What lies beneath the Language of Thought

Steve Piantadosi UC Berkeley, Psychology July, 2018

Humans learn many formal systems

- Basic logic
- Natural language logic
- First-order logic quantifiers
- Second-order quantification
- Generalized quantifiers
- Grammars
- Programming languages
- Tree structures and relations
- Dominance hierarchies/relations
- Physics
- Arbitrary graphs
- Games
- Simulations
- Mathematics
- Reasoning

- (e.g., and, or, not, iff)
- (e.g. "and", "or")
- (e.g. \forall , \exists)
- (e.g. there exists a property P ...)
- (e.g. natural language "most")
- (e.g. context-free grammars)
- (e.g. python, haskell, prolog)
- (e.g. kinship systems)
- (e.g. Putin > Trump)
- (e.g. block stacking)
- (e.g. subway map)
- (e.g. tic-tac-toe, nim, battleship)
- (e.g. hypotheticals)
- (e.g. calculus, algebra)
- (e.g. knights and naves)

Where does all of this come from?

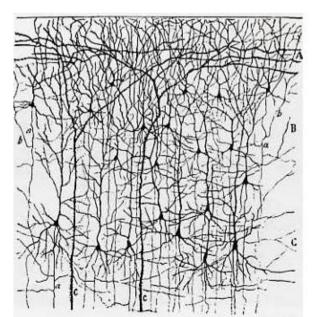
and, or not, if, while, for, pair, define, plus, times, empty, equals, etc. 

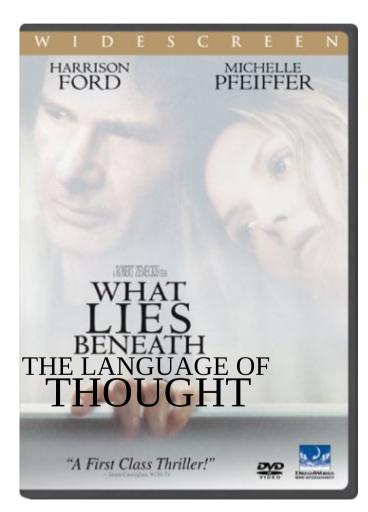
Bayesian program induction



Two central problems

- What can we learn without presupposing the necessary concepts?
 - For instance: do we need to assume logic? What if newborns don't know logic? If they don't, what could it possibly be built from? Are we committed to newborns having the full power of a compiler?
- How can we talk about programs (if statements, logic, sets, etc.) when real brains look like this crazy stuff?







The key idea

- View the LOT as a first and foremost a system for encoding
 - a general model learner
 - grounded in simple underlying dynamics not a normal programming language (e.g. not a C compiler – or even a scheme compiler)
 - informed by cognitive psychology about what is natural (composition, structure, recursion)
- Learning is creating a representation that is <u>isomorphic</u> to some thing in the world.
 - LOT expressions must "act like" stuff in the world
 - The primitives must let it "act like" anything we can comprehend

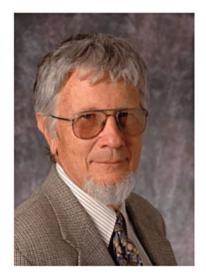
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Isomorphism as the heart of representation

A mental representation is a functioning isomorphism between a set of processes in the brain and a behaviorally important aspect of the world. This way of defining a representation is taken directly from the mathematical definition of a representation. To establish a representation in mathematics is to establish an isomorphism (formal correspondence) between two systems of mathematical investigation (for example, between geometry and algebra) that permits one to use one system to establish truths about the other (as in analytic geometry, where algebraic methods are used to prove geometric theorems). [Gallistel]



Int. J. Systems Sci., 1970, vol. 1, No. 2, 89-97

EVERY GOOD REGULATOR OF A SYSTEM MUST BE A MODEL OF THAT SYSTEM¹

Roger C. Conant

Department of Information Engineering, University of Illinois, Box 4348, Chicago, Illinois, 60680, U.S.A.

and W. Ross Ashby

Biological Computers Laboratory, University of Illinois, Urbana, Illinois 61801, U.S.A.²

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The design of a complex regulator often includes the making of a model of the system to be regulated. The making of such a model has hitherto been regarded as optional, as merely one of many possible ways.

m this paper a theorem is presented which shows, under very broad conditions, that any regulator that is maximally both successful and simple *must* be isomorphic with the system being regulated. (The exact assumptions are given.) Making a model is thus necessary.

