Efficient multi-variate abstraction using an array representation for combinators

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1. Introduction

Turner [9] showed how a pure functional programming language could be implemented using combinatory logic in a practicable way. This method is currently out of fashion, but for some time I have believed that its full potential is still to be realized. This belief was partially vindicated by Stevens [7] who developed a family of interesting algorithms which use a novel notation for combinators. The work reported here was inspired by that of Stevens, but is significantly different. (Additional information about the motivation behind the current research can be found elsewhere [2, pp. 2–5].)

2. Fixing terminology

There are several systems of combinatory logic. The one used here is weak combinatory logic. On the whole, standard terminology is used [4].

Assume given an infinite sequence of symbols called variables and two constants, K and S, called basic combinators. The letters v, w, x, y and z, sometimes decorated with subscripts or superscripts, are used for variables. A term is defined thus:

(a) Every variable is a term;
(b) Every constant is a term;
(c) If P and Q are terms, then so is (PQ).

The letters P, Q, R, S, T and X, sometimes decorated with subscripts or superscripts, are used for terms. An atom is a variable or a constant. A term of the form (PQ) is an application, but the outermost pair of parentheses is usually omitted. Normally, no space is left between the terms of an application, but sometimes one will be inserted for clarity and readability. Application associates to the left, so PQRST is the same as ((PQ)R)ST. The symbol ≡ represents syntactic identity: P ≡ Q means that P and Q are exactly the same term.

A subterm is defined thus: P is a subterm of P; P is a subterm of QR if P is a subterm of Q or P is a subterm of R. Every term P can be uniquely expressed in the form P1P2...Pm, where P1 is an atom and m ≥ 1. The Pi are known as the primal components of P. The non-standard notion of a subprimal component

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is defined thus: $P$ is a subprimal component of $P'$; $P$ is a subprimal component of $Q$ if $P$ is a subprimal component of one of the primal components of $Q$. For example, the subprimal components of $vw(x(yz))$ are: $vw(x(yz)), v, w, x(yz), x, yz, y, z$.

Because combinatory logic contains no variable-binding operators every variable in a term is free: $FV(P)$ represents the set of free variables in $P$. The length of $P$, written $|P|$, is the number of occurrences of atoms in $P$. Substituting $P$ for every occurrence of $x$ in $X$, written $[P/x]X$, is defined thus:

(a) $[P/x]x \equiv x$;
(b) $[P/x]Y \equiv Y$, if $Y$ is an atom distinct from $x$;
(c) $[P/x]QR \equiv ([P/x]Q)([P/x]R)$.

A term of the form $K P Q$ or $S P Q R$ is a redex. Contracting an instance of a redex in a term $S$ means replacing one occurrence of $K P Q$ or $P Q R$ by $P R Q$. Let the result be $T$. Then we say that $S$ contracts to $T$, written $S \rightarrow T$, and that $T$ is the contractum. $S$ is said to reduce to $T$, written $S \rightarrow T$, iff $T$ results from $S$ by carrying out a finite (possibly zero) number of contractions. Combinators B, C and I can be defined in terms of K and S:

$$B \triangleq S(KSK), \quad C \triangleq S(BBS)(KK) \quad \text{and} \quad I \triangleq SKK.$$

These contract thus:

$$\begin{align*}
BPQ & \rightarrow P(R), \\
CPQR & \rightarrow PRQ \quad \text{and} \quad \langle P \rightarrow P \rangle.
\end{align*}$$

Uni-variate bracket abstraction is a syntactic operation which removes a variable $x$ from a term $X$, written $[x]X$, satisfying the property that $([x]X)P \rightarrow [P/x]X$. If $[x]X \equiv Q$, then $X$ is the input term and $Q$ the abstract. Usually, in combinatory logic, multi-variate abstraction $\{x_1, x_2, \ldots, x_n\}X$ is defined to be the same as $\{x_1\}(\{x_2\}(\ldots(\{x_n\}[X])\ldots))$. In this paper, however, it represents an abstraction that removes several variables simultaneously. Furthermore, all the variables in the bracket prefix $\{x_1, x_2, \ldots, x_n\}$ are assumed to be distinct.

