

Stochastic Lambda-Calculus

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Pidgin Curry?

Combinatory logic is an abstract science dealing with objects called combinators. What their objects are need not be specified; the important thing is how they act upon each other.

One is free to-choose for one's "combinators" **anything one likes** (for example, computer programs). Well, I have chosen **birds** for my combinators — motivated, no doubt, by the memory of the late **Professor Haskell Curry**, who was both a great combinatory logician and an avid bird-watcher.

The main reason I chose combinatory logic for the central theme of this book was not for its practical applications, of which there are many, but for its great entertainment value. Here is a field considered highly technical, yet perfectly available to the general public; it is chock-full of material from which one can cull excellent recreational puzzles, and at the same time it ties up with fundamental issues in modern logic.

What could be better for a puzzle book? (Preface, p. x.)

Raymond M. Smullyan. *To Mock a Mockingbird and Other Logic Puzzles*
Alfred A. Knopf, 1985, x + 256 pp.

Some Other Quotations

There is, however, one feature that I would like to suggest should be incorporated in the machines, and that is a ***random element***.

– Alan Turing, ***Intelligent Machinery, A Heretical Theory***

83. What is the difference between a Turing machine and the modern computer? It's the same as that between Hillary's ascent of Everest and the establishment of a Hilton hotel on its peak.

60. Dana Scott is the Church of the Lattice-Way Saints.

30. Simplicity does not precede complexity, but follows it.

– Alan Perlis, ***Epigrams on Programming***

Church's λ -Calculus

Definition. λ -calculus — as a formal theory — has rules for the *explicit definition* of functions via well known equational axioms:

α -conversion

$$\lambda X. [\dots X \dots] = \lambda Y. [\dots Y \dots]$$

β -conversion

$$(\lambda X. [\dots X \dots]) (T) = [\dots T \dots]$$

η -conversion

$$\lambda X. F(X) = F$$

NOTE: The third axiom will be dropped in favor of a theory employing properties of a **partial ordering**.

The Enumeration Operator Model

Definitions. (1) *Pairing*: $(n, m) = 2^n(2m+1)$.

(2) *Sequence numbers*: $\langle \rangle = 0$ and

$$\langle n_0, n_1, \dots, n_{k-1}, n_k \rangle = (\langle n_0, n_1, \dots, n_{k-1} \rangle , n_k).$$

(3) *Sets*: $\text{set}(0) = \emptyset$ and $\text{set}((n, m)) = \text{set}(n) \cup \{m\}$.

(4) *Kleene star*: $X^* = \{n \mid \text{set}(n) \subseteq X\}$, for sets $X \subseteq \mathbb{N}$.

Definition. The *model* is given by these definitions on the *powerset* of the set integers, $\mathcal{P}(\mathbb{N})$:

Application:

$$F(X) = \{ m \mid \exists n \in X^* . (n, m) \in F \}$$

Abstraction:

$$\lambda X . [\dots X \dots] =$$

$$\{0\} \cup \{ (n, m) \mid m \in [\dots \text{set}(n) \dots] \}$$

What is the Secret?

- (1) The powerset $\mathcal{P}(\mathbb{N}) = \{X \mid X \subseteq \mathbb{N}\}$ is a **topological space** with the sets $\mathcal{U}_n = \{X \mid n \in X^*\}$ as a **basis** for the topology.
- (2) Functions $\Phi: \mathcal{P}(\mathbb{N})^n \rightarrow \mathcal{P}(\mathbb{N})$ are **continuous** iff, for all $m \in \mathbb{N}$, we have $m \in \Phi(X_0, X_1, \dots, X_{n-1})$ iff there are $k_i \in X_i^*$ for each of the $i < n$, such that $m \in \Phi(\text{set}(k_0), \text{set}(k_1), \dots, \text{set}(k_{n-1}))$.
- (3) The application operation $F(X)$ is continuous as a function of **two** variables.
- (4) If the function $\Phi(X_0, X_1, \dots, X_{n-1})$ is continuous, then the abstraction term $\lambda X_0. \Phi(X_0, X_1, \dots, X_{n-1})$ is continuous in all of the **remaining variables**.
- (5) If $\Phi(X)$ is continuous, then $\lambda X. \Phi(X)$ is the **largest set** F such that for all sets T , we have $F(T) = \Phi(T)$. And, therefore, generally $F \subseteq \lambda X. F(X)$.

NOTE: This model could easily have been defined in 1957!!

It clearly satisfies the rules of **α , β -conversion** (but not **η**).

