, n = \varsigma(y)0].n \Leftarrow \varsigma(y)2) \rightsquigarrow [pos = \varsigma(p)p.n, n = \varsigma(y)2]$	RED UPD 0
$pos \in \{pos, n\}$	
$advFor(\langle IVK, \epsilon, \{pos, n\}, pos \rangle, A) = \bullet$	
$\left\{ \begin{array}{l} ib(\{pos, n\}, pos) \vdash \diamond \\ \downarrow \\ advFor(\langle VAL, ib(\{pos, n\}, pos), \{pos, n\}, \epsilon \rangle, A) = \bullet \end{array} \right.$	
$ib(\{pos, n\}, pos) \vdash [pos = \varsigma(p)p.n, n = \varsigma(y)2] \rightsquigarrow [pos = \varsigma(p)p.n, n = \varsigma(y)2]$	RED VAL 0
$n \in \{pos, n\}$	
$advFor(\langle IVK, ib(\{pos, n\}, pos), \{pos, n\}, n \rangle, A) = \bullet$	
$\left\{ \begin{array}{l} ib(\{pos, n\}, n) \cdot ib(\{pos, n\}, pos) \vdash \diamond \\ \downarrow \\ advFor(\langle VAL, ib(\{pos, n\}, n) \cdot ib(\{pos, n\}, pos), 2, \epsilon \rangle, A) = \bullet \end{array} \right.$	
$ib(\{pos, n\}, n) \cdot ib(\{pos, n\}, pos) \vdash 2 \rightsquigarrow 2$	RED VAL 0
$ib(\{pos, n\}, pos) \vdash [pos = \varsigma(p)p.n, n = \varsigma(y)2].n \rightsquigarrow 2$	RED SEL 0
$\epsilon \vdash ([pos = \varsigma(p)p.n, n = \varsigma(y)0].n \Leftarrow \varsigma(y)2).pos \rightsquigarrow 2$	RED SEL 0

Without advice on update, just on values:

$$\begin{aligned}
 A &\triangleq \langle VAL \wedge S = \{pos, n\} \triangleright \varsigma(a_4) \rangle \\
 a_1 &\triangleq [\color = \varsigma(s)0, pos = \varsigma(s)s.orig.pos, orig = \varsigma(s)a_2, n = \varsigma(y)2] \\
 a_2 &\triangleq \Pi_{VAL}\{\bullet, a_3\}() \\
 a_3 &\triangleq [pos = \varsigma(p)p.n, n = \varsigma(y)0] \\
 a_4 &\triangleq [\color = \varsigma(s)0, pos = \varsigma(s)s.orig.pos, orig = \varsigma(s)\text{proceed}_{VAL}(), n = \varsigma(s)s.orig.n] \\
 a_5 &\triangleq [\color = \varsigma(s)0, pos = \varsigma(s)s.orig.pos, orig = \varsigma(s)a_2, n = \varsigma(s)s.orig.n] \\
 a_6 &\triangleq [\color = \varsigma(s)0, pos = \varsigma(s)s.orig.pos, orig = \varsigma(s)a_7, n = \varsigma(s)s.orig.n] \\
 a_7 &\triangleq \Pi_{VAL}\{\bullet, a_3\}()
 \end{aligned}$$

