Optimizing a Non-Deterministic Abstract Machine with Environments
Malgorzata Biernacka, Dariusz Biernacki, Sergueï Lenglet, Alan Schmitt

To cite this version:
Malgorzata Biernacka, Dariusz Biernacki, Sergueï Lenglet, Alan Schmitt. Optimizing a Non-Deterministic Abstract Machine with Environments. 2024. hal-04568253v3

HAL Id: hal-04568253
https://inria.hal.science/hal-04568253v3
Preprint submitted on 8 Jul 2024

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L’archive ouverte pluridisciplinaire HAL, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d’enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Distributed under a Creative Commons Attribution 4.0 International License
Optimizing a Non-Deterministic Abstract Machine with Environments

Małgorzata Biernacka  
Institute of Computer Science, University of Wrocław, Poland

Dariusz Biernacki  
Institute of Computer Science, University of Wrocław, Poland

Sergueï Lenglet  
Université de Lorraine, France  
Université Sorbonne Paris Nord, France

Alan Schmitt  
INRIA, France

Abstract

Non-deterministic abstract machine (NDAM) is a recent implementation model for programming languages where one must choose among several redexes at each reduction step, like process calculi. These machines can be derived from a zipper semantics, a mix between structural operational semantics and context-based reduction semantics. Such a machine has been generated also for the λ-calculus without a fixed reduction strategy, i.e., with the full non-deterministic β-reduction.

In that machine, substitution is an external operation that replaces all the occurrences of a variable at once. Implementing substitution with environments is more low-level and more efficient as variables are replaced only when needed. In this paper, we define a NDAM with environments for the λ-calculus without a fixed reduction strategy. We also introduce other optimizations, including a form of refocusing, and we show that we can restrict our optimized NDAM to recover some of the usual λ-calculus machines, e.g., the Krivine Abstract Machine. Most of the improvements we propose in this work could be applied to other NDAMs as well.

2012 ACM Subject Classification Theory of computation → Abstract machines

Keywords and phrases Abstract machine, Explicit substitutions, Refocusing.

Funding This work is partially funded by the National Science Centre of Poland under grant no. 2019/33/B/ST6/00289, and PHC Polonium.

Acknowledgements We thank the anonymous reviewers for their comments.

1 Introduction

Abstract machines are first-order tail-recursive transition systems for term reduction. Known examples of machines for languages based on the λ-calculus include SECD [38], CEK [31], and the KAM [37]. They serve as an operational semantics [38, 31, 30] or an implementation model [35, 40] of programming languages, but are also relevant in proof theory [37], higher-order model checking [45], or as cost models [10], to name but a few applications.

Abstract machines for the λ-calculus usually implement a deterministic strategy which selects a specific redex at each step [26, 30]. In contrast, a Non-Deterministic Abstract Machine (NDAM) [14] may choose any of the available redexes: it explores the term, making an arbitrary choice when several are possible, e.g., going left or right of an application. If the machine focuses on a subterm which cannot be further reduced, it annotates it and backtracks to make a different choice. Annotations prevent the machine from visiting an already explored subterm.

NDAMs have been originally developed as an implementation model for non-deterministic languages, in particular, for process calculi. However, considering an NDAM for the λ-calculus
without a fixed reduction strategy is worthwhile too, at least from the theoretical point of view – one can expect that such an NDAM should subsume various deterministic reduction strategies or abstract machines, and, therefore, play a role of a unifying framework in which one can study and compare them. Moreover, such an abstract machine should be readily adjustable to account for the \( \lambda \)-calculus extended with concurrency primitives.

Defining an NDAM can be hard, so the framework in [14] comes with a sound and complete automatic derivation procedure from another semantics format called zipper semantics, which is a middle ground between a Structural Operational Semantics (SOS) [43] and a context-based reduction semantics [31, 25]. Like an SOS, a zipper semantics explores a term with structural rules, but it remembers its position in the term using an evaluation context, i.e., a term with a hole [31]. Zipper semantics and their derived NDAMs have been presented for process calculi [46, 39], but also for the \( \lambda \)-calculus without a fixed reduction strategy.

As a result of the generic derivation procedure, the NDAM for the \( \lambda \)-calculus cannot be perfectly tailored for it, and, moreover, it inherits some of the artifacts from the underlying zipper semantics. In particular, it relies on substitution, an external operation which replaces all the occurrences of a variable at once, traversing the entire term. A more elementary and efficient implementation of \( \beta \)-reduction that is traditionally used in abstract machines uses environments, i.e., delayed substitutions, where a variable is looked up only when needed, and closures are built [38]. The goal of this work is to obtain an environment-based NDAM for the \( \lambda \)-calculus without a fixed reduction strategy.

To this end we first define a zipper semantics with environments for the \( \lambda \)-calculus without a fixed reduction strategy. It is inspired by \( \lambda \sigma \hat{\tau} \), a classic calculus with explicit substitutions [24]. It turns out that we only need to implement a small subset of the rewrite rules of Curien et al.’s calculus in our zipper semantics. Then, the NDAM automatically derived from the zipper semantics itself uses environments.

Unfortunately, the derived NDAM still suffers from performance issues. For example, it is not in refocused form [26, 27]: after a \( \beta \)-reduction, the NDAM reconstructs the whole term before looking for the next redex. The reason is that an NDAM should be able to find any redex, even the ones next to the root. As a compromise between efficiency and completeness, we let the machine reconstruct an arbitrary part of the term from the contraction site. It does not reconstruct the whole term after each \( \beta \)-reduction, but it may still reconstruct enough of it so that it can find \( \beta \)-redexes closer to the root. We also propose other optimizations, in particular related to the way the NDAM manages annotations.

The end product is an optimized NDAM for the \( \lambda \)-calculus with environments. By restricting its behavior, we are then able to express some of the existing \( \lambda \)-calculus machines or strategies, as we expected. For example, if we forbid the exploration step which goes under a \( \lambda \)-abstraction, we restrict ourselves to weak strategies. In addition, if we always go left of an application, we simulate the KAM [37]. We can also encode more complex strategies like full normal order by enforcing priorities on the machine steps of the optimized NDAM. We thus demonstrate that NDAMs can be brought closer to usual, more efficient abstract machines designed for a specific reduction strategy.

**Contributions.** Our contributions are as follows:

- We define a zipper semantics with environments for the \( \lambda \)-calculus, using a subset of a known calculus with explicit substitutions.
- From the zipper semantics we derive a correct-by-construction non-deterministic abstract machine for the \( \lambda \)-calculus with environments and without a fixed reduction strategy.
- We optimize, in a provably-correct way, the derived NDAM with optimization techniques
that we devise to improve the performance of NDAMs in general, which are either generic,
like a form of refocusing, or language-specific, like a better management of annotations.

We provide an OCaml implementation of the optimized machine to experiment with.

We show that the optimized NDAM we obtain is able to express many of the existing
machines designed for a specific reduction strategy.

Structure of the paper. Section 2 recalls the definition of the zipper semantics with
substitution, which we characterize with a zipper semantics with environments. In Section 3,
we first present the NDAM derived from the environment-based zipper semantics, and then
define a more optimized machine. In Section 4, we show how to obtain previously known
machines from the optimized NDAM. Section 5 discusses related work and Section 6 concludes
the paper. The appendix contains the proofs missing from the paper.

2 Zipper Semantics

We present the zipper semantics of the $\lambda$-calculus with environments and prove its equivalence
with the substitution-based variant.

2.1 Syntax and Substitution-based Zipper Semantics

We present the syntax of the $\lambda$-calculus where variables are represented with de Bruijn
indices. We let $n, m, p$ range over indices, i.e., natural numbers, and define the syntax of
$\lambda$-calculus terms, ranged over with $t$ and $s$, as follows.

$$t, s ::= \lambda.t \mid n \mid ts$$

A de Bruijn index counts how many $\lambda$-abstractions separate the variable from its binder. For
example, using more familiar named variables, $\lambda.0$ stands for $\lambda x.x$, and $\lambda.\lambda.1$ for $\lambda x.\lambda y.x$. We
use a de Bruijn representation for the environment-based calculus (Section 2.2), to prevent
unwanted name captures when moving an environment inside a $\lambda$-abstraction. We write
$t\{s/n\}$ for the capture-avoiding substitution of $n$ with $s$ in $t$, defined as follows, using the
auxiliary renaming operation on indices $\langle t\rangle^n_p$ [28].

$$
\begin{align*}
(t_1 t_2)\{s/n\} & \equiv t_1\{s/n\} t_2\{s/n\} \\
(\lambda.t)\{s/n\} & \equiv \lambda.t\{s/n + 1\} \\
(m\{s/n\}) & \equiv \begin{cases} m - 1 & \text{if } m > n \\
\langle s\rangle^n_0 & \text{if } m = n \\
m & \text{if } m < n \end{cases} \\
\langle m\rangle^n_p & \equiv \begin{cases} m + n & \text{if } m \geq p \\
m & \text{if } m < p \end{cases}
\end{align*}
$$

The goal of zipper semantics is to go through a term looking for a redex, using an
evaluation context to remember the current focus in the term. We represent evaluation
contexts $\mathcal{E}, \mathcal{F}$ as stacks of elementary contexts $\mathfrak{f}$, called frames.

$$\mathcal{E}, \mathcal{F} ::= \mathfrak{f} :: \mathcal{E} \mid \mathcal{f} \quad \mathfrak{f} ::= \mathfrak{a} t \mid \mathfrak{a} \mid \lambda$$

In defining the zipper semantics, it is convenient to interpret contexts inside-out: the topmost
frame represents the innermost construct. We define a plugging operation for frames $\mathfrak{f}\langle t\rangle$
environments $\epsilon$ shift (environments, context frames are built from terms or closures: based calculus, we use a context to remember the current position when exploring as source a term or a closure proofs, in particular to prove termination results. can be expressed with the other constructs [1, 24]: we use it because it is more convenient in that the lift construct is not mandatory in calculi with explicit substitutions, as its behavior is mapped to 0, and any $\epsilon$ mapped by Closures extend terms by pairing them with an environment $\lambda\sigma$ from an arbitrary $\beta$-reduce with the axiom $\text{lam}$. To avoid index clashes. Informally, the shift ($\uparrow$) is the lift $\uparrow\epsilon$ to avoid index clashes. Informally, $\epsilon$ should not be applied to 0 in $l$, as 0 is bound by the $\lambda$, and the occurrences of 0 in the image of $\epsilon$ should not be captured by that $\lambda$. To avoid this, we $\text{shift}$ ($\uparrow$) the result of the application of $\epsilon$. To summarize, $\uparrow\epsilon$ acts like $0 \cdot \epsilon \circ \uparrow$: the variable 0 is mapped to 0, and any $n \geq 1$ is mapped by $\epsilon$ and the result is then shifted. This reminds that the lift construct is not mandatory in calculi with explicit substitutions, as its behavior can be expressed with the other constructs [1, 24]: we use it because it is more convenient in proofs, in particular to prove termination results.

The zipper semantics for the environment-based $\lambda$-calculus is defined in Figure 2. It takes as source a term or a closure $l$, but always results in a closure $c$. As in the substitution-based calculus, we use a context to remember the current position when exploring $l$. With environments, context frames are built from terms or closures: $\mathfrak{F} ::= \circ l \mid l \circ \mid \lambda$. An evaluation

![Figure 1 Zipper semantics of the substitution-based lambda-calculus](image)
context $\mathcal{E}$, $\mathcal{F}$ is still a stack of frames interpreted inside-out, and we plug terms or closures inside a frame or a context. The plugging operation is defined as in Section 2.1.

The zipper semantics uses an extra argument to record what the current environment is while exploring $l$. To this end, we define an option type $\mathcal{E}$, $\mathcal{F}$, which we call local environment, signalling whether the environment is defined or not.

$\mathcal{E}, \mathcal{F} ::= \star | !e$

When $\mathcal{E} = \star$, the local environment is not (yet) defined. Although $\star$ could be seen as $\text{id}$, we distinguish the two to make explicit the absence or presence of an environment. We extend the following constructs to local environments and overload some of the symbols. The rationale behind these definitions is explained when we detail the rules of the zipper semantics.

$$l[\star] \triangleq l \quad l \cdot \star \triangleq l \cdot \text{id} \quad \epsilon \cdot \star \triangleq !\epsilon \quad \uparrow \star \triangleq \star$$

$$l[!e] \triangleq l[e] \quad l \cdot !e \triangleq l \cdot \epsilon \quad \epsilon_1 \cdot !e_2 \triangleq !\epsilon_1 \circ \epsilon_2 \quad \uparrow !e \triangleq !\uparrow !\epsilon$$

Zipper semantics transitions are of two kinds: $l \mapsto_{ev} \mathcal{E}, \mathcal{F}$ means that $l$ with local environment $\mathcal{E}$ inside $\mathcal{F}$ is reducing to $c$, i.e., $\mathcal{F}(l[\mathcal{E}])$ is $\beta$-reducing to $c$. The transition $\epsilon \mapsto_{\text{env}} \mathcal{E}$ means that solving $n$ in the environment $\epsilon \cdot \mathcal{E}$ results in a term or closure which itself reduces within $\mathcal{E}$ to $c$: $\mathcal{E}(n[\epsilon \cdot \mathcal{E}])$ is $\beta$-reducing to $c$. The rules defining the semantics implement some of the rewriting rules of $\lambda\sigma_\Downarrow$.

The transition $l \mapsto_{ev} \mathcal{E}, \mathcal{F}$ is going through $l$ until it finds a $\beta$-redex, keeping track of the current environment $\mathcal{E}$ while doing so. We start with an empty context and an undefined environment $l \mapsto_{ev} \mathcal{F}$. When we focus on a construct $l[\epsilon]$ (rule env), we set the local environment to $!e$ if it was undefined; otherwise, we compose it with $\epsilon$. The definition of $\text{id}$ reflects these two cases.

In the application case (rules $\text{appL}$ and $\text{appR}$), we pack the local environment with the term or closure pushed on the context. In the body of a $\lambda$-abstraction, the local environment becomes $\uparrow \mathcal{E}$ (rule $\text{lamb}$): if it is undefined ($\mathcal{E} = \star$), it remains so in the body of the abstraction, which explains why $\uparrow \star$ is defined as $\star$. Otherwise, $\mathcal{E} = !e$, and $\epsilon$ is lifted. For a variable $n$ (rule $\text{var}$), $\mathcal{E}$ should be defined as $!e$ for some $\epsilon$, so that we can resolve $n$ in $\epsilon$ using the $\text{var}$ rule.
We state the main properties of the environment-based zipper semantics: the proofs are in Appendix A. We first show it corresponds to the substitution-based semantics. To this end, Curien et al. prove a similar result when relating $\lambda \sigma^\eta$ to the plain $\lambda$-calculus [24, Proposition 4.2], and relies on the fact that we use lift; the proof would be more complex without lift [1].

Lemma 1. For all $l$, $k$, $\epsilon$, $[[k \cdot \epsilon]] = [[\epsilon]](\hat{0}l)/0$.

Curien et al. prove a similar result when relating $\lambda \sigma^\eta$ to the plain $\lambda$-calculus [24, Proposition 4.2].
rephrased in our setting, they prove that given two terms \( t \) and \( s \), \([t \cdot \text{id}] = t\{s/0}\).