Two non-standard notations for combinators are introduced here, namely as strings of the letters $y$ and $n$, called $yn$-strings, and as arrays or matrices of the letters $y$ and $n$, called $yn$-arrays. The letters $\gamma$ and $\delta$ are used for arbitrary $yn$-arrays and $\beta$ for a $yn$-string. $size(\beta)$ is the number of occurrences of $y$ and $n$ in $\beta$ and $\beta_i$, for $1 \leq i \leq size(\beta)$, is the $i$th letter in $\beta$.

String concatenation is represented by juxtaposition. Then we say that $S$ contracts thus:

- $\langle x, y \equiv \gamma \rangle, \ldots Q, \ldots$.
- $\langle b \equiv \delta \rangle, \ldots P, \ldots$.

Let $\beta$ be a $yn$-string. Then a $\beta$-redex is any term of the form $\beta P_1 P_2 \ldots P_{m+1}$, where $m = size(\beta)$. Contracting an instance of a $\beta$-redex in a term $S$ means replacing one occurrence of $\beta P_1 P_2 \ldots P_{m+1}$ by $Q_1 Q_2 \ldots Q_m$, where, for $1 \leq i \leq m$,

$$Q_i = \begin{cases}
   P_i, & \text{if } \beta_i = y; \\
   P_{m+1}, & \text{if } \beta_i = n.
\end{cases}$$

Let $[x] = [x_1, x_2, \ldots, x_n]$. Then $rpv([x], P)$ is the number of distinct non-atomic subprimal components of $P$, other than $P$ itself, which contain at least one of the variables $x_1, x_2, \ldots, x_n$. There is an alternative characterization of $rpv$. Let $P$ be represented using the fewest possible parentheses. Then $rpv([x], P)$ is half the number of parentheses that enclose subterms containing at least one of the variables $x_1, x_2, \ldots, x_n$.

Thus,

$$rpv([x, y, z], x(yz)(uv)) = 3 \quad \text{and} \quad rpv([x, y], x(y(xv))(uv)) = 2.$$

Let $P = P_1 P_2 \ldots P_n$, where $P_1$ is an atom. Then

$$\begin{align*}
\text{if } (\forall i \in 1..a)[x_i \notin \Gamma(f) \text{ or } \langle 3i \in 1..a \rangle x_i \equiv P_j] \quad \text{then } 0 \quad \text{else } 1 + rpv([\bar{x}], P_j),
\end{align*}$$

where $(\forall i \in 1..a)$ means ‘for all $i$ such that $1 \leq i \leq a$’ and $(\exists i \in 1..a)$ means ‘for some $i$ such that $1 \leq i \leq a$’. Putting a conditional inside a summation may be unusual, but its meaning is straightforward. Let $\Gamma(j)$ be a Boolean-valued function and let $f(j)$ and $g(j)$ be integer-valued ones. Then

$$\begin{align*}
\sum_{j=1}^{m} & \text{ if } \Gamma(j) \text{ then } f(j) \text{ else } g(j) \\
= & \begin{cases}
   \text{if } \Gamma(1) \text{ then } f(1) \text{ else } g(1) & \text{if } \Gamma(2) \text{ then } f(2) \text{ else } g(2) + \cdots & \text{if } \Gamma(m) \text{ then } f(m) \text{ else } g(m).
\end{cases}
\end{align*}$$
3. Contraction

In order to explain how yn-arrays are contracted two functions have to be defined on yn-strings: \( yc(\beta) \) is the number of occurrences of the letter \( y \) in \( \beta \) and \( posy(i, \beta) \) is the position of the \( i \)th occurrence of \( y \) in \( \beta \), where \( 1 \leq i \leq yc(\beta) \); if \( i > yc(\beta) \), then \( posy(i, \beta) \) is not defined. For example, \( yc(yynyyny) = 4 \), \( posy(1, yynyyny) = 1 \) and \( posy(2, yynyyny) = 4 \). Let \( y \) be an \( a \times m \) yn-array. Then it contracts thus:

\[
\gamma P_1 P_2 \ldots P_m P_{m+1} \ldots P_{m+a} \rightarrow_1 Q_1 Q_2 \ldots Q_m,
\]

where, for \( 1 \leq j \leq m \), \( Q_j \equiv P_j P_{g_j(1)} P_{g_j(2)} \ldots P_{g_j(n_j)} \), where \( s_j = yc(\gamma_{1,j} \gamma_{2,j} \ldots \gamma_{a,j}) \) and, for \( 1 \leq k \leq s_j \), \( g_j(k) = m + posy(k, \gamma_{1,j} \gamma_{2,j} \ldots \gamma_{a,j}) \). For example,

\[
\begin{array}{cccc}
| & n & y & n \\
| y & n & y & n \\
| n & n & y & y \\
& 1 & P_1 P_2 P_3 P_4 P_5 P_6 P_7 & (P_3 P_4).
\end{array}
\]

