### CS447: Natural Language Processing

http://courses.engr.illinois.edu/cs447

# Lecture 16:

More on Compositional Semantics, Verb Semantics

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# Combinatory Categorial Grammar (CCG)

### **Admin**

### Midterm:

Regrade requests for midterm accepted until Nov 9 Points available on Compass. 22 points = 100%

### Project/Literature review proposals:

Due at the end of day on Monday on Compass One page PDF (in LaTeX, not Word) is sufficient Include your names and NetIDs Include all references (ideally with hyperlinks) Explain what you want to do and why.

Include a to-do list For projects: describe what resources you have or need.

(Use existing datasets, don't annotate your own data)

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2

# CCG categories

Simple (atomic) categories: NP, S, PP

### **Complex categories** (functions):

Return a result when combined with an argument

VP, intransitive verb SNP

Transitive verb (S\NP)/NP

Adverb (S\NP)\(S\NP)

Prepositions ((S\NP)\(S\NP))/NP

(NP\NP)/NP

PP/NP

# CCG categories are functions

CCG has a few atomic categories, e.g

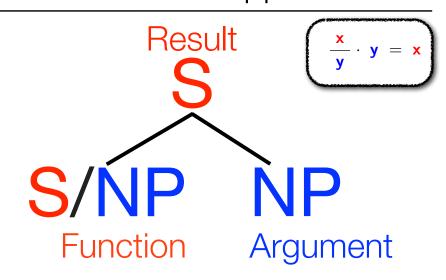
S, NP, PP

All other CCG categories are **functions**:

S / NP Result Dir. Argument

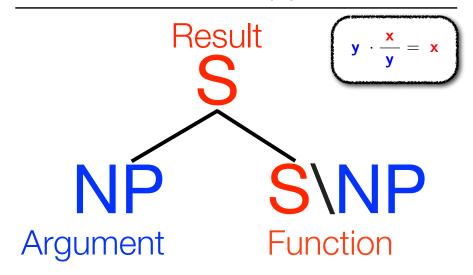
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Rules: Function application

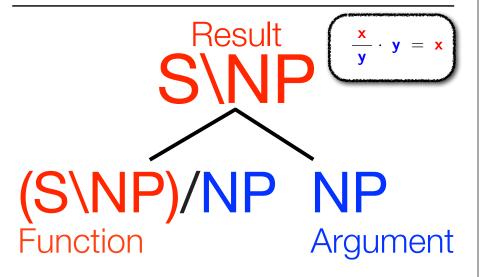


6

Rules: Function application



Rules: Function application



# A (C)CG derivation

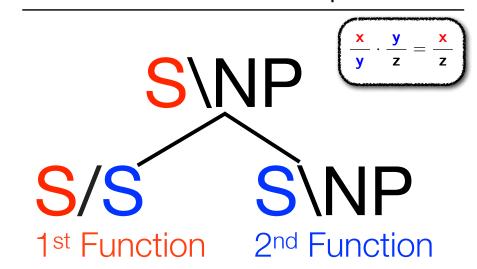
$$\begin{tabular}{lll} \hline John & eats & tapas \\ \hline \hline NP & \hline \hline \hline \hline (S \backslash NP) / NP & \hline \hline NP \\ \hline \hline \hline S \backslash NP & < \\ \hline \hline S & \\ \hline \hline \end{array}$$

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9

11

# Rules: Function Composition



Rules: Type-Raising

S/(S|P)  $y = \frac{x}{x} \cdot y = \frac{x}{(\frac{x}{y})}$ 

# Type-raising and composition

Type-raising:  $X \to T/(T\backslash X)$ 

Turns an argument into a function.

 $NP \rightarrow S/(S\NP)$  (subject)  $NP \rightarrow (S\NP)\((S\NP)/NP)$  (object)

Harmonic composition:  $X/Y \quad Y/Z \rightarrow X/Z$ 

Composes two functions (complex categories)

 $(S\NP)/PP PP/NP \rightarrow (S\NP)/NP S/(S\NP) (S\NP)/NP \rightarrow S/NP$ 

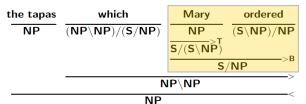
Crossing function composition:  $X/Y Y \setminus Z \rightarrow X \setminus Z$ Composes two functions (complex categories)

 $(S\NP)/S$   $S\NP$   $\rightarrow$   $(S\NP)\NP$ 

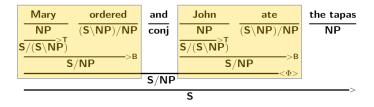
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# Type-raising and composition

### Wh-movement (relative clause):



### Right-node raising:



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13

15

# Using Combinatory Categorial Grammar (CCG) to map sentences to predicate logic

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14

# λ-Expressions

We often use **λ-expressions** to construct complex logical formulas:

