

The Essence of Eta-Expansion in Partial Evaluation

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Abstract. Selective eta-expansion is a powerful “binding-time improvement”, *i.e.*, a source-program modification that makes a partial evaluator yield better results. But like most binding-time improvements, the exact problem it solves and the reason why have not been formalized and are only understood by few.

In this paper, we describe the problem and the effect of eta-redexes in terms of monovariant binding-time propagation: eta-redexes preserve the static data flow of a source program by interfacing *static higher-order values in dynamic contexts* and *dynamic higher-order values in static contexts*. They contribute to two *distinct* binding-time improvements.

We present two extensions of Gomard’s monovariant binding-time analysis for the pure λ -calculus. Our extensions annotate *and* eta-expand λ -terms. The first one eta-expands static higher-order values in dynamic contexts. The second also eta-expands dynamic higher-order values in static contexts.

As a significant application, we show that our first binding-time analysis suffices to reformulate the traditional formulation of a CPS transformation into a modern one-pass CPS transformer. This binding-time improvement is known, but it is still left unexplained in contemporary literature, *e.g.*, about “cps-based” partial evaluation.

We also outline the counterpart of eta-expansion for partially static data structures.

Keywords: 2-level λ -calculus, binding-time analysis, coercions.

1. Introduction

Partial evaluation is a program-transformation technique for specializing programs [10], [15]. In the last decade it has been described using the notion of *binding times* [18]. Essentially the computations in a source program are divided into “static” or specialization-time computations (performed by the partial evaluator) and “dynamic” or run-time computations (to be performed in the specialized program). Partial evaluation amounts to performing the static computations and constructing the specialized program so that running it performs the dynamic computations. Thus a partial evaluator evaluates static expressions (*i.e.*, expressions that only depend on partial-evaluation time data) and reconstructs dynamic expressions (*i.e.*, expressions that depend on run-time data). For this to work, the binding-time

division must be congruent (also called consistent) [14], [21], [22], *i.e.*, no static computation may depend on the result of a dynamic computation.

In this setting, two sorts of expressible values coexist: static values and dynamic values (*i.e.*, residual expressions); correspondingly two sorts of contexts coexist: static contexts and dynamic contexts. Recall that a context is an expression with one hole [1]. A higher-order (resp. partially static) context is an expression with a higher-order (resp. partially static) hole. A static (resp. dynamic) context is an expression with a static (resp. dynamic) hole. A hole is static (resp. higher-order, partially static, and dynamic) whenever the expression fitting this hole is static (resp. higher-order, partially static, and dynamic).

To obtain consistency, Mix-style partial evaluators [15] coerce static values and contexts to be respectively dynamic values and dynamic contexts, when they encounter a clash. This is acceptable if source programs are first-order and values are either fully static or fully dynamic. However these coercions are excessive for higher-order programs with partially static values and contexts.

Lacking better interface between higher-order and dynamic, source programs must often be modified “to improve their binding times” and thus “to make them specialize better”. In Section 12.4 of their textbook [15], Jones, Gomard, and Sestoft list *eta-expansion* as an effective binding-time improvement but give only a brief idea of *why* it works.

In the following section, we use the term *dynamize*, with the meaning “make dynamic”, to characterize the effect of eta-expansion. We explain how eta-redexes *prevent* a binding-time analysis from

- dynamizing static values in dynamic contexts, and
- dynamizing static contexts around dynamic values

when the values are higher-order. Preventing static values and contexts from being dynamized improves the annotation in case the static values are used elsewhere or in case other static values may also occur in the same context. In Section 3 we present two binding-time analyses that insert eta-redexes automatically, and we illustrate them with two continuation-based program transformations. Section 4 outlines the counterpart of eta-expansion for partially static data structures. After a comparison with related work, we conclude.

2. The essence of eta-expansion

We show three examples, where

- a number occurs both in a static and in a dynamic context,
- a *higher-order* value occurs both in a static and in a dynamic context, and
- a function is applied to both a static and a dynamic *higher-order* argument.

After the examples, we summarize why eta-expansion improves binding times, given a monovariant binding-time analysis.² We use “@” (pronounced “apply”) to denote applications, and we abbreviate $(e_0@e_1)@e_2$ by $e_0@e_1@e_2$ and $e_0@(\lambda x.e)$ by $e_0@\lambda x.e$.

Reminder: eta-expanding a higher-order expression e yields the expression

$$\lambda v.e@v$$

where v does not occur free in e [1].

2.1. First-order static values in dynamic contexts

The following expression is partially evaluated in a context with y dynamic.

$$(\lambda x.(x + y) \times (x - 1))@42$$

Assume that this β -redex will be reduced. The addition depends on the dynamic operand y , so it should be reconstructed (in other words, x occurs in a dynamic context, $[\cdot] + y$). Both subtraction operands are static, so the subtraction can be performed (in other words, x occurs in a static context, $[\cdot] - 1$). The multiplication should be reconstructed since its first operand is dynamic. Overall, binding-time analysis yields the following two-level term.

$$(\overline{\lambda x}.\underline{(x+y)}\underline{\times}(x-1))\overline{@}42$$

(Consistently with Nielson and Nielson [21], overlined means static and underlined means dynamic.)