The theorem has the interesting corollary that the living brain, so far as it is to be successful and efficient as a regulator for survival, *must* proceed, in learning, by the formation of a model (or models) of its environment.

1. INTRODUCTION

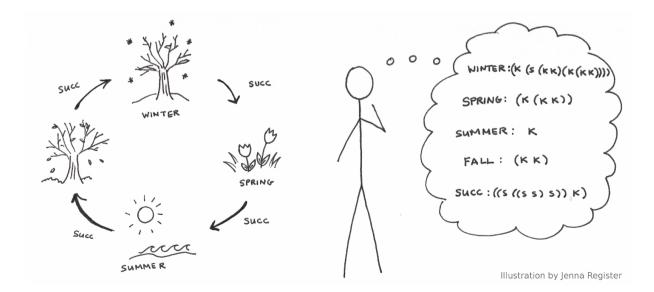
Today, as a step towards the control of complex dynamic systems, models are being used ubiquitously. Being modelled, for instance, are the air traffic flow around New york, the endocrine balances of the pregnant sheep, and the flows of money among the banking centres.

Combinatory logic (or something like it)

• A solution to several problems:

- CL allows encoding of arbitrary logical systems (is Turing-complete)
- CL is based in very simple dynamics (which themselves specify only how CL terms interact)

(no *cognitive* primitives)



Combinatory logic



Moses Schönfinkel



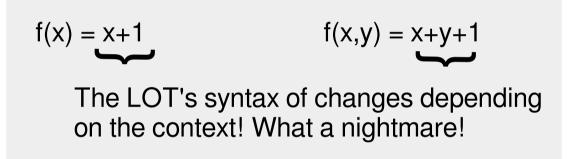


John von Neumann

Haskell Curry

The LOT does not need (explicit) variables

• Variable binding is a problem in neuroscience / cognitive science In fact, central in debates about representation (Marcus 2003)



- Not a problem in CL: variables emerge *only implicitly* through how primitives treat their arguments
 - f(x)=x+1 can be written as (S + (K 1)), no need for x.

Combinatory logic is super simple

Rule 1: $(\mathbf{K} \times \mathbf{y}) \rightarrow \mathbf{x}$ or in other notation $\mathbf{K}(\mathbf{x},\mathbf{y}) \rightarrow \mathbf{x}$

Rule 2: $(S \times y z) \rightarrow ((x z) (y z))$ or in other notation, $S(x,y,z) \rightarrow x(z,y(z))$

Currying – a function can take the next arguments in line e.g. $((K x) y) \rightarrow (K x y) \rightarrow x$

For example f(x)=x+1

- Let f := (S + (K 1))
- Then,

$$(f 7) := ((S + (K 1)) 7) \rightarrow (S + (K 1) 7) \rightarrow ((+ 7) ((K 1) 7)) \rightarrow (+ 7 ((K 1) 7)) \rightarrow (+ 7 (K 1 7)) \rightarrow (+ 7 1)$$

- ; Definition of f
- ; Currying
- ; Definition of S
- ; Currying
- ; Currying
- ; Definition of K

But we can do better

- But what about terms like "+" and "1"? These are still "standard" LOT operations whose meaning must be determined elsewhere.
- **Pure combinatory logic** permits computation <u>without</u> any primitives other than **S**, **K**.

Motivating question

- Is there any learning system that can acquire the most primitive computational concepts?
 - True/false
 - If/then
 - Quantification
 - Iteration/Recursion
 - Data structures (e.g. lists/trees)
 - Identity function

true := (K K)
false := K
and := ((S (S (S S))) (K (K K)))
or := ((S S) (K (K K)))
not := ((S ((S K) S)) (K K))

(or true false) = $(((\mathbf{S} \ \mathbf{S}) \ (\mathbf{K} \ (\mathbf{K} \ \mathbf{K}))) \ (\mathbf{K} \ \mathbf{K}) \ \mathbf{K})$

true := (K K)
false := K
and := ((S (S (S S))) (K (K K)))
or := ((S S) (K (K K)))
not := ((S ((S K) S)) (K K))

(or true false) = (((S S) (K (K K))) (K K) K) \rightarrow (((S S) (K (K K)) (K K)) K) ; Currying rule

true := (K K)
false := K
and := ((S (S (S S))) (K (K K)))
or := ((S S) (K (K K)))
not := ((S ((S K) S)) (K K))