An \( a \times m \) yn-array \( \gamma \) has to be followed by at least \( a + m \) terms in order for a contraction to be possible. Informally, the first \( m \) terms following \( \gamma \) can be thought of as functions and the next \( a \) can be seen as their possible arguments. The yn-array tells us which of these are going to be passed to the functions to become their actual arguments. Thus, the \( j \)th column tells us which of the \( P_{m+1}, P_{m+2}, \ldots, P_{m+a} \) follow \( P_j \) in the contractum of the yn-array: \( P_{m+i} \) only occurs if \( \gamma_{j, i} \) is \( y \). For example, the 4th column of the yn-array used in the example is nny. This tells us which of the terms \( P_5, P_6 \) and \( P_7 \) follow \( P_4 \) in the contractum. As only the last letter of nny is \( y \), only \( P_7 \) does. This means that \( (P_5 P_7) \) occurs in the contractum.

4. Translation

Many people, on first encountering yn-strings, think that they are similar to director strings [6]. Director strings, however, are not combinators. They are elements of a new formal system called ‘the director string calculus’ whose properties have to be established from scratch, yn-strings and yn-arrays, by contrast, are just alternative notations for combinators, just as Roman and Arabic numerals are alternative representations for numbers. This can be established by translating them into the usual notation for combinators. This is achieved by the function \( Trans \) which employs \( trans \) which translates yn-strings into standard combinators. The function \( trans \) uses the series of combinators \( B_i \), for \( i \geq 1 \), defined thus:

\[
B_i = \begin{cases} 
B, & \text{if } i = 1, \\
B B_{i-1} B, & \text{if } i > 1.
\end{cases}
\]

Note that each \( B_i \), for \( i \geq 1 \), has the same effect as the \( B' \) defined in [1, pp. 163–164]. The function \( trans \) is defined thus:

\[
trans(y) = B_l, 
\]

\[
trans(n) = K, 
\]

\[
trans(\beta y) = B_1 B_3 \times trans(\beta), \quad \text{if } size(\beta) \geq 1, 
\]

\[
trans(\beta n) = B_1 B_3 \times trans(\beta), \quad \text{if } size(\beta) \geq 1, 
\]

where \( i = size(\beta) \). For example,

\[
\begin{array}{l}
trans(nnyy) = B_2 S (B_1 C \times trans(n)) \\
= B_2 S (B_1 C K).
\end{array}
\]

The function \( trans \) is correct if \( trans(\beta) P_1 P_2 \ldots P_m P_{m+1} \rightarrow Q \), where \( m = size(\beta) \) and \( Q \) is the result of contracting \( \beta P_1 P_2 \ldots P_m P_{m+1} \). That \( trans \) is correct is proved elsewhere [2, pp. 9–10].

The function \( Trans \) is defined thus:

\[
Trans(\gamma) = \frac{Trans(\gamma_1 \gamma_2 \ldots \gamma_{1,m})}{a-1 \text{ times}} \\
\frac{Trans(\gamma_2 \gamma_2 \ldots \gamma_{2,m})}{a-2 \text{ times}} \\
\frac{Trans(\gamma_{a-1} \gamma_{a-1} \ldots \gamma_{a-1,m})}{a \text{ times}}
\]

\[
trans(y_{a,1} y_{a,2} \ldots y_{a,m}).
\]