We can summarize some of the binding-time information by giving the binding-time types of variables, as in Lambda-Mix [12], [15]. Here, x has type s (static) and y has type d (dynamic). After specialization (*i.e.*, two-level reduction), the residual term reads as follows.

$$(42 + y) \times 41$$

Lambda-Mix’s binding-time analysis is able to give an appropriate annotation of the above program because the argument to $\lambda x.(x + y) \times (x - 1)$ is a *first-order* value. Inserting the static value in the dynamic context $([\cdot] + y)$ poses no problem. We now move on to the case where the inserted value is higher-order.

2.2. Higher-order static values in dynamic contexts

The following expression is partially evaluated in a context with g dynamic.

$$(\lambda f.f@g@f)@a.a$$

Again, assume that this β -redex is to be reduced. f occurs twice: once as the function part of an application (which here is a static context), and once as the argument of $f@g$ (which here is a dynamic context). The latter occurrence forces the binding-time analysis to classify f , and thus the rightmost λ -abstraction, to be dynamic (see Section 5 for a detailed motivation of this classification). Overall, binding-time analysis yields the following two-level term.

$$(\overline{\lambda}f.f@g@f)\overline{\lambda}a.a$$

Here, f has type d , and g has also type d . After specialization, the residual term reads as follows.

$$(\lambda a.a)@g@la.a$$

So unlike the first-order case, the fact that f , the static value, occurs in the dynamic context $f@g@[\cdot]$ “pollutes” its occurrence in the static context $[\cdot]@g@f$, so that neither is reduced statically.

NB: Since f is dynamic and occurs twice, a cautious binding-time analysis would reclassify the outer application to be dynamic: there is usually no point in duplicating residual code. In that case, the expression is totally dynamic and so is not simplified at all.

In this situation, a binding-time improvement is possible since $\lambda a.a$ will occur in a dynamic context. We can coerce this occurrence by eta-expanding the occurrence of f in the dynamic context (the eta-redex is boxed).

$$(\lambda f.f@g@(\boxed{\lambda y.f@y}))@la.a$$

Binding-time analysis now yields the following two-level term.

$$(\overline{\lambda}f.f@g@(\overline{\lambda}y.f@y))\overline{\lambda}a.a$$

Here, f has type $d \rightarrow d$, and both g and y have type d . Specialization yields the residual term

$$g@ly.y$$

which is more reduced statically.

In this case, the eta-redex effectively protects the static higher-order expression $\lambda a.a$ from being dynamized in the remainder of the computation. Instead, only the occurrence in the dynamic context is affected.

2.3. Higher-order dynamic values in static contexts

The following expression is partially evaluated in a context with d_0 and d_1 dynamic.

$$(\lambda f.f@d_0@(f@(\lambda x_1.x_1)@d_1))@la.a$$

f is applied twice: once to d_0 and something else, and once to $\lambda x_1.x_1$ and d_1 . In a monovariant higher-order binding-time analysis, d_0 dynamizes $\lambda x_1.x_1$, since the first parameter of f can only have one binding time. Overall, binding-time analysis yields the following two-level term.

$$(\bar{\lambda}f.f@d_0@(f@(\lambda x_1.x_1)@d_1))@\bar{\lambda}a.a$$

Here, f has type $d \rightarrow d$, x_1 has type d , and a has type d (corresponding to the type of d_0). Specialization yields the following residual term.

$$d_0@((\lambda x_1.x_1)@d_1)$$

The context $f@[\cdot]$ occurs twice in the source term. The dynamic value d_0 appears in the first occurrence, and the static value $\lambda x_1.x_1$ appears in the second occurrence. Since the context can only have one binding time (since it is the same f), d_0 pollutes $f@[\cdot]$, which in turn pollutes $\lambda x_1.x_1$. Since f is in fact $\lambda a.a$, the result of the application becomes dynamic. So the two potentially static applications of this result, respectively $[\cdot]@(f@(\lambda x_1.x_1)@d_1)$ and $[\cdot]@d_1$, become dynamic.

In this situation, a binding-time improvement is possible since both d_0 and $\lambda x_1.x_1$ occur (as results) in a potentially static context. We coerce d_0 by eta-expanding it (the eta-redex is boxed).

$$(\lambda f.f@(\lambda x_0.d_0@x_0)@(f@(\lambda x_1.x_1)@d_1))@\lambda a.a$$

Binding-time analysis now yields the following two-level term.

$$(\bar{\lambda}f.f@(\bar{\lambda}x_0.d_0@x_0)@(f@(\bar{\lambda}x_1.x_1)@d_1))@\bar{\lambda}a.a$$

Here, f has type $(d \rightarrow d) \rightarrow (d \rightarrow d)$, corresponding to statically applying f to both arguments. x_0 and x_1 both have type d , and a has type $d \rightarrow d$ (corresponding to the type of $\lambda x_1.x_1$). Specialization yields the residual term

$$d_0@d_1$$

which is more reduced statically.

In this case, the eta-redex effectively prevents the dynamic expression d_0 from being propagated to f and dynamizing $\lambda x_1.x_1$ in the remainder of the computation. Instead, only the occurrence in the static context is affected.

2.4. Summary

In a monovariant binding-time analysis, each time a higher-order static value occurs both in a potentially static context and in a dynamic context, the dynamic context dynamizes the higher-order value, which in turn dynamizes the potentially static context.

Conversely, each time a higher-order static value and a dynamic value occur in the same potentially static context, the dynamic value dynamizes the context, which in turn dynamizes the higher-order value.