$\epsilon \vdash (a_3.n \Leftarrow \varsigma(y)2) \rightsquigarrow a_1$	see lemma 1
$pos \in \{\color, pos, orig, n\}$	
$advFor(\langle IVK, \epsilon, \{\color, pos, orig, n\}, pos \rangle, A) = \bullet$	
$\left\{ \begin{array}{l} ib(\{\color, pos, orig, n\}, pos) \vdash \diamond \\ \downarrow \\ advFor(\langle VAL, ib(\{\color, pos, orig, n\}, pos), \{\color, pos, orig, n\}, \epsilon \rangle, A) = \bullet \end{array} \right.$	
$ib(\{\color, pos, orig, n\}, pos) \vdash a_1 \rightsquigarrow a_1$	RED VAL 0
$orig \in \{\color, pos, orig, n\}$	
$advFor(\langle IVK, ib(\{\color, pos, orig, n\}, pos), \{\color, pos, orig, n\}, orig \rangle, A) = \bullet$	
$\left\{ \begin{array}{l} ib(\{\color, pos, orig, n\}, orig) \cdot ib(\{\color, pos, orig, n\}, pos) \vdash \diamond \\ \downarrow \\ ib(\{\color, pos, orig, n\}, orig) \cdot ib(\{\color, pos, orig, n\}, pos) \vdash a_2 \rightsquigarrow a_3 \end{array} \right.$	
$ib(\{\color, pos, orig, n\}, pos) \vdash a_1.orig \rightsquigarrow a_3$	RED VPRCD 0
$pos \in \{\color, pos, orig, n\}$	
$advFor(\langle IVK, ib(\{\color, pos, orig, n\}, pos), \{pos, n\}, pos \rangle, A) = \bullet$	
$ib(\{\color, pos, orig, n\}, pos) \cdot ib(\{\color, pos, orig, n\}, pos) \vdash a_3.n \rightsquigarrow 0$	RED SEL 0
$ib(\{\color, pos, orig, n\}, pos) \vdash a_1.orig.pos \rightsquigarrow 0$	see lemma 2 RED SEL 0

$$\downarrow \epsilon \vdash (a_3.n \Leftarrow \varsigma(y)2).pos \rightsquigarrow 0 \quad \text{RED SEL 0}$$

where lemma 1 is:

$$\begin{array}{l}
\left(\begin{array}{l}
\epsilon \vdash \diamond \\
advFor(\langle VAL, \epsilon, \{pos, n\}, \epsilon \rangle, A) = \langle \varsigma()a_4 \rangle \\
close_{VAL}(a_4, \{\bullet, a_3\}) = a_5 \\
\left(\begin{array}{l}
va \vdash \diamond \\
advFor(\langle VAL, va, \{color, pos, orig, n\}, \epsilon \rangle, A) = \bullet \\
va \vdash a_5 \rightsquigarrow a_5
\end{array} \right) \\
\epsilon \vdash a_3 \rightsquigarrow a_5 \\
advFor(\langle UPD, \epsilon, \{color, pos, orig, n\}, n \rangle, A) = \bullet \\
n \in \{color, pos, orig, n\} \\
\epsilon \vdash (a_3.n \Leftarrow \varsigma(y)2) \rightsquigarrow a_1
\end{array} \right) \\
\text{RED VAL 0} \\
\text{RED VAL 1} \\
\text{RED UPD 0}
\end{array}$$

lemma 2 is:

$$\begin{array}{l}
\left(\begin{array}{l}
ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash \diamond \\
advFor(\langle VAL, ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), \{pos, n\}, \epsilon \rangle, A) = \langle \varsigma()a_4 \rangle \\
close_{VAL}(a_4, \{\bullet, a_3\}) = a_6 \\
\left(\begin{array}{l}
va \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash \diamond \\
advFor(\langle VAL, va \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), \{color, pos, orig, n\}, \epsilon \rangle, A) = \bullet \\
va \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_6 \rightsquigarrow a_6
\end{array} \right) \\
ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_3 \rightsquigarrow a_6 \\
n \in \{color, pos, orig, n\} \\
advFor(\langle IVK, ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), \{color, pos, orig, n\}, n \rangle, A) = \bullet \\
\left(\begin{array}{l}
ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash \diamond \\
advFor(\langle VAL, ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), \{color, pos, orig, n\}, \epsilon \rangle, A) = \bullet \\
ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_6 \rightsquigarrow a_6
\end{array} \right) \\
orig \in \{color, pos, orig, n\} \\
advFor(\langle IVK, ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), \{color, pos, orig, n\}, orig \rangle, A) = \bullet \\
\left(\begin{array}{l}
ib(\{color, pos, orig, n\}, orig) \cdot ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash \diamond \\
ib(\{color, pos, orig, n\}, orig) \cdot ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_7 \rightsquigarrow a_3
\end{array} \right) \\
\text{RED VPRCD 0} \\
ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_6.orig \rightsquigarrow a_3 \quad \text{RED SEL 0} \\
n \in \{pos, n\} \\
advFor(\langle IVK, ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), \{pos, n\}, n \rangle, A) = \bullet \\
\left(\begin{array}{l}
ib(\{pos, n\}, n) \cdot ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash \diamond \\
advFor(\langle VAL, ib(\{pos, n\}, n) \cdot ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos), 0, \epsilon \rangle, A) = \bullet \\
ib(\{pos, n\}, n) \cdot ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash 0 \rightsquigarrow 0
\end{array} \right) \\
\text{RED VAL 0} \\
ib(\{color, pos, orig, n\}, n) \cdot ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_6.orig.n \rightsquigarrow 0 \quad \text{RED SEL 0} \\
ib(\{pos, n\}, pos) \cdot ib(\{color, pos, orig, n\}, pos) \vdash a_3.n \rightsquigarrow 0 \quad \text{RED SEL 0}
\end{array} \right)$$

