

Cognitive Compositional Semantics using Continuation Dependencies

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Goal: model how brains represent complex scoped quantified propositions

- ▶ Use only **cued associations** (dependencies from cue to target state)
[Marr, 1971, Anderson et al., 1977, Murdock, 1982, McClelland et al., 1995, Howard and Kahana, 2002]
(no direct implementation of unconstrained beta reduction)
- ▶ Interpret by traversing cued associations in sentence, match to memory
(assume learned traversal process, sensitive to up/down entailment)
- ▶ Despite austerity, can model scope using ‘continuation’ dependencies
- ▶ Seems to make reassuring predictions:
 - ▶ conjunct matching is easy, even in presence of quantifiers
 - ▶ quantifier upward/downward entailment (monotone incr/decr) is hard
 - ▶ disjunction is as hard as quantifier upward/downward entailment
- ▶ Empirical evaluation shows no coverage or learnability gaps
 - ▶ cognitively motivated model is about as accurate as state of art

Background: why dependencies?

Model connections in associative memory w. matrix [Anderson et al., 1977]:

$$v = M u \quad (1)$$

$$(M u)_{[i]} \stackrel{\text{def}}{=} \sum_{j=1}^J M_{[i,j]} \cdot u_{[j]} \quad (1')$$

Build cued associations using outer product [Marr, 1971]:

$$M_t = M_{t-1} + v \otimes u \quad (2)$$

$$(v \otimes u)_{[i,j]} \stackrel{\text{def}}{=} v_{[i]} \cdot u_{[j]} \quad (2')$$

Merge results of cued associations using pointwise / diagonal product:

$$w = \text{diag}(u) v \quad (3)$$

$$(\text{diag}(v) u)_{[i]} \stackrel{\text{def}}{=} v_{[i]} \cdot u_{[i]} \quad (3')$$

Background: why dependencies?

Dependency relations with label ℓ_j from u_j to v_j can be stored as vectors r_j :

$$R \stackrel{\text{def}}{=} \sum_i v_i \otimes r_i \quad (4a)$$

$$R' \stackrel{\text{def}}{=} \sum_i r_i \otimes \ell_i \quad (4b)$$

$$R'' \stackrel{\text{def}}{=} \sum_i r_i \otimes u_i \quad (4c)$$

And retrieved/traversed using accessor matrices R, R', R'' [Schuler, 2014]:

$$v_i \approx R \text{diag}(R' \ell_i) R'' u_i \quad (5)$$

This cue sequence can be simplified as dependency function:

$$v_i = (\mathbf{f}_{\ell_i} u_i) \quad (6)$$

Background: predications and graph matching

Dependencies can combine into *predications* [Copestake et al., 2005]:

$$(f \ u \ v_1 \ v_2 \ v_3 \ \dots) \Leftrightarrow (f_0 \ u)=v_f \wedge (f_1 \ u)=v_1 \wedge (f_2 \ u)=v_2 \wedge (f_3 \ u)=v_3 \wedge \dots \quad (7)$$

For example:

$$(\text{CONTAIN } u \ v_1 \ v_2) \Leftrightarrow (f_0 \ u)=v_{\text{CONTAIN}} \wedge (f_1 \ u)=v_1 \wedge (f_2 \ u)=v_2 \quad (8)$$

Dependencies incrementally matched to memory during comprehension:

$$v_t = R \ R'' \ v_{t-1} \quad (9a)$$

$$A_t = A_{t-1} + R \ \text{diag}(R' \ R'^T \ R'' \ v_{t-1}) \ R'' \ A_{t-1} \ v_{t-1} \otimes v_t \quad (9b)$$

(or reverse, during production).