Both problems can be circumvented by inserting eta-redexes in source programs. The eta-redex serves as “padding” around a value and inside a context, keeping one from dynamizing or from being dynamized by the other.

Eta-expanding a higher-order static expression f (when it occurs in a dynamic context) into

$$\underline{\lambda}v.f\bar{\@}v$$

creates a value that can be used for replacement. This prevents the original expression from being dynamized by a dynamic context. Instead, the new abstraction is dynamized.

Eta-expanding a higher-order dynamic expression g (when it occurs in a potentially static context) into

$$\bar{\lambda}v.g\@v$$

creates a value that can be used for replacement. This prevents a potentially static context from being dynamized by g . Instead, the new application is dynamized.

Informally, eta-expansion changes the *two-level type* [21] of a term as follows. Assume that f and g have type $t_1 \rightarrow t_2$, where t_1 and t_2 are ground types. The first eta-expansion coerces the type $t_1 \twoheadrightarrow t_2$ to be $t_1 \rightarrow t_2$. The second eta-expansion coerces the type $t_1 \rightarrow t_2$ to be $t_1 \twoheadrightarrow t_2$. Note that *inside* the redexes, the type of f is still $t_1 \twoheadrightarrow t_2$ and the type of g is still $t_1 \rightarrow t_2$.

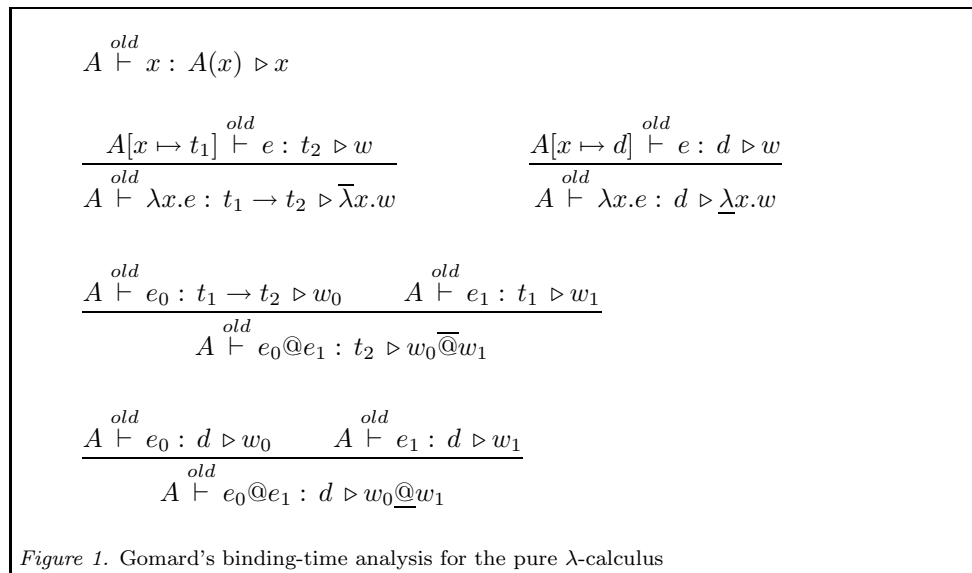
Further eta-expansion is necessary if t_1 or t_2 are not ground types. In fact, both kinds of eta-redex synergize. For example, if a higher-order static expression h has type $(t_1 \twoheadrightarrow t_2) \twoheadrightarrow t_3$ then its associated eta-redex reads as follows.

$$\underline{\lambda}v.h\bar{\@}\bar{\lambda}w.v\@w$$

In this example, the outer eta-expansion (of a static value in a dynamic context) creates the occurrence of a dynamic expression in a static context — hence the inner eta-redex.

To make our approach applicable to untyped languages, we will in the rest of the paper give dynamic entities the ground type d , as in Lambda-Mix [12], [15], rather than a two-level type such as $t_1 \rightarrow t_2$.

Information to guide the insertion of eta-redexes can *not* be obtained directly from the output of a binding-time analysis: at that point all conflicts have been resolved. Moreover, it would be naïve to insert, say, one eta-redex around every subterm: sometimes more than one is needed for good results, as in the last example and in the CPS-transformation example in Section 3.3.2. Alternatively, programs could be required to be simply typed. Then the type of each subterm determines the maximal number of eta-redexes that might be necessary for that subterm. However this type-driven eta-redex insertion yields many unnecessary eta-redexes.



In the following section we demonstrate how to insert a small and appropriate number of eta-redexes automatically. Our approach does not require programs to be typed and both for the example in Section 2.2 and for Plotkin's CPS transformation we show that it gives a good result.

3. Automatic insertion of eta-redexes

3.1. Binding-time analysis for the pure λ -calculus

Our starting point is the binding-time analysis in Figure 1. The analysis is that of Gomard [12], restricted to the pure λ -calculus. Types are finite and generated from the following grammar.

$$t ::= d \mid t_1 \rightarrow t_2$$

The type d denotes the type of dynamic entities. The judgement $A \vdash e : t \triangleright w$ means that under hypothesis A , the λ -term e can be assigned type t with annotated term w .

