Ling 151/Psych 156A: Acquisition of Language II

Lecture 14 Syntactic categorization II

Announcements

HW5 available (due 2/16/18)

Be working on review questions for syntactic categorization

Acquisition task syntactic categorization idea glitter unicorn Noun penguin owl kitty



Nouns behave similarly:

They can combine with certain types of words to make larger units (like Noun Phrases).

Determiner + Noun ("the kitty")

 $[NP \rightarrow Det + N]$





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They can combine with certain types of words to make larger units (like Noun Phrases).



Rule with category Noun = new phrases with words of category Noun

Categories give us expressive power

This is very handy for generating new expressions we haven't heard before.



")



Determiner + Noun ("the dax ")

 $[NP \rightarrow Det + N]$

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Determiner + Noun ("the dax ")

 $[NP \rightarrow Det + N]$

Rule with category Noun = new phrases with words of category Noun

Categories give us expressive power

Because we can do this as adults, we use this expressive power as evidence that we as adults have categories.





Determiner + Noun ("the dax ")

 $[NP \rightarrow Det + N]$

Rule with category Noun = new phrases with words of category Noun

Categories give us expressive power

This expressive power is sometimes called **productivity**.





We have many categories in human language.

Some are open-class — it's easy to add new words to them.



We have many categories in human language.

Some are open-class — it's easy to add new words to them.