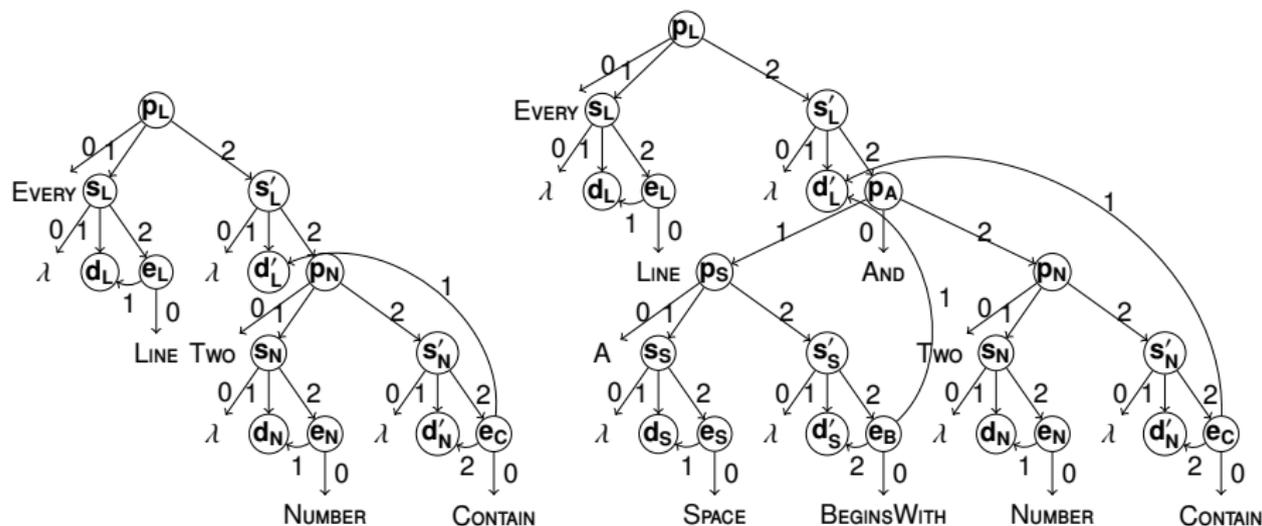
Need conditional traversal for entailment [MacCartney and Manning, 2009].

Scoped quantified predications: 'direct' style

Can implement a 'direct' semantics based on lambda calculus [Koller, 2004]:

$(\text{EVERY } p_L s_L s'_L) \wedge (\text{SET } s_L d_L e_L) \wedge (\text{LINE } e_L d_L) \wedge (\text{SET } s'_L d'_L p_N) \wedge$

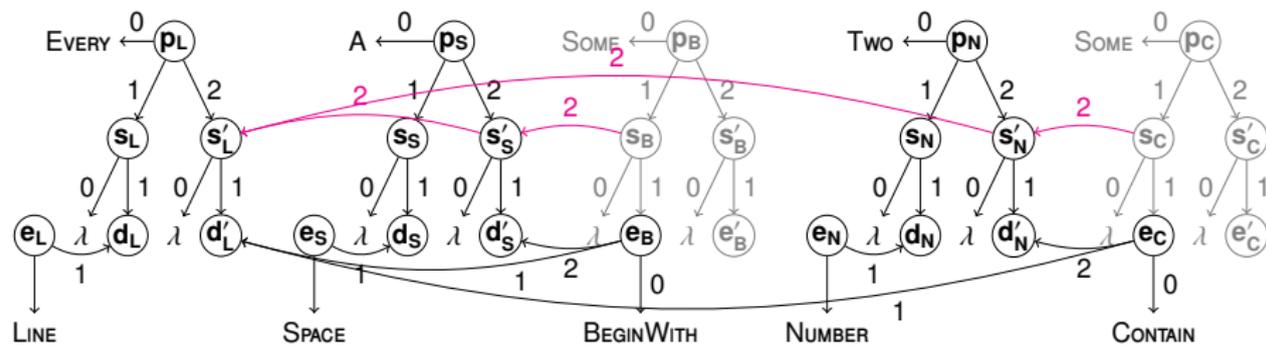
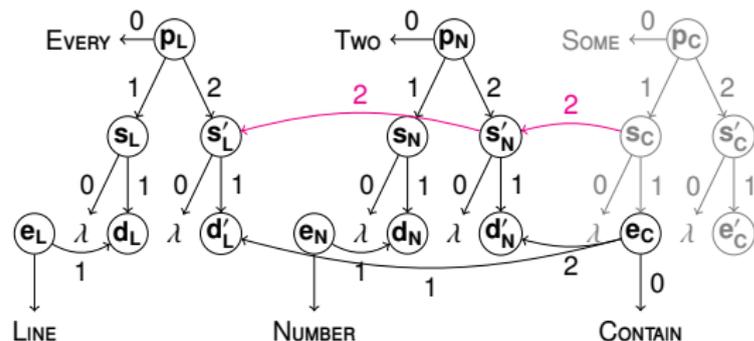
$(\text{TWO } p_N s_N s'_N) \wedge (\text{SET } s_N d_N e_N) \wedge (\text{NUMBER } e_N d_N) \wedge (\text{SET } s'_N d'_N e_C) \wedge (\text{CONTAIN } e_C d'_L d'_N)$