THIS LECTURE IS DEDICATED TO THE MEMORIES OF

John R. Myhill

Born: 11 August 1923, Birmingham, UK

Died: 15 February 1987, Buffalo, NY

John Shepherdson

Born: 7 June 1926, Huddersfield, UK

Died: 8 January 2015, Bristol, UK

Hartley Rogers, Jr.

Born: 6 July, 1926, Buffalo, NY

Died: 17 July, 2015, Waltham, MA

- John Myhill and John C. Shepherdson, *Effective operations on partial recursive functions*, **Zeitschrift für Mathematische Logik und Grundlagen der Mathematik**, vol. 1 (1955), pp. 310-317.
- Richard M. Friedberg and Hartley Rogers Jr., *Reducibility and completeness for sets of integers*, **Mathematical Logic Quarterly**, vol. 5 (1959), pp. 117-125. Some earlier results are presented in an abstract in **The Journal of Symbolic Logic**, vol. 22 (1957), p. 107.
- Hartley Rogers, Jr., **Theory of Recursive Functions and Effective Computability**, McGraw-Hill, 1967, xix + 482 pp.

Some Lambda Properties

Theorem. For all sets of integers F and G we have:

$$\lambda X.F(X) \subseteq \lambda X.G(X) \text{ iff } \forall X.F(X) \subseteq G(X),$$

$$\lambda X.(F(X) \cap G(X)) = \lambda X.F(X) \cap \lambda X.G(X),$$

and

$$\lambda X.(F(X) \cup G(X)) = \lambda X.F(X) \cup \lambda X.G(X).$$

Definition. A continuous operator $\Phi(X_0, X_1, \dots, X_{n-1})$

is **computable** iff in the model this set is **RE**:

$$F = \lambda X_0 \lambda X_1 \dots \lambda X_{n-1} . \Phi(X_0, X_1, \dots, X_{n-1}).$$

How to do Recursion?

Three Basic Theorems.

- All pure λ -terms define *computable* operators.
- If $\Phi(X)$ is continuous and if we let $\nabla = \lambda X. \Phi(X(X))$, then the set $P = \nabla(\nabla)$ is the *least fixed point* of Φ .
- The least fixed point of a *computable* operator is computable.

A Principal Theorem. These computable operators:

$$\text{Succ}(X) = \{n+1 \mid n \in X\},$$

$$\text{Pred}(X) = \{n \mid n+1 \in X\}, \text{ and}$$

$$\text{Test}(Z)(X)(Y) = \{n \in X \mid 0 \in Z\} \cup \{m \in Y \mid \exists k. k+1 \in Z\},$$

together with λ -calculus, suffice for defining **all RE sets**.

Gödel Numbering

Theorem. There is a computable $V = \lambda x. V(x)$ where

- (i) $V(\{0\}) = \lambda y. \lambda x. y,$
- (ii) $V(\{1\}) = \lambda z. \lambda y. \lambda x. z(x)(y(x)),$
- (iii) $V(\{2\}) = \text{Test},$
- (iv) $V(\{3\}) = \text{Succ},$
- (v) $V(\{4\}) = \text{Pred},$ and
- (vi) $V(\{4 + (n, m)\}) = V(\{n\})(V(\{m\})).$

Theorem. Every *recursively enumerable set* is of the form $V(\{n\})$.

NOTE: The operator V is the analogue of the Universal Turing Machine.

Inseparable Sets?

Definition. Modify the definition of \mathbf{V} via *finite approximations*:

- (i) $\mathbf{V}_k(\{n\}) = \mathbf{V}(\{n\}) \cap \{i \mid i < k\}$ for $n < 5$, and
- (ii) $\mathbf{V}_k(\{4 + (n, m)\}) = \mathbf{V}_k(\{n\}) (\mathbf{V}_k(\{m\}))$.

Theorem. Each $\mathbf{V}_k(\{n\}) \subseteq \mathbf{V}_{k+1}(\{n\})$ is *finite*,
the predicate $j \in \mathbf{V}_k(\{n\})$ is *recursive*,
and we have:

$$\mathbf{V}(\{n\}) = \bigcup_{k < \infty} \mathbf{V}_k(\{n\}).$$

Theorem. The sets \mathcal{L}_0 and \mathcal{L}_1 are *recursively enumerable*,
disjoint, and *recursively inseparable*:

$$\mathcal{L}_0 = \{n \mid \exists j [0 \in \mathbf{V}_j(\{n\})(\{n\}) \wedge 1 \notin \mathbf{V}_j(\{n\})(\{n\})]\}$$

$$\mathcal{L}_1 = \{n \mid \exists k [1 \in \mathbf{V}_k(\{n\})(\{n\}) \wedge 0 \notin \mathbf{V}_k(\{n\})(\{n\})]\}$$

How to Randomize?