We generalize their result to any closures \( l \), \( k \) and any environment \( \epsilon \). With this result, we can relate the two zipper semantics.

\begin{itemize}

\item \textbf{Theorem 2.} If \( l \xrightarrow{\star} \text{ev} c \), then \( [l] \xrightarrow{\text{ev}} [c] \).

For all \( t \xrightarrow{\text{ev}} s \), there exists \( c \) such that \( t \xrightarrow{\star} \text{ev} c \) and \([c] = s\).

\end{itemize}

The environment-based semantics is equivalent to the substitution-based one, which itself characterizes a context-based semantics of the \( \lambda \)-calculus [14].

The environment-based zipper semantics also satisfies the properties required so it can be derived into an NDAM [14]. In particular, it must be \textit{machine constructive}: in each rule, it is possible to construct the terms in the premise from those of the conclusion. The semantics must also be \textit{terminating}, ensuring that the search for a redex in the abstract machine does not loop infinitely. To prove it, it is enough to exhibit a strictly decreasing size of transitions such that the size of the premise of each inductive rule is smaller than that of its conclusion. The definition of such a size is already necessary to prove Theorem 2, and can be found in Appendix A on page 19. We point out that the size is strictly decreasing because we distinguish \( \star \) from \text{id}, which allows us to give different sizes for both. Replacing \( \star \) with \text{id} would make the zipper rule \text{id}! in Figure 2 not terminating.

The original theory [14] requires the zipper semantics to be \textit{reversible}, which is the reverse of being machine constructive: for each rule, the terms in the conclusion can be deduced from those in the premise. This property makes it easier to design the backtracking mechanism of the derived machine. The zipper semantics of Figure 2 is not reversible: in rules \text{consN} and \text{consZ}, respectively \( l \) and \( \epsilon \) in the conclusion do not occur in the premise. We also need to inverse the operator \( \bullet \) in rules \text{env}, \text{liftN}, and \text{comp}. We show in Section 3.1 that we only need a small change in the derivation procedure to allow backtracking for these rules. However, we argue that the derivation procedure could easily be adapted to account for non-reversible zipper semantics in many cases.

\section{Non-Deterministic Abstract Machines}

We discuss the shortcomings of the NDAM derived from the environment-based semantics, and present an optimized abstract machine dealing with these issues.

\subsection{Derived Abstract Machine}

A zipper semantics is not yet an implementation, as it does not explain what to do when several rules can be applied during the search for a redex, like in the application or \( \lambda \)-abstraction cases. The automatically derived abstract machine (DAM in short) augments the zipper semantics with a backtracking mechanism and annotations to handle these cases.

Roughly, each zipper rule becomes a \textit{forward step}, which focuses on a subterm. When several rules can be applied, the machine chooses arbitrarily one of them, e.g., going left or right of an application. Because of this non-determinism, it can make a wrong choice and focus on a subterm which cannot reduce. In such a case, the machine \textit{annotates} the subterm as a normal form, and then switches to a \textit{backward} mode, in which it undoes enough steps to return to the last configuration it made a choice, to make a different one if possible. The forward steps may focus only on subterms without annotations, ensuring that the machine cannot make the same (wrong) choice twice. The DAM relies on a stack \( \pi \) to remember in which order the forward steps have been applied. The backward mode then uses this stack to undo the steps in the reverse order.
Formally, we let $\Sigma$, $\Delta$ range over annotations, and we extend the syntax of terms, closures, and environments with annotations.

\[
\begin{align*}
\Sigma, \Delta &::= \emptyset \mid \text{ev} \mid \text{var} \\
\epsilon, \phi &::= \text{id}^{\Sigma} \mid \epsilon^{\Sigma} \mid l \cdot \epsilon \mid \epsilon \circ \phi \mid \epsilon^{\Sigma} \\
\rho &::= \Sigma^c \mid \lambda \Sigma^t \mid t @^{\Sigma} s \\
c, d &::= l[\epsilon]^{\Sigma} \mid \lambda \Sigma^c \mid c @^{\Sigma} d
\end{align*}
\]

We use $l$ and $k$ to range over (annotated) terms and closures. The machine annotates a term or closure with $\text{ev}$ and an environment with $\text{var}$. We write $\text{an}(l)$ or $\text{an}(\epsilon)$ for the topmost annotation of $l$ or $\epsilon$: for instance $\text{an}(l @^{\Sigma} k) = \Sigma$. We write $l^+\text{ev}$ and $\epsilon^{+\text{var}}$ for the operation of annotating the topmost construct of $l$ or $\epsilon$: for example, $(l @^{\Sigma} k)^+\text{ev} = l @^{\text{ev}} k$ independently of $\Sigma$. Contexts in machine configurations are built of annotated terms, closures, and environments.

The syntax of the stack $\pi$ is of the shape $\pi ::= \rho(\tilde{e}) \::= \pi | \triangledown$ where $\triangledown$ represents the empty stack, $\rho$ ranges over the names of the zipper rules in Figure 2, and $\tilde{e}$ represents information stored in the stack, which might be needed to make backtracking possible. In most cases, $\tilde{e}$ is empty.

The DAM is defined in Figure 4; we discuss some of the steps here. Machine configurations $\langle l | E, \varepsilon, \pi \rangle_{\text{ev}}$ and $\langle \epsilon | E, n, \varepsilon, \pi \rangle_{\text{var}}$ reflect the transitions of the zipper semantics. Each zipper rule is turned into a forward step, where the conclusion and premise of the rule are respectively the source and target of the step. The stack $\pi$ also records the rule being applied. For example, the rules $\text{appl}$, $\text{lam}$ and $\text{consN}$ translate as steps D1, D3, and D14, respectively. Each forward step has a side condition which checks that the new subterm under focus is not annotated, thus ensuring it has not already been visited.

If no forward step applies to the current term, then a switching step annotates it and changes to the backward mode $\text{bev}$ or $\text{bvar}$. The configurations $\langle \pi | l, E, \varepsilon \rangle_{\text{bev}}$ and $\langle \epsilon | E, n, \varepsilon \rangle_{\text{var}}$ then discriminate on $\pi$ to undo the last applied rule. Each forward step has a mirroring backward step to undo it. The switching steps are steps D7 and D21 for the $\text{ev}$ and $\text{var}$ modes respectively, and the backward steps corresponding to $\text{appl}$, $\text{lam}$ and $\text{consN}$ are steps D8, D11, and D22, respectively.

Finally, when the machine finds a redex, it applies the axiom $\text{lam}\beta$ (step D6). The machine constructs the result of the $\beta$-reduction plugged inside $E$. After reduction, some annotations may no longer be valid, so to be safe, the DAM erases all of them: the operation $|l|$ sets every annotation in $l$ to $\emptyset$.

**Reversibility.** In the original derivation procedure [14], the role of $\pi$ is mainly to record which rules are being applied, to know in which order to undo them if needed. We go beyond that design in this paper, and use $\pi$ to also store some information to be able to backtrack even in the case of non-reversible zipper rules. We argue that what we do is generic enough to be added to the original derivation procedure.

The rules $\text{consN}$, $\text{consZ}$, and $\text{liftZ}$ are not reversible because some entities of the conclusion of the rule do not occur in its premise. It is easy to automatically collect such missing entities, and record them in the stack $\pi$. For example, the steps D14 and D15—corresponding to $\text{consN}$ and $\text{consZ}$—remember in $\pi$ the part of the environment which is discarded, either its head or its tail.

A rule may be not reversible also because it relies on an external operator which is not injective, like $\circ$. In that case, we need to provide extra operations and remember some information in the stack to be able to invert it. By its definition, $\epsilon \circ \varepsilon = \epsilon \circ \epsilon_1 \circ \epsilon_2$ because either $\epsilon = \epsilon_1$ and $\varepsilon = \epsilon_2$, or $\epsilon = \epsilon_1 \circ \epsilon_2$ and $\varepsilon = \epsilon$. To distinguish between the two possibilities, we remember in the stack $\pi$ a boolean $b$ defined by an operation $\varepsilon \equiv b$ which evaluates to $\top$.
\[
\langle l_1 \circ^0 l_2 \mid E, \varepsilon, \pi \rangle_{ev} \rightarrow \langle l_1 \mid \circ (l_2[\varepsilon]) \rangle_{ev} \langle E, \varepsilon, \text{app} \pi \rangle_{ev} \quad \text{if } an(l_1) = \emptyset \quad (D1)
\]
\[
\langle l_1 \circ^0 l_2 \mid E, \varepsilon, \pi \rangle_{ev} \rightarrow \langle l_2 \mid (l_1[\varepsilon]) \circ E, \varepsilon, \text{app} \pi \rangle_{ev} \quad \text{if } an(l_2) = \emptyset \quad (D2)
\]
\[
\langle \lambda^0, l \mid E, \varepsilon, \pi \rangle_{ev} \rightarrow \langle l \mid \lambda : E, \top \rangle_{ev} \quad \text{if } an(l) = \emptyset \quad (D3)
\]
\[
\langle l[l[\varepsilon] \mid \circ k \mid E, \varepsilon, \pi \rangle_{ev} \rightarrow \langle l[l[l[k \cdot \varepsilon, \text{env} \varepsilon \cdot \pi \rangle_{ev} \quad \text{if } an(l) = \emptyset \quad (D4)
\]
\[
\langle n^0 \mid E, l, \pi \rangle_{ev} \rightarrow \langle e \mid E, n, \pi, \text{var} \rangle_{ev} \quad \text{if } an(e) = \emptyset \quad (D5)
\]
\[
\langle \lambda^0, l \mid \circ k \mid E, \varepsilon, \pi \rangle_{ev} \rightarrow \langle \lambda^0, l \mid E, \varepsilon, \pi \rangle_{ev} \quad \text{otherwise (cf steps D4, D18, and D20).}
\]
\[
\langle l \mid E, \varepsilon, \pi \rangle_{ev} \rightarrow \langle l \mid l \cdot \pi \rangle_{ev} \langle E, \varepsilon, \pi \rangle_{ev} \quad \text{otherwise (D7)}
\]

\[
\langle \text{app} \pi \mid l_1 \circ (l_2[\varepsilon]) \rangle_{ev} \rightarrow \langle l_1 \circ^0 l_2 \mid E, \varepsilon, \pi \rangle_{ev} \quad (D8)
\]
\[
\langle \text{app} \pi \mid l_1 \circ (l_1[\varepsilon]) \rangle_{ev} \rightarrow \langle l_1 \circ^0 l_2 \mid E, \varepsilon, \pi \rangle_{ev} \quad (D9)
\]
\[
\langle \text{env}(b) \pi \mid l \mid E, \varepsilon \rangle_{ev} \rightarrow \langle l[l[l[b \cdot \varepsilon, \text{env} \varepsilon \cdot \pi \rangle_{ev} \quad \text{if } \bullet^{-1}(b, \varepsilon) = (e, \varepsilon) \quad (D10)
\]
\[
\langle \text{lam} \pi \mid l \mid \lambda : E, \top \rangle_{ev} \rightarrow \langle \lambda^0, l \mid E, \varepsilon, \pi \rangle_{ev} \quad (D11)
\]
\[
\langle \text{var} \pi \mid e \mid E, n, \pi \rangle_{ev} \rightarrow \langle n^0 \mid E, l, \pi \rangle_{ev} \quad (D12)
\]
\[
\langle \bullet \mid l \mid \pi \rangle_{ev} \rightarrow \langle l[l[\varepsilon] \rangle_{nf} \quad (D13)
\]

\[
\langle l^0 \varepsilon \mid E, n + 1, \varepsilon, \pi \rangle_{\text{var}} \rightarrow \langle e \mid E, n, \varepsilon, \text{cons} \pi \rangle_{\text{var}} \quad \text{if } an(e) = \emptyset \quad (D14)
\]
\[
\langle l^0 \varepsilon \mid E, 0, \varepsilon, \pi \rangle_{\text{var}} \rightarrow \langle l \mid E, \varepsilon, \text{cons} \pi \rangle_{ev} \quad \text{if } an(l) = \emptyset \quad (D15)
\]
\[
\langle 1^0 d^0 \mid E, n, l, \pi \rangle_{\text{var}} \rightarrow \langle e \mid E, n, \pi, \text{id} \rangle_{\text{var}} \quad \text{if } an(e) = \emptyset \quad (D16)
\]
\[
\langle 1^0 \varepsilon \mid E, n, l, \pi \rangle_{\text{var}} \rightarrow \langle e \mid E, n + 1, \pi, \text{shift} \rangle_{\text{var}} \quad \text{if } an(e) = \emptyset \quad (D17)
\]
\[
\langle 1^0 \varepsilon \mid E, n + 1, \varepsilon, \pi \rangle_{\text{var}} \rightarrow \langle e \mid E, n, \text{lift} \rangle_{\text{var}} \quad \text{if } an(e) = \emptyset \quad (D18)
\]
\[
\langle 1^0 \varepsilon \mid E, 0, \pi \rangle_{\text{var}} \rightarrow \langle e \mid E, 0, \pi, \text{lift} \rangle_{\text{var}} \quad \text{if } an(e) = \emptyset \quad (D19)
\]
\[
\langle e_1 \circ e_2 \mid E, n, \varepsilon, \pi \rangle_{\text{var}} \rightarrow \langle e_1 \mid E, e_2 \cdot \pi \rangle_{\text{var}} \quad \text{if } an(e_1) = \emptyset \quad (D20)
\]
\[
\langle e \mid E, n, \varepsilon, \pi \rangle_{\text{var}} \rightarrow \langle e \mid E, n, \varepsilon, \pi \rangle_{\text{var}} \quad \text{otherwise (D21)}
\]

\[
\langle \text{cons} \pi \rangle_{\text{var}} \rightarrow \langle l^0 \varepsilon \mid E, n + 1, \varepsilon, \pi \rangle_{\text{var}} \quad (D22)
\]
\[
\langle \text{cons} \pi \rangle_{\text{var}} \rightarrow \langle l^0 \varepsilon \mid E, 0, \varepsilon, \pi \rangle_{\text{var}} \quad (D23)
\]
\[
\langle \text{id} \rangle_{\text{var}} \rightarrow \langle e \mid E, n, \pi \rangle_{\text{var}} \quad (D24)
\]
\[
\langle \text{shift} \rangle_{\text{var}} \rightarrow \langle l \mid E, n, \pi \rangle_{\text{var}} \quad (D25)
\]
\[
\langle \text{lift} \rangle_{\text{var}} \rightarrow \langle e \mid E, n + 1, \varepsilon, \pi \rangle_{\text{var}} \quad \text{if } \bullet^{-1}(b, \varepsilon) = (e, \varepsilon) \quad (D26)
\]
\[
\langle \text{lift} \rangle_{\text{var}} \rightarrow \langle e \mid E, 0, \varepsilon, \pi \rangle_{\text{var}} \quad \text{if } \bullet^{-1}(b, \varepsilon) = (e, \varepsilon) \quad (D27)
\]

\[
\langle \text{comp} \rangle_{\text{var}} \rightarrow \langle e_1 \circ e_2 \mid E, n, \varepsilon, \pi \rangle_{\text{var}} \quad \text{if } \bullet^{-1}(b, \varepsilon') = (e_1, \varepsilon) \quad (D28)
\]

**Figure 4** Derived Abstract Machine

if $\varepsilon = \pi$, and to $\perp$ otherwise (cf steps D4, D18, and D20). We then define the invert of $\bullet$, written $\bullet^{-1}$, such that $\bullet^{-1}(\pi, e_1 \circ e_2) = (e_1 \circ e_2, \pi)$ and $\bullet^{-1}(\perp, e_1 \circ e_2) = (e_1, e_2)$, and use it.
Optimizing a Non-Deterministic Abstract Machine with Environments

3.2 Optimized Abstract Machine

Being the result of a generic procedure, the DAM is not optimized for the particular case of the λ-calculus with environments. We explain how to optimize it, resulting in the Optimized Abstract Machine, defined in Figure 5.

