

Non-deterministic Characterisations*

Cynthia Kop

Department of Computer Science, University of Copenhagen (DIKU)
kop@di.ku.dk

Abstract

In this paper, we extend Jones’ result—that cons-free programming with k^{th} -order data and a call-by-value strategy characterises EXP^kTIME —to a more general setting, including pattern-matching and non-deterministic choice. We show that the addition of non-determinism is unexpectedly powerful in the higher-order setting. Nevertheless, we can obtain a non-deterministic parallel to Jones’ hierarchy result by appropriate restricting rule formation.

Digital Object Identifier 10.4230/LIPIcs...

1 Introduction

In [4], Jones introduces *cons-free programming*. Working with a small functional programming language, cons-free programs are defined to be *read-only*: recursive data cannot be created or altered (beyond taking sub-expressions), only read from the input. By imposing further restrictions on data order and recursion style, classes of cons-free programs turn out to characterise various deterministic classes in the time and space hierarchies of computational complexity. Most relevantly to this work, cons-free programs with data order k characterise the class EXP^kTIME of decision problems decidable in $\mathcal{O}(\exp_2^k(a \cdot n^b))$ on a Turing Machine.

The classes thus characterised are all *deterministic*: they concern the time and space to solve decision problems on a deterministic Turing Machine. As the language considered by Jones is deterministic, a natural question is whether adding non-deterministic choice to the language would increase expressivity accordingly. The answer, at least in the base case, is *no*: following an early result by Cook [2], Bonfante shows [1] that adding a non-deterministic choice operator to cons-free programs with data order 0 makes no difference in expressivity: whether with or without non-deterministic choice, such programs characterise P.

In this paper, we consider the generalisation of this question: does adding non-deterministic choice give more expressivity when data of order greater than 0 is admitted? Surprisingly, the answer is yes! However, we do not obtain the non-deterministic classes; rather, non-deterministic cons-free programs of any data order ≥ 1 characterise **ELEMENTARY**, the class $\text{EXP}^0\text{TIME} \cup \text{EXP}^1\text{TIME} \cup \text{EXP}^2\text{TIME} \cup \dots$. As this is less useful for complexity arguments, we amend cons-freeness with a further restriction—unary variables—which allows us to obtain the expected generalisation: that (thus restricted) cons-free programs of data order k characterise EXP^kTIME , whether or not non-deterministic choice is allowed.

We also generalise Jones’ language with pattern matching and user-defined constructors.

2 Cons-free programming

For greater generality—and greater ease of expressing examples—we extend Jones’ language to a limited functional programming language with pattern matching. We will use terminology from the term rewriting world, but very little of the possibilities of this world.

* Supported by the Marie Skłodowska-Curie action “HORIP”, program H2020-MSCA-IF-2014, 658162.



2.1 Higher-order Programs

We consider programs using simple types, including product types. The *type order* $o(\sigma)$ of a type σ is defined as follows: $o(\kappa) = 0$ for κ a *sort* (base type), $o(\sigma \times \tau) = \max(o(\sigma), o(\tau))$ and $o(\sigma \Rightarrow \tau) = \max(o(\sigma) + 1, o(\tau))$.

Assume given three disjoint set of identifiers: \mathcal{C} of *constructors*, \mathcal{D} of *defined symbols* and \mathcal{V} of *variables*; each symbol is equipped with a type. Following Jones, we limit interest to constructors with a type $\iota_1 \Rightarrow \dots \Rightarrow \iota_m \Rightarrow \kappa$ where all ι_i are types of order 0 and κ is a *sort*. *Terms* are expressions s such that $s : \sigma$ can be derived for some type σ using the clauses:

- $c s_1 \cdots s_m : \kappa$ if $c : \iota_1 \Rightarrow \dots \Rightarrow \iota_m \Rightarrow \kappa \in \mathcal{C}$ and each $s_i : \iota_i$
- $a s_1 \cdots s_n : \tau$ if $a : \sigma_1 \Rightarrow \dots \Rightarrow \sigma_n \Rightarrow \tau \in \mathcal{V} \cup \mathcal{D}$ and each $s_i : \sigma_i$
- $(s, t) : \sigma \times \tau$ if $s : \sigma$ and $t : \tau$

Thus, constructors cannot be partially applied, while variables and defined symbols can be. If $s : \sigma$, we say σ is the type of s , and let $\text{Var}(s)$ be the set of variables occurring in s . A term s is *ground* if $\text{Var}(s) = \emptyset$. We say t is a subterm of s , notation $s \triangleright t$, if either $s = t$ or $s = a s_1 \cdots s_n$ with $a \in \mathcal{C} \cup \mathcal{F} \cup \mathcal{V}$ and $s_i \triangleright t$ for some i , or $s = (s_1, s_n)$ and $s_i \triangleright t$ for some i . Note that the head of an application is *not* a subterm of the application.

A *rule* is a pair of terms $f \ell_1 \cdots \ell_k \rightarrow r$ such that (a) $f \in \mathcal{D}$, (b) no defined symbols occur in any ℓ_i , (c) no variable occurs more than once in $f \ell_1 \cdots \ell_k$, (d) $\text{Var}(r) \subseteq \text{Var}(f \ell_1 \cdots \ell_k)$, and (e) r has the same type as $f \ell_1 \cdots \ell_k$. A *substitution* γ is a mapping from variables to ground terms of the same type, and $s\gamma$ is obtained by replacing variables x in s by $\gamma(x)$.

We fix a set \mathcal{R} of rules, which are *consistent*: if $f \ell_1 \cdots \ell_k \rightarrow r$ and $f q_1 \cdots q_n \rightarrow s$ are both in \mathcal{R} , then $k = n$; we call k the *arity* of f . The set \mathcal{DA} of *data terms* consists of all ground constructor terms. The set \mathcal{VA} of *values* is given by: (a) all data terms are values, (b) if v, w are values, then (v, w) is a value, (c) if $f \in \mathcal{D}$ has arity k , $n < k$ and s_1, \dots, s_n are values, then $f s_1 \cdots s_n$ is a value if it is well-typed. Note that values whose type is a sort are data terms. The *call-by-value* reduction relation on ground terms is defined by:

- $(s, t) \rightarrow^* (v, w)$ if $s \rightarrow^* v$ and $t \rightarrow^* w$
- $a s_1 \cdots s_n \rightarrow^* a v_1 \cdots v_n$ if each $s_i \rightarrow^* v_i$ and either $a \in \mathcal{C}$, or $a \in \mathcal{D}$ and $n < \text{arity}(a)$
- $f s_1 \cdots s_m \rightarrow^* w$ if there are values v_1, \dots, v_m and a rule $f \ell_1 \cdots \ell_n \rightarrow r$ with $n \leq m$ and substitution γ such that each $s_i \rightarrow^* v_i = \ell_i \gamma$ and $(r\gamma) v_{n+1} \cdots v_m \rightarrow^* w$

Note that rule selection is non-deterministic; a choice operator might for instance be implemented by having rules **choose** $x y \rightarrow x$ and **choose** $x y \rightarrow y$.

2.2 Cons-free Programs

Since the purpose of this research is to find groups of programs which can handle *restricted* classes of Turing-computable problems, we must impose certain limitations. In particular, we will limit interest to *cons-free* programs:

► **Definition 1.** A rule $\ell \rightarrow r$ is *cons-free* if for all subterms $r \triangleright s$ of the form $s = c s_1 \cdots s_n$ with $c \in \mathcal{C}$, we have: $s \in \mathcal{DA}$ or $\ell \triangleright s$. A program is *cons-free* if all its rules are.