Hard to learn to match conjunct (left) in conjoined representation (right).

Scoped quantified predications: 'continuation' style

Change redundant dependency '2' at lambdas to instead point up to context:



Upward dependencies look like 'continuation-passing' style [Barker, 2002].

Bestiary of referential states

Set referents are now context-sensitive. . .

- ▶ **ordinary discourse referents** $d \in D$ [Karttunen, 1976]:
 - ▶ referents with no arguments
- ▶ **eventualities** $e \in E$ [Davidson, 1967, Parsons, 1990]:
 - ▶ referents with beginning, end, duration
 - ▶ one argument for each participant, ordered arbitrarily
- ▶ **reified sets or groups** $s \in S$ [Hobbs, 1985]:
 - ▶ referents with cardinalities, can be co-referred by plural anaphora
 - ▶ has **iterator** argument d_1
 - ▶ has **scope** argument s_2 , sim. to continuation parameters [Barker, 2002]
 - ▶ has **superset** argument s_3 specifying superset
- ▶ **propositions** $p \in P$ [Thomason, 1980]:
 - ▶ referents that can be believed or doubted
 - ▶ form of generalized quantifier [Barwise and Cooper, 1981]
 - ▶ has **restrictor** argument s_1
 - ▶ has **nuclear scope** argument s_2

Translation to lambda calculus

Lambda calculus terms Δ can be derived from predications Γ :

- Initialize Δ with lambda terms (sets) that have no outscoped sets in Γ :

$$\frac{\Gamma, (\text{SET } s \ i \ _); \Delta}{\Gamma, (\text{SET } s \ i \ _); (\lambda_i \text{ TRUE}), \Delta} \quad (\text{SET } _ _ \ s \ _) \notin \Gamma$$

- Add constraints to appropriate sets in Δ :

$$\frac{\Gamma, (f \ i_0 \ .. \ i \ .. \ i_N); (\lambda_i \ o), \Delta}{\Gamma; (\lambda_i \ o \wedge (\mathbf{h}_f \ i_0 \ .. \ i \ .. \ i_N)), \Delta} \quad i_0 \in E$$

- Add constraints of supersets as constraints on subsets in Δ :

$$\frac{\Gamma, (\text{SET } s \ i \ _), (\text{SET } s' \ i' \ s'' \ s); (\lambda_i \ o \wedge (\mathbf{h}_f \ i_0 \ .. \ i \ .. \ i_N)), (\lambda_{i'} \ o'), \Delta}{\Gamma, (\text{SET } s \ i \ _), (\text{SET } s' \ i' \ s'' \ s); (\lambda_i \ o \wedge (\mathbf{h}_f \ i_0 \ .. \ i \ .. \ i_N)), (\lambda_{i'} \ o' \wedge (\mathbf{h}_f \ i_0 \ .. \ i' \ .. \ i_N)), \Delta}$$