Bypassing the bvar mode. The DAM does not exploit the fact that solving n in ϵ is syntax-directed: at most one zipper rule applies to a given n and ϵ. Suppose the machine

\[
\begin{align*}
\langle l_1 \cdot \epsilon \mid E, E, P \rangle_{\text{var}} &\rightarrow \langle l_1 \mid \epsilon \mid E, E, P \rangle_{\text{var}} \\
\langle l \cdot \epsilon \mid E, 0, E, P \rangle_{\text{var}} &\rightarrow \langle l \mid E, E, P \rangle_{\text{ev}} \\
\langle \lambda x \cdot E[n, l, P] \rangle_{\text{var}} &\rightarrow \langle \lambda x \mid E[n, l, P] \rangle_{\text{var}} \\
\langle \lambda x \cdot E[n, l, P] \rangle_{\text{var}} &\rightarrow \langle \lambda x \mid E[n, l, P] \rangle_{\text{var}} \\
\langle \lambda \cdot E \mid E \rangle_{\text{rec}} &\rightarrow \langle \lambda \mid E \rangle_{\text{rec}} \\
\langle E \mid c \rangle_{\text{rec}} &\rightarrow \langle E \mid c \rangle_{\text{rec}} \\
\langle \lambda x \cdot c \rangle_{\text{rec}} &\rightarrow \langle \lambda x \mid E \rangle_{\text{rec}} \\
\langle E \mid c \rangle_{\text{rec}} &\rightarrow \langle E \mid c \rangle_{\text{rec}} \\
\end{align*}
\]

Figure 5 Optimized Abstract Machine

in the backtracking steps D10, D26, and D28. Defining \( \cdot^{-1} \) or \( = \) cannot be done for any non injective operator, but if they exist, such operations could be provided to the automatic derivation procedure with the zipper semantics.
configuration $D = \langle n^\Sigma | E, \epsilon, \pi \rangle_{ev}$ initiates a sequence of steps $D \rightarrow^m D'_{var}$ such that no forward step applies to $D'_{var}$. The DAM then backtracks: starting with $D'_{var}$ and for each step, it applies the switching step to add the annotation, and then a backward step. In the end, we need $2m$ steps to add the annotations and get back to $D$. Instead, the switching step from $D_{var}$ could restore $D$ directly, meaning that we no longer need backtracking steps for the var mode, nor annotations for environments.

To implement this change, whenever we start solving $n$ inside $\epsilon$ in the OAM, we remember the initial values of the variable and the environment as a pair $(n, \epsilon)$ in the var mode (step O5). We let $P$ range over such pairs. The pair is passed along in steps O13 to O19, and is discarded if the variable can in fact be solved in step O14. If variable resolution fails, the “otherwise” step restores the initial values as a closure $n[\epsilon]^ev$ annotated as a normal form (step O20).

Refocusing. The DAM is not in refocused form: after $\beta$-reduction, it plugs the resulting closure into $E$ before looking for the next redex. This is necessary for the machine to be complete w.r.t. the zipper semantics, so that it is able to reach any available redex, including those in $E$. A middle ground between performance and completeness is to reconstruct an arbitrary part of the context $E$, while still having the possibility to recreate it completely.

In the OAM, $\beta$-reduction produces a closure in the context $E$ (step O6). The OAM then enters the rec mode dedicated to partially reconstruct the full closure using $E$. The steps O21 to O23 push the current top frame of the context on the closure. At any point, step O24 may conclude the reconstruction and restart the machine in the ev mode to find the next redex. The reconstruction being non-deterministic itself, the machine may reconstruct fully using the whole context, meaning that any redex in $E$ is still reachable, and the machine is still complete w.r.t. the zipper semantics.

Removing the stack $\pi$. The stack $\pi$ is used for backtracking, mainly to remember in which order to undo forward steps. To simplify the machine, we would like to use the context $E$ for that role, as it uniquely changes when the rules appL, appR, and lam are applied. However, it stays the same when env or var are applied (cf. Figure 2). The idea is then to undo the former steps, but not the latter ones. Such a change entails significant differences in the behavior of the OAM compared to the DAM, that we illustrate on an example in Figure 6. The first picture represents a possible exploration path in a closure, which goes through several constructs $l_1[\epsilon_1] \ldots l_m[\epsilon_m]$. Along this path, the two machines collect $\epsilon_1 \ldots \epsilon_m$ and
combine them into an environment $\epsilon$ (second picture).

If solving $n$ in $\epsilon$ produces a normal form $d$, the DAM restores the closure in its original shape while adding annotations, up to some application where it stops backtracking and starts exploring another path. Concretely, the DAM undoes the variable resolution steps, annotates $n$ as a normal form, and puts back the environments $(\epsilon_1, \ldots, \epsilon_m)$ at their original position. The OAM backtracks up to the same application, annotating all the constructs on the way, but without undoing the variable resolution steps—leaving $d$ in place—or restoring $(\epsilon_1, \ldots, \epsilon_m)$.

In practice, the OAM does not feature the stack $\pi$, and there are only three backtracking steps (O9 to O11) which mirror the steps O1 to O3. The steps O4 and O5 cannot be reversed. A construct $\Sigma^i$ in the OAM is therefore either yet to be traversed, or of the form $n[\epsilon]^\star$ after a failed variable resolution. If we go through a plain term (without any $\Sigma^i$), we might reach a variable $n$ with the local environment $\star$. To be uniform, we create an annotated closure $n[\text{id}]^\star$ (step O7), turning the plain term itself into a closure. Unlike in the DAM, plain terms cannot be annotated in the OAM: only closures are annotated.

**Keeping annotations.** Instead of erasing all the annotations after $\beta$-reduction as in the DAM, we would like to keep as much of them as possible. Suppose the $\beta$-redex is $(\lambda^0. l) @^0 k$ in some context $E$ and current environment $\Sigma$. The normal forms in $E$ are still normal form after the $\beta$-reduction, so if any of them are annotated as such, we should keep these annotations. In contrast, the annotations in the body of the $\lambda$-abstraction $l$ may no longer be valid: a term $0^\star @^\epsilon t$ is no longer a normal form if the $\beta$-reduction replaces 0 with a $\lambda$-abstraction. As an approximation, we ignore all the annotations inside $l$.

We therefore need a way to know if an annotation is in the body of a $\beta$-redex. We notice that if solving a variable in a local environment $\Sigma$ produces a normal form, then the machine adds an annotation and resets $E$ in the configuration to $\star$ (steps O20 or O7). We then propagate the annotation upward from the leaves with the combinations of step O8 and steps O9 to O11. In that process, the local environment remains $\star$. We have $E \neq \star$ only when we explore a closure for which we do not know yet if it is a normal form. This observation is enough to know if an annotation is meaningful after a $\beta$-reduction. Let $c$ such that $\text{an}(c) = \text{ev}$ in some environment $E$; the annotation is valid only if $E \equiv \star$. Otherwise, the environment has changed after $c$ has been annotated and $c$ is in the body of a $\beta$-redex.

On top of that, we need to consider a particular case. Suppose we look for a redex in $(\lambda^0. 0 @^0 l) @^0 \lambda^0 k$, starting with the $\lambda$-abstraction on the right, annotating it as $(\lambda^0. 0 @^0 l) @^0 \lambda^0 k$. After $\beta$-reduction and solving 0, we get $(\lambda^0 k) @^0 \Sigma^i$ in the environment $\star$ (with $\epsilon = \lambda^0 k :: \text{id}$); according to the previous criterion, the annotation of the $\lambda$-abstraction is valid, preventing the machine to $\beta$-reduce further. For a $\lambda$-abstraction, to be a normal form means it is not in the function position of an application. If a $\lambda$-abstraction ends up in such a position after a $\beta$-reduction, we need to ignore its annotation.

Consider a local environment $\Sigma$ and a term or closure $l$. We use booleans $\top, \bot$, ranged over by $b$, to indicate whether $l$ is in the function position of an application. Given a context $E$, the function $\text{infun}(E)$ returns $\top$ if the hole of the context is in function position ($E = \circ \ l :: E'$), and $\bot$ otherwise. We write $\Sigma, b \vdash l$ if the annotation of $l$ is still valid, i.e., with $\Sigma \equiv \star$, and $l$ is a closure $c$ such that $\text{an}(c) = \text{ev}$, and in the case where $l$ is a $\lambda$-abstraction, with $b = \bot$. Otherwise, we write $\Sigma, b \not\vdash l$. We use this predicate as a side condition of the steps O1 to O3 and step O14, meaning that we can take these steps if the new subterm under focus is not annotated, or its annotation is meaningless. There is no side condition for step O4, because for this step, the local environment $\epsilon \bullet \Sigma$ cannot be $\star$ by definition of $\epsilon$. 
Example. We show a machine run starting from \(t = (\lambda^0.0 \oplus^0 \lambda^0.1) \oplus^0 (\lambda^0.0 \oplus^0 1\), i.e., \((\lambda x. x) \oplus (\lambda y. x)\). To simplify, we omit empty annotations and the symbol \(\oplus\) if it is not annotated, writing for instance \(t\) as \((\lambda 0.1) (\lambda 0)\).

The machine may go right of the topmost application, focusing on an unsolvable variable, since \(E\) is \(*\) for now. The machine annotates it and backtracks.

\[
(t | *, *)_{ev} \xrightarrow{O^2} (1 | (\lambda 0.1) (\lambda 0) \oplus *)_{ev} \xrightarrow{O^7} ((\lambda 0.1) (\lambda 0) \oplus *)_{ev} \xrightarrow{O^7} \Delta \mathcal{O}_1
\]

The machine has no choice but to go left of the application. Then it might go left again, and inside the body of the \(\lambda\)-abstraction.

\[
\mathcal{O}_1 \xrightarrow{O} \xrightarrow{O} (0.1 | \lambda \:: \cdot 0.0 :: 1[\text{id}]^{ev} :: *)_{ev} \xrightarrow{\Delta} \mathcal{O}_2
\]

The term 0 \(\lambda.1\) is a normal form, so the machine annotates it and backtracks.

\[
\mathcal{O}_2 \xrightarrow{O^8} \xrightarrow{O} (\lambda :: 0.0 :: 1[\text{id}]^{ev} :: 0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev})_{bev}
\]

\[
\mathcal{O}_3 \xrightarrow{O^1} (\lambda :: 0.0 :: 1[\text{id}]^{ev} :: 0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev})_{bev}
\]

At this point, the machine \(\beta\)-reduces and switches to the \(\text{rec}\) mode, in which it may decide to reconstruct the frame in the context. It then restarts in \(\text{ev}\) mode to find the next redex.

\[
\mathcal{O}_3 \xrightarrow{O^6} ([0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev}]_{rec}
\]

\[
\mathcal{O}_4 \xrightarrow{O^2} ([0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev}]_{rec}
\]

The machine cannot go right of the application, because we have \(*, \perp \vdash 1[\text{id}]^{ev}\); the annotation is still valid. The machine therefore has to go left.

\[
\mathcal{O}_4 \xrightarrow{O^1} ([0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev}]_{rec}
\]

Because the local environment is no longer \(*\), the annotations of the subterm under focus are not valid and can be ignored. For instance, we can go right of the application and under the \(\lambda\)-abstraction to reach the variable.

\[
\mathcal{O}_5 \xrightarrow{O^2} ([0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev}]_{rec}
\]

Solving the variable in the \(\text{var}\) mode produces \(\lambda 0\), a normal form, annotated as such by the machine, which then goes back to the first application in the context.

\[
\mathcal{O}_6 \xrightarrow{\cdot \cdot \cdot \cdot \lambda 0.1 [\text{id}]^{ev} \cdot 1[\text{id}]^{ev} :: \lambda 0.0 :: 1[\text{id}]^{ev} :: (*)_{ev}
\]

\[
\mathcal{O}_7 \xrightarrow{\cdot \cdot \cdot \cdot \lambda 0.1 [\text{id}]^{ev} \cdot 1[\text{id}]^{ev} :: \lambda 0.0 :: 1[\text{id}]^{ev} :: (*)_{ev}
\]

The machine has no choice but to go left of the application, until it reaches the variable, which can be solved.

\[
\mathcal{O}_7 \xrightarrow{O^1} \xrightarrow{O^2} ([0[\text{id}]^{ev} \oplus \lambda \text{id}. [\text{id}]^{ev} \cdot 1[\text{id}]^{ev}]_{rec}
\]

\[
\mathcal{O}_8 \xrightarrow{O^3} \xrightarrow{O^4} \xrightarrow{O^5} \xrightarrow{O^6} \xrightarrow{O^7} \xrightarrow{O^8} \xrightarrow{\Delta}
\]
The machine may then β-reduce and not reconstruct anything from the context.

\[ O_8 \overset{OBO24}{\to} (0 | \lambda^e \cdot \lambda^e.0[\uparrow]_e \cdot \text{id}) | \circ 1[\text{id}]^e \cdot \text{id} \]  

The variable can be solved immediately, producing a λ-abstraction. A β-reduction is then possible, without any reconstruction because the context is empty.

\[ O_9 \to^* (\lambda^e \cdot \lambda^e.0[\uparrow]_e \cdot \text{id}) | \circ 1[\text{id}]^e \cdot \text{id} \]  

The machine has to go through the resulting closure to annotate it as a normal form, pushing the environment \(1[\text{id}]^e \cdot \text{id}\) in the body of the λ-abstraction while doing so. The run ends with \((\lambda^e.0[\uparrow]_e \cdot \text{id}) | \circ 1[\text{id}]^e \cdot \text{id}\).

**Implementation.** We have implemented the OAM as a skeletal semantics [19] from which we extract an OCaml program that evaluates all possible reductions from a starting term. The implementation is very close to the formal definition of the OAM and is available online [13].

### 3.3 Correspondence between the DAM and the OAM

To relate the OAM and the DAM, we define an equivalence \(\approx\) between OAM and DAM machine states, and prove that from a sequence of OAM steps, we can build a sequence of equivalent DAM steps and conversely. We let \(O\) and \(D\) range over the machines states of the optimized and the derived machines, respectively. We informally explain the ingredients needed to define \(\approx\); the formal definitions and proofs are in Appendix B.

Consider two similar runs of the OAM and DAM from the same starting point: after doing the same β-reductions, they are at the same position in the closure. As explained in Figure 6 some environments may be pushed to the leaves and some variables may be solved in the OAM, but not in the DAM. Besides the closures on the OAM side may contain more annotations than on the DAM one, but some of these annotations could be meaningless.