This definition follows those for cons-free term rewriting in [3, 5] in generalising Jones' definition in [4]; the latter fixes the constructors in the program and therefore simply requires that the only non-constant constructor, **cons**, does not occur in any right-hand side.

In a cons-free program, if v_1, \dots, v_n, w are all data terms, then any data term occurring in the derivation of $f v_1 \cdots v_n \rightarrow^* w$ is a subterm of some v_i . This includes the result w .

3 Turing Machines and decision problems

In this paper, we particularly consider complexity classes of *decision problems*. A decision problem is a set $A \subseteq \{0, 1\}^+$. A deterministic Turing Machine *decides* A in time $P(n)$ if every evaluation starting with a tape $\sqcup x_1 \dots x_n \sqcup \dots$ completes in at most $P(n)$ steps, ending in the **Accept** state if $x_1 \dots x_n \in A$ and in the **Reject** state otherwise.

Let $\exp_2^0(m) = m$ and $\exp_2^{k+1}(m) = \exp_2^k(2^m) = 2^{\exp_2^k(m)}$. The class EXP^kTIME consists of those decision problems which can be decided in $P(n) \leq \exp_2^k(a \cdot n^b)$ steps for some a, b .

► **Definition 2.** A program $(\mathcal{C}, \mathcal{D}, \mathcal{R})$ with constructors **true**, **false** : **bool**, \square : **list** and $::$ (denoted infix) of type **bool** \Rightarrow **list** \Rightarrow **list**, and a defined symbol **start** : **list** \Rightarrow **bool** *accepts* a decision problem A if for all $\vec{x} = x_1 \dots x_n \in \{0, 1\}^+$: $\vec{x} \in A$ iff **start** $(\overline{x_1} :: \dots :: \overline{x_n} :: \square) \rightarrow^*$ **true**, where $\overline{x_i} = \mathbf{true}$ if $x_i = 1$ and **false** if $x_i = 0$. (Note that it is not required that *all* evaluations end in **true**, just that there is at least one—and none if $x \notin A$).

4 A lower bound for expressivity

To give a lower bound on expressivity, we consider the following result paraphrased from [4]:

► **Lemma 3.** *Suppose that, given an input list $cs ::= \overline{x_1} :: \dots :: \overline{x_n} :: \square$ of length n , we have a representation of $0, \dots, P(n)$, symbols **seed**, **pred**, **zero** $\in \mathcal{D}$, and cons-free rules \mathcal{R} with:*

- **seed** $cs \rightarrow^* v$ for v a value representing $P(n)$
- if v represents $i > 0$, then **pred** $cs v \rightarrow^* w$ for w a value representing $i - 1$
- if v represents i , then **zero** $cs i \rightarrow^* \mathbf{true}$ iff $i = 0$ and **zero** $cs i \rightarrow^* \mathbf{false}$ iff $i \neq 0$

Then any problem which can be decided in time $P(n)$ is accepted by a cons-free program whose data order is the same as that of \mathcal{R} , and which is deterministic iff \mathcal{R} is.

Proof Idea. By simulating an evaluation of a Turing Machine. This simulation encodes all transitions of the machine as rules; a transition from state i to state j , reading symbol r , writing w and moving to the right is encoded by a rule **transition** $i r \rightarrow (j, (w, R))$. In addition, there are rules for **state** $cs n$ —which returns the state the machine is in at time n —, **position** $cs n$ —which returns the position of the tape reader—and **tape** $cs n p$ —for the symbol on the tape at position p and time n . Rules are for instance:

state $cs n \rightarrow \mathbf{ifthenelse} (\mathbf{zero} \text{ } cs \text{ } n) \mathbf{Start} (\mathbf{fst} (\mathbf{transitionat} \text{ } cs (\mathbf{pred} \text{ } cs \text{ } n)))$

This returns **Start** at time 0, and otherwise the state reduced to in the last transition. ◀