- Add quantifiers over completely constrained sets in Δ :

$$\frac{\Gamma, (\text{SET } s \ i \ _), (f \ p \ s' \ s''), (\text{SET } s' \ i' \ s \ _), (\text{SET } s'' \ i'' \ s' \ s'); (\lambda_i \ o), (\lambda_{i'} \ o'), (\lambda_{i''} \ o''), \Delta}{\Gamma, (\text{SET } s \ i \ _); (\lambda_i \ o \wedge (\mathbf{h}_f (\lambda_{i'} \ o') (\lambda_{i''} \ o''))), \Delta} \quad \begin{array}{l} p \in P, (f' \ .. \ i' \ ..) \notin \Gamma, \\ (f'' \ .. \ i'' \ ..) \notin \Gamma. \end{array}$$

For example: $(\text{EVERY } (\lambda_{d_L} \text{ SOME } (\lambda_{e_L} \text{ BEINGALINE } e_L \ d_L))$
 $(\lambda_{d'_L} \text{ TWO } (\lambda_{d_N} \text{ SOME } (\lambda_{e_N} \text{ BEINGANUM } e_N \ d_N))$
 $(\lambda_{d'_N} \text{ SOME } (\lambda_{e_H} \text{ HAVING } e_H \ d'_L \ d'_N))))$

Predictions

This model makes reassuring predictions (to be evaluated in future work)...

- ▶ Conjunct matching is easy, automatic, learned early.
Evidence: errors until about 21 months [Gertner and Fisher, 2012].
- ▶ Upward/downward entailment on 1st/2nd argument is much harder:
More than two perl scripts work. ⊢ *More than two scripts work.*
Fewer than two scripts work. ⊢ *Fewer than two perl scripts work.*
Not simple matching; speaker must learn conditional matching rules.
Evidence: 'quantifier spreading' [Inhelder and Piaget, 1958, Philip, 1995]
(children until ~10yrs don't reliably constrain restrictor with noun, etc.).
- ▶ Disjunction is similarly difficult:
Every line begins with at least 1 space or contains at least 2 dashes.
Can be translated to conjunction using de Morgan's law:
No line begins with less than 1 space and contains less than 2 dashes.
Yields downward-entailing quantifiers, requiring conditional matching.
- ▶ Other phenomena? Evaluation shows no coverage/learnability gaps.

Dependency graph composition: lexical items

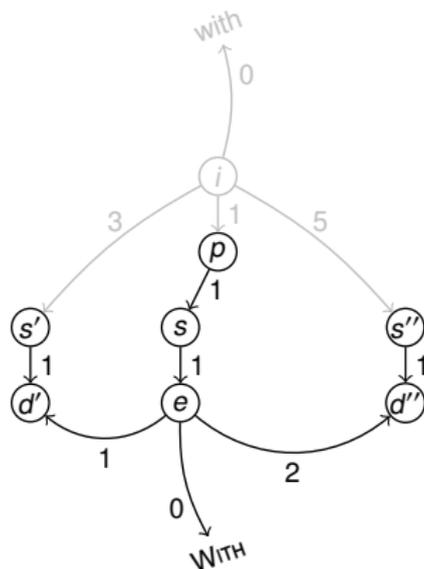
Semantics here extends categorial grammar of [Nguyen et al., 2012]...

Lexical items associate syntactic arguments with semantic arguments:

$$\begin{aligned}x &\Rightarrow u\varphi_1 \dots \varphi_n : \lambda_i (\mathbf{f}_0 i) = x \\ &\wedge (\mathbf{f}_0 (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 i)))) = x \\ &\wedge (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 i)))) = (\mathbf{f}_1 (\mathbf{f}_3 i)) \wedge \dots \\ &\wedge (\mathbf{f}_n (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 i)))) = (\mathbf{f}_1 (\mathbf{f}_{2n+1} i))\end{aligned}$$

For example:

$$\begin{aligned}\text{with} &\Rightarrow \mathbf{A-aN-bN} : \lambda_i (\mathbf{f}_0 i) = \text{with} \\ &\wedge (\mathbf{f}_0 (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 i)))) = \text{WITH} \\ &\wedge (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 i)))) = (\mathbf{f}_1 (\mathbf{f}_3 i)) \\ &\wedge (\mathbf{f}_2 (\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_1 i)))) = (\mathbf{f}_1 (\mathbf{f}_5 i)).\end{aligned}$$