\( Trans \) produces quite complicated terms as the following example shows:

\[
\begin{array}{cccc}
| & y & n & y \\
| n & y & y & \\
| n & n & y & \\
\end{array}
\]

\[
= trans(nnyy) \times trans(nnyy) \times trans(nny) \\
= B_3 S (B_1 C (B_2 S (B_1 C K))) \\
( B_3 S (B_2 S (B_1 C K))) (B_2 S (B_1 C K)).
\]
**Proposition 1.** The function $\text{Trans}$ is correct in the sense that if

$$\gamma P_1 P_2 \ldots P_m P_{m+1} \ldots P_{m+a} \to Q_1 Q_2 \ldots Q_m,$$

where $\gamma$ is a $\gamma$-array and the $Q_j$, for $1 \leq j \leq m$, are as specified by the way $\gamma$ contracts, then

$$\text{Trans}(\gamma) P_1 P_2 \ldots P_m P_{m+1} \ldots P_{m+a} \to Q_1 Q_2 \ldots Q_m.$$

**Proof.** Let $t_i$, for $1 \leq i \leq a$, be as shown above and let $P_1^i \equiv P_i$, for $1 \leq i \leq m$. Then

$$t_o \ldots t_1 P_1^1 P_2^1 \ldots P_m^1 P_{m+1} \ldots P_{m+a} \to t_o - 1 \ldots t_1 P_2^2 P_2^2 \ldots P_m^2 P_{m+2} \ldots P_{m+a},$$

where $P_2^2 \equiv P_1^1 P_{m+1}$, if $\gamma_{1,j} = y$, and $P_2^j \equiv P_j^1$, if $\gamma_{1,j} = n$, because $t_o$ is a $\gamma$-string which contracts in this way,

$$t_o - 2 \ldots t_1 P_3^3 P_2^3 \ldots P_m^3 P_{m+3} \ldots P_{m+a},$$

where $P_3^3 = P_2^2 P_{m+2}$, if $\gamma_{2,j} = y$, and $P_j^3 \equiv P_j^2$, if $\gamma_{2,j} = n$, because $t_o - 1$ is a $\gamma$-string.

$$\to \ldots \to P_1^{a+1} P_2^{a+1} \ldots P_m^{a+1},$$

where $P_j^{a+1} \equiv P_j^a P_{m+a}$, if $\gamma_{a,j} = y$, and $P_j^{a+1} \equiv P_j^a$, if $\gamma_{a,j} = n$, because $t_1$ is a $\gamma$-string. Thus,

$$P_j^{a+1} \equiv P_j^1 P_h(j)_1 P_h(j)_2 \ldots P_h(j)(r_j),$$

where $P_h(j)_1$ is the $h_j(1)$th term in the list $P_{m+1}, P_{m+2}, \ldots, P_{m+a}$, where $h_j(1)$ is the position of the first occurrence of the letter $y$ in $\gamma_1, i \gamma_2, j \ldots \gamma_{a,j}$, and $P_h(j)_2$ is the $h_j(1)$th term in the same list, where $h_j(2)$ is the position of the second occurrence of the letter $y$ in $\gamma_{1,j} \gamma_2, j \ldots \gamma_{a,j}$ and so on. $h_j(r_j)$ is the position of the final occurrence of $y$ in $\gamma_1, i \gamma_2, j \ldots \gamma_{a,j}$. Thus, $r_j$ is the total number of occurrences of $y$ in $\gamma_1, i \gamma_2, j \ldots \gamma_{a,j}$. Thus, $r_j = \gamma c(y) \gamma_{1,j} \gamma_2, j \ldots \gamma_{a,j}$ and, for $1 \leq k \leq r_j$, $h_j(k) = g_j(k)$, where $g_j(k)$ is the function defined in the context of explaining how $\gamma$-arrays contract, namely $g_j(k) = a + \text{posy}(k, \gamma_1, \gamma_2, j \ldots \gamma_{a,j})$. \(\square\)

5. Abstraction

In order to present an abstraction algorithm that produces abstracts containing $\gamma$-arrays two functions have to be defined: $tv(\{\vec{x}\}, P)$ returns the total number of variables in the list $\vec{x}$ occurring in $P$ and $\text{inx}(i, [\vec{x}], P)$ returns the index of the $i$th variable in the list $\vec{x}$ occurring in $P$. For example, $tv(\{x_1, x_2, x_3\}, x_1 x_2) = 2$ and $\text{inx}(1, [x_1, x_2, x_3], x_2 x_3(x_1 x_2)) = 2$. Algorithm (M) is shown in Fig. 1. Note that a different algorithm would result if it was not a requirement for $P_1$ to be an atom. The element $\gamma_{n,j}$ of the $\gamma$-array $\gamma$ tells us whether or not $x_n$ occurs in $P_j$. A letter $y$ says that it does and $n$ that it does not.