We relate two states \(O\) and \(D\) by pushing all the environments to the leaves, and then solving all the variables, which is exactly what the translation \(\llbracket\llbracket\) of Figure 3 is doing. We define two similar translations on annotated closures or terms \(\llbracket\llbracket_0\) (for the OAM) and \(\llbracket\llbracket_1\) (for the DAM) which produce annotated terms. The translation \(\llbracket\llbracket_0\) also removes meaningless annotations, by using the same criterion as in Section 3.2. We then compare the results \(r_0\) and \(r_d\) of these translations using a predicate \(\triangleright\), which checks that the two annotated terms are the same, except that \(r_0\) may contain more annotations than \(r_d\).

We extend \(\llbracket\llbracket_0\), \(\llbracket\llbracket_1\) and \(\triangleright\) to contexts frame by frame, and we write \(O \approx D\) if either:

- \(O = \langle l | E, \llbracket_0E\rangle, D = \langle k | F, \llbracket_0F\rangle\) and \(\llbracket_0E\llbracket_0F\triangleright \llbracket_1E\llbracket_1F\);  
- \(O = \langle E | C_1\rangle_{\llbracket_0C_1\rangle}, D = \langle F | C_2\rangle_{\llbracket_0C_2\rangle}\) and \(\llbracket_0E\llbracket_0F\triangleright \llbracket_1E\llbracket_1F\);  
- \(O = \langle c | \triangleright, D = \langle d | \triangleright\rangle\) and \(\llbracket_0c\llbracket_0\triangleright \llbracket_1d\llbracket_1\).

The equivalence is defined only on the modes which are common to both machines. Equivalent machine states can be considered as synchronization points between the OAM and DAM machine runs. In particular, the contexts \(E\) and \(F\) of two equivalent states have the same shape, so the machines are at the same position in the term or closure being explored.

We extend the equivalence to sequences of machines states, writing \((O_0, O_1, \ldots, O_i) \approx (D_0, D_1, \ldots, D_i)\) if \(O_0 = \llbracket_0E_0\llbracket_0F_0\), \(D_0 = \llbracket_1E_0\llbracket_1F_0\) for the same \(l, F_0 \to^* O_j \to^* O_{j+1}\) and \(D_j \to^* F_{j+1}\) for all \(0 \leq j \leq i\), and \(O_j \approx D_j\) for all \(0 \leq j \leq i\). The correspondence theorems then state that given two equivalent sequences, if one of the machines is doing a step, then we can find equivalent states to continue the sequences.
**Theorem 3** (OAM implies DAM). If \((O_0, O_1, \ldots, O_i) \approx (D_0, D_1, \ldots, D_i)\) and \(O_i \rightarrow O'_i\), there exist \(O_{i+1}\) and \(D_{i+1}\) such that \(O'_i \rightarrow^* O_{i+1}\), \(D_i \rightarrow^* D_{i+1}\), and \(O_{i+1} \approx D_{i+1}\).

**Theorem 4** (DAM implies OAM). If \((O_0, O_1, \ldots, O_i) \approx (D_0, D_1, \ldots, D_i)\) and \(O_i \rightarrow D'_i\), there exist \(O_{i+1}\) and \(D_{i+1}\) such that \(D'_i \rightarrow^* D_{i+1}\), \(O_i \rightarrow^* O_{i+1}\), and \(O_{i+1} \approx D_{i+1}\).

We sketch some of the cases of Theorem 3, the proofs for Theorem 4 are similar. Step O20 of the OAM (Figure 5) is matched in the DAM by a sequence of bvar steps which produces an equivalent configuration. Applying the axiom (step O6) leads to the rec mode, in which a number \(m\) of frames of \(E\) are reconstructed. If \(E\) contains \(n\) frames in total, the DAM matches by applying its \(\beta\)-reduction step (step D6 in Figure 4), which rebuilds the whole context, and then it gets to the same position as in the OAM with \(n-m\) forward steps.

A backward step of the OAM, e.g. step O9, is not necessarily matched by the corresponding step in the DAM—step D8 in Figure 4—because the DAM may have to undo steps corresponding to the applications of the rules env or var. We prove that we can undo them first and then apply step D8 in the DAM. Finally, suppose a valid annotation in the OAM has been erased at the same position in the DAM. The subterm under focus is still a normal form, so the sequence which added the annotation on the DAM side before its erasure can still be applied to that subterm to restore the annotation.

### 4 Recovering Known Strategies and Abstract Machines

The OAM can be used to obtain specialized abstract machines that implement various reduction strategies. A reduction strategy can be seen as a way to choose a particular redex. In the OAM, the transitions O1, O2, O3 are navigating through the term or closure by going left (\(\triangleright\)) or right (\(\triangleright\)) of the application, or down under the lambda (\(\downarrow\)), respectively. By controlling which of these navigation rules to apply and in which order, we can define various deterministic or non-deterministic strategies. To define reduction strategies succinctly, we use the phased format of strategies [17]. For example, the call-by-name strategy can be defined as cbn = \(\triangleright\cbn; \beta\), a strategy that navigates to the left of an application as much as possible (iterating itself), and then tries to perform \(\beta\)-reduction. Formally, a strategy can be defined as a set of term decompositions. The sequence of strategies \(r; s\) is a strategy containing any decomposition in \(r\), or a \(r\)-normal form that is in \(s\). Furthermore, \(r \cup s\) is a strategy that contains any decomposition that is in \(r\) or in \(s\). If a term admits multiple decompositions, according to \(r\) or to \(s\), the strategy \(r \cup s\) is non-deterministic.

**Krivine Abstract Machine.** The Krivine Abstract Machine [37] implements exactly call-by-name evaluation as defined above. In the OAM, it corresponds to restricting navigation to only \(\triangleright\), and removing \(\triangleright\) and \(\downarrow\). Thus, we go left as much as possible, and if we reach a \(\lambda\)-abstraction, we try to apply step O6 to \(\beta\)-reduce. If it fails because the context is empty, we are done, and we apply steps O8 and O12: the (annotated) lambda abstraction will be returned as the normal form. If, on the other hand, we reach a variable, the machine tries to solve it by rule O5. If it is a free variable, the machine cannot solve it and backtracks, thus reconstructing the call-by-name normal form—an application of a (annotated) variable to a number of closures (possibly none). An inspection of reachable machine configurations reveals that excluding the two navigation rules further entails that rules O10, O11, and also O15-O18, O22, O23 are not reachable: environments are constructed using only two operators, and thus the machine is significantly simplified. Moreover, the call-by-name strategy being deterministic, the reconstruction transitions can be short-circuited and eliminated altogether.
Right-to-left open call by value. We consider a variant of the call-by-value strategy that operates on open terms, and evaluates arguments in an application from right to left: it is defined as \( \text{rcbw} = \downarrow \text{rcbw}; \uparrow \text{rcbw}; \beta \). This strategy can be realized in the OAM by keeping \( \downarrow \) and \( \uparrow \) and forbidding \( \downarrow \), but with the priority given to \( \downarrow \), applying it whenever possible before we try \( \uparrow \). Whenever we can apply the \( \beta \)-rule, the argument on top of the stack must be a value, because of the precedence of \( \downarrow \) over \( \uparrow \): the machine only performs a restricted \( \beta \)-contraction, as in call-by-value. The resulting machine is closest to the variant presented in [18], or it can be seen as a right-to-left variant of the CEK machine [31].

In contrast, we cannot obtain the left-to-right variant of call-by-value in a similar way. To do it, we would need to prioritize \( \uparrow \) over \( \downarrow \), which in turn requires further inspection of the top of the stack to make sure that \( \beta \)-contraction is only applied when the argument is a value. Other similar call-by-value strategies [42, 7] cannot be expressed. It is due to the fact that OAM only looks at the top-level constructor and no further.

Normal order. The normal-order strategy is a fully reducing strategy that can be seen as an iteration of call-by-name until no redex is left anywhere in the term. Its phased form is \( \text{no} = (\beta; \downarrow \text{no}; \downarrow \text{no}) \cup \downarrow \text{no} \). It is worth noting that in this case the sum of strategies \( \cup \) does not produce a non-deterministic strategy, because \( (\beta; \downarrow \text{no}; \downarrow \text{no}) \) and \( \downarrow \text{no} \) each contains decompositions of structurally different terms. Normal order is an example of a hybrid strategy that does not behave uniformly, but the navigation steps depend on the context: the \( \downarrow \) rule can only be applied when the lambda abstraction is not in the operator position. A canonical abstract machine that implements normal order is Crécut’s KN machine [22]. In the OAM, we first try to apply \( \beta \), then \( \downarrow \), then \( \downarrow \). Otherwise, we can try \( \downarrow \) (only when we know that \( \beta \) is not applicable). This adjustment of OAM results in a variant of KN.

Weak reduction. A simple restriction that results in a non-deterministic strategy can be obtained by removing the rule \( \downarrow \) from the OAM. This strategy can be described in the phased form as \( \text{weak} = \downarrow \text{weak} \cup \downarrow \text{weak} \cup \beta \).

Head reduction. Consider the following two strategies in phased form: \( \text{head} = (\beta; \downarrow \text{head}) \cup \downarrow \text{head} \) and \( \text{ihead} = (\downarrow \text{ihead}; \beta) \cup \downarrow \text{ihead} \). They both perform head reduction and compute head normal forms (i.e., terms of the form \( \lambda x_1 \ldots \lambda x_n . t_1 \ldots t_n \)), but the first strategy is “outermost”, and the second is “innermost”. The two strategies are deterministic, because for each term only one of the \( \cup \)-substrategies can be used. If we restrict OAM in that we remove the rule \( \downarrow \) and order the remaining evaluation rules: first \( \downarrow \), \( \downarrow \), and then \( \beta \), we obtain the implementation of the \( \downarrow \text{head} \) strategy. If, on the other hand, we put these rules in the order: \( \beta \) first, then \( \downarrow \), then \( \downarrow \), we obtain the machine for the \( \downarrow \text{head} \) strategy. In both cases, we can further trim the machine by removing all the transitions made unreachable by removing \( \downarrow \).

Unexpressible strategies. We have remarked above that the left-to-right call by value is not expressible in an instance of OAM obtained by restriction and prioritization only. The same is true for any strategy that hinges on the same issue, like the fireball calculus [42, 7], the non-deterministic variant of CbV where we can reduce freely to the left or to the right of the application as long as it is possible, and only then we try to \( \beta \)-reduce. This strategy in phased form is defined as \( \text{cbw} = (\downarrow \text{cbw} \cup \downarrow \text{cbw}); \beta \), but in fact this definition enforces that \( \beta \)-contraction will be applied only when the argument is a weak normal form. So it is not possible to directly obtain this kind of behavior in the OAM without further inspection of the stack in O6, as in the left-to-right call by value.
5 Related Work

Non-deterministic abstract machines for the $\lambda$-calculus. In [14], Biernacka et al. present a general framework for deriving non-deterministic abstract machines from zipper semantics. One of the example derivations in their work leads to a non-deterministic abstract machine for the $\lambda$-calculus, where no reduction strategy is imposed and each redex in a given term can be selected for contraction thanks to a loop-free backtracking mechanism encoded in the transition system of the machine. The abstract machines of the present work follow the same principles, but they are optimized crucially in two ways: they replace substitution with environments and they are in the refocused form. Both these advancements greatly improve the performance measures of the machines since they influence, respectively, how substitution is implemented [6] and how term decomposition is resumed after a redex has been contracted [26, 27].

A lightweight form of non-determinism is present in [3], a recent work by Accattoli and Barenbaum who introduce an abstract machine for normal-order reduction in the $\lambda$-calculus that has a diamond property. Their machine is non-deterministic in that when dealing with a partially normalized term of the form $x t_1 \ldots t_n$, it can choose arbitrarily which argument $t_i$ to reduce next. (The machine can also jump from one job to another when a normal form has been reached, without a step-by-step backtracking.) The machine is diamond non-deterministic, i.e., it satisfies the strongest, one-step form of confluence, which implies that the choice of a reduction site does not affect the result, nor the length of the machine run. The machines considered in the present work, as more general, obviously are not diamond non-deterministic in this sense.

Explicit substitutions and environment-based abstract machines. In their pioneering work on explicit substitutions, Abadi et al. [1] present an abstract machine for weak-head normalization in the $\lambda\sigma$-calculus that generalizes the KAM (originally featuring de Bruijn indices and closures) to open terms. Full normalization is then achieved as an iteration of the weakly normalizing machine. Substitutions from the calculus play the role of environments in the machine, which makes the relationship between the calculus and the machine particularly explicit. The $\lambda\sigma$-calculus is also used by Crégut [22] for designing and establishing the correctness of his strongly normalizing abstract machines KN and KNL. Call-by-value and call-by-name reduction strategies and their implementation on an environment-based abstract machine are discussed in the context of a variant of $\lambda\sigma$ with names and garbage collection by Rose [44].

In [23], Curien introduces a minimal weak non-deterministic calculus of closures $\lambda\rho$, requiring a big-step reduction rule for term application, that allows him to express the reduction strategies for the KAM and the Categorical Abstract Machine (CAM) [21], taking advantage of the presence of closures in the calculus. The author also sketches a generalization of the weak calculus to a strong version that accounts for full $\beta$-reduction, and is formalized in [24] as $\lambda\sigma_\delta$.

The full power of the $\lambda\sigma_\delta$-calculus is then used in Hardin et al.’s framework for reasoning about the correspondence between several weak reduction strategies in $\lambda\sigma_w$, a subcalculus of $\lambda\sigma_\delta$, and known abstract machines such as the KAM, SECD, CAM, and Cardelli’s FAM [20]. Curien’s weak calculus of closures is, in turn, studied by Biernacka and Danvy [18] who introduce its modification $\lambda\hat{\rho}$ that, unlike $\lambda\rho$, makes it possible to define one-step reduction strategies underlying abstract machines such as the KAM, CEK, and Leroy’s ZINC [40]. It is shown that $\lambda\hat{\rho}$ and the abstract machines are in an intimate relation established by an
enriched refocusing. The calculus $\lambda \hat{\rho}$ is then used by García-Pérez et al. [34, 33] to reconstruct and prove Cregut’s KN correct in a very precise sense. In our case, the underlying small-step semantics is zipper semantics, rather than the more traditional structural operational semantics.

Accattoli et al. [4] study the relationship between the linear substitution calculus (LSC) [11] and a number of abstract machines for weak normalization, including the KAM, CEK, a simplified version of SECD, the lazy KAM [22], and Sestoft’s call-by-need machine [47] (ZINC is also sketched). It is argued that a weak version of LSC, a calculus with an explicit representation of reduction contexts and a notion of (explicit) substitution at a distance, can abstractly and effectively represent or distill, in the authors’ words, abstract machines for weak normalization. Interestingly, the authors also introduce new machines inspired by LSC: the Milner Abstract Machine (MAM) that uses only one, global environment and therefore avoids the concept of closure at a cost of explicit $\alpha$-renaming, and the Milner Abstract machine by-neeD (MAD) that takes advantage of the single global environment introduced in the MAM that facilitates sharing of computed values.

In [5], the above ideas are used to design the Strong Milner Abstract Machine, a machine for normal-order strong normalization that features a single environment and is shown to have a linear overhead both in the number of steps in a reduction sequence in LSC and in the size of the initial term – Strong MAM is distilled (decoded) into LSC and analysed for complexity. Such issues are central in Accattoli’s Useful MAM [2], that is the first reasonable, i.e., polynomially related to the cost model of RAM or Turing machines, machine for strong normalization in the $\lambda$-calculus in the literature. Useful MAM, however, is not studied in the context of LSC or another calculus of explicit substitutions. In the same line of work, Accattoli and Barras provide a thorough complexity analysis and comparison between machines with local and global environments [6], whereas Accattoli et al. [7, 10], present reasonable abstract machines for open call-by-value that rely on a global environment.