- $-\lambda x. \varphi(...x...)$  is a **function** where x is a variable, and  $\varphi$  some FOL expression.
- -**β-reduction** (called λ-reduction in textbook): Apply  $\lambda x. \varphi(..x...)$  to some argument a:  $(\lambda x. \varphi(..x...) a) \Rightarrow \varphi(..a...)$  Replace all occurrences of x in  $\varphi(..x...)$  with a
- **n-ary functions** contain embedded  $\lambda$ -expressions:  $\lambda x. \lambda y. \lambda z. give(x,y,z)$

# **CCG** semantics

Every syntactic constituent has a semantic interpretation:

Every **lexical entry** maps a word to a syntactic category and a corresponding semantic type:

```
John=(NP, john') Mary=(NP, mary') loves: ((S\NP)/NP \lambda x.\lambda y.loves(x,y))
```

Every **combinatory rule** has a syntactic and a semantic part:

Function application:  $X/Y:\lambda x.f(x)$  Y:a  $\rightarrow X:f(a)$ 

Function composition:  $X/Y:\lambda x.f(x)$   $Y/Z:\lambda y.g(y) \rightarrow X/Z:\lambda z.f(\lambda y.g(y).z)$ 

Type raising:  $X:a \rightarrow T/(T\setminus X) \lambda f. f(a)$ 

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# An example with semantics

$$\begin{array}{c|c} \textit{John} & \textit{sees} & \textit{Mary} \\ \textbf{NP}: \textit{John} & \overline{(\textbf{S} \backslash \textbf{NP}) / \textbf{NP}: \lambda x. \lambda y. sees(x,y)} & \overline{\textbf{NP}: \textit{Mary}} \\ \hline & \overline{\textbf{S} \backslash \textbf{NP}: \lambda y. sees(\textit{Mary},y)} \\ \hline & \overline{\textbf{S}: sees(\textit{Mary},\textit{John})} \\ \end{array} \\ <$$

Supplementary material: quantifier scope ambiguities in CCG

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17

19

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### 18

# Quantifier scope ambiguity

"Every chef cooks a meal"

### -Interpretation A:

For every chef, there is a meal which he cooks.

$$\forall x [chef(x) \rightarrow \exists y [meal(y) \land cooks(y, x)]]$$

### -Interpretation B:

There is some meal which every chef cooks.

$$\exists y [meal(y) \land \forall x [chef(x) \rightarrow cooks(y, x)]]$$

# Interpretation A

Every	chef	cooks	a	meal	
$ \overline{(S/(S\backslash NP))/N} \lambda P\lambda Q.\forall x[Px \to Qx] $	$\overline{\mathbf{N}}$ $\lambda z.chef(z)$	$\frac{(S\backslashNP)/NP}{\lambda u.\lambda v.cooks(u,v)}$	$\frac{\overline{((S\backslashNP)\backslash((S\backslashNP)/NP))/N}}{\lambda P\lambda Q\exists y[Py\land Qy]}$	$\overline{\mathbf{N}}$ $\lambda z.meal(z)$	
$ \frac{S/(S\backslashNP)}{\lambda Q.\forall x[\lambda z.chef(z)x \to Qx]}  \equiv \lambda Q.\forall x[chef(x) \to Qx] $		$ \frac{(S \backslash NP) \backslash ((S \backslash NP) / NP)}{\lambda Q \exists y [\lambda z. meal(z) y \land Q y]} \\ \equiv \lambda Q \lambda w. \exists y [meal(y) \land Q y w] $			
			S\NP	<	

 $\lambda w.\exists y [meal(y) \land \lambda u \lambda v.cooks(u, v)yw]$  $\equiv \lambda w.\exists y [meal(y) \land cooks(y, w)]$ 

 $\mathbf{S} : \forall x [chef(x) \to \lambda w. \exists y [meal(y) \land cooks(y, w)] x] \\ \equiv \forall x [chef(x) \to \exists y [meal(y) \land cooks(y, x)]]$ 

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# Interpretation B

Every $\frac{(S/(S\backslash NP))/N}{\lambda P\lambda Q. \forall x[Px \to Qx]}$	chef N	$\frac{\text{cooks}}{(S\backslashNP)/NP}$	$\frac{a}{(S\backslash(S/NP))/N}$	meal N		
$\lambda P\lambda Q.\forall x[Px\to Qx]$	$\lambda z.chef(z)$	$\lambda u.\lambda v.cooks(u,v)$	$\lambda P \lambda Q \exists y [Py \wedge Qy]$	$\lambda z.meal(z)$		
$S/(S \setminus NF)$ $\lambda Q \forall x [\lambda z.chef(z)]$ $\equiv \lambda Q \forall x [chef(x)]$	$)x \rightarrow Qx$	>B	$ \begin{array}{c} S \setminus (S/NI) \\ \lambda Q \exists y [\lambda z.meal(\\ \equiv \lambda Q \exists y [meal(\end{aligned})] $	$z)y \wedge Qy$		
$\lambda w. \forall x [chef(x)] $ $\equiv \lambda w. \forall x [chef(x)] $	$S/NP x) \rightarrow \lambda u \lambda v. co chef(x) \rightarrow cool$	poks(u,v)wx				
$ \mathbf{S} \exists y [meal(y) \land \lambda w. \forall x [chef(x) \to cooks(y, w)] x] \\ \equiv \exists y [meal(y) \land \forall x [chef(x) \to cooks(y, x)]] $						

To summarize...