3.2. Eta-expansion of static values in dynamic contexts

Figure 2 presents the first of our new binding-time analyses. It both inserts eta-redexes and binding-time annotates λ -terms. The judgement $A \vdash e : t \triangleright w$ means

$$\begin{array}{l}
A \vdash x : A(x) \triangleright x \quad (1) \\
\\
\frac{A[x \mapsto t_1] \vdash e : t_2 \triangleright w}{A \vdash \lambda x.e : t_1 \rightarrow t_2 \triangleright \overline{\lambda x}.w} \quad (2) \\
\\
\frac{A[x \mapsto d] \vdash e : t_2 \triangleright w \quad t_2 \vdash z \Rightarrow m \quad \emptyset[z \mapsto t_2] \overset{old}{\vdash} m : d \triangleright w'}{A \vdash \lambda x.e : d \triangleright \underline{\lambda x}.w'[w/z]} \quad (3) \\
\\
\frac{A \vdash e_0 : t_1 \rightarrow t_2 \triangleright w_0 \quad A \vdash e_1 : t_1 \triangleright w_1}{A \vdash e_0 @ e_1 : t_2 \triangleright w_0 @ w_1} \quad (4) \\
\\
\frac{A \vdash e_0 : d \triangleright w_0 \quad A \vdash e_1 : t_1 \triangleright w_1 \quad t_1 \vdash z \Rightarrow m \quad \emptyset[z \mapsto t_1] \overset{old}{\vdash} m : d \triangleright w'_1}{A \vdash e_0 @ e_1 : d \triangleright w_0 @ (w'_1[w_1/z])} \quad (5)
\end{array}$$

z is always a fresh variable.

Figure 2. Binding-time analysis with eta-expansion of static values in dynamic contexts

$$\frac{d \vdash e \Rightarrow e \quad \frac{t_1 \vdash x \Rightarrow x' \quad t_2 \vdash e @ x' \Rightarrow e'}{t_1 \rightarrow t_2 \vdash e \Rightarrow \lambda x.e'}}{}$$

Figure 3. Full eta-redex expansion

that under hypothesis A , the λ -term e can be assigned type t with annotated term w , where eta-redexes may have been inserted into w .

We use the judgement $t \vdash e \Rightarrow m$ to insert eta-redexes (see Figure 3). Intuitively, eta-redexes are inserted when the analysis meets a dynamic abstraction with a static body (Rule 3), and a dynamic application with a static argument (Rule 5). Only when the value is higher-order does eta-expansion takes place — see the first rule of Figure 3. When the value is first-order, eta-expansion is of course not possible. In the case of static values occurring in static contexts, it is not necessary.

Note that our analysis generalizes Gomard's analysis: if in Rule 3 we always choose $t_2 = d$, and in Rule 5 we always choose $t_1 = d$, then we obtain Gomard's analysis.

If w is an annotated term, then \widehat{w} denotes the underlying λ -term. If $A \vdash \widehat{w} : t \triangleright w$ for some A and t , then w is said to be well-annotated. A well-annotated term has

a consistent binding-time division [22]. To prove that our analysis produces only well-annotated terms, we need the following lemma about Gomard's analysis.

LEMMA 1 *Suppose z is the only free variable of e_2 .*

*If $A \vdash^{old} e_1 : t \triangleright w_1$ and $\emptyset[z \mapsto t] \vdash^{old} e_2 : t' \triangleright w_2$,
then $A \vdash^{old} e_2[e_1/z] : t' \triangleright w_2[w_1/z]$.*

Proof: Consider the following more general property.

*If $A \vdash^{old} e_1 : t \triangleright w_1$ and $A' \vdash^{old} e_2 : t' \triangleright w_2$,
then $A'' \vdash^{old} e_2[e_1/z] : t' \triangleright w_2[w_1/z]$,*

where A can be obtained from A'' by removing the binding for the free variables of e_2 except the one for z , and where A' can be obtained from A'' removing the bindings in A and adding the binding $z \mapsto t$. From this property the lemma immediately follows. The general property is proved by induction on the structure of the proof of $A' \vdash^{old} e_2 : t' \triangleright w_2$. ■

We can then prove correctness: our analysis produces only well-annotated terms.

THEOREM 1 *If $A \vdash e : t \triangleright w$, then $A \vdash^{old} \widehat{w} : t \triangleright w$.*

Proof: By induction on the structure of the proof of $A \vdash e : t \triangleright w$, using Lemma 1 for Rules 3 and 5. ■

As a corollary we get that even though eta-expansion allows more static reductions to take place, specialization will terminate since types are finite. In another setting, eta-expansion and generalization are used both to improve binding times and to ensure termination [6], [19], [20].

Future work includes finding an efficient implementation of our binding-time analysis.

3.3. Examples

3.3.1. Higher-order static values in dynamic contexts

We now demonstrate that the new binding-time analysis inserts the expected eta-redex in the example program $(\lambda f.f@g@f)@la.a$ from Section 2.2. We derive

$$\emptyset[g \mapsto d] \vdash (\lambda f.f@g@f)@la.a : d \triangleright (\overline{\lambda}f.f\overline{@}g\overline{@}\lambda y.f\overline{@}y)\overline{@}\lambda a.a.$$

Consider the following fragment of the derivation, using Rules 2 and 4.