Open classes:

$$\begin{aligned}
A &\triangleq \langle \text{VAL} \wedge S = \{\text{pos}, n\} \triangleright \varsigma()a_4, \text{UPD} \wedge S = \{\text{color}, \text{pos}, \text{orig}, n\} \wedge k = n \vee k = \text{pos} \triangleright \varsigma(t, r)a_2 \rangle \\
a_1 &\triangleq [\text{color} = \varsigma(s)a_5.\text{color}, \text{pos} = \varsigma(s)s.\text{orig}.pos, \text{orig} = \varsigma(s)\Pi_{\text{UPD}}\{\bullet, n\}(a_5.\text{orig}, \varsigma(tt)2), n = \varsigma(s)s.\text{orig}.n] \\
a_2 &\triangleq [\text{color} = \varsigma(s)t.\text{color}, \text{pos} = \varsigma(s)s.\text{orig}.pos, \text{orig} = \varsigma(s)\text{proceed}_{\text{UPD}}(t.\text{orig}, \varsigma(t)r), n = \varsigma(s)s.\text{orig}.n] \\
a_3 &\triangleq [\text{color} = \varsigma(s)t.\text{color}, \text{pos} = \varsigma(s)s.\text{orig}.pos, \text{orig} = \varsigma(s)\Pi_{\text{UPD}}\{\bullet, n\}(t.\text{orig}, \varsigma(t)r), n = \varsigma(s)s.\text{orig}.n] \\
a_4 &\triangleq [\text{color} = \varsigma(s)0, \text{pos} = \varsigma(s)s.\text{orig}.pos, \text{orig} = \varsigma(s)\text{proceed}_{\text{VAL}}(), n = \varsigma(s)s.\text{orig}.n] \\
a_5 &\triangleq [\text{color} = \varsigma(s)0, \text{pos} = \varsigma(s)s.\text{orig}.pos, \text{orig} = \varsigma(s)\Pi_{\text{VAL}}\{\bullet, a_9\}(), n = \varsigma(s)s.\text{orig}.n] \\
a_6 &\triangleq \Pi_{\text{UPD}}\{\bullet, n\}(a_8, \varsigma(tt)2) \\
a_7 &\triangleq [\text{pos} = \varsigma(p)p.n, n = \varsigma(tt)2] \\
a_8 &\triangleq a_{10}.\text{orig} \\
a_9 &\triangleq [\text{pos} = \varsigma(p)p.n, n = \varsigma(y)0] \\
a_{10} &\triangleq [\text{color} = \varsigma(ss)0, \text{pos} = \varsigma(ss)ss.\text{orig}.pos, \text{orig} = \varsigma(ss)\Pi_{\text{VAL}}\{\bullet, a_9\}(), n = \varsigma(ss)ss.\text{orig}.n] \\
a_{11} &\triangleq [\text{color} = \varsigma(s)0, \text{pos} = \varsigma(s)s.\text{orig}.pos, \text{orig} = \varsigma(s)a_{13}, n = \varsigma(s)s.\text{orig}.n] \\
a_{12} &\triangleq a_{11}.\text{orig}.n \\
a_{13} &\triangleq \Pi_{\text{VAL}}\{\bullet, a_7\}()
\end{aligned}$$