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21

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### 22

# Understanding sentences

"Every chef cooks a meal"

$$\forall x [chef(x) \rightarrow \exists y [meal(y) \land cooks(y, x)]] \\ \exists y [meal(y) \land \forall x [chef(x) \rightarrow cooks(y, x)]]$$

We translate sentences into (first-order) predicate logic.

Every (declarative) sentence corresponds to a proposition, which can be true or false.

### But...

... what can we do with these representations?

Being able to translate a sentence into predicate logic is not enough, unless we also know what these predicates mean. Semantics joke (B. Partee): The meaning of life is *life*'

Compositional formal semantics tells us how to fit together pieces of meaning, but doesn't have much to say about the meaning of the basic pieces (i.e. lexical semantics)

... how do we put together meaning representations of multiple sentences?

We need to consider discourse (there are approaches within formal semantics, e.g. Discourse Representation Theory)

... Do we really need a *complete* analysis of each sentence? This is pretty brittle (it's easy to make a parsing mistake) Can we get a more shallow analysis?

# Semantic Role Labeling/ Verb Semantics

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25

### What do verbs mean?

Verbs describe events or states ('eventualities'):

Tom broke the window with a rock.

The window broke.

The window was broken by Tom/by a rock.

We want to translate verbs to predicates.

But: a naive translation (e.g. subject = first argument, object = second argument, etc.) does not capture the differences in meaning

break(Tom, window, rock)
break(window)
break(window, Tom)
break(window, rock)

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26

# Semantic/Thematic roles

Verbs describe events or states ('eventualities'):

Tom broke the window with a rock.

The window broke.

The window was broken by Tom/by a rock.

**Thematic roles** refer to participants of these events:

Agent (who performed the action): Tom

Patient (who was the action performed on): window

Tool/Instrument (what was used to perform the action): rock

Semantic/thematic roles (agent, patient) are different from grammatical roles (subject or object).

# The inventory of thematic roles

We need to define an inventory of thematic roles

To create systems that can identify thematic roles automatically, we need to create labeled training data.

It is difficult to give a formal definition of thematic roles that generalizes across all verbs.

# PropBank and FrameNet

Proposition Bank (PropBank):

Very coarse argument roles (arg0, arg1,...), used for all verbs (but interpretation depends on the specific verb)

Arg0 = proto-agent

Arg1 = proto-patient

Arg2...: specific to each verb

ArgM-TMP/LOC/...: temporal/locative/... modifiers

### FrameNet:

Verbs fall into classes that define different kinds of frames (change-position-on-a-scale frame: rise, increase,...). Each frame has its own set of "frame elements" (thematic roles)

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29

31

# **PropBank**

agree.01 Arg0: Agreer Arg1: Proposition

Arg2: Other entity agreeing

 $[Argo\ The\ group]$  agreed  $[Arg1\ it\ wouldn't\ make\ an\ offer]$ 

[Argo John] agrees with [Arg2 Mary]

fall.01 Arg1: patient/thing falling Arg2: extent/amount fallen

Arg3: start point Arg4: end point

[Arg1 Sales] fell [Arg4 to \$251 million] [Arg1 Junk bonds] fell [Arg2 by 5%]

Semantic role labeling: Recover the semantic roles of verbs (nowadays typically PropBank-style)

Machine learning; trained on PropBank Syntactic parses provide useful information

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30

### **Diathesis Alternations**

Active/passive alternation:

Tom broke the window with a rock. (active voice)

The window was broken by Tom/by a rock. (passive voice)

Causative alternation:

Tom broke the window. ('causative'; active voice)

The window broke. ('anticausative'/'inchoative'; active voice)

Dative alternation

Tom gave the gift to Mary.

Tom gave Mary the gift.

Locative alternation:

Jessica loaded boxes into the wagon.

Jessica loaded the wagon with boxes.

### Verb classes

Verbs with similar meanings undergo the same syntactic alternations, and have the same set of thematic roles (Beth Levin, 1993)

**VerbNet** (<u>verbs.colorado.edu</u>; Kipper et al., 2008) A large database of verbs, their thematic roles and their alternations