► **Example 4.** For $P(n) = (n+1)^2 - 1$, we can represent $i \in \{0, \dots, P(n)\}$ as any pair (l_1, l_2) of lists, where $i = |l_1| \cdot (n+1) + |l_2|$. For the counting functions, we define:

seed $cs \rightarrow (\square, \square)$	zero $cs (\square, \square) \rightarrow \mathbf{true}$
pred $cs (xs, y::ys) \rightarrow (xs, ys)$	zero $cs (xs, y::ys) \rightarrow \mathbf{false}$
pred $cs (x::xs, \square) \rightarrow (xs, cs)$	zero $cs (x::xs, \square) \rightarrow \mathbf{false}$

► **Lemma 5.** *For any $a, b > 0, k \geq 0$, there are cons-free, deterministic rules $\mathcal{R}_{a,b}^k$ defining counting functions as in Lemma 3 such that, for $P(n) = \exp_2^k(a \cdot n^b) - 1$, the numbers $\{0, \dots, P(n)\}$ can be represented. All function variables in $\mathcal{R}_{a,b}^k$ have a type $\sigma \Rightarrow \mathbf{bool}$.*

Proof Idea. For $k = 0$, we can count to $a \cdot n^b - 1$ using an approach much like Example 4. Given $\mathcal{R}_{a,b}^k$, which represents numbers as a type σ , we can define $\mathcal{R}_{a,b}^{k+1}$ by representing a number i with bit vector $b_0 \dots b_M$ (with $M = \exp_2^k(a \cdot n^b)$) as the function in $\sigma \Rightarrow \mathbf{bool}$ which maps a “number” j to **true** if $b_i = 1$ and to **false** otherwise. ◀

The observation that the functional variables take only one input argument will be used in Lemma 8 below. The counting techniques from Example 4 and Lemma 5 originate from Jones' work. However, in a non-deterministic system, we can do significantly more:

► **Lemma 6.** *Let $P_0(n) := n$, and for $k \geq 0$, $P_{k+1}(n) := 2^{P_k(n)} - 1$. Then for each k , we can represent all $i \in \{0, \dots, P_k(n)\}$ as a term of type $\text{bool}^k \Rightarrow \text{list}$, and accompanying counting functions seed_k , pred_k and zero_k can be defined.*

Proof. The base case ($k = 0$) is Example 4. For larger k , let $i \in \{0, \dots, 2^{P_k(n)} - 1\}$ have bit vector $b_1 \dots b_{P_k(n)}$; we say $s : \text{bool}^k \Rightarrow \text{list}$ represents i at level k if for all $1 \leq j \leq P_k(n)$: $b_j = 1$ iff $s \text{ true} \rightarrow^* v$ for some v which represents j at level $k-1$, and $b_j = 0$ iff $s \text{ false} \rightarrow^* v$ for such v . This relies on non-determinism: $s \text{ true}$ reduces to a representation of *every* j with $b_j = 1$. A representation O of 0 at level $k-1$ is used as a default, e.g. $s \text{ false} \rightarrow^* O$ even if each $b_j = 0$. The zero and pred rules rely on testing bit values, using:

$$\begin{aligned} \text{bitset}_k \text{ cs } F \ j &\rightarrow \text{bshelp}_k \text{ cs } F \ j \ (\text{equal}_{k-1} \text{ cs } (F \ \text{true}) \ j) \ (\text{equal}_{k-1} \text{ cs } (F \ \text{false}) \ j) \\ \text{bshelp}_k \text{ cs } F \ j \ \text{true } b &\rightarrow \text{true} \quad \text{bshelp}_k \text{ cs } F \ j \ b \ \text{true} \rightarrow \text{false} \\ \text{bshelp}_k \text{ cs } F \ j \ \text{false } \text{false} &\rightarrow \text{bitset}_k \text{ cs } F \ j. \end{aligned}$$

These rules are non-terminating, but if F represents a number at level k , and j at level $k-1$, then $\text{bitset}_k \text{ cs } F \ j$ reduces to exactly one value: true if $b_j = 1$, and false if $b_j = 0$. ◀

Thus, we can count up to arbitrarily high numbers; by Lemma 3, every decision problem in ELEMENTARY is accepted by a non-deterministic cons-free program of data order 1.