(or true false) = (((S S) (K (K K))) (K K) K) → (((S S) (K (K K)) (K K)) K) ; Currying rule → ((S S (K (K K)) (K K)) K) ; Currying rule twice

true := (K K)
false := K
and := ((S (S (S S))) (K (K K)))
or := ((S S) (K (K K)))
not := ((S ((S K) S)) (K K))

(or true false) = $(((\mathbf{S} \ \mathbf{S}) \ (\mathbf{K} \ (\mathbf{K} \ \mathbf{K}))))$	$(\mathbf{K} \ \mathbf{K}) \ \mathbf{K})$
\rightarrow (((SS) (K (KK)) (KK)) K)	; Currying rule
\rightarrow ((SS(K(KK)) (KK)) K)	; Currying rule twice
\rightarrow ((S (K K) ((K (K K)) (K K))) K)	; Definition of S

true := (K K)
false := K
and := ((S (S (S S))) (K (K K)))
or := ((S S) (K (K K)))
not := ((S ((S K) S)) (K K))

(or true false) = $(((\mathbf{S} \ \mathbf{S}) \ (\mathbf{K} \ (\mathbf{K} \ \mathbf{K}))) \ (\mathbf{K} \ \mathbf{K}) \ \mathbf{K})$ \rightarrow (((S S) (K (K K)) (K K)) K) ; Currying rule ; Currying rule twice \rightarrow ((S S (K (K K)) (K K)) K) \rightarrow ((S (K K) ((K (K K)) (K K))) K) ; Definition of S \rightarrow ((S (K K) (K (K K) (K K))) K) ; Currying ; Definition of K \rightarrow ((S (K K) (K K)) K) \rightarrow (S (K K) (K K) K) ; Currying : Definition of S \rightarrow ((K K) K ((K K) K)) ; Currying \rightarrow ((K K K) ((K K) K)) ; Definition of K \rightarrow (K ((K K) K)) \rightarrow (K (K K K)) ; Currying Definition of K \rightarrow (**K K**)

Church encoding

• This technique is known as <u>church encoding</u>.

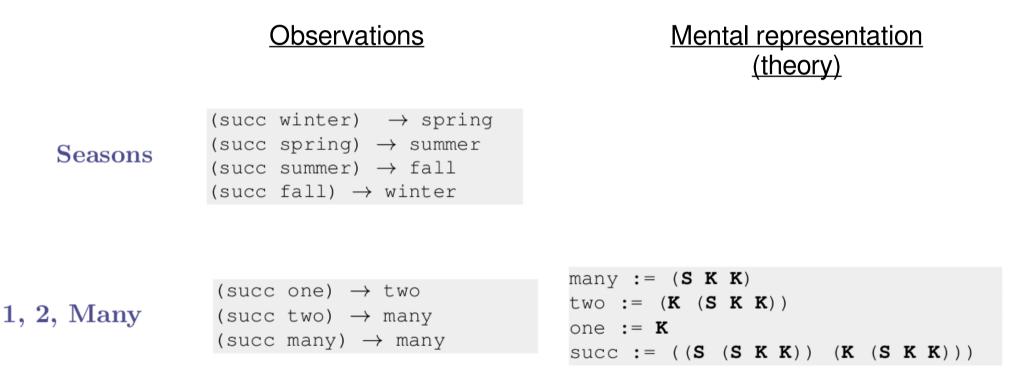


... suppose we have a program that does some complicated calculation with numbers to yield a boolean result. If we replace all the numbers and arithmetic operations with [combinator]-terms representing them and evaluate the program, we will get the same result. Thus, in terms of their effects on the overall result of programs, there is no observable difference between the real numbers and their Church-[encoded]numeral representations. (Pierce 2002)



Churiso

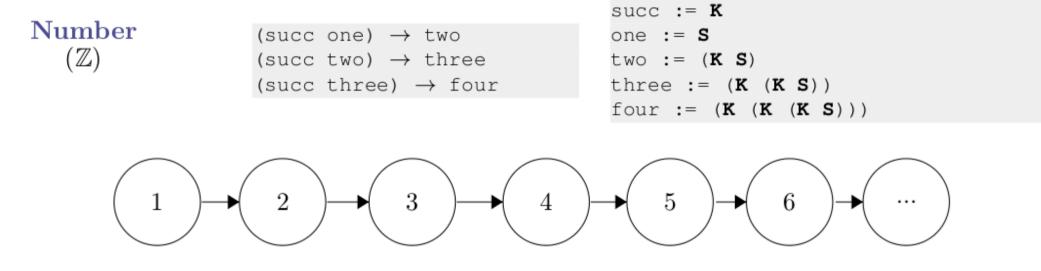
 My lab has been working on a library to infer church encodings from simple relational information.