$$\frac{\frac{A \vdash f@g@f : d \triangleright f@g@ly.f@y}{\emptyset[g \mapsto d] \vdash \lambda f.f@g@f : t \rightarrow d \triangleright \bar{\lambda}f.f@g@ly.f@y} \quad \frac{\emptyset[g \mapsto d][x \mapsto d] \vdash x : d \triangleright x}{\emptyset[g \mapsto d] \vdash \lambda a.a : t \triangleright \bar{\lambda}a.a}}{\emptyset[g \mapsto d] \vdash (\lambda f.f@g@f)@la.a : d \triangleright (\bar{\lambda}f.f@g@ly.f@y)@la.a}$$

where t abbreviates $d \rightarrow d$ and A abbreviates $\emptyset[g \mapsto d][f \mapsto t]$. We need to derive $A \vdash f@g@f : d \triangleright f@g@ly.f@y$. Here follows the last step of the derivation, using Rule 5.

$$\frac{A \vdash f@g : d \triangleright f@g \quad A \vdash f : t \triangleright f \quad t \vdash z \Rightarrow \lambda y.z@y \quad A' \overset{old}{\vdash} \lambda y.z@y : d \triangleright \underline{\lambda}y.z@y}{A \vdash (f@g)@f : d \triangleright f@g@ly.f@y}$$

where A' abbreviates $\emptyset[z \mapsto t]$. The last of the four assumptions is derived as follows.

$$\frac{\frac{A'[y \mapsto d] \overset{old}{\vdash} z : t \triangleright z \quad A'[y \mapsto d] \overset{old}{\vdash} y : d \triangleright y}{A'[y \mapsto d] \overset{old}{\vdash} z@y : d \triangleright z@y}}{A' \overset{old}{\vdash} \lambda y.z@y : d \triangleright \underline{\lambda}y.z@y}$$

Thus, our analysis inserts exactly the same eta-redex that we inserted by hand in Section 2.2.

3.3.2. The CPS transformation

Let us now turn to the transformation of λ -terms into continuation-passing style (CPS). This example is significant because historically, the virtue of eta-redexes became apparent in connection with partial evaluation of CPS interpreters and with CPS transformers [2], [11]. It also has practical interest since the pattern of construction and use of higher-order values in the CPS transform is prototypical. Figure 4 displays Plotkin's original CPS transformation for the call-by-value lambda-calculus [23], written as a two-level term.

$$\begin{aligned} [\cdot] &: \text{syntax} \Rightarrow \text{CPSsyntax} \Rightarrow \text{CPSsyntax} \\ [x] &= \underline{\lambda}k.k@x \\ [\lambda x.e] &= \underline{\lambda}k.k@(\underline{\lambda}x.[e]) \\ [e_0@e_1] &= \underline{\lambda}k.[e_0]@(\underline{\lambda}v_0.[e_1]@(\underline{\lambda}v_1.v_0@v_1@k)) \\ [e] &\text{ is the CPS counterpart of the expression } e. \end{aligned}$$

Figure 4. Two-level formulation of Plotkin's CPS transformation

Since the transformation is a syntax constructor, *all* occurrences of @ and λ are dynamic. And in fact, Gomard's binding-time analysis does classify all occurrences to be dynamic.

$$\begin{aligned}
[\cdot] & : \text{syntax} \Rightarrow (\text{CPSyntax} \Rightarrow \text{CPSyntax}) \Rightarrow \text{CPSyntax} \\
[x] & = \overline{\lambda}k.k\overline{@}x \\
[\lambda x.e] & = \overline{\lambda}k.k\overline{@}\lambda x.\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v \\
[e_0\overline{@}e_1] & = \overline{\lambda}k.[e_0]\overline{@}\lambda v_0.[e_1]\overline{@}\lambda v_1.v_0\overline{@}v_1\overline{@}\lambda v_2.k\overline{@}v_2
\end{aligned}$$

$\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v$ is the CPS counterpart of the expression e .

Figure 5. Two-level formulation of Plotkin's CPS transformation after eta-expansion

But CPS terms resulting from this transformation contain redundant “administrative” beta-redexes, which have to be post-reduced [26]. These beta-redexes can be avoided by inserting eta-redexes in the CPS transformation, allowing some beta-redexes in the transformation to become static.³

Figure 5 shows the revised transformation containing three extra eta-redexes: one for the CPS transformation of applications, and two for the CPS transformation of abstractions.

As analyzed elsewhere [11], the eta-redex $\lambda k.[e]\overline{@}k$ prevents the outer $\lambda k\dots$ from being dynamized. The two other eta-redexes $\lambda v.k\overline{@}v$ and $\lambda v_2.k\overline{@}v_2$ enable k to be kept static. The types of the transformations (shown in the figures) summarize the binding-time improvement.

Our new analysis inserts exactly these three eta-redexes, given Plotkin's original specification. We now show the derivation for the case of abstraction.