To obtain a more fine-grained characterisation which still admits non-deterministic choice, we will therefore consider a restriction of cons-free programming which avoids Lemma 6.

► **Definition 7.** A cons-free program *has unary variable* if all variables occurring in any rule in \mathcal{R} have a type ι or $\sigma \Rightarrow \iota$, with $o(\iota) = 0$.

Intuitively, in a program with unary variables, functional variables cannot be *partially applied*; thus, such variables represent a function mapping to *data*, and not to some complex structure. Note that the input type σ of a unary variable $x : \sigma \Rightarrow \iota$ is allowed to be a product $\sigma_1 \times \dots \times \sigma_n$. Lemma 6 relies on non-unary variables, but Lemma 5 does not. We obtain:

► **Lemma 8.** *Any problem in EXP^kTIME is accepted by a (non-deterministic) extended cons-free program of data order k .*

5 An upper bound for expressivity

To see that extended cons-free programs *characterise* the EXPTIME hierarchy, it merely remains to be seen that every decision problem that is accepted by a call-by-value cons-free program with unary variables and of data order k , can be solved by a deterministic Turing Machine—or, equivalently, an algorithm in pseudo code—running in polynomial time.

► **Algorithm 9** (Finding the values for given input in a fixed extended cons-free program \mathcal{R}).

Input: a term $\text{start } v_1 \dots v_n : \iota$ with each v_i a data term and $o(\iota) = 0$.

Output: all data terms w such that $\text{start } v_1 \dots v_n \rightarrow^* w$.

Let $\mathcal{B} := \bigcup_{1 \leq i \leq m} \{w \in \mathcal{DA} \mid v_i \triangleright w\} \cup \bigcup_{\ell \rightarrow r \in \mathcal{R}} \{w \in \mathcal{DA} \mid r \triangleright w\}$.

For all types σ occurring as data in \mathcal{R} , generate $\llbracket \sigma \rrbracket$ and a relation \sqsupseteq , as follows:

- $\llbracket \kappa \rrbracket = \{s \in \mathcal{B} \mid s : \kappa\}$ if κ is a sort; for $A, B \in \llbracket \kappa \rrbracket$, let $A \sqsupseteq B$ if $A = B$
- $\llbracket \sigma \times \tau \rrbracket = \{(A, B) \mid A \in \llbracket \sigma \rrbracket \wedge B \in \llbracket \tau \rrbracket\}$; $(A_1, A_2) \sqsupseteq (B_1, B_2)$ if $A_1 \sqsupseteq B_1$ and $A_2 \sqsupseteq B_2$
- $\llbracket \sigma \Rightarrow \tau \rrbracket = \mathcal{P}(\{(A, B) \mid A \in \llbracket \sigma \rrbracket \wedge B \in \llbracket \tau \rrbracket\})$; for $A, B \in \llbracket \sigma \Rightarrow \tau \rrbracket$ let $A \sqsupseteq B$ if $A \supseteq B$

For all $f : \sigma_1 \Rightarrow \dots \Rightarrow \sigma_m \Rightarrow \iota \in \mathcal{D}$, note that we can safely assume that $\text{arity}(f) \geq m - 1$. For all such f , and all $A_1 \in \llbracket \sigma_1 \rrbracket, \dots, A_m \in \llbracket \sigma_m \rrbracket, v \in \llbracket \iota \rrbracket$, note down a statement: $f A_1 \cdots A_m \approx v$. If $\text{arity}(f) = m - 1$, also note down $f A_1 \cdots A_{m-1} \approx O$ for all $O \in \llbracket \sigma_m \Rightarrow \iota \rrbracket$.