An example of (M) should clarify its operation:

$$[x_1, x_2, x_3](x_1 x_2)(x_1 x_3) x_2 \equiv \begin{array}{llll}
| & y & y & y & n \\
| & n & n & y & n \\
| & n & n & n & n \\
\end{array} \equiv 1\left(\left(\mid x_1, x_2 \mid x_1, x_3 \right)\right) \equiv 1 \left(\left(\mid x_1, x_3 \mid x_3, x_1 \right)\right) \equiv 1 \left(1\mid x_1, x_3 \mid 1\right).$$

The top row $\gamma y y y$ of the $3 \times 4$ $\gamma$-array shows the pattern of occurrences of the variable $x_1$ in the primal components of the input term. Similarly, the second row $y n n n$ shows the pattern of occurrences of the variable $x_2$ in the primal components of the input term and the third row does the same for the variable $x_3$.

Algorithm (M) has the property that $(x_{\bar{i}})P_{\bar{x}} \to P$. This is Proposition 2 and the proof is by induction on $\phi(a, rpv([\vec{x}]), P)$, where $\phi : N_1 \times N \to N_1$ is a total bijection. $(N)$ is the set of all non-negative whole numbers and $N_1$ is the set of all positive whole numbers.) The function $\phi$ is defined thus:

$$\phi(x, y) \equiv \begin{cases} x^2, & x = y + 1; \\
(x - 1)^2 + y + 1, & x > y + 1; \\
y^2 + y + x, & x < y + 1. \end{cases}$$

**Proposition 2.** $(x_{\bar{i}})P_{\bar{x}} \to P$.

**Proof.** Let $[\vec{x}] = \{x_1, x_2, \ldots, x_o\}$ and $P \equiv P_1 P_2 \ldots P_m$, where $P_1$ is an atom. The proof is by induction on $\phi(a, rpv([\vec{x}]), P)$.

In the base case $\phi(a, rpv([\vec{x}]), P)) = 1$. Thus, $a = 1$ and $rpv([\vec{x}]), P) = 0$. We have that $(x_j, P)_{x_j} \equiv \gamma Q_1 Q_2 \ldots Q_m x_1$, where $\gamma$ and the $Q_j$, for $1 \leq j \leq m$, are...
In this algorithm $P_1$ must be an atom.

\[ [x_1, x_2, \ldots, x_a] P_1 P_2 \ldots P_m \equiv \gamma Q_1 Q_2 \ldots Q_m, \]

where $\gamma$ is a $\gamma$-array and, for $1 \leq i \leq a$ and $1 \leq j \leq m$,

\[ \gamma_{i,j} = \begin{cases} y, & \text{if } x_i \in \text{FV}(P_j), \\ \gamma_{i,j} & \text{otherwise} \end{cases} \]

and, for $1 \leq j \leq m$,

\[ Q_j \equiv \begin{cases} P_j, & \text{if } P_j \equiv x_i, \text{ for some } i \text{ such that } 1 \leq i \leq a, \\ \{x_{f_1(1)}, x_{f_2(2)}, \ldots, x_{f_j(q_j)}\} P_j, & \text{if } P_j \not\equiv x_i, \text{ for any } i \text{ such that } 1 \leq i \leq a, \\ \{x_{f_j(1)}, x_{f_j(2)}, \ldots, x_{f_j(q_j)}\} P_j, & \text{otherwise} \end{cases} \]

where $q_j = t v([x_1, \ldots, x_a], P_j)$ and, for $1 \leq k \leq q_j$, $f_j(k) = \text{ins}(k, [x_1, \ldots, x_a], P_j)$.

Therefore, $R_j \rightarrow P_j$, by the inductive hypothesis. The result follows by induction. $\square$

The proof of Proposition 4 makes use of a lemma. Informally, this states that the length of an abstract produced by (M) is not affected by adding extra variables to the bracket prefix which do not occur in the input term.