 How we infer: use ideas from the inductive LOT – prefer encodings with <u>short running time</u> + <u>simple structure</u>.

Combinator generalization

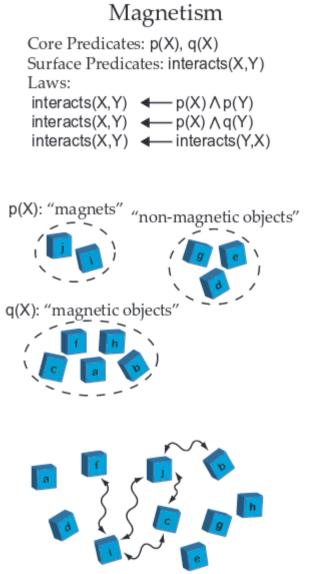
 The key feature is that the best CL encoding of some relations will extend to novel, unseen relations.



 $(succ one) \rightarrow two$ $(succ two) \rightarrow three$ $(succ three) \rightarrow four$

 $\frac{\textbf{Even-Odd}}{(\mathbb{Z}_2)}$

(even one) \rightarrow True (even two) \rightarrow True (even three) \rightarrow True (even four) \rightarrow True

(odd one) \rightarrow True (odd two) \rightarrow True (odd three) \rightarrow True (odd four) \rightarrow True 



Ullman, Goodman, & Tenenbaum (2012)

(attract	p1	p2)	\rightarrow	True
(attract	p2	p1)	\rightarrow	True
(attract	p1	n1)	\rightarrow	True
(attract	p1	n2)	\rightarrow	True
(attract	p2	n1)	\rightarrow	True
(attract	p2	n2)	\rightarrow	True
(attract	n1	n2)	\rightarrow	True
(attract	n2	n1)	\rightarrow	True
(attract	n1	p1)	\rightarrow	True
(attract	n1	p2)	\rightarrow	True
(attract	n2	p1)	\rightarrow	True
(attract	n2	p2)	\rightarrow	True

attract := ((**S S**) (**K I**)) n1 := **K** n2 := **K** p2 := (**K K**) p1 := (**K K**)

(attract	p1	p2)	\rightarrow	True
(attract	p2	p1)	\rightarrow	True
(attract	p1	n1)	\rightarrow	True
(attract	p1	n2)	\rightarrow	True
(attract	p2	n1)	\rightarrow	True
(attract	p2	n2)	\rightarrow	True
(attract	n1	n2)	\rightarrow	True
(attract	n2	n1)	\rightarrow	True
(attract	n1	p1)	\rightarrow	True
(attract	n1	p2)	\rightarrow	True
(attract	n2	p1)	\rightarrow	True
(attract	n2	p2)	\rightarrow	True
; and one	e si	ingle	e e	xample
(attract	n1	x) -	\rightarrow 1	True

```
attract := ((S S) (K I))
n1 := K
n2 := K
p2 := (K K)
p1 := (K K)
```

(attract	p1	p2)	\rightarrow	True
(attract	p2	p1)	\rightarrow	True
(attract	p1	n1)	\rightarrow	True
(attract	p1	n2)	\rightarrow	True
(attract	p2	nl)	\rightarrow	True
(attract	p2	n2)	\rightarrow	True
(attract	n1	n2)	\rightarrow	True
(attract	n2	n1)	\rightarrow	True
(attract	n1	p1)	\rightarrow	True
(attract	n1	p2)	\rightarrow	True
(attract	n2	p1)	\rightarrow	True
(attract	n2	p2)	\rightarrow	True
; and one	e si	ingl	e e	xample
(attract	n1	x) ·	\rightarrow 1	Frue

attract	$:= ((\mathbf{S} \ \mathbf{S}) \ (\mathbf{K} \ \mathbf{I}))$
nl	:= K
n2	:= K
p2	:= (K K)
p1	:= (K K)
х	:= (K K)