A special form of environments, called crumbled environments, is introduced in [9], where Accattoli et al. study call-by-value evaluation of both closed and open $\lambda$-terms. Such environments stem from the representation of terms considered by the authors, i.e., a representation with explicit sharing or explicit substitutions (in a new formulation), and they encode evaluation contexts besides storing delayed substitutions. The crumbling technique is then used by Accattoli et al. to define a reasonable abstract machine, called SCAM, for strong call-by-value [8]. Their work takes advantage of yet another calculus of explicit substitutions – the value substitution calculus (VSC) and of the concept of implosive sharing known from the literature on call-by-need [48]. Other reasonable abstract machines for strong call-by-value and strong call-by-need have been derived through the functional correspondence [12] from higher-order evaluators with heap-based sharing and local environments by Biernacka et al. [15, 16].

In contrast to the above collection of results by Accattoli et al. and Biernacka et al., in the present work we do not consider any sharing-oriented optimizations that go beyond the introduction of local environments, and we focus solely on the interaction of such environments with non-determinism. Whether such optimizations can be incorporated in NDAMs is a worthwhile question that requires further research.

6 Conclusion

We present an environment-based abstract machine for the $\lambda$-calculus without a fixed reduction strategy. To this end, we define a zipper semantics—inspired by a calculus of
explicit substitutions \[24\]—and derive an NDAM from it. We then optimize the NDAM so that it requires fewer steps to find a redex. In the resulting OAM, we can recognize existing deterministic machines by restricting and/or prioritizing steps.

Our work demonstrates that zipper semantics can be defined with environments instead of substitution. It also shows that the derived machine can be used as a stepping stone towards more optimized machines. Some optimizations are language specific, like knowing which annotations to keep after reduction, but others could be made generic in the NDAM derivation procedure. For example, it is always possible to bypass the backtracking steps of a deterministic mode of the machine, like we do with the bvar mode in the OAM. Refocusing is also possible as long as the stack \(\pi\) can be replaced by another element for backtracking, such as the context \(E\). In fact, the non-deterministic machines for the \(\lambda\)-calculus with substitution and for HOcore \[14\] could be automatically derived in the refocused form from their respective zipper semantics.

We believe that our approach of defining a zipper semantics, optimizing the derived machine, and possibly restricting it to make it deterministic is generic and can be used to define non-deterministic and deterministic abstract machines for other calculi. In particular, we would like to apply this technique to the \(\lambda\)-calculus with futures \[32, 41\], and see whether we can recover the different task creation strategies of this calculus, such as the eager strategy \[36\] or the lazy one \[29\].

---

**References**


A Properties of the Translation

We recall the definition of the translation.

\[ [I][e_1][e_2] ≅ [I][e_1][e_2] \] (1)
\[ [I][k \cdot e] ≅ [k] \] (6)
\[ [I_1][I_2][e] ≅ [I_1][I_2][e] \] (2)
\[ [n + 1][k \cdot e] ≅ [n][e] \] (7)
\[ [\lambda][e] ≅ \lambda[e] \] (3)
\[ [0][e] ≅ 0 \] (8)
\[ [n](\text{id}) ≅ n \] (4)
\[ [n][\uparrow] ≅ n + 1 \] (5)
\[ [n][e \circ e_2] ≅ [[n][e_1]][e_2] \] (10)

We extend the translation to frames, contexts, and local environments as follows.

\[ [\cdot][I] = [I] \] \[ [I][\cdot] = [I] \] \[ [\lambda] = \lambda \]
\[ [\square] = \square \] \[ [[\mathbb{S}} : \mathbb{E}] = [[\mathbb{S}} : \mathbb{E}] \]
\[ [\text{[I]}(\star) = [I] \] \[ [I](e) = [I](e) \]

Lemma 5. For all \( I, e, [I](e) \) is a term.

Proof. We define two sizes \( f_1, f_2 \) on \( I, e \), and \([I](e)\).

\[ f_1(I)(e) \triangleq f_1(I)f_1(e) \] \[ f_2(I)(e) \triangleq f_2(I)f_2(e) \]
\[ f_1(I)[e] \triangleq f_1(I)f_1(e) + 1 \] \[ f_2(I)[e] \triangleq f_2(I)f_2(e) + 1 \]
\[ f_1(l_1 l_2) \triangleq f_1(l_1)f_1(l_2) + 1 \] \[ f_2(l_1 l_2) \triangleq f_2(l_1)f_2(l_2) + 1 \]
\[ f_1(\lambda l) \triangleq f_1(l) + 2 \] \[ f_2(\lambda l) \triangleq f_2(l) + 2 \]
\[ f_1(n) \triangleq 2^{n+1} \] \[ f_2(n) \triangleq 1 \]
\[ f_1(\text{id}) \triangleq 2 \] \[ f_1(\uparrow) \triangleq 2 \]
\[ f_1(k \cdot e) \triangleq f_1(k) + f_1(e) \] \[ f_2(k \cdot e) \triangleq f_2(k) + f_2(e) \]
\[ f_1(\uparrow e) \triangleq f_1(e) \] \[ f_2(\uparrow e) \triangleq 4f_2(e) \]
\[ f_1(e_1 \circ e_2) \triangleq f_1(e_1)f_1(e_2) \] \[ f_2(e_1 \circ e_2) \triangleq f_2(e_1)(f_2(e_2) + 1) \]

The size \( f_1 \) is strictly decreasing on 1, 2, 3, 4, 6, 7, 8, and remains equal on the others. The size \( f_2 \) is strictly decreasing on the remaining ones, so the lexicographic ordering of \( f_1, f_2 \) is strictly decreasing. By induction on that size, we easily prove the result. \( \blacksquare \)

Lemma 6. For all \( I, j, h \), \([I](\uparrow^j)[\uparrow^h] = [I](\uparrow^{j+h})\).

Proof. We prove that for all \( i \), \([I](\uparrow^i)[\uparrow^j]) = [I](\uparrow^{i+j}) \) by induction on \( I \), starting with terms, and then generalizing it to all configurations. The only interesting case is the variable case \( n \). Suppose \( n \geq i \). We can prove by induction on \( i \) that \([n][\uparrow^i][\uparrow^j][\uparrow^h] = [n - i][\uparrow^j][\uparrow^h] \) by induction on \( I \), starting with terms, and then generalizing it to all configurations. The only interesting case is the variable case \( n \). Suppose \( n \geq i \). We can prove by induction on \( i \) that \([n][\uparrow^i][\uparrow^j][\uparrow^h] = [n - i + j][\uparrow^h] = [n + j][\uparrow^h] = n + j + h \) and \([n][\uparrow^i][\uparrow^j][\uparrow^h] = n + j + h \).
If \( n < i \), we can prove that \( [[[n]](\hat{t}^{i} \uparrow^{j})](\hat{t}^{j} \uparrow^{h}) = \prod [0](\hat{t}^{i-n} \uparrow^{j}) \uparrow^{h} \) on termination. The result then extends easily on all configurations by induction. In the case \( l[e] \), we conclude by applying the result on terms on \( [t](e) \).

\[ \blacksquare \]

\textbf{Lemma 7.} For all \( l \), \( [t](\hat{t} \uparrow \text{id}) = [t] \).

\textbf{Proof.} We prove that for all \( h \), \( [t](\hat{t}^{h+1} \text{id}) = [t](\hat{t}^{h} \text{id}) \), following the same principle as in the proof of Lemma 6.

\[ \blacksquare \]

\textbf{Lemma 8.} For all \( t \), \( [t] = t \).

\textbf{Proof.} By induction on \( t \), using Lemma 7 in the \( \lambda \)-abstraction case.

\[ \blacksquare \]

\textbf{Lemma 9.} For all \( n \), \( h \), \( m \), \( \epsilon \) such that \( n \geq m \) and \( h \geq m \), we have \( [n](\hat{t}^{h} \epsilon) = \prod [n-m](\hat{t}^{h-m} \epsilon) \uparrow^{m} \).

\textbf{Proof.} By induction on \( m \). For \( m = 0 \), we have to prove that \( [n](\hat{t}^{h} \epsilon) = \prod [n](\hat{t}^{h} \epsilon) \), which holds by Lemma 8.

Suppose \( n \geq m + 1 \), \( h \geq m + 1 \), and the hypothesis holds for \( m \): then \( [n](\hat{t}^{h} \epsilon) = \prod [n-m](\hat{t}^{h-m} \epsilon) \uparrow^{m} \). Because \( n-m \geq 1 \), we get

\[ \prod [n-m](\hat{t}^{h-m} \epsilon) \uparrow^{m} = \prod [n-m-1](\hat{t}^{h-m-2} \epsilon) \uparrow^{(m+1)} \]

The result holds for \( m+1 \), as wished.

We make a heavy use of Lemma 9 in the following proofs, and do not necessarily refer to it each time.

The proofs of many of the following lemmas are along the same pattern: by induction on terms, then extend it to all configurations, and only the variable case is interesting. We only detail the variable case for such lemmas.

\textbf{Lemma 10.} For all \( l \), \( i \), and \( h \), \( [t](\hat{t}^{h} \epsilon) = [t](\hat{t}^{i} \epsilon) \) in the variable case.

\textbf{Proof.} We prove that for all \( j \), \( [t](\hat{t}^{j} \epsilon) = [t](\hat{t}^{j} \epsilon) \) in the variable case, if \( n < j \), we get \( n \) on both sides.

If \( n \geq j \), then \( [n](\hat{t}^{j} \epsilon) = [n-j](\hat{t}^{j} \epsilon) \) in the variable case. We conclude with Lemma 10.

\[ \blacksquare \]

\textbf{Lemma 11.} For all \( l \), \( \epsilon \), and \( i \), \( [t](\hat{t}^{i} \epsilon) = [t](\hat{t}^{i} \epsilon) \).

\textbf{Proof.} We prove that for all \( h \), \( [n](\hat{h} \epsilon) = [n](\hat{h} \epsilon) \) in the variable case, if \( n < h \), we get \( n \) on both sides.

Suppose \( n \geq h \).

\[ [n](\hat{h} \epsilon) = [n-h](\hat{h} \epsilon) \] in the variable case, and we have \( [n](\hat{h} \epsilon) = [n-h](\hat{h} \epsilon) \).

We conclude with Lemma 10.

\[ \blacksquare \]

\textbf{Lemma 12.} For all \( l \), \( \epsilon \), and \( \phi \), \( [t](\epsilon \circ \phi) = [t](\epsilon \circ \phi) \).

\textbf{Proof.} We prove that for all \( i \), \( [t](\hat{i} \circ \phi) = [t](\hat{i} \circ \phi) \).

If \( n < i \), then we get \( n \) on both sides. If \( n \geq i \), then \( [n](\hat{i} \circ \phi) = [n-i](\epsilon \circ \phi) \).

We conclude with Lemma 11.

\[ \blacksquare \]
Lemma 13. For all $E$, $\llbracket E \rrbracket = \llbracket [E] \rrbracket$.

Proof. By induction on $E$. The case $E = k$ is easy.

We have $\llbracket (k :: E) \rrbracket = \llbracket E(k) \rrbracket$ by induction. The case $k :: E \llbracket [E] \rrbracket$. The case $k :: E$ is similar.

We have $\llbracket (\lambda :: E) \rrbracket = \llbracket [E] \rrbracket$ by Lemma 9.

Lemma 14. For all $E$, $\llbracket E \rrbracket = \llbracket [E] \rrbracket$.

Proof. We prove that for all $E$, $\llbracket E \rrbracket = \llbracket [E] \rrbracket$, starting with terms. We proceed by induction: the application case is easy.

Lemma 15. For all $l$ and $h$, $\llbracket l \rrbracket (\uparrow h) = \llbracket [l] \rrbracket (\uparrow h)$.

Proof. We prove that for all $l$ and $h$, $\llbracket l \rrbracket (\uparrow h) = \llbracket [l] \rrbracket (\uparrow h)$, starting with terms. The proof then extends easily to configurations by induction.

Lemma 16. For all $l$, $k$, and $\epsilon$, $\llbracket [l \cdot k \cdot \epsilon] \rrbracket = \llbracket [l \cdot \epsilon] \rrbracket (\llbracket k \rrbracket)$.

Proof. We prove that for all $l$, $\llbracket [l \cdot k \cdot \epsilon] \rrbracket = \llbracket [l \cdot \epsilon] \rrbracket (\llbracket k \rrbracket)$, starting with terms. The interesting case is for a variable $n$, where we distinguish three cases.

Suppose $n < h$. Then

$$\llbracket n \rrbracket (\uparrow h \cdot k \cdot \epsilon) = \llbracket 0 \rrbracket (\uparrow h - n \cdot k \cdot \epsilon) (\uparrow n)$$

by Lemma 9.

and

$$\llbracket n \rrbracket (\uparrow h \cdot \epsilon) (\llbracket k \rrbracket) = \llbracket 0 \rrbracket (\uparrow h - n \cdot \epsilon) (\llbracket k \rrbracket) (\uparrow n)$$

by Lemma 9.

Suppose $n = h$. Then

$$\llbracket h \rrbracket (\uparrow h \cdot k \cdot \epsilon) = \llbracket 0 \rrbracket (\uparrow h \cdot \epsilon) (\uparrow h)$$

by Lemma 9.

and

$$\llbracket h \rrbracket (\uparrow h \cdot \epsilon) (\llbracket k \rrbracket) = \llbracket 0 \rrbracket (\uparrow h \cdot \epsilon) (\llbracket k \rrbracket) (\uparrow h)$$

by Lemma 9.
We conclude with Lemma 14.

Suppose $n > h$. Then
\[
[n](\cdot^h (k \cdot \epsilon)) = [[n - h](k \cdot \epsilon)](\cdot^h) \quad \text{by Lemma 9}
\]
and
\[
[n](\cdot^{h+1} \epsilon)\{[k]/h\} = [[n - h - 1](\epsilon)](\cdot^{h+1})\{[k]/h\} \quad \text{by Lemma 9}
\]
We conclude with Lemma 15.

It concludes the proof on terms, and we extend the proof easily to all configurations by induction.

► Lemma 17. For all $l$, $\epsilon$, and $\mathcal{E}$, $[[l](\epsilon \mathbin{\bullet} \mathcal{E})] = [[[l](\epsilon)](\mathcal{E})]$. 

Proof. By case analysis on $\mathcal{E}$, using Lemmas 12 or 8 to conclude.

► Lemma 18. If $l \overset{E, \mathcal{E}}{\not\rightarrow}_{\mathit{var}} c$, then $[[l](\mathcal{E})] \overset{E}{\not\rightarrow}_{\mathit{ev}} [c]$. 
If $\epsilon \overset{E, n, l, \mathcal{E}}{\not\rightarrow}_{\mathit{var}} c$, then $[[n](\epsilon)](\mathcal{E}) \overset{E}{\not\rightarrow}_{\mathit{ev}} [c]$. 