**Lemma 3.** Let $[\tilde{x}] = [x_1, x_2, \ldots, x_a]$ and $[\tilde{\gamma}] = [y_1, y_2, \ldots, y_b]$. Then if $b \leq a$ and $(\forall k \in 1..b)(\exists i \in 1..a)(x_i \equiv y_k)$ and $(\forall k, l \in 1..b)(\forall i, j \in 1..a)$ (if $x_i \equiv y_k$ and $x_j \equiv y_l$ and $i < j$, then $k < l$) and $(\forall i \in 1..a)$ (if $x_i \in \text{FV}(P)$, then $(\exists k \in 1..b)(x_i \equiv y_k)$, then $\#([\tilde{x}] P) = \#([\tilde{\gamma}] P)$.

**Proof.** Let $P \equiv P_1 P_2 \ldots P_m$, where $P_1$ is an atom. Then $[\tilde{x}] P \equiv [\tilde{\gamma}] Q_1 Q_2 \ldots Q_m$, where $\gamma$ and the $Q_j$, for $1 \leq j \leq m$, are as specified by (M),

\[ \rightarrow R_1 R_2 \ldots R_m, \]

where the $R_j$, for $1 \leq j \leq m$, are determined by the way $\gamma$-arrays contract. First, consider the case when $(\forall i \in 1..a)(\forall j \in 1..m)x_i \not\in \text{FV}(P_j)$. Then, $\gamma_{i,j} = \gamma_{i,j} \equiv Q_j \equiv P_j$ and $R_j \equiv Q_j$. Thus, $R_j \equiv P_j$. Next, consider the case when $(\exists i \in 1..a)(\exists j \in 1..m)x_i \in \text{FV}(P_j)$. Then, $\gamma_{i,j} = \gamma_{i,j} \equiv P_j$ is a variable or a term. If $P_j \equiv x_i$, for some $i$, then $Q_j \equiv P_j$. If $P_j$ is a term, then $Q_j \equiv \{x_{f_j(1)}, \ldots, x_{f_j(q_j)}\} P_j$. When $\gamma_{i,j} = \gamma_{i,j} \equiv x_k$, then $R_j \equiv Q_j x_{g_j(1)}, \ldots, x_{g_j(r_j)}$, where $f_j(k) = \text{ins}(k, \tilde{x}, P_j)$ and $g_j(k) = \text{posy}(k, \gamma_{i,j} \gamma_{j,j} \ldots \gamma_{a,j})$. Thus, $f_j(k) = g_j(k)$, for all $k$. Also, $f_j(q_j) \leq a$, for all $j$, and $\text{rpv}([x_{f_j(1)}, \ldots, x_{f_j(q_j)}], P_j) < \text{rpv}([\tilde{x}], P)$. Thus,

\[ \phi(f_j(q_j), \text{rpv}([x_{f_j(1)}, \ldots, x_{f_j(q_j)}], P_j)) \]

\[ < \phi(a, \text{rpv}([\tilde{x}], P)). \]

To establish that $(\forall j \in 1..m) Q_j \equiv R_j$ we consider two cases. (1) If $P_j$ contains none of the abstraction variables, then $Q_j \equiv P_j$ and $R_j \equiv P_j$, both by (M). (2) When (M) is applied recursively only those variables that actually occur in the primal component $P_j$ are included in the bracket prefix that is passed to the recursive call of the algorithm. Thus, again, $Q_j \equiv R_j$.

The only difference between the abstracts is that $\gamma$ is an $a \times m$ $\gamma$-array, whereas $\delta$ is an $b \times m$ $\gamma$-array. If $b < a$, then $\gamma$ has $a - b$ extra rows each of which consists entirely of occurrences of $\gamma$. These correspond to the extra variables in the bracket prefix $[\tilde{x}]$ which,
ex hypothesi, do not occur in \( P \). As \(|\gamma| = \# \delta\), we have that \(|\gamma(x) P| = |\gamma(y) P|\). □

Proposition 4. \(|(\vec{x}) P| = 1 + |P| + rpv(\vec{x}, P)\).

Proof. Let \([\vec{x}] = [x_1, x_2, \ldots, x_a] \) and \( P \equiv P_1 P_2 \ldots P_m \), where \( P_i \) is an atom. The proof is by induction on \( rpv(\vec{x}, P) \).