Consider the following fragment of the derivation, using Rules 2 and 4 and abbreviating $\emptyset[[e] \mapsto (d \rightarrow d) \rightarrow d] \vdash [k \mapsto d \rightarrow d]$ by A .

$$\frac{A \vdash k : d \rightarrow d \triangleright k \quad A \vdash \lambda x.[e] : d \triangleright \underline{\lambda}x.\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v}{A \vdash k\overline{@}\lambda x.[e] : d \triangleright k\overline{@}\lambda x.\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v}$$

$$\frac{}{\emptyset[[e] \mapsto (d \rightarrow d) \rightarrow d] \vdash \lambda k.k\overline{@}\lambda x.[e] : (d \rightarrow d) \rightarrow d \triangleright \lambda k.k\overline{@}\lambda x.\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v}$$

We need to derive $A \vdash \lambda x.[e] : d \triangleright \underline{\lambda}x.\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v$. Let us abbreviate $(d \rightarrow d) \rightarrow d$ by t and $\lambda k.z\overline{@}(\lambda v.k\overline{@}v)$ by E . Here follows the last step of the derivation, using Rule 3.

$$\frac{A[x \mapsto d] \vdash [e] : t \triangleright [e] \quad t \vdash z \Rightarrow E \quad \emptyset[z \mapsto t] \vdash \overset{old}{E} : d \triangleright \underline{\lambda}k.z\overline{@}\lambda v.k\overline{@}v}{A \vdash \lambda x.[e] : d \triangleright \underline{\lambda}x.\underline{\lambda}k.[e]\overline{@}\lambda v.k\overline{@}v}$$

The other two cases are left to the reader.

Thus the two-level λ -calculus proves particularly useful for specifying CPS transformations — something that was done so far by sheer insight [27] or by hand [11]. This mode of specification has direct applications to continuation-based program transformation [5], [16], [17].

3.3.3. Improved “cps-based” cogen

Bondorf and Dussart’s new work [5] relies on two key eta-expansions that are analogous to those of Section 3.3.2. These eta-expansions come for free with the binding-time analysis of Figure 2.

3.4. Eta-expansion of dynamic values in static contexts

Figure 6 presents the second of our new binding-time analyses. It is an extension of the binding-time analysis in Figure 2. Again, the judgement $A \vdash e : t \triangleright w$ means that under hypothesis A , the λ -term e can be assigned type t with annotated term w , where eta-redexes may have been inserted into w .

Intuitively, eta-redexes are inserted when the analysis meets a static abstraction with a dynamic body (Rule 8), a dynamic abstraction with a static body (Rule 9), a static application with a dynamic argument (Rule 11), and a dynamic application with a static argument (Rule 12).

Again, the analysis generalizes Gomard’s analysis: if we never use Rule 8 and Rule 11, and in Rule 9 we always choose $t_2 = d$, and in Rule 12 we always choose $t_1 = d$, then we obtain Gomard’s analysis.

Theorem 1 holds also for this analysis, so the analysis produces only well-annotated terms.

Again, future work includes finding an efficient implementation of the binding-time analysis.

4. Partially static data structures

For data structures, we present a technique similar to eta-redexes that maintains the static data flow of source programs. As a prototypical example, we consider pairing. The ideas extend to other data structures in a straightforward manner.

Essentially, we delta-expand a pair-type expression p into the following expression.⁴

$$\text{Pair}(\text{Fst } p, \text{Snd } p)$$

4.1. Partially static values in dynamic contexts

The following expression is partially evaluated in a context with g and d dynamic.

$$(\lambda p. g @ (10 + \text{Fst } p) @ p) @ \text{Pair}(1, d)$$

Again, let us assume that this β -redex is to be reduced. p occurs twice: once as the argument of the first projection Fst (a static context), and once as the argument part of an application (a dynamic context). The latter occurrence forces the

$$A \vdash x : A(x) \triangleright x \quad (6)$$

$$\frac{A[x \mapsto t_1] \vdash e : t_2 \triangleright w}{A \vdash \lambda x.e : t_1 \rightarrow t_2 \triangleright \overline{\lambda x}.w} \quad (7)$$

$$\frac{A[x \mapsto t_1] \vdash e : d \triangleright w \quad t_2 \vdash z \Rightarrow m \quad \emptyset[z \mapsto d] \vdash^{old} m : t_2 \triangleright w'}{A \vdash \lambda x.e : t_1 \rightarrow t_2 \triangleright \overline{\lambda x}.w'[w/z]} \quad (8)$$

$$\frac{A[x \mapsto d] \vdash e : t_2 \triangleright w \quad t_2 \vdash z \Rightarrow m \quad \emptyset[z \mapsto t_2] \vdash^{old} m : d \triangleright w'}{A \vdash \lambda x.e : d \triangleright \underline{\lambda x}.w'[w/z]} \quad (9)$$

$$\frac{A \vdash e_0 : t_1 \rightarrow t_2 \triangleright w_0 \quad A \vdash e_1 : t_1 \triangleright w_1}{A \vdash e_0 @ e_1 : t_2 \triangleright w_0 @ w_1} \quad (10)$$

$$\frac{A \vdash e_0 : t_1 \rightarrow t_2 \triangleright w_0 \quad A \vdash e_1 : d \triangleright w_1 \quad t_1 \vdash z \Rightarrow m \quad A' \vdash^{old} m : t_1 \triangleright w'_1}{A \vdash e_0 @ e_1 : t_2 \triangleright w_0 @ (w'_1[w_1/z])} \quad (11)$$

where $A' = \emptyset[z \mapsto d]$.

$$\frac{A \vdash e_0 : d \triangleright w_0 \quad A \vdash e_1 : t_1 \triangleright w_1 \quad t_1 \vdash z \Rightarrow m \quad \emptyset[z \mapsto t_1] \vdash^{old} m : d \triangleright w'_1}{A \vdash e_0 @ e_1 : d \triangleright w_0 @ (w'_1[w_1/z])} \quad (12)$$

z is always a fresh variable.