Argument composition: constrain nuclear scope

Arguments apply constraints of predicates to nuclear scope of arguments:

$$d : g \quad c\text{-}ad : h \Rightarrow c : \lambda_i (g (\mathbf{f}_{2n} i)) \wedge (h i) \wedge (\mathbf{f}_{2n+1} i) = (\mathbf{f}_2 (\mathbf{f}_1, (\mathbf{f}_{2n} i))) \quad (\text{Aa})$$

$$c\text{-}bd : g \quad d : h \Rightarrow c : \lambda_i (g i) \wedge (h (\mathbf{f}_{2n} i)) \wedge (\mathbf{f}_{2n+1} i) = (\mathbf{f}_2 (\mathbf{f}_1 (\mathbf{f}_{2n} i))) \quad (\text{Ab})$$

For example:

$$\frac{\frac{\text{with}}{\mathbf{A}\text{-}\mathbf{a}\mathbf{N}\text{-}\mathbf{b}\mathbf{N} : \lambda_i (\mathbf{f}_0 i) = \text{with}} \quad \frac{\text{a number}}{\mathbf{N} : \lambda_i (\mathbf{f}_0 i) = \text{num}}}{\mathbf{A}\text{-}\mathbf{a}\mathbf{N} : \lambda_i (\mathbf{f}_0 i) = \text{with} \wedge (\mathbf{f}_0 (\mathbf{f}_4 i)) = \text{num} \wedge (\mathbf{f}_5 i) = (\mathbf{f}_2 (\mathbf{f}_1 (\mathbf{f}_4 i)))} \text{Ab}$$

Modifier composition: constrain restrictor

Modifiers apply constraints of modifier to restrictor of modificand:

$$u\text{-ad} : g \quad c : h \Rightarrow c : \lambda_j \exists_i (\mathbf{f}_2 i)=j \wedge (g i) \wedge (h j) \wedge (\mathbf{f}_3 i)=(\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_2 i))) \quad (\text{Ma})$$

$$c : g \quad u\text{-ad} : h \Rightarrow c : \lambda_j \exists_i (\mathbf{f}_2 i)=j \wedge (g j) \wedge (h i) \wedge (\mathbf{f}_3 i)=(\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_2 i))) \quad (\text{Mb})$$

For example:

$$\frac{\frac{\text{lines}}{\mathbf{N} : \lambda_i (\mathbf{f}_0 i)=\text{lines}} \quad \frac{\text{with a number}}{\mathbf{A-aN} : \lambda_i (\mathbf{f}_0 i)=\text{with ...}}}{\mathbf{N} : \lambda_i (\mathbf{f}_0 i)=\text{lines} \wedge \exists_j (\mathbf{f}_0 j)=\text{with ...} \wedge (\mathbf{f}_2 j)=i \wedge (\mathbf{f}_3 j)=(\mathbf{f}_1 (\mathbf{f}_1 (\mathbf{f}_2 j)))} \text{Mb}$$

Scope dependencies calculated (non-incrementally)

First define a partition of the set of group referents in a sentence into sets $\{s, s', s''\}$ of referents s whose iterators ($\mathbf{f}_1 s$) are connected by semantic dependencies.