Dominance



A | B | C | D

$\begin{array}{c} \mathbf{Dominance} \\ (a \succ b \succ c \succ d) \end{array}$

True	:=	K			
(dom	а	b)	\rightarrow	True	
(dom	а	C)	\rightarrow	True	
; No	ir	nfoi	rma	tion a,d	relation
(dom	b	C)	\rightarrow	True	
(dom	b	d)	\rightarrow	True	
(dom	С	d)	\rightarrow	True	
(dom	b	a)	\rightarrow	True	
(dom	С	a)	\rightarrow	True	
(dom	С	b)	\rightarrow	True	
(dom	d	b)	\rightarrow	True	
(dom	d	C)	\rightarrow	True	
(dom	b	a)	\rightarrow	True	
(dom	С	b)	\rightarrow	True	
(dom	d	C)	\rightarrow	True	

 $\begin{array}{rcl} a & := & (K & (K & K)) \\ b & := & (S & (S & K)) \\ c & := & (S & K & K) \\ d & := & (K & K) \\ dom & := & (& ((S & (K & (S & (K & (S & (K & K))) & K)) & S) \\ & & \hookrightarrow) & (S & (K & (S & (K & (S & (K & K))) & K)) & S) \\ & & \hookrightarrow) & (K) \end{array}$

Formal languages

Language	Facts	Representation
$\frac{\mathbf{Regular}}{((ab)^n)}$	(a start) \rightarrow state_a (b state_a) \rightarrow accept (a accept) \rightarrow state_a (b accept) \rightarrow invalid	start := $(K (K (S K K)))$ a := $((S (S K K)) K)$ b := $((S (S K K)) ((S (K (S (K (S S (K \hookrightarrow K))) K)) S) (S K K)))$
	(a invalid) \rightarrow invalid (b invalid) \rightarrow invalid	<pre>invalid := (((S (K (S (K (S S (K K)))) K →)) S) (S K K)) (S K K)) accept := (S K K)</pre>
	(a start) \rightarrow got_a (b got_a) \rightarrow accept (a got_a) \rightarrow got_aa (b got_aa) \rightarrow want_b	
$\begin{array}{c} \textbf{Context-free}\\ (a^n b^n) \end{array}$	(b got_aa) \rightarrow want_b (b want_b) \rightarrow accept (a got_aa) \rightarrow got_aaa (b got_aaa) \rightarrow want_bb (b want_bb) \rightarrow want_b	<pre>start := S a := (S (K K)) b := (((S (K (S (K (S S (K K))) K)) S)</pre>
	(a got_aaa) \rightarrow got_aaaa (b got_aaaa) \rightarrow want_bbb (b want_bbb) \rightarrow want_bb	

Quantifiers

(start True) \rightarrow accept
(start False) \rightarrow reject
(reject True) \rightarrow accept
(reject False) \rightarrow reject
(accept True) \rightarrow accept