Figure 6. Binding-time analysis with both eta-expansion of static values in dynamic contexts and eta-expansion of dynamic values in static contexts

binding-time analysis to classify p , and thus the occurrence of Fst , to be dynamic. Overall, binding-time analysis yields the following two-level term.

$$(\overline{\lambda p}.g @ (10 + \underline{Fst} p) @ p) @ \overline{Pair}(1, d)$$

After specialization, the residual term reads as follows.

$$g @ (10 + Fst (Pair(1, d))) @ Pair(1, d)$$

The fact that p , the partially static value, occurs in a dynamic context “pollutes” its occurrence in the partially static context, so that neither is reduced statically.

NB: Since p occurs twice, a cautious binding-time analysis would reclassify the outer application to be dynamic: there is usually no point in duplicating residual code. In that case, the expression is totally dynamic and so is not simplified at all.

In this situation, a binding-time improvement is possible since $Pair(1, d)$ will occur in a dynamic context. We can coerce this occurrence by inserting a delta-redex in the dynamic context (the redex is boxed).

$$(\lambda p.g@(10 + Fst p)@\boxed{Pair(Fst p, Snd p)})@Pair(1, d)$$

Binding-time analysis now yields the following two-level term.

$$(\overline{\lambda}p.g@(\overline{10 + Fst p})@\overline{Pair(Fst p, Snd p)})@\overline{Pair}(1, d)$$

Specialization yields the residual term

$$g@11@Pair(1, d)$$

which is more reduced statically.

In this case, the delta-redex effectively prevents the partially static expression from being dynamized in the remainder of the computation. Instead, only the occurrence in the dynamic context is affected.

4.2. Dynamic values in partially static contexts

It is simple to construct an example analogous to Section 2.3. Just have a partially static value and a dynamic value coexist in a context: the dynamic value dynamizes the context, which in turn, dynamizes the partially static value. Delta-expanding the corresponding dynamic expression in the source program prevents this approximation for the same (monovariant) binding-time analysis.

5. Related work

Our work is concerned with binding-time improvements and thus *off-line* partial evaluation of procedural programs. Eta-expansion is specific to the λ -calculus. We are not aware of any counterpart in partial evaluation of logic programs.

5.1. Mix-style partial evaluation

Mix-style partial evaluators developed at DIKU, such as Similix and Lambda-Mix [2], [12], [15], process procedural programs. When the first-order version of Similix [4] was extended to process higher-order programs [2], it was observed that naïve syntax reconstruction led to the occurrence of Scheme closures in residual programs. Two solutions were possible:

- lifting closures into syntax to construct the residual program (at specialization time); and
- dynamizing closures occurring in dynamic contexts (at binding-time analysis time).

The latter solution — coercing static values and contexts to be respectively dynamic values and dynamic contexts, in case of binding-time clash — was chosen in Similix-2, to avoid potential code duplication.⁵ Thus one is forced to state this code duplication explicitly by garnishing one’s source programs with eta-redexes (see Section 10.1.4, Item (2) and Section 12.4 of Jones, Gomard, and Sestoft’s textbook [15] for two separate explanations). This solution has been consistently maintained in the later versions of Similix [3], [5], and adopted in Lambda-Mix [12], [15].

It seems that this decision, together with the forward nature of binding-time analysis [9], have created the need for binding-time improvements:

- Eta-expansion prevents the dynamization of higher-order values and contexts. Delta-expansion prevents the dynamization of partially static values and contexts. Thus they both improve the binding times of source programs.
- Tail-recursive style in general (typically CPS) prevents the dynamization of intermediate results [9]. Thus it improves the static data flow of source programs.

5.2. Schism

Schism [8] does not dynamize static values whenever they occur in dynamic contexts. Instead, it inserts a “freeze” annotation coercing their result to be dynamic. For example, the term of Section 2.2 would be annotated as follows

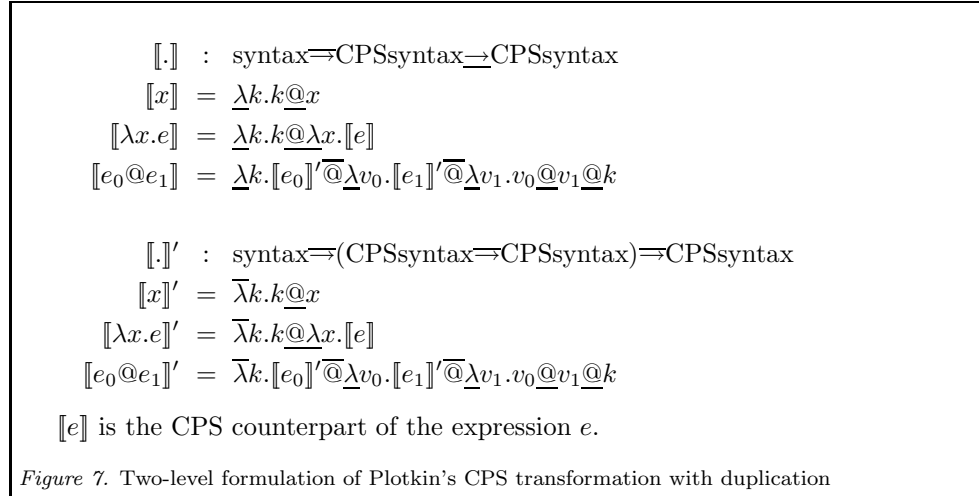
$$(\overline{\lambda}f.f\overline{@}g\overline{@}(\text{freeze } f))\overline{@}\lambda a.a$$

and its specialization would yield the same good result as in Section 2.2. The freeze operator acts like eta-expansion, and enables Schism to deal with higher-order and partially static values in dynamic contexts without dynamizing them. Independently, Schism’s polyvariant binding-time analysis [7] deals with dynamic values in static contexts, though currently, higher-order parameters are treated in a monovariant way.