Proof. Most cases are straightforward by induction. We detail some for which we need extra results.

\begin{itemize}
  \item env By induction, we get $[[l](\epsilon \mathbin{\bullet} \mathcal{E})] \overset{E}{\not\rightarrow}_{\mathit{ev}} [c]$, but we want $[[l](\epsilon)](\mathcal{E}) \overset{E}{\not\rightarrow}_{\mathit{ev}} [c]$. We get the result by Lemma 12 if $\mathcal{E} = \epsilon$, and by Lemma 8 if $\mathcal{E} = \mathcal{E}$. 
  \item lam In the case $\mathcal{E} = \star$, we need Lemma 8 to conclude.
  \item lam$\beta$ We want $[[\lambda.l](\epsilon)](\mathcal{E}(\langle l[k \cdot \epsilon] \rangle))$ (where we might have $\epsilon = \mathit{id}$, i.e., $\lambda.[l](\cdot^\epsilon) \overset{E}{\not\rightarrow}_{\mathit{ev}} \mathcal{E}(\langle l[k \cdot \epsilon] \rangle))$. We have $\lambda.[l](\cdot^\epsilon) \overset{E}{\not\rightarrow}_{\mathit{ev}} \mathcal{E} \langle l \cdot \epsilon \rangle \langle \mathcal{E} \rangle \langle [[k]/0] \rangle$. We conclude with Lemmas 13 and 16.
\end{itemize}

► Lemma 19. Let $t \overset{E}{\not\rightarrow}_{\mathit{ev}} s$. For all $l$, $\epsilon$, and $\mathcal{E}$ such that $[[l](\mathcal{E})] = t$, for all $\mathcal{E}'$ such that $[[\mathcal{E}']] = \mathcal{E}$, there exists $c$ such that $l \overset{E, n, \mathcal{E}}{\not\rightarrow}_{\mathit{ev}} c$ and $[c] = s$. 

For all $n$, $\epsilon$, $\mathcal{E}$ such that $[[n](\epsilon)](\mathcal{E}) = t$, for all $\mathcal{E}'$ such that $[[\mathcal{E}']] = \mathcal{E}$, there exists $c$ such that $\epsilon \overset{E, n, \mathcal{E}}{\not\rightarrow}_{\mathit{var}} c$ and $[c] = s$.

Proof. We define an ordering on couples $(l, \mathcal{E})$ and triples $(n, \epsilon, \mathcal{E})$ as follows.

- $(l, \mathcal{E}) > (k, \mathcal{F})$ if $f_1(l)(f_2(\mathcal{E}) + 1) > f_1(k)(f_2(\mathcal{F}) + 1)$;
- $(l, \mathcal{E}) > (n, \epsilon, \mathcal{F})$ if $f_1(l)(f_2(\mathcal{E}) + 1) > f_2(\epsilon)(f_2(\mathcal{F}) + 1)$;
- $(n, \epsilon, \mathcal{E}) > (l, \mathcal{F})$ if $f_2(\epsilon)(f_2(\mathcal{E}) + 1) > f_1(l)(f_2(\mathcal{F}) + 1)$;
- $(n, \epsilon, \mathcal{E}) > (n, \phi, \mathcal{F})$ if $f_2(\epsilon)(f_2(\mathcal{E}) + 1) > f_2(\phi)(f_2(\mathcal{F}) + 1)$.

where

\[
\begin{align*}
f_1(l[\epsilon]) & \overset{\Delta}{=} f_1(l_1)(f_2(\epsilon) + 1) & f_2(\star) & \overset{\Delta}{=} 1 \\
f_1(l_1 l_2) & \overset{\Delta}{=} f_1(l_1) + f_1(l_2) + 1 & f_2(l[\epsilon]) & \overset{\Delta}{=} f_2(\epsilon) \\
f_1(l \cdot l) & \overset{\Delta}{=} 4f_1(l) & f_2(l[\mathit{id}]) & \overset{\Delta}{=} 2 \\
f_1(\mathit{id}) & \overset{\Delta}{=} 3 & f_2(l[\uparrow]) & \overset{\Delta}{=} 2 \\
f_1(n) & \overset{\Delta}{=} 3 & f_2(l \cdot \epsilon) & \overset{\Delta}{=} f_1(l) + f_2(\epsilon) \\
f_2(\mathit{id}) & \overset{\Delta}{=} 2 & f_2(\uparrow\epsilon) & \overset{\Delta}{=} 3f_2(\epsilon) \\
f_2(l \circ c_2) & \overset{\Delta}{=} f_2(c_1)(f_2(c_2) + 1)
\end{align*}
\]
The sizes have been chosen to be strictly decreasing between the conclusion and the premise of a ZS rule.

We prove the two items simultaneously by induction on \(<\). Let \( t \xrightarrow{E} ev \ s \).

Let \( l, \mathcal{E} \) such that \([l](\mathcal{E}) = t\), and \( F \) such that \([F] = E\).

If \( l = n \), then \( \mathcal{E} = \lambda \epsilon \) for some \( \epsilon \), because \([l](\mathcal{E}) \) makes a transition and there is no transition for variable in the substitution world. Because \( f_1(n)(f_2(\epsilon) + 1) > f_2(\epsilon)(f_2(\epsilon) + 1) \), we can apply the induction hypothesis on \((n, \epsilon, *)\), and there exists \( c \) such that \( \epsilon \xrightarrow{F,n,\mathcal{E}_{var}} c \) and \([c] = s\). We then conclude by applying rule \( \text{cons} \).

If \( l = l' \epsilon \), then \( t = [[l'](\epsilon)](\mathcal{E}) = [[l'](\epsilon)](\mathcal{E}) = [[l'](\epsilon)(\cdot)\mathcal{E}) \). Since \( f_1(l'[\epsilon])(f_2(\epsilon) + 1) = f_1(l'[\epsilon])(f_2(\epsilon) + 1) > f_2(\epsilon)(f_2(\epsilon) + 1) + 1 \), we can apply the induction hypothesis on \((l', \epsilon)\mathcal{E})\); there exists \( c \) such that \( l' \xrightarrow{F,l',\mathcal{E}} c \) and \([c] = s\); we conclude using rule \( \text{env} \).

Suppose \( l = l_1 l_2 \); then \( t = [[l_1](\mathcal{E})][l_2](\mathcal{E}) \). There are two possible premises for \( t \xrightarrow{E} ev \ s \).

Suppose \( [[l_1](\mathcal{E})] \xrightarrow{[l_2](\mathcal{E})] : E} ev \ s \). Because \((l_1, l_2, \mathcal{E}) > (l_1, \mathcal{E})\) and \([([l_2](\mathcal{E})) : E] = [([l_2](\mathcal{E})) : E] \), we can apply the induction hypothesis: there exists \( c \) such that \( l_1 \xrightarrow{F,\mathcal{E}_{var}} c \) and \([c] = s\). Applying \( \text{appL} \), we get \( l_1 l_2 \xrightarrow{F,\mathcal{E}_{var}} c \), as wished. The other case is treated similarly.

If \( l = \lambda \cdot l' \); then \( [\lambda \cdot l'](\mathcal{E}) = \lambda \cdot [[l'](\mathcal{E})] = \lambda \cdot t \). We have two possibilities. The inductive case is treated like the application case. Otherwise, the transition \( t \xrightarrow{E} ev \ s \) comes from \( \lambda \cdot l' \xrightarrow{s' :: F} ev \mathcal{E}(l'(s'/0)) \) with \( t = \lambda \cdot t' \) and \( E = \circ s' :: E \). Because \([F] = E \), there exist \( k, F' \) such that \( F \circ k :: F', [k] = s', \) and \([F'] = E' \). By \( \lambda \text{m} \), we have \( \lambda \cdot l' \xrightarrow{k :: F' \mathcal{E}_{ev}} F'(l'((k :: \mathcal{E})) \). To conclude, we need \([E'(l'((k :: \mathcal{E})))](\mathcal{E})) = \mathcal{E}(l'(s'/0)) \). By Lemmas 13 and 16, \([E'(l'((k :: \mathcal{E})))](\mathcal{E})) = \mathcal{E} [E'(l'((k :: \mathcal{E})))](\mathcal{E}) = \mathcal{E}(l'(s'/0)) \), hence the result holds.

Let \( n, \epsilon, \mathcal{E} \) such that \([[n](\epsilon)](\mathcal{E}) = t \). We proceed by case analysis on \( \epsilon \).

If \( \epsilon = \text{id} \), then \( \mathcal{E} = \lambda \epsilon' \) for some \( \epsilon' \), because \( t \) cannot be a variable. Because \((n, \text{id}, \lambda \epsilon') > (n, \epsilon', *)\) and \([[n](\epsilon')](\mathcal{E})) = [[n](\epsilon')](\mathcal{E}) = [[n](\text{id})](\mathcal{E}) = t \), we can apply the induction hypothesis on \((n, \epsilon', *)\) to obtain \( c \) such that \( \epsilon \xrightarrow{F,n,\mathcal{E}_{var}} c \) and \([c] = s\). We apply rule \( \text{id} \) to conclude.

Suppose \( \epsilon = k \cdot \epsilon' \) and \( n = 0 \). Because \((0, \epsilon, \mathcal{E}) > (k, \mathcal{E})\) and \([[0](\epsilon)](\mathcal{E})) = [k](\mathcal{E}) = t \), we can apply the induction hypothesis on \((k, \mathcal{E})\) to obtain \( c \) such that \( k \xrightarrow{F,E_{ev}} c \) and \([c] = s\).

We apply rule \( \text{cons} \) to conclude.

The remaining cases are similar.

### B Correspondence Between the DAM and the OAM

The main differences between closures in the OAM and DAM machine states is that some environments \( l \cdot x \) in the DAM may be pushed to the leaves in the OAM, and that some variables \( n \) may be solved in the OAM but not in the DAM. These discrepancies can be erased by adapting the translation to terms \( [\cdot] \) of Figure 3, which pushes all the environments to the leaves and solve all the variables. Closures in OAM states may also have more annotations than in their DAM counterparts, because annotations are erased after \( \beta \)-reduction in the DAM, but some of these extra annotations may be meaningless.

We adapt the translation of Figure 3 to our current problem. We define two translations \( [l]_o(\mathcal{E}) \) on the OAM side \( [l]_d(\mathcal{E}) \) and the DAM side which go through \( l \) to push the envi-
environments to the leaves. On the OAM side, it also keeps the meaningful annotations, and discards the others. To do so, we use a translation \(\llbracket n \rrbracket_o(\epsilon)\) where \(b \in \{\top, \bot\}\) is a boolean remembering whether the environment \(\mathcal{E}\) is \(\star\) or not. We define \(\mathcal{S}\) as \(\mathcal{S} = \epsilon\) and \(b = \top,\) and as \(\emptyset\) otherwise.

\[
\begin{align*}
\llbracket n \rrbracket_o(\star) & \triangleq \llbracket n \rrbracket_o(\text{id}) \\
\llbracket n \rrbracket_o(\epsilon) & \triangleq \llbracket n \rrbracket_o^L(\epsilon) \\
\llbracket [\epsilon_1]^{\Sigma}_l(\epsilon_2) & \triangleq \llbracket [\epsilon_1]^L_l(\epsilon_1 \circ \epsilon_2) \text{ if } l \neq n \\
\llbracket \pi \rangle^L_n(\epsilon) & \triangleq (n, \epsilon \circ \epsilon_2, ?\Sigma) \\
\llbracket n \rrbracket_o(\epsilon) & \triangleq (n, \epsilon, \emptyset) \\
\llbracket [\epsilon_1]^{\Sigma}_l \llbracket \epsilon \rrbracket_o & \triangleq \llbracket [\epsilon_1]^L_l \llbracket \epsilon \rrbracket_o^L \\
\llbracket \lambda^{\Sigma} \llbracket \epsilon \rrbracket_o & \triangleq \lambda^{\Sigma} \llbracket \epsilon \rrbracket_o \\
\llbracket \pi \rangle^L_n(\epsilon) & \triangleq (n, \epsilon, \emptyset)
\end{align*}
\]

The two translations produce similar results, denoted \(r_o\) and \(r_d\), which mirrors the initial annotated term or closure \(l\), except the variables are turned into triples \((n, \epsilon, \Sigma)\) containing the variable itself, the environment for it, and its annotation. We recall that variables are not directly annotated in the OAM, but may be annotated when paired with an environment \(n(e)\).

We check that \(r_o\) contains more annotations than \(r_d\) with the predicate \(r_o \triangleright r_d\), defined below. We use two operations \(\llbracket (n, \epsilon, \Sigma) \rrbracket_o\) and \(\llbracket (n, \epsilon, \Sigma) \rrbracket_d\) which selects \(n\) and \(\epsilon\) in the usual steps, the result being a pair \((l, \mathcal{E})\) or a variable \(m\). The operation \(\llbracket (n, \epsilon, \Sigma) \rrbracket_d\) also checks that the annotations of \(\epsilon\) correspond to \(\Sigma: an(\epsilon) = \text{var} \iff \Sigma = \epsilon\).

\[
\begin{align*}
\begin{array}{c}
\frac{r_o \triangleright r_d}{r_o \triangleright r_d \triangleright r_d} \quad \Sigma \supseteq \Delta \\
r_o \triangleright r_d \triangleright r_d \quad \Sigma \supseteq \Delta \\
r_o \triangleright r_d \quad \Sigma \supseteq \Delta \\
r_o \triangleright r_d \quad \Sigma \supseteq \Delta \\
r_o \triangleright r_d \quad \Sigma \supseteq \Delta
\end{array}
\end{align*}
\]

We extend the translations \(\llbracket \cdot \rrbracket_o\) and \(\llbracket \cdot \rrbracket_d\) and relation \(\triangleright\) to contexts frame by frame.

We define an equivalence between machine states. We have \(\mathcal{O} \approx D\) if either:

\[
\begin{align*}
\mathcal{O} & = (\llbracket [\epsilon_1]^{\mathcal{E}}_l \llbracket \mathcal{E} \rrbracket_o, \llbracket [\epsilon_1]^{\mathcal{E}}_l \llbracket \mathcal{E} \rrbracket_d) & \llbracket [\epsilon_1]^{\mathcal{E}}_l \llbracket \mathcal{E} \rrbracket_d \triangleright [\epsilon_1]^{\mathcal{E}}_l \llbracket \mathcal{E} \rrbracket_d \\
\mathcal{O} & = (\llbracket \mathcal{E} \rrbracket_o, D = (\llbracket \epsilon \rrbracket_o, \llbracket \epsilon \rrbracket_d, \llbracket \epsilon \rrbracket_o \triangleright [\epsilon_1]^{\mathcal{E}}_l \llbracket \mathcal{E} \rrbracket_d)
\end{align*}
\]

We write \((\mathcal{O}_0, \mathcal{O}_1, \ldots, \mathcal{O}_t) \approx (\mathcal{D}_0, \mathcal{D}_1, \ldots, \mathcal{D}_t)\) if \(\mathcal{O}_0 = \mathcal{I}_t\) and \(\mathcal{D}_0 = \mathcal{I}_t\) for the same \(t\), \(\mathcal{O}_j \triangleright^* \mathcal{O}_{j+1}\) and \(\mathcal{D}_j \triangleright^* \mathcal{D}_{j+1}\) for all \(0 \leq j < t\), \(\mathcal{O}_j \approx \mathcal{D}_j\) for all \(0 \leq j \leq t\), and for all switching step \(c \rightarrow c'\) in the sequence \(\mathcal{O}_0 \triangleright^* \mathcal{O}_t\), there exists \(\mathcal{O}_j\) such that \(\mathcal{O}_j = c'\).