(accept False) \rightarrow accept

Existential $(\exists z \dots)$

Domain	Facts	Representation
Reversal	(reverse $x \ y$) \rightarrow ($y \ x$)	reverse := ((S (K (S (S K K)))) K)
If-else	True := (K K) False := K (ifelse True $x \ y$) $\rightarrow x$ (ifelse False $x \ y$) $\rightarrow y$	ifelse := ((S ((S K) S)) (S K)))
Identity	(identity x) $\rightarrow x$	identity := (S K K)
Repetition	(repeat $f(x) \rightarrow (f(f(x)))$	repeat := $((\mathbf{S} (\mathbf{K} \mathbf{S}) \mathbf{K})) (\mathbf{S} \mathbf{K} \mathbf{K})$ $\hookrightarrow)$
Recursion	$(\texttt{Y} \ f) \dashrightarrow (f \ (\texttt{Y} \ f))$	$Y := (((\mathbf{S} (\mathbf{K} \mathbf{S}) \mathbf{K}) ((\mathbf{S} ((\mathbf{S} (\mathbf{K} (\mathbf{S} (\mathbf{S} (\mathbf{K} (\mathbf{S} (\mathbf{S} (\mathbf{K} (\mathbf{S} (\mathbf{S} (\mathbf{K} (\mathbf{S} (\mathbf{S} (\mathbf{K} ($
Mutual recursion	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{rcl} \texttt{Y} \star & := & (\; (\; (\; \texttt{S} \; \; (\; \texttt{K} \; \texttt{S}) \; \; \texttt{K}) \; \; (\; (\; (\; \texttt{S} \; \; (\; \texttt{K} \; \texttt{S}) \; \; \texttt{K})) \\ & \hookrightarrow (\; (\; \texttt{S} \; \; (\; \texttt{K} \; \texttt{S}) \; \; \texttt{K})) \; \; (\; (\; \texttt{S} \; \; (\; \texttt{K} \; \; \texttt{S}) \; \; \texttt{K})) \\ & \hookrightarrow \texttt{K} \; \; (\; \texttt{S} \; \texttt{S} \; \; (\; \texttt{K} \; \texttt{S}) \; \; \texttt{K})) \; \; \texttt{S}) \; \; (\; \texttt{S} \; \; (\; \texttt{K} \; \; \texttt{S} \; (\\ \hookrightarrow \texttt{K} \; \; (\; \texttt{S} \; \texttt{S} \; \; (\; \texttt{K} \; \texttt{K}))) \; \; \texttt{K})) \; \; \texttt{S}) \; \; (\texttt{S} \; (\; \texttt{K} \; (\; \texttt{S} \; \; (\\ \hookrightarrow \texttt{S} \; \; \texttt{K} \; \texttt{K}))))) \; \; (\texttt{S} \; (\; (\; \texttt{S} \; \; (\; \texttt{K} \; \; (\texttt{S} \; \; (\\ \iff \texttt{S} \; \; \iff \texttt{K} \; \texttt{K})))))) \\ & \hookrightarrow \; (\; \texttt{S} \; \; (\; \texttt{K} \; \texttt{S}) \; \; \texttt{K})) \end{array} $
Apply	(apply $f \ x$) \rightarrow ($f \ x$)	apply = (S K K)
Tree, List	(first (pair $x \ y$)) $\rightarrow x$ (rest (pair $x \ y$)) $\rightarrow y$	pair := $(((S (K S) K) (S (K (S (K \hookrightarrow (S S (K K))) K)) S)) ((S (K \hookrightarrow (S (K (S S (K K))) K)) S)) ((S (K \hookrightarrow (K (S (K (S S (K K))) K)) S) (S \hookrightarrow (K (S (K (S S (K K))) K)) S \hookrightarrow)))first := ((S (S K K)) (K (S K)))rest := ((S (S K K)) (K K))$

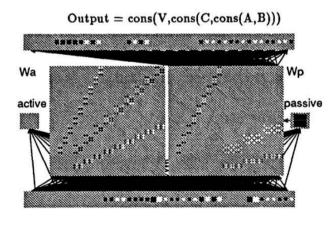
A wonderful feature of CL

• Combinatory logic requires only two simple rules:

 $(\mathbf{K} \times \mathbf{y}) \rightarrow \mathbf{x}$

 $(\mathbf{S} \times \mathbf{y} \ \mathbf{z}) \rightarrow ((\mathbf{x} \ \mathbf{z}) \ (\mathbf{y} \ \mathbf{z}))$

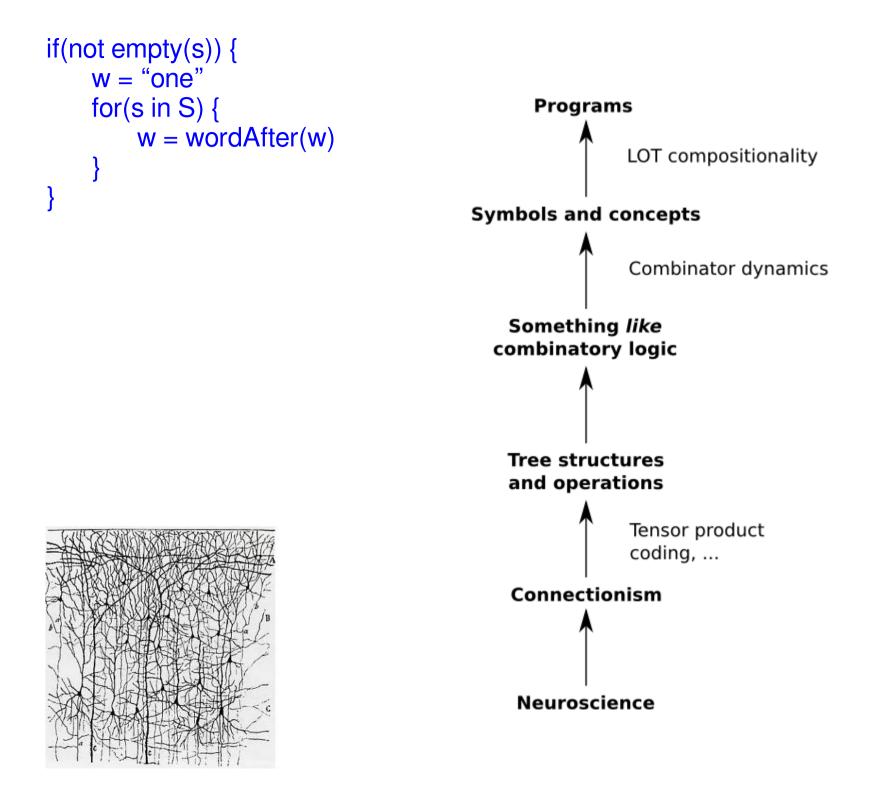
- Both are tree transformations
- Many neural implementations of trees, and simple manipulations (e.g. Smolensky's tensor product coding, Boltzcons, etc.)