Thus Schism’s extra power makes it possible to interface partially static and higher-order values and contexts smoothly, without loss of static information. Eta-expansion enables such a smooth interface for Mix-style partial evaluators.

5.3. Polyvariance and duplication

As explained in Footnote 1, a monovariant binding-time analysis maintains one binding-time description for each source expression whereas a polyvariant analysis



may maintain several binding-time descriptions for each source expression. Given a monovariant analysis, duplicating source expressions simulates some polyvariance. The effect is the same if the target language of the analyses is the two-level λ -calculus, since the analyzed program contains duplicated parts of the source program, with different annotations.⁶

Figure 7 illustrates the effect of polyvariance and duplication. It can be obtained either by the monovariant analysis of a source program where the definition of $\llbracket \cdot \rrbracket$ has been hand-duplicated, or by a polyvariant binding-time analysis (that duplicates the definition of $\llbracket \cdot \rrbracket$). As testified by their types, the first variant approximates the second.

In this example, polyvariance does not give the same effect as eta-expansion. This of course suggests to mix both — a future work.

5.4. Online partial evaluation

An online partial evaluator such as FUSE [28] is inherently polyvariant over binding times [24] and thus meets no problem when dynamic values reach static contexts. The converse situation can be handled in specializers that carry two representations of each closure. Such systems include FUSE and Schism.

6. Conclusion

Inserting eta-redexes in source programs has until now been listed as black magic in the literature on Mix-style partial evaluation [15]. We have described the effect of eta-redexes in terms of binding-time coercions: eta-redexes offer a syntactic

representation of coercions and thus they prevent the binding-time analysis from approximating higher-order values and higher-order contexts to be dynamic. This effect is of prime importance for contemporary monovariant binding-time analyses since it enables one to carry out specialization as two-level β -reduction. It is also useful for contemporary polyvariant binding-time analyses, since eta-redexes can reduce the multiplication of variants. We also demonstrated how to integrate eta-expansion in an offline partial evaluator, by extending an existing binding-time analysis. Finally, we have outlined the counterpart of eta-expansion for partially static data structures.

Future work naturally includes developing a partial evaluator with better coercions, to eliminate the need of binding-time improvements by eta-expansion.

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Notes

1. Also in Proc. PEPM'94, pages 11-20.
2. A binding-time analysis is “monovariant” if it associates *one* binding-time description to any source expression. (It is “polyvariant” if it may associate several binding-time descriptions to any expression.) For consistency [14], [21], [22], in case of clash, a monovariant binding-time analysis approximates the clashing descriptions with an encompassing dynamic description, as illustrated in the following table.

binding-time description x	binding-time description y	least encompassing dynamic description of x and y
<i>static</i>	<i>dynamic</i>	<i>dynamic</i>
<i>dynamic</i>	<i>dynamic</i> \rightarrow <i>dynamic</i>	<i>dynamic</i>
(<i>static</i> , <i>static</i>)	<i>dynamic</i>	<i>dynamic</i>
(<i>static</i> , <i>static</i>)	(<i>static</i> , <i>dynamic</i>)	(<i>static</i> , <i>dynamic</i>)
(<i>static</i> , <i>dynamic</i>)	(<i>dynamic</i> , <i>static</i>)	(<i>dynamic</i> , <i>dynamic</i>)

3. In fact, in the particular case of the call-by-value CPS transformation, these static beta-redexes *precisely* coincide with Plotkin's administrative redexes [11]. However, this coincidence only happens for call-by-value and not, *e.g.*, for the call-by-name CPS transformation — an observation independently made by John Hatcliff at Kansas State University and by Ray McDowell at the University of Pennsylvania in fall 1993 (personal communication to the first author).
4. Primitive operations are known as “delta rules” in the lambda-calculus [1].
5. For example, Similix specializes the source program

```
(define (main1 d)
  ((lambda (f) (cons f (f 2)))
   (lambda (a) (- (* 3 (- a d) 1))))
```

into the following residual program.

```
(define (main1-0 d_0)
  (let ([f_2 (lambda (a_1) (- (* 3 (- a_1 d_0)) 1))])
    (cons f_2 (f_2 2))))
```

The decision to residualize is symptomatic of the tension inherent in partial evaluation: unfolding calls exposes opportunities for static computation, but if there are not many opportunities, or if they do not amount to much, unfolding just leads to code duplication. For example, Similix transforms the source program

```
(define (main2 d)
  ((lambda (f) (cons (f 1) (f 2)))
   (lambda (a) (- (* 3 (- a d) 1))))
```

into the following residual program.

```
(define (main2-0 d_0)
  (cons (- (* 3 (- 1 d_0)) 1) (- (* 3 (- 2 d_0)) 1)))
```

6. As pointed out by one of the referees, this duplication may be criticized but it happens only at BTA-time.

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