Definition. By a *random variable* we mean a function

$$\mathbf{X}: [0, 1] \rightarrow \mathcal{P}(\mathbb{N}),$$

where, for $n \in \mathbb{N}$, the set $\{t \in [0, 1] \mid n \in \mathbf{X}(t)\}$ is always *Lebesgue measurable*.

Theorem. The random variables over $\mathcal{P}(\mathbb{N})$ are closed under (pointwise) application and form a model for the λ -calculus — expanding the original model.

This idea is the beginning of putting a **Boolean-valued Logic** on random variables using the complete Boolean algebra of measurable sets **modulo** sets of measure zero. This new model gives us a programming language with **randomized parameters**.

Randomized Coin Tossing

Definition. A *coin flip* is a random variable

$$\mathbf{F}: [0, 1] \rightarrow \{\{0\}, \{1\}\},$$

It is *fair* iff $\mu[\mathbf{F} = \{0\}] = 1/2$.

Definition. *Pairing functions* for sets in $\mathcal{P}(\mathbb{N})$ can be defined by these enumeration operators:

$$\mathbf{Pair}(X)(Y) = \{2n \mid n \in X\} \cup \{2m+1 \mid m \in Y\}$$

$$\mathbf{Fst}(Z) = \{n \mid 2n \in Z\} \text{ and } \mathbf{Snd}(Z) = \{m \mid 2m+1 \in Z\}.$$

Definition. A *tossing process* is a random variable \mathbf{T} where $\mathbf{Fst}(\mathbf{T})$ is a fair coin flip and where $\mathbf{Snd}(\mathbf{T})$ is *another* tossing — with the successive flippings all being *mutually independent*.

The problem with **using** a coin-tossing process \mathbf{T} in an algorithm is that once $\mathbf{Fst}(\mathbf{T})$ has been looked at, then **that toss** should be **discarded**, and only the new coins from $\mathbf{Snd}(\mathbf{T})$ should be used in the **future**.

A Prototype Algorithm Language

Perhaps a solution is always to evaluate programs **in the order** in which expressions are written. Let's try a very sparse language.

V_i — a *variable*

$M(N)$ — an *application*

$\lambda V_i . M$ — an *abstraction*

$M \oplus N$ — a *stochastic choice*

Let $V_i = M$ **in** N — a *direct valuation*

The idea here is that the **text** M is evaluated in an **environment** giving the values of free variables. Then the result is passed on to a **continuation**. In case a random choice is needed, the **tossing** process is called. We will try to employ a **continuation semantics** where the **denotation** of a program uses the λ -calculus formulation:

$\langle M \rangle (\text{env}) (\text{cont}) (\text{toss})$

The Semantical Equations

- $\langle V_i \rangle(E)(C)(T) = C(E(\{i\}))(T)$
- $\langle M(N) \rangle(E)(C)(T) = \langle M \rangle(E)(\lambda X. \langle N \rangle(E)(\lambda Y. C(X(Y))))(T)$
- $\langle \lambda V_i. M \rangle(E)(C)(T) = C(\lambda X. \langle M \rangle(E[X/\{i\}]))(T)$
- $\langle M \oplus N \rangle(E)(C)(T) = \text{Test}(\text{Fst}(T))(\langle M \rangle(E))(\langle N \rangle(E))(C)(\text{Snd}(T))$
- $\langle \text{Let } V_i = M \text{ in } N \rangle(E)(C)(T) = \langle N \rangle(E[\langle M \rangle(E)/\{i\}])(C)(T)$

Running a (closed) program means evaluating:

$$\langle M \rangle(\emptyset)(\lambda X. \lambda Y. X)(T)$$

The semantics and model as presented here, however, are only **sketches**. Examples of randomized algorithms need to be worked out, as well as good methods of **proving probabilistic properties** of programs.

An Absoluteness Theorem

Theorem. If a closed program has a *non-random* value, then the value is the same for all tossing processes.

Proof Idea: Working within Boolean-valued logic over the measure algebra of Lebesgue sets modulo sets of measure zero, all tossing processes are the same up to a measure-preserving automorphism of the measure algebra.

A PLEA FOR HELP !

Let's find some good
applications for this model
with random variables!