Input = cons(cons(A,B),cons(cons(Aux,V),cons(by,C)))

Legendre, Miyata, Smolensky (1990)

Figure 1: Recursive tensor product network processing a passive sentence



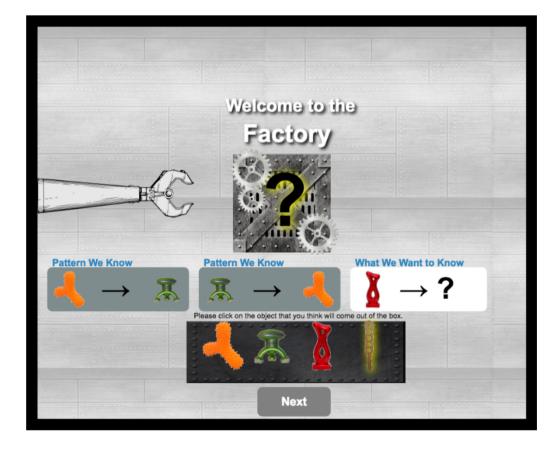
Lessons from CL

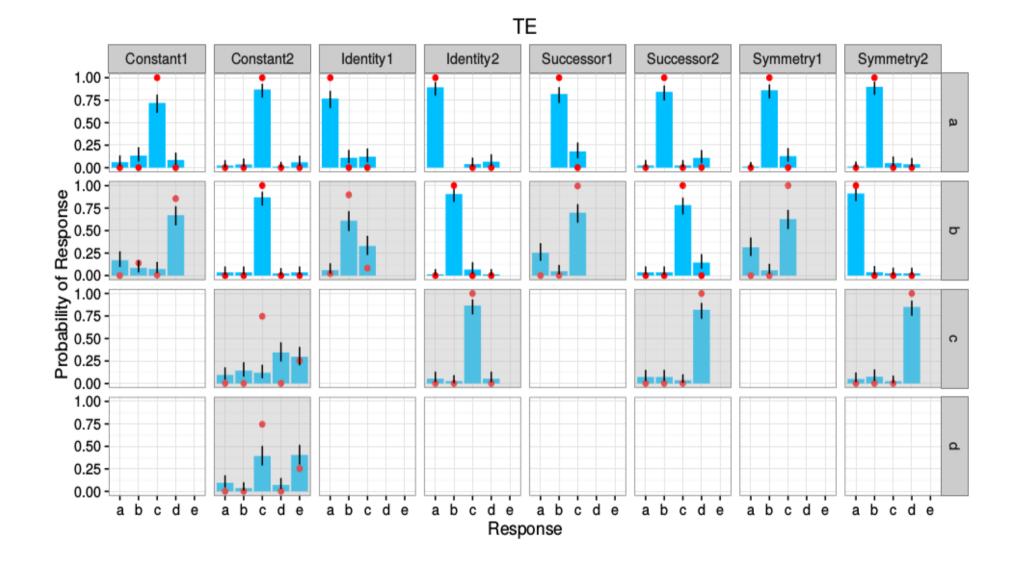
- There is a real sense in which learners can construct almost all logical representations from a primitive, dynamical basis.
- **Overarching idea:** a language for isomorphisms that is built from pieces with simple, non-cognitive dynamics (no logic, control flow, numbers, etc.)

This encoding system is:

dynamical Turing-complete symbolic sub-symbolic deductive inductive structured compositional variable-free simplicity-driven emergent parallelizable

Encoding+CL as a psychological theory





Register & Piantadosi, in prep