In the base case \( rpv(\vec{x}, P) = 0 \). Thus, for \( 1 \leq j \leq m \), either \( P_j \equiv x_i \), for some \( i \) such that \( 1 \leq i \leq a \), or \( x_i \not\in FV(P_j) \), for any \( i \) such that \( 1 \leq i \leq a \). Also, \([\vec{x}] P \equiv \gamma Q_1 Q_2 \ldots Q_m \), where \( \gamma \) and the \( Q_j \), for \( 1 \leq j \leq m \), are as specified by (M). If \( P_j \equiv x_i \), then \( Q_j \equiv 1 \). If \( x_i \not\in FV(P_j) \), then \( Q_j \equiv P_j \). Thus,

\[ |(\vec{x})P| = 1 + \sum_{j=1}^{m} |P_j| = 1 + |P| + rpv(\vec{x}, P). \]

This establishes the base case.

In the inductive step, we have that \( [\vec{x}]P \equiv \gamma Q_1 Q_2 \ldots Q_m \), where \( \gamma \) and the \( Q_j \), for \( 1 \leq j \leq m \), are as specified by (M). For \( 1 \leq j \leq m \), if \( P_j \equiv x_i \), for some \( i \) such that \( 1 \leq i \leq a \), or \( x_i \not\in FV(P_j) \), for any \( i \) such that \( 1 \leq i \leq a \), then \( Q_j \equiv \#P_j \). Thus,

\[ |(\vec{x})P| = 1 + \sum_{j=1}^{m} \text{if } \exists i \in \ldots a \text{ such that } \forall i \in \ldots a \text{ and if } \exists i \in \ldots a \text{ such that } \forall i \in \ldots a \text{ then } \end{equation}

\[ |(\vec{x})P| = 1 + \sum_{j=1}^{m} \text{if } \exists i \in \ldots a \text{ such that } \forall i \in \ldots a \text{ and if } \exists i \in \ldots a \text{ then } \end{equation}

\[ = 1 + \sum_{j=1}^{m} |P_j| + \sum_{j=1}^{m} \text{if } \exists i \in \ldots a \text{ such that } \forall i \in \ldots a \text{ and if } \exists i \in \ldots a \text{ then } \end{equation}

\[ = 1 + |P| + rpv(\vec{x}, P) \]

6. Conclusion

The most popular way of judging the efficiency of an abstraction algorithm is by considering the length of the abstract produced. It has been correctly argued, in my opinion, that by itself this is a very crude measure of efficiency [8, pp. 148–159]. It is, therefore, only one of the factors that we need to take into account when comparing algorithms. If (M) is applied to \( P \), the length of the abstract produced is \( 1 + |P| + rpv(\vec{x}, P) \). This assumes that the length of a \( yn \)-array is 1. This is reasonable if \( a \) and \( m \) are small, as they usually are when the algorithm is used to implement a functional language, but the larger \( a \) and \( m \) become the more problematic this assumption becomes. The number of \( yn \)-arrays in an abstract is \( 1 + rpv(\vec{x}, P) \) and the largest of these is an \( a \times m \) bit array. The maximum value that \( rpv(\vec{x}, P) \) can take, if \( |P| \geq 2 \), is \( |P| - 2 \). Thus, the space required to store these arrays is not greater than \((|P| - 1)(a \times m)\) bits. Joy, Rayward-Smith and Burton [5, Table 1, p. 216] present the lengths of abstracts produced by various algorithms and (M) is comparable to the best of them. The way in which (M) operates is, however, considerably simpler than its rivals.

It should be noted that \( 1 + rpv(\vec{x}, P) \) is also the number of times that (M) is called. This is relevant when considering the length of time needed to produce the abstract. Similar information for other algorithms is not readily available, but my experience with some of the best known suggests that they are called many more times than this when applied to the same input terms. This is an area where more research needs to be done. It would also be useful to know how (M) performs in practice and how it compares empirically.
with other abstraction methods and with other ways of implementing a functional language.

Even if it turns out that $\nu$-arrays are not a practicable way of implementing a functional language, they do have a certain theoretical interest and many fascinating and unusual properties, as I am beginning to discover [3].

References