The last condition ensures that all switching steps (where annotations are added) of the OAM are matched by the DAM, adding an annotation at the same position (because
contexts between related states must have the same shape). However, there are switching steps from the DAM which are not matched by the OAM.

**Lemma 20.** Let $O \cong D$ such that $O = (l \mid E, \epsilon)_{ev}$.

- If $[\llangle E \rrangle] = l_1 \cdot l_2$, there exists $D'$ such that $D' = \langle k_1 \cdot k_2 \mid \forall, \forall', \pi' \rangle_{ev}, D \rightarrow^* D'$, and $O \cong D'$.
- A similar result holds if $[\llangle E \rrangle] = \lambda^0 l'$ or $[\llangle E \rrangle] = (n, \epsilon, \emptyset)$.

**Proof.** Let $D = \langle k \mid \forall, \forall', \pi \rangle_{ev}$. Assume $[\llangle E \rrangle] = l_1 \cdot l_2$. Because $[\llangle E \rrangle] \triangleright [k]_{ev}$, there exist $k_1$ and $k_2$ such that $\llangle k \rrangle = k_1 \cdot k_2$. We prove that the derivation exists by induction on the number of steps required to compute $\llangle k \rrangle = k_1 \cdot k_2$. If that number is 0, then $k$ is an application and the result holds by picking $D' = D$.

Otherwise, we distinguish cases based on $k$. If $k = k' \cdot \alpha$, then $[k]_{ev} = [k']_{ev}$. The step $D \rightarrow \langle k \mid \forall, \forall', \pi' \rangle_{ev}$ is possible for some $\pi''$, otherwise $k'$ would be annotated, and $k_1 \cdot k_2$ as well. We have $O \cong \langle k' \mid \forall, \forall', \pi'' \rangle_{ev}$. If $k = m^\Delta$, then $\langle (m, \forall, \Delta) \rangle_{ev} = (d, \forall'')$ and $\llangle d \rrangle_{ev} = k_1 \cdot k_2$. Therefore $d$ is not annotated, $\Delta$ must be empty, and the environment in $\forall''$ is not annotated: the sequence $D \rightarrow^* \langle d \mid \forall, \forall'', \pi'' \rangle_{ev}$ is possible for some $\pi''$. We have $O \cong \langle d \mid \forall, \forall'', \pi'' \rangle_{ev}$, and by induction, there exists $D'$ as wished such that $\langle d \mid \forall, \forall'', \pi'' \rangle_{ev} \rightarrow^* D'$. We then conclude $D \rightarrow^* D'$.

The abstraction and variable cases are similar. ▶

**Lemma 21.** Let $O \cong D$ such that $O = (E \mid c)_{bev}$.

- If $E = c \cdot l' :: E'$, there exists $D'$ such that $D' = \langle \text{appl} :: \pi' \mid l_1, (l_2[\forall']) :: \forall', \forall' \rangle_{bev}, D \rightarrow^* D'$, and $O \cong D'$.
- A similar result holds if $E = l' :: E'$, $E = \lambda :: E'$, or $E = \epsilon$.

**Proof.** We detail only the first item, the other cases are similar. Let $D = \langle \pi \mid l, \forall, \forall' \rangle_{bev}$. Because $[E]_{bev} \triangleright [\forall]_{bev} = c \cdot (l_2[\forall']) :: \forall'$ and $\pi$ is of the shape $\pi'' + \text{appl} :: \pi'$. When we apply the rule $\text{appl}$ in the DAM, the annotation set $\Delta$ must be empty for a valid configuration, so $\Delta = \emptyset$. We proceed by induction on the size of $\pi''$. If $\pi''$ is empty, then pick $D' = D$ and the result holds. Otherwise, we distinguish two cases.

1. Suppose $\pi = \text{env}(\epsilon) :: \pi''$. Then $D \rightarrow \langle [E]_{bev} \mid \forall, \forall', \pi'' \rangle_{ev} \rightarrow^* \llangle D \rrangle$ with $\forall = \epsilon \cdot \forall'$. Because $D$ is a backward configuration, we know $l$ is annotated, and $D_1$ can only do a switching step $D_1 \rightarrow \langle \pi'' \mid l_1[\forall'], \forall', \forall' \rangle_{bev} \rightarrow^* \llangle D_2 \rrangle$. We have $\llangle E \rrangle_{bev} = \llangle E \rrangle_{ev} = [l_1]_{ev} \cdot [\forall']_{ev}$, and hence $\llangle E \rrangle_{bev} = [l_1]_{ev} \cdot [\forall']_{ev}$ and $O \cong D_2$. We apply the induction hypothesis to $D_2$ to conclude.

2. Suppose $\pi = \text{cons}(\epsilon) :: \pi''$. Then $D \rightarrow \langle (l, \Delta') \mid \forall, \forall', \pi'' \rangle_{bev} \rightarrow^* \llangle D \rrangle$. Because $D$ is a backward configuration, we know $l$ is annotated, and $D_1$ can only do a switching step $D_1 \rightarrow \langle \pi'' \mid l_1[\forall'], \forall', \forall' \rangle_{bev} \rightarrow^* \llangle D_2 \rrangle$. In fact, there exists a sequence $D_1 \rightarrow^* \langle \pi'' \mid l_1[\forall'], \forall', \forall' \rangle_{bev}$ which undoes all the variable steps. Because $\llangle (n, \epsilon, \forall') \rrangle = (l, \forall)$ and $\llangle E \rrangle_{bev} = [l_1]_{ev} \cdot [\forall']_{ev}$ and therefore $O \cong D_2$. We apply the induction hypothesis to $D_2$ to conclude.

We write $\llangle r_1 \rrangle$ or $\llangle r_2 \rrangle$ for the forcing of $r_1$ or $r_2$, where the variable triples are solved. We extend $[\cdot]$ to annotated terms and closures so that it ignores the annotations.

**Lemma 22.** For all $l, E$, we have $\llangle [l]_E \rrangle_{ev} = [l](E)$ and $\llangle [l]_E \rrangle_{ev} = [l](E)$. ▶
Proof. The definitions of \([\cdot]_0\), \([\cdot]_d\), and \([\cdot]\) are almost the same if we ignore annotations. ▶

Lemma 23. If \([\|l\|_0(\mathcal{E})\] ▷ \([k]_d(\mathcal{F})\), then \([\|l\|_0(\mathcal{E})\] ▷ \([k](\mathcal{F})\).

Proof. We first prove the result on terms without the \(l[\cdot]_0\) construct, which is easy by induction on the derivation of ▷. We then show that \([\cdot]_0\) and \([\cdot]_d\) also preserve [\cdot]. ▶

Lemma 24. If \(r_o ▷ r_o^1\), \(\|r_o^1\| = \|r_o\|\), and the annotations of \(r_o^1\) and \(r_o\) are the same, then \(r_o ▷ r_o^2\).

If \(r_o ▷ r_o,\) \(\|r_o^1\| = \|r_o^2\|,\) and the annotations of \(r_o^1\) and \(r_o^2\) are the same, then \(r_o^2 ▷ r_o\).

Proof. By induction on the derivation of the ▷ judgements. ▶

Lemma 25. If \([\|l\|_0(\mathcal{E})\] ▷ \([k]_d(\mathcal{F})\), then \([\|l]_0(\mathcal{E})\] ▷ \([k][\mathcal{F}]_d(\mathcal{E})\)\).

If \(r_o ▷ \|k](\mathcal{F})\), then \(r_o ▷ \|k](\mathcal{F})\).

Proof. We have \([\|l\|_0(\mathcal{E})\] = \([l](\mathcal{E})\) = \([l]_0(\mathcal{E})\)\) and similarly for \([k]_d(\mathcal{F})\), so we can conclude with Lemma 24. ▶

Lemma 26. Let \([\|l\|_0(\mathcal{E})\] ▷ \([F]_d\), \(\|l'\|_0(\mathcal{F})\] ▷ \([k']_0(\mathcal{E})\), and \(\|l\|_0(\mathcal{E})\] ▷ \([k]_d(\mathcal{F})\) and \(\|l'\|_0(\mathcal{F})\) ▷ \([k']_d(\mathcal{E})\) such that all the annotation sets are empty on the DAM side. Then \(\mathcal{E}(\|l'\|_0(\mathcal{F})\] ▷ \([F]_d\)) ▷ \([k]_d(\mathcal{F})\).

Proof. We can prove that for all \(l, k, \mathcal{E}, \mathcal{F}, l', k',\) and \(n, \|l\|_0(\mathcal{E})\] ▷ \([k]_d(\mathcal{F})\) and \(\|l'\|_0(\mathcal{F})\) ▷ \([k']_d(\mathcal{E})\) imply \([l]_0(\mathcal{E})\) ▷ \([k]_d(\mathcal{F})\) \([k']_d(\mathcal{E})\)|\(n\). The proof is by induction on the derivation of \([\|l\|_0(\mathcal{E})\] ▷ \([k]_d(\mathcal{F})\). The inductive cases are straightforward, and in the variable case, we either change the variable the same way on both side, or we replace it with related configurations, so the result holds.

So we have \([\|l\|_0(\mathcal{E})[l]_0[\mathcal{E}]_0(\mathcal{F})]_0[\mathcal{F}]_0(\mathcal{E})_0)\] ▷ \([k]_d(\mathcal{F})[k]_d(\mathcal{E})_0(\mathcal{F})_0)\), but then \([\|l\|_0(\mathcal{E})[l']_0(\mathcal{F})]_0[\mathcal{F}]_0(\mathcal{E})_0)\] ▷ \([k]_d(\mathcal{F})[k']_d(\mathcal{E})_0(\mathcal{F})_0)\) by Lemma 16, and similarly on the DAM side. By Lemma 23, we have \([l']_d(\mathcal{F})_0[\mathcal{E}]_0[\mathcal{F}]_0)\). Because there are no annotation on the DAM side, we have \([l']_d(\mathcal{F})_0[\mathcal{F}]_0)\). By induction on the size of the contexts, we then prove that \(\mathcal{E}(\|l'\|_0(\mathcal{F})\] ▷ \([F]_d\)) ▷ \([k]_d(\mathcal{F})\), as wished. ▶

Theorem 27 (OAM implies DAM). If \((O_0, O_1, \ldots, O_t) \approx (D_0, D_1, \ldots, D_t)\) and \(O_i \rightarrow O'\), there exist \(O_{i+1}\) and \(D_{i+1}\) such that \(O' \rightarrow O_{i+1}, D_1 \rightarrow O_{i+1},\) and \(O_{i+1} \approx D_{i+1}\).

Proof. We proceed by case analysis on \(O_i \rightarrow O'\).

\(O_1\) We have \(O_i = (l_1 |\rightarrow l_2 \| E, E)_{ev} \rightarrow (l_1 |\rightarrow l_2, E)_{ev} = O'\) with \(E, \nmid l_1\). We pick \(O_{i+1} = O'\). Because \(O_i\) is valid, \(\Sigma\) cannot be a blocking annotation, and we have \([l_1 |\rightarrow l_2 \| E]_0(\mathcal{E}) = [l_1 \| E]_0(\mathcal{E}) \rightarrow [l_2 \| E]_d(\mathcal{E})\). By Lemma 20, there exists \(D'\) such that \(D' = (k' \| E, E', \cdot \pi)'_ev, D_i \rightarrow O',\) and \(O_i \approx D'\). Because \(O_i \approx D'\), we have in particular \([l_1 \| E]_0(\mathcal{E}) \rightarrow [k]_d(\mathcal{F})\) and \([l_2 \| E]_d(\mathcal{E}) \rightarrow [k]_d(\mathcal{F})\), and since \(E, \nmid l_1\), there is no blocking annotation in \(k_1\). Therefore we have \(D' \rightarrow (k_1 |\rightarrow (k_1 \| E)) : F, F', \pi''_ev = D_{i+1}\). From \([l_2 \| E]_0(\mathcal{E}) \rightarrow [k]_d(\mathcal{F})\), we deduce \([l_2 \| E]_d(\mathcal{E}) \rightarrow [k]_d(\mathcal{F})\) by Lemma 25, so we have \(O_{i+1} \approx D_{i+1}\).

\(O_2\) to \(O_3\) Similar to the case above.

\(O_4\) We have \(O_i = (l_1 \| E, E)_{ev} \rightarrow (l_1 \| E, E, \cdot \pi)'_ev = O'\). We pick \(O_{i+1} = O'\) and \(D_{i+1} = D_i\). Because \(O_i \approx D_i\), we have \([l_1 \| E, \cdot \pi]'_d(\mathcal{E}) \rightarrow [k]_d(\mathcal{F})\), i.e., \([l_1 \| E, \cdot \pi]'_d(\mathcal{E}) \rightarrow [k]_d(\mathcal{F})\), therefore \(O_{i+1} \approx D_{i+1}\) holds.
(O5) Because \([\pi|\alpha]\) by Lemma 20, there exists \(D'\) such that \(D' = (n^0|F'|\pi')_D\), \(D_1 \rightarrow^* D'\), and \(O_i \approx D'\). In particular, we have \([\pi|\alpha](F') = (n, \pi, \phi)\) for some \(\phi\) with \((n, \pi, \phi) \triangleright (n, \phi, \phi)\). From there, we distinguish three cases.

Suppose \(||(n, \pi, \phi)||_d = ||(n, \phi, \phi)||_d = m\). There is a one-to-one correspondence between the equations defining \(\pi|\alpha\) and \(\pi|\alpha\) and the var steps of the OAM and the DAM, except for the equations returning a variable, which can be replaced by the switching step of the var mode. On the DAM side, by definition of \(\pi|\alpha\), we are sure that the environment is not annotated because we start from a variable with no annotation. As a result, we can build sequences \(O_i \rightarrow^* (l|\Gamma, E)_{D}\) \(D_{i+1}\), and check that we have \(O_{i+1} \approx D_{i+1}\), as wished.

Suppose \(||(n, \pi, \phi)||_o = (l, E)\) with \(E, \text{infun}(\Gamma) \triangleright l\). Because \((n, \pi, \phi) \triangleright (n, \phi, \phi)\), there exists \((k, F'')\) such that \(||(n, \phi, \phi)||_o = (k, F'')\) and \([l]_o(\phi) \triangleright [k]_d(F'')\). In particular, \(K\) is not annotated. As in the previous subcase, we can build sequences \(O_i \rightarrow^* (l|\Gamma, E)_{D}\) \(D_{i+1}\) with \(O_{i+1} \approx D_{i+1}\), as wished.