For all rules $\ell \rightarrow r$, all $s : \tau$ with $r \supseteq s$ or $s = r x$, all $O \in \llbracket \tau \rrbracket$ and all substitutions γ mapping the variables $x : \sigma \in \text{Var}(s)$ to elements of $\llbracket \sigma \rrbracket$, note down a statement $s\gamma \approx O$. Mark all statements $x\gamma \approx O$ such that $x\gamma \sqsupseteq O$ as *confirmed*, and all other statements *unconfirmed*. Repeat the following steps until no new statements are confirmed anymore.

- For every unconfirmed statement $f A_1 \cdots A_n \approx O$, determine all rules $f \ell_1 \cdots \ell_k \rightarrow r$ (with $k = n$ or $k = n - 1$) and substitutions γ mapping $x : \sigma \in \text{Var}(f \ell_1 \cdots \ell_k)$ to an element of $\llbracket \sigma \rrbracket$, such that each $A_i = \ell_i \gamma$, and mark the statement as confirmed if $(r x_{k+1} \cdots x_n) \gamma [x_{k+1} := A_{k+1}, \dots, x_n := A_n] \approx O$ is confirmed.
- For every unconfirmed statement $(F s)\gamma \approx O$, mark the statement as confirmed if there exists A with $(A, O) \in \gamma(F)$ and $s\gamma \approx A$ is confirmed.
- For every unconfirmed statement $(f s_1 \cdots s_n)\gamma \approx O$, mark it as confirmed if there are A_1, \dots, A_n such that both $f A_1 \cdots A_n \approx O$ and each $s_i \gamma \approx A_i$ are confirmed.

Then return all w such that **start** $v_1 \cdots v_n \approx w$ is marked confirmed.

► **Lemma 10.** *Algorithm 9 is in EXP^kTIME —where k is the data order of \mathcal{R} —and returns the claimed output.*

Proof Idea. The complexity of Algorithm 9 is determined by the size of each $\llbracket \sigma \rrbracket$. The proof of soundness and completeness of the algorithm is more intricate; this fundamentally relies on replacing the values $f v_1 \cdots v_n$ with $n < \text{arity}(f)$ by subsets of the set of all tuples (A, w) with the property that, intuitively, $f v_1 \cdots v_n A \rightarrow^* w$. ◀

6 Conclusion

Thus, we obtain the following variation of Jones' result:

► **Theorem 11.** *A decision problem A is in EXP^kTIME if and only if there is a cons-free program \mathcal{R} of data order k and with unary variables, which accepts A . This statement holds whether or not the program is allowed to use non-deterministic choice.*

In addition, we have adapted Jones' language to be more permissive, admitting additional constructors and pattern matching. This makes it easier to specify suitable programs.

Using non-deterministic programs is a step towards further characterisations; in particular, we intend to characterise $\text{NEXP}^k\text{TIME} \subseteq \text{EXP}^{k+1}\text{TIME}$ using restricted non-deterministic cons-free programs of data order $k + 1$.

References

- 1 G. Bonfante. Some programming languages for logspace and ptime. In *AMAST '06*, volume 4019 of *LNCS*, pages 66–80, 2006.
- 2 S.A. Cook. Characterizations of pushdown machines in terms of time-bounded computers. *ACM*, 18(1):4–18, 1971.
- 3 D. de Carvalho and J. Simonsen. An implicit characterization of the polynomial-time decidable sets by cons-free rewriting. In *RTA-TLCA '14*, volume 8560 of *LNCS*, pages 179–193, 2014.
- 4 N. Jones. Life without cons. *JFP*, 11(1):5–94, 2001.
- 5 C. Kop and J. Simonsen. Complexity hierarchies and higher-order cons-free rewriting. In *FSCD '16*, volume 52 of *LIPICs*, pages 23:1–23:18, 2016.