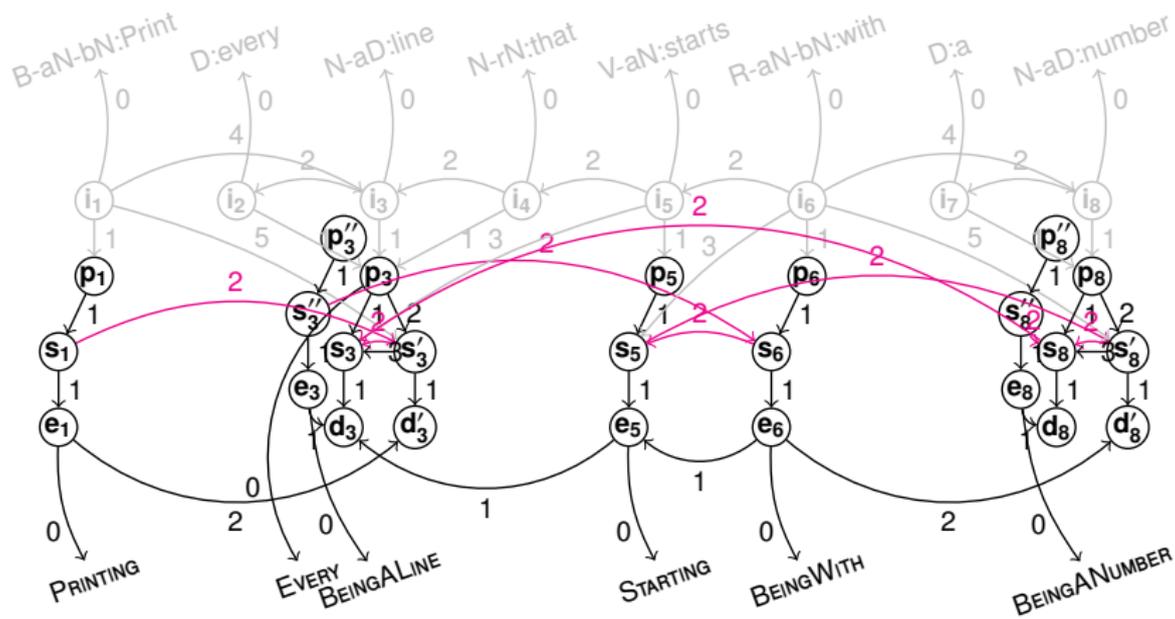
Construct scope dependencies from these partitions using a greedy algorithm:

1. start with an arbitrary referent from this partition
2. select the highest-ranked referent of that partition that is not yet attached
3. designate it as the new highest-scoping referent in that partition
4. attach it as outscoping the previous highest-scoping referent (if exists)
5. if referent has superset/subset that was not yet a highest-scoping referent:
 - ▶ switch to the partition of superset/subset referent and carry on
6. if referent has superset/subset referent that *is* the highest-scoping referent:
 - ▶ connect it to its subset/superset with a scope dependency
 - ▶ merge the two referents' partitions

Eventually you'll have one partition of connected scope dependencies.

Complete representation

Automatically generated from categorial grammar [Nguyen et al., 2012]:



Any coverage or learnability gaps?

Compare model predictions to [Manshadi and Allen, 2011] scripting corpus:

Print [₁ *every line*] *that starts with* [₂ *a number*] .

scoping relations: 1 > 2

Nice domain b/c quantifiers are frequent and natural!

350 training sentences, 94 non-duplicate test sentences.

Then introduce lexicalization into preference rankings using training data:

- ▶ bilexical weights based on frequency $\tilde{F}(h, h')$ head h' outscoped by h (e.g. *lines* often outscoped by *files*, b/c files contain multiple lines)

Per-sentence scope accuracy (perfect recall), given gold-standard parse:

System	AR
This system, w/o lexicalization	60*
[Manshadi and Allen, 2011] baseline	63
[Manshadi et al., 2013]	72
This system, w. lexicalization	72*

* statistically significant difference ($p = 0.001$ by two-tailed McNemar's test)

Lexicalized system gets about state of the art accuracy!

Cognitive compositional semantics using continuation dependencies

- ▶ seems to make reassuring predictions:
 - ▶ conjunct matching is easy, even in presence of quantifiers
 - ▶ quantifier upward/downward entailment is hard
 - ▶ disjunction is as hard as quantifier upward/downward entailment
- ▶ empirical evaluation shows no coverage or learnability gaps

Future work:

- ▶ incremental interpreter, similar to [van Schijndel and Schuler, 2013]
- ▶ this will essentially treat quantifier scope as coreference
- ▶ experiments: look for coreference-like behavior in quantifier scope

Thanks!

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