Suppose \(||(n, \pi, \phi)||_o = (l, E)\) with \(E, \text{infun}(\Gamma) \triangleright l\), i.e., \(E = *\). Because \((n, \pi, \phi) \triangleright (n, \phi, \phi)\), there exists \((k, F'')\) such that \(||(n, \phi, \phi)||_o = (k, F'')\) and \([l]_o(\phi) \triangleright [k]_d(F'')\). If \(K\) is not annotated, we conclude as in the first subcase. Otherwise, there exists \(O_j\) such that the annotation is added to \(l\) at that point. By construction of the sequences of machine states, \(D_j = (\pi_j|k_j, F_j, \phi_j)_D\) also adds the annotation. Because the DAM is sound w.r.t. the ZS, \([k]_j(\phi_j)\) is a normal form. We have \([l]_o(\phi) \triangleright [k]_d(\phi_j)\), which implies \([l]_o(\phi) = [k]_d(\phi_j)\) by Lemma 23. We also have \([l]_o(\phi) = [k]_d(\phi_j)\), so \([k]_d(\phi_j)\) is a normal form. Because the DAM is complete w.r.t. the ZS, there exists a sequence \((k|\pi', F', \pi')_D \rightarrow^* \langle \pi'| k+i, \pi'| \rangle_{D_{i+1}}\), and then \(\langle \pi'| k+i, \pi'| \rangle_{D_{i+1}} \rightarrow^* \langle \pi'| n+i, \pi'| \rangle_{D_{i+1}}\). We also have \(O_i \rightarrow^* (\langle \pi'| n+i \rangle)_{D_{i+1}} \approx O_{i+1}\) with \(O_{i+1} \approx D_{i+1}\).

(O6) We have \(O_i = \langle \pi'| l_1 | \phi' | (\Gamma, E)_{\pi} \rangle \rightarrow (l|\Gamma|E)^0_{\pi} \approx O'\). Because \(O_i\) is valid, \(\Sigma\) cannot be a blocking annotation, and we have \([\lambda|\Sigma]^0_\lambda(E) = \lambda^0_\lambda([\lambda]_\lambda(E))\). By Lemma 20, there exists \(D'\) such that \(D' = \langle \lambda|\lambda, k| \phi', \pi'\rangle_{D_{i+1}} \rightarrow^* D'\), and \(O_i \approx D'\). Because \(O_i \approx D'\), we have in particular \([l]_o(\phi) \triangleright [k]_d(\phi_j)\), \(F' = \sigma k' \rightarrow F\), \(\pi' = \sigma k' \rightarrow \pi'\), and \([l]_o(\phi) \triangleright [k]_d(\phi_j)\), and \([E]\_o \triangleright [F]\_d\). Then \(D' \rightarrow (\langle F|k'|\phi'|\lambda\rangle_\lambda(\Gamma,F)^0) | [\lambda, \lambda, \lambda]_\lambda(\Gamma,F)^0 \approx D'\).

Since \([E]_o \triangleright [F]_d\), the contexts are of the same size \(a\). Let \(O' \rightarrow^* \langle E_a | [l_1 | \phi' \rangle | \Gamma_a, \pi' \rangle_{\pi'} \approx O'\). The contexts of \(b\) reconstruction steps taken by the OAM. Because \([E]_o \triangleright [F]_d\), the contexts have the same shape, and because the annotations have been erased on the DAM side, nothing prevents the DAM to match with \(D' \rightarrow^* \langle E_b | k_0 | \phi' \rangle | [\lambda, \lambda, \lambda]_\lambda(\Gamma,F)^0 \approx D_{i+1}\). Because \([E]_o \triangleright [F]_d\) frame by frame, we also have \([E_b]_o \triangleright [F_b]_d\) and \([E_a-b]_o \triangleright [F_a-b]_d\). We conclude that \(O_{i+1} \approx D_{i+1}\) with Lemma 26.

(O7) Similar to the first subcase of (O5).

(O8) The OAM does a switching step because all the subterms are annotated. We sketch the application case, the abstraction one is similar. We have \(\langle c_1 \hat{\alpha} c_2 | E\rangle_{\pi} \rightarrow (\langle c_1 \hat{\alpha} c_2 | E\rangle_{\pi} \approx O_{i+1}\). Because \(O_i\) is valid, \(\Sigma\) cannot be a blocking annotation, and we have \([c_1 \hat{\alpha} c_2]_{\pi} = \lambda^0([c_1]_{\lambda} \hat{\alpha} [c_2]_{\lambda})\). By Lemma 20, there exists \(D'\) such that \(D' = \langle k_0 \hat{\alpha} k_2 | \phi', \pi'\rangle_{D_{i+1}} \rightarrow^* D'\), and \(O_i \approx D'\). If \(k_1\) and \(k_2\) are annotated, then the DAM directly does the switching step. Otherwise, we prove as in the third subcase of (O5) that the DAM adds the missing annotations, and then does the switching step to \(D_{i+1}\). We can then prove that \(O_{i+1} \approx D_{i+1}\).

(O9) We have \(O_i = \langle c | \phi' | \pi' \rangle_{D_{i+1}} \rightarrow (\langle c | \phi' | \pi' \rangle_{D_{i+1}} \approx O_{i+1}\). We pick \(O_{i+1} \approx O'\). By Lemma 21, there exists \(D'\) such that \(D' = \langle \text{app} | \pi | l_1 | c_2 | l_2 | \phi' \rangle_{D_{i+1}} \rightarrow^* D'\), and \(O \approx D'\). Then \(D' \rightarrow (\langle l_1 | \phi' | \pi' \rangle_{D_{i+1}} \approx D_{i+1}\). Because \(O \approx D'\), we know \([c_2]_{\pi} \triangleright [l_1]_d(\phi)\) and
(O10)-(O12) Similar to the case above.

Lemma 28. Let $O \approx D$ such that $D = (k \mid F, F, \pi)_{ev}$.
- If $[k]_{d}(F) = k_1 \oplus k_2$, there exists $O'$ such that $O' = (l_1 \mathrel{\oplus} l_2 \mid F, F', \pi')_{ev}, O \rightarrow^* O'$, and $O' \approx D$.
- A similar result holds if $[k]_{d}(F) = \lambda \cdot l'$, or if $[l]_{o}(\varepsilon) = (n, \epsilon, \emptyset)$ and $[k]_{d}(F) = (n, \phi, \emptyset)$.

Proof. Let $O = (l \mid E, E')_{ev}$. Assume $[k]_{o}(F) = k_1 \oplus k_2$. Because $[l]_{o}(E) \triangleright [k]_{d}(F)$, there exist $l_1$ and $l_2$ such that $[l]_{o}(E) = l_1 \mathrel{\oplus} l_2$. We prove that the derivation exists by induction on the number of steps required to compute $[l]_{o}(E) = l_1 \mathrel{\oplus} l_2$. If that number is 0, then $l$ is an application and the result holds by picking $O' = O$.

Otherwise, the only other case is $l' = l'['\varepsilon']$; then $[l']_{o}(E) = [l']_{d}(\varepsilon)\triangleright [k]_{d}(F)$ holds. The step $O \rightarrow (l' \mid F, \varepsilon\varepsilon)_{ev}$ is always possible. We have $(l' \mid E, \varepsilon\varepsilon)_{ev} \approx D$, and $[l']_{d}(\varepsilon) = l_1 \mathrel{\oplus} l_2$ with one less step, so by induction, there exists $O'$ as wished such that $(l' \mid E, \varepsilon\varepsilon)_{ev} \rightarrow^* O'$, from which we conclude $O \rightarrow^* O'$.

The abstraction and variable cases are similar.

Theorem 29 (DAM implies OAM). If $(O_0, O_1 \ldots O_i) \approx (D_0, D_1 \ldots D_i)$ and $D_i \rightarrow D'$, there exist $O_{i+1}$ and $D_{i+1}$ such that $D' \rightarrow^* D_{i+1}, O_i \rightarrow^* O_{i+1}$, and $O_{i+1} \approx D_{i+1}$.

Proof. By case analysis on the step $D_i \rightarrow D'$.

(D1) We have $D_i = (k_1 \mathrel{\oplus} k_2 \mid F, F, \pi)_{ev} \rightarrow (k_1 \mid \varepsilon (k_2])_{ev} = D'$ with $ev \notin an(k_1)$.

By Lemma 28, there exists $O'$ such that $O' = (l_1 \mathrel{\oplus} l_2 \mid E, F', \pi)_{ev}, O_i \rightarrow^* O'$, and $O' \approx D_i$.

Because $O' \approx D_i$, we have in particular $[l_1]_{o}(E') \triangleright [k_1]_{d}(F)$ and $[l_2]_{o}(E') \triangleright [k_2]_{d}(F)$. We distinguish two cases.

If $E, \top \upharpoonright l_1$, we have $O' \rightarrow (k_1 \mid \varepsilon (k_2])_{ev} = O_{i+1}$, and we pick $D_{i+1} = D'$. From $[l_1]_{o}(E') \triangleright [k_2]_{d}(F)$, we deduce $[l_2]_{o}(E') \triangleright [k_2]_{d}(F)$ by Lemma 25, so we have $O_{i+1} \approx D_{i+1}$.

Otherwise, there is a blocking annotation on $l_1$. By the same reasoning as in the proof of Theorem 27 (subcase O5), we can prove that the DAM generates an annotation on $k_1$: there exists a sequence $D' \rightarrow^* (k' \mathrel{\oplus} k_2 \mid F, F, \pi)_{ev} = D_{i+1}$ such that $ev \in an(k'_1)$.

In fact, all the subterms of $k'_1$ must be annotated, and it must also be the case of the subterms of $l_1$. By picking $O_{i+1} = O'$, we have $O_{i+1} \approx D_{i+1}$ as wished.

(D2)-(D3) Similar to the case above.

(D4) We have $D_i = (k_1 \mid F, F, \pi)_{ev} \rightarrow (k_1 \mid \varepsilon (F \triangleright F'))_{ev} = D'$. We pick $D_{i+1} = D'$ and $O_{i+1} = O_i$.

Because $O_i \approx D_i$, we have $[l]_{o}(E) \triangleright [k]_{d}(F)$, i.e., $[l]_{o}(E) \triangleright [k]_{d}(\phi\cdot F)$, therefore $O_{i+1} \approx D_{i+1}$ holds.

(D5) Because $D_i$ is valid, we have $D_i = (n_0 \mid E, \top \emptyset, \pi)_{ev}$. If $[l_0]_{o}(E) \triangleright (n, \phi, \emptyset)$ with $[l_0]_{o}(E) \neq (n, \epsilon, \emptyset)$, then $[l_0]_{o}(E) \triangleright (k, F')$ and $[l_0]_{o}(E) \triangleright [k]_{d}(F')$. By definition of $\| \cdot \|_{d}$, the environment $\phi$ does not have blocking annotations, and there is a one-to-one corresponding between the definition of $\| \cdot \|_{d}$ and the var step, so there exists a sequence $D_i \rightarrow^* (k \mid F, F')_{ev} = D_{i+1}$. By choosing $O_{i+1} = O_i$, we have the required result.

If $[l_0]_{o}(E) = (n, \epsilon, \emptyset)$, then $l = n[\epsilon]_{ev}$ and $E = \varepsilon$: we have a valid forward machine state with a blocking annotation, a contradiction. If $[l_0]_{o}(E) = (n, \emptyset, \emptyset)$, by Lemma 28, there exists $O'$ such that $O' = (n \mid E, \top \emptyset, \pi)_{ev}, O_i \rightarrow^* O'$, and $O' \approx D_i$. From there, we distinguish three cases, and we conclude for each of them as in the proof of Theorem 27.
(D6) We have \( D_i = (\lambda^A, k \mid \sigma \to k') :: \mathbb{F}, \mathbb{F}, \pi)_{ev} \to (\mathbb{F}(k[k', \mathbb{F}], \sigma)_{ev} = D' \). By Lemma 28, there exists \( O' \) such that \( O' = (\lambda^D, k \mid \sigma', \mathbb{E})_{ev} \), which gives \( O_i \to O' \), and \( O' \approx D_i \). Because \( O' \approx D_i \), we have in particular \( [\mathbb{I}]_o(\sigma) \triangleright [k]_d(\sigma) \triangleright [\mathbb{F}]_d(\sigma) \), and \( [\mathbb{I}]_o \triangleright [\mathbb{F}]_d \); these relations also hold after erasing on the DAM side. Then we have \( O' \to (\mathbb{E} \mid \mathbb{I}[\mathbb{I}]_{ev} \to^* (\mathbb{E} \mid \mathbb{I}[\mathbb{I}]_{ev})_{ev} \triangleq O_{i+1} \), and we pick \( D_{i+1} = D' \). We conclude that \( O_{i+1} \approx D_{i+1} \) with Lemma 26.

(D7) In that case, the step \( D_i \to D' \) is a switching step: the DAM has to backtrack because of the annotations. We pick \( D_{i+1} = D' \). A possible case is that the DAM cannot do step (D5) because of a \( \sigma \) annotation on the environment. However, by definition of \( \| \cdot \|_d \), the environment is annotated only if the variable is also annotated, which is in contradiction with the fact that \( D_i \) is valid. Therefore, the DAM backtracks only because of \( \mathbb{E} \) annotations, which are also present on the OAM side because \( D_i \approx O_i \). The OAM also does a switching step \( O_i \to O_{i+1} \), and we have \( O_{i+1} \approx D_{i+1} \).

(D10) We have \( D = (\text{env}(\sigma) :: \pi \mid k, \mathbb{F}, \sigma \cdot \mathbb{F})_{bev} \to (\mathbb{F}(k[k, \mathbb{F}], \sigma)_{bev} = D' \). Because \( D \) is a backward configuration, we know \( k \) is annotated, and \( D' \) can only do a switching step \( D' \to (\pi \mid k[k, \mathbb{F}], \sigma \cdot \mathbb{F})_{bev} \triangleq D_{i+1} \). We pick \( O_{i+1} = O_i \). Because \( O_i \approx D_i \), we have \( [\sigma]_o(\sigma) \triangleright [k]_d(\sigma \cdot \mathbb{F}) \), i.e., \( [\sigma]_o(\sigma) \triangleright [k]_d(\sigma \cdot \mathbb{F}) \). Therefore we have \( O_{i+1} \approx D_{i+1} \).

(D8) We have \( D_i = (\text{appl} : \pi \mid \mathbb{I}_o \circ (\mathbb{I}_d[\mathbb{F}]) :: \mathbb{F}, \mathbb{F})_{bev} \to (\mathbb{I}_o \circ (\mathbb{I}_d[\mathbb{F}]) \mathbb{F}, \mathbb{F})_{ev} \triangleq D_{i+1} \). Because \( O_i \approx D_i \), we must have \( O_i = (\mathbb{I}_o :: \mathbb{F}, \mathbb{F})_{bev} \to (\mathbb{I}_o \circ (\mathbb{I}_d[\mathbb{F}]) \mathbb{F}, \mathbb{F})_{ev} \triangleq O_{i+1} \), and we know \( [\sigma]_o(\sigma) \triangleright [\mathbb{I}_o(\sigma)] \mathbb{I}_d(\sigma) \) and \( [\mathbb{I}_o(\sigma)] \mathbb{I}_d(\sigma) \triangleright [\mathbb{I}_d(\sigma)] \mathbb{I}_d(\sigma) \). The latter implies \( [\mathbb{I}_o(\sigma)] \mathbb{I}_d(\sigma) \triangleright [\mathbb{I}_d(\sigma)] \mathbb{I}_d(\sigma) \), and we have \( [\mathbb{I}_o(\sigma)] \mathbb{I}_d(\sigma) \triangleright [\mathbb{I}_d(\sigma)] \mathbb{I}_d(\sigma) \) for \( O_{i+1} \approx D_{i+1} \).

(D9)-(D13) Similar to the case above.