$$\begin{array}{lcl}
\left. \begin{array}{l} \epsilon \vdash (a_9.n \Leftarrow \varsigma(y)2) \rightsquigarrow a_1 \\ \text{pos} \in \{\text{color}, \text{pos}, \text{orig}, n\} \\ \text{advFor}(\langle \text{IVK}, \epsilon, \{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos} \rangle, A) = \bullet \\ \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_1.\text{orig}.pos \rightsquigarrow 2 \end{array} \right\} & & \text{see lemma 1} \\
\epsilon \vdash (a_9.n \Leftarrow \varsigma(y)2).\text{pos} \rightsquigarrow 2 & & \\
& & \left. \begin{array}{l} \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_1.\text{orig}.pos \rightsquigarrow 2 \\ \epsilon \vdash (a_9.n \Leftarrow \varsigma(y)2).pos \rightsquigarrow 2 \end{array} \right\} & & \text{see lemma 2} \\
& & & & \text{RED SEL 0}
\end{array}$$

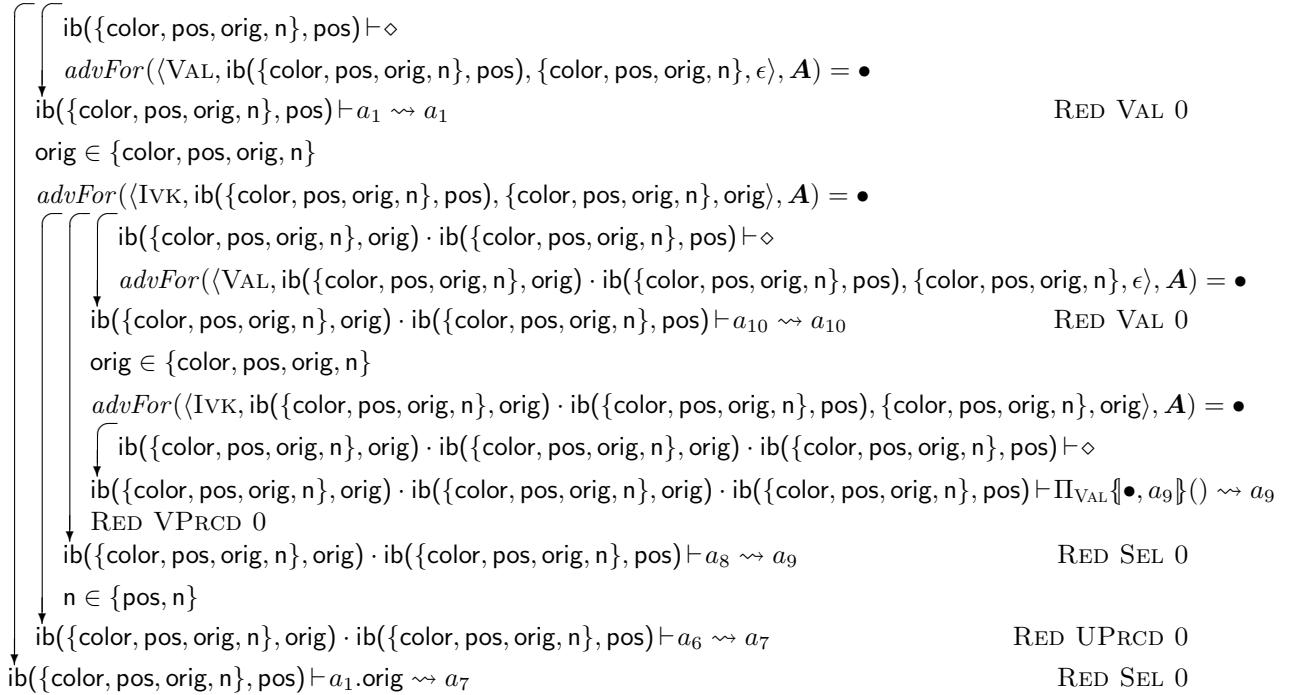
where lemma 1 is:

$$\begin{array}{lcl}
\left. \begin{array}{l} \epsilon \vdash \diamond \\ \text{advFor}(\langle \text{VAL}, \epsilon, \{\text{pos}, n\}, \epsilon \rangle, A) = \langle \varsigma()a_4 \rangle \\ \text{close}_{\text{VAL}}(a_4, \{\bullet, a_9\}) = a_5 \\ \left. \begin{array}{l} \text{va} \vdash \diamond \\ \text{advFor}(\langle \text{VAL}, \text{va}, \{\text{color}, \text{pos}, \text{orig}, n\}, \epsilon \rangle, A) = \bullet \\ \text{va} \vdash a_5 \rightsquigarrow a_5 \end{array} \right\} \\ \epsilon \vdash a_9 \rightsquigarrow a_5 \end{array} \right\} & & \\
& & \left. \begin{array}{l} \text{advFor}(\langle \text{VAL}, \text{va}, \{\text{color}, \text{pos}, \text{orig}, n\}, \epsilon \rangle, A) = \bullet \\ \text{va} \vdash a_5 \rightsquigarrow a_5 \end{array} \right\} & & \text{RED VAL 0} \\
& & \epsilon \vdash a_9 \rightsquigarrow a_5 & & \text{RED VAL 1} \\
& & \left. \begin{array}{l} \text{advFor}(\langle \text{UPD}, \epsilon, \{\text{color}, \text{pos}, \text{orig}, n\}, n \rangle, A) = \langle \varsigma(t, r)a_2 \rangle \\ \text{close}_{\text{UPD}}(a_2, \{\bullet, n\}) = a_3 \\ \left. \begin{array}{l} \text{ua} \vdash \diamond \\ \text{advFor}(\langle \text{VAL}, \text{ua}, \{\text{color}, \text{pos}, \text{orig}, n\}, \epsilon \rangle, A) = \bullet \\ \text{ua} \vdash a_1 \rightsquigarrow a_1 \end{array} \right\} \\ \epsilon \vdash (a_9.n \Leftarrow \varsigma(y)2) \rightsquigarrow a_1 \end{array} \right\} & & \\
& & \left. \begin{array}{l} \text{advFor}(\langle \text{VAL}, \text{ua}, \{\text{color}, \text{pos}, \text{orig}, n\}, \epsilon \rangle, A) = \bullet \\ \text{ua} \vdash a_1 \rightsquigarrow a_1 \end{array} \right\} & & \text{RED VAL 0} \\
& & \epsilon \vdash (a_9.n \Leftarrow \varsigma(y)2) \rightsquigarrow a_1 & & \text{RED UPD 1}
\end{array}$$

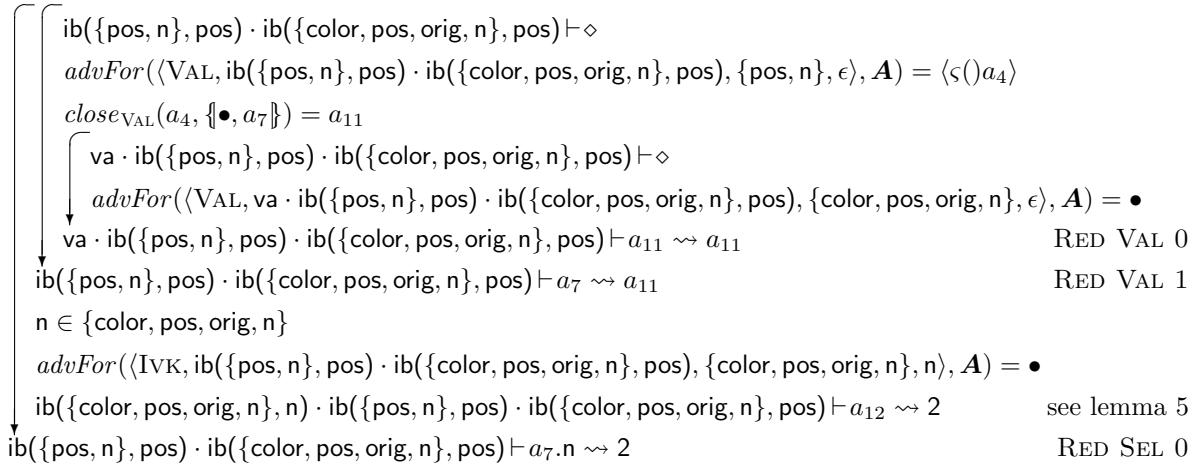
lemma 2 is:

$$\begin{array}{lcl}
\left. \begin{array}{l} \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_1.\text{orig} \rightsquigarrow a_7 \\ \text{pos} \in \{\text{pos}, n\} \\ \text{advFor}(\langle \text{IVK}, \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}), \{\text{pos}, n\}, \text{pos} \rangle, A) = \bullet \\ \text{ib}(\{\text{pos}, n\}, \text{pos}) \cdot \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_7.n \rightsquigarrow 2 \end{array} \right\} & & \text{see lemma 3} \\
\text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_1.\text{orig}.pos \rightsquigarrow 2 & & \\
& & \left. \begin{array}{l} \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_1.\text{orig}.pos \rightsquigarrow 2 \\ \text{ib}(\{\text{pos}, n\}, \text{pos}) \cdot \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_7.n \rightsquigarrow 2 \end{array} \right\} & & \text{see lemma 4} \\
& & \text{ib}(\{\text{color}, \text{pos}, \text{orig}, n\}, \text{pos}) \vdash a_1.\text{orig}.pos \rightsquigarrow 2 & & \text{RED SEL 0}
\end{array}$$

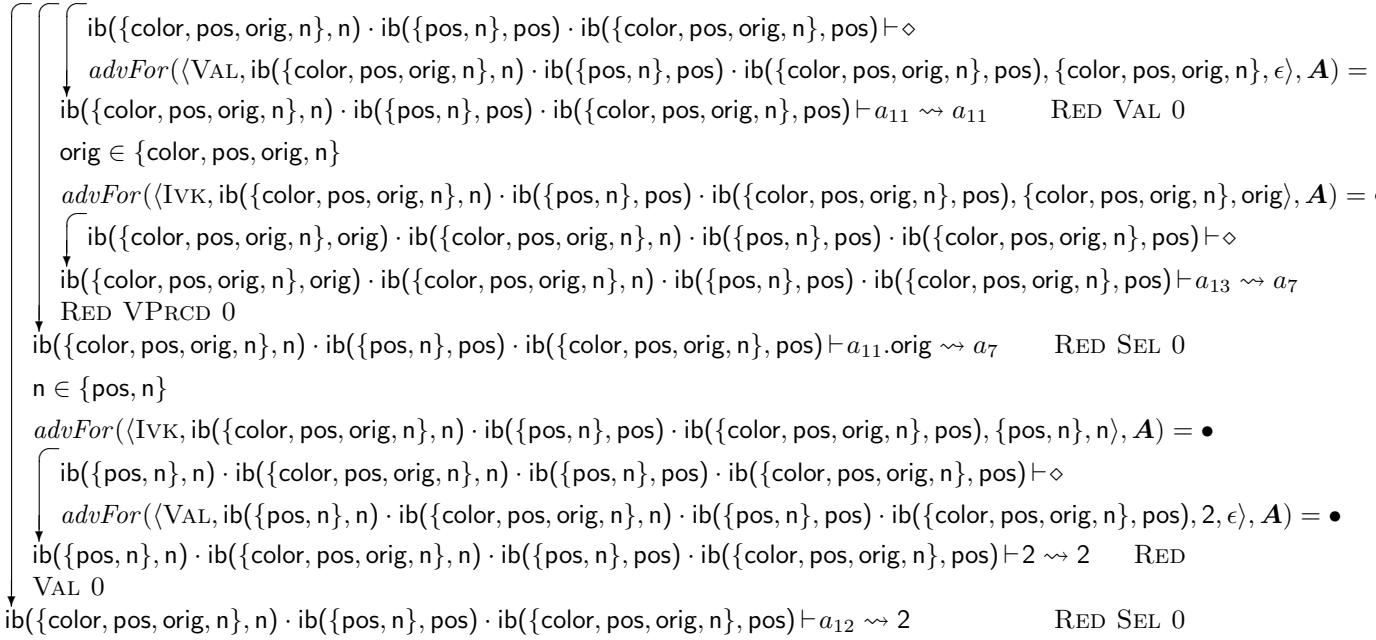
lemma 3 is:



lemma 4 is:



lemma 5 is:



4.5 HyperJ and Adaptive Methods

The reduction proofs for HyperJ and adaptive methods are quite tedious and so are omitted here. Nevertheless, these proofs have been mechanically checked and are included as test cases with the interpreter (in the class `sasp.MainTest` available from <http://www.cs.iastate.edu/~cclifton/sasp>).

5 Acknowledgments

The work of Clifton and Leavens was supported in part by the US National Science Foundation under grants CCR-0097907 and CCR-0113181.

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