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[VP \rightarrow Negation + V]
```

It's not daxing	surprise		stand
- it's dancing!	find	Verb	dance
		adore	



We have many categories in human language.

Some are open-class — it's easy to add new words to them.





We have many categories in human language.

Some are closed-class — the words in them are fixed.





We have many categories in human language.

Some are closed-class — the words in them are fixed.





There's significant debate on when these categories develop.



Some studies suggest that syntactic category knowledge may already be in place around the age of two

- Determiners (like "the"), Nouns: Valian 1986, Valian, Solt, & Stewart 2008
- Auxiliary verbs: Stromswold 1989, Rispoli, Hadley, & Holt 2009, Rissman, Legendre, & Landau 2013
- Verbs: Kowalski & Yang 2012





Other studies suggest that it may appear significantly later:

- Determiners (like "the"), Nouns: Pine & Lieven 1997, Meylan et al. 2017
- Auxiliary verbs: Wilson 2003, Theakston & Lieven 2005, Theakston, Lieven, Pine, & Rowland 2005, Theakston & Lieven 2008, Theakston & Rowland 2009
- Verbs: Tomasello 1992, Tomasello 2006





Easy to observe: When children know individual words.





Harder to observe: When children have recognized these words belong to categories.



























This causes the child to combine words of the same category with similar words, so that there's overlap in usage within a category.



How can we tell? idea didn't glitter unicorn not Negation Noun would penguin might can't wouldn't kitty could won't Auxiliary will can should

This overlap is something we can quantitatively assess to gauge productivity with respect to categories (Tomasello 1992, Pine & Lieven 1997, Naigles, Hoff, & Vear 2009, Yang 2010, 2011, 2013, Goldin-Meadow & Yang 2016).





Premise: If children's usage shows enough productivity, as measured by overlap, this suggests they have rules that are based on the more abstract symbols like *Noun*, *Verb*, *Auxiliary*, and *Negation*.



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Productive rules based on categories



This contrasts with other alternatives for how to generate these combinations.

Semi-productive rules where some words come from categories and some words don't.



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This contrasts with other alternatives for how to generate these combinations.

Non-productive rules where word combinations are just amalgams memorized directly from the input because the child doesn't have categories.







Representation options



How much overlap do we expect to see if a child has categorybased productive rules?



Representation options



For example, should we expect every verb to combine with every auxiliary?



Representation options



In fact, it turns out naturalistic linguistic output shows power law behavior (a Zipfian distribution)...


...where a few things are said very frequently...



... and most things are said very infrequently.







Bates, Pearl & Braunwald, in prep.

What we can do: Computational-level analysis of children's productions, using formal metrics that take this into account and describe how children generate their utterances given their underlying representations frequency VP \rightarrow Aux + dance rank VP \rightarrow will + dance $VP \rightarrow will + V$ VP \rightarrow Aux + V ...will dance... ...will dance... ...will dance... ...will dance...

Bates, Pearl & Braunwald, in prep.





Basic idea:

Compare the observed overlap in children's produced combinations against the expected overlap if a specific underlying representation were what children used to generate that combination.

Bates, Pearl & Braunwald, in prep.





Calculating observed overlap

This is based on the child's productions.



If a word combines with more than one lexical item (ex: a verb combining with more than one auxiliary verb), overlap for that word = 1.

...will dance... ...would dance... ...will dance... ...will dance...

overlap for *dance* = 1

Bates, Pearl & Braunwald, in prep.





Calculating observed overlap

This is based on the child's productions.



If a word combines with only one lexical item (ex: a verb combining with only one auxiliary verb), overlap for that word = 0.

...will dance... ...will dance... ...will dance... ...will dance...

overlap for *dance* = 0

Bates, Pearl & Braunwald, in prep.





Calculating observed overlap

This is based on the child's productions.



Observed overlap for words potentially from one category

total overlap from all words

total number of words

Bates, Pearl & Braunwald, in prep.





Calculating observed overlap

This is based on the child's productions.

Observed overlap for words potentially from one category



$=\frac{\text{total overlap from all words}}{\text{total number of words}}$

5

Bates, Pearl & Braunwald, in prep.





Calculating observed overlap

This is based on the child's productions.

Observed overlap for words potentially from one category



ns.

total overlap from all words total number of words

1 + 1 + 0 + 1 + 0 = 3

Bates, Pearl & Braunwald, in prep.





Calculating observed overlap

This is based on the child's productions.

Observed overlap for words potentially from one category



total overlap from all words total number of words

$$\frac{3}{5} = 0.6$$

Bates, Pearl & Braunwald, in prep.



observedAuxVerboverlap0.50.6

Underlying representations



Calculating expected overlap

This is based on what we think the child's underlying representations are.

Bates, Pearl & Braunwald, in prep.



observed Verb Aux 0.5 0.6 overlap

Calculating expected overlap

If both words come from categories, we can calculate expected overlap using a formula developed by Yang (2011) that takes into account how frequently the child produces words of these categories.



Bates, Pearl & Braunwald, in prep.



observed Verb Aux 0.5 overlap 0.6

Calculating expected overlap

If both words come from categories, we can calculate expected overlap using a formula developed by Yang (2011) that takes into account how frequently the child produces words of these categories.



$$Exp_{Prod} = \frac{\sum_{a_j \in Anch} expprod_{a_j}}{|Anch|}$$

$$expprod_{a_j} = 1 + (Unk|-1)(1-p_{a_{j_{out}}})^S - \sum_{w_i \in Unk} (p_{w_{i_{out}}} p_{a_{j_{out}}} + 1 - p_{a_{j_{out}}})^S$$

Bates, Pearl & Braunwald, in prep.



- observed Verb Aux 0.5 0.6 overlap
- Calculating expected overlap
- If both words come from categories, we can calculate expected overlap using a formula developed by Yang (2011) that takes into account how frequently the child produces words of these categories.



Bates, Pearl & Braunwald, in prep.

Underlying representations

			$VP \rightarrow Aux + V$		VP \rightarrow will + dance
		expected overlap	Aux 0.7	Verb 0.8	will dance
observed	Διιχ	Verh	$VP \rightarrow W$	vill + V	VP \rightarrow Aux + dance
overlap	0.5	0.6			

Calculating expected overlap

Bates, Pearl & Braunwald, in prep.



observed	Aux	Verb
overlap	0.5	0.6

Calculating expected overlap

If one word comes from a category and one word doesn't, we can calculate expected overlap using a formula adapted from Yang's that takes into account how frequently the child produces words of the one category and how frequently she heard words of the other category combine with it.



Bates, Pearl & Braunwald, in prep.

ntations	100

	$v_{P} \rightarrow$	VP \rightarrow will + dance	
expected overlap	Aux 0.7	Verb 0.8	will dance

observed	Aux	Verb
overlap	0.5	0.6

Calculating expected overlap

If one word comes from a category and one word doesn't, we can calculate expected overlap using a formula adapted from Yang's that takes into account how frequently the child produces words of the one category and how frequently she heard words of the other category combine with it.



 $VP \rightarrow will + V$

Bates, Pearl & Braunwald, in prep.

		0
ns		

Underlying representations

	$v_{P} \rightarrow$	VP \rightarrow will + dance	
expected overlap	Aux 0.7	Verb 0.8	will dance

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Calculating expected overlap

If one word comes from a category and one word doesn't, we can calculate expected overlap using a formula adapted from Yang's that takes into account how frequently the child produces words of the one category and how frequently she heard words of the other category combine with it.



Bates, Pearl & Braunwald, in prep.

tions	

			$VP \rightarrow Aux + V$		VP \rightarrow will + dance	
		expected overlap	Aux 0.7	Verb 0.8	will dance	
observed	Aux	Verb			VP \rightarrow will + V	
overlap	0.5	0.6			VP \rightarrow Aux + dance	

Calculating expected overlap

If one word comes from a category and one word doesn't, we can calculate expected overlap using a formula adapted from Yang's that takes into account how frequently the child produces words of the one category and how frequently she heard words of the other category combine with it.







Bates, Pearl & Braunwald, in prep.

tell? Underlying representations

> $Exp_{Not} = Obs(Phr_{Exp_{Not}})$ $Phr_{Exp_{Not}} = S \ phrases \ where$

> > $phr_s \in S = w_i a_i$

with probability $p_{w_i a_{j_{in}}}$

			VP \rightarrow	Aux + V	
		expected overlap	Aux 0.7	Verb 0.8	will dance
observed Arms		Marila	$VP \rightarrow will + V$		VP \rightarrow Aux + dance
overlap	0.5	0.6	Aux	Verb	Aux Verb
Calculating expected overlap		0.9	0.7	0.5 0.6	
					VP \rightarrow will + dance

If neither word comes from a category, we can calculate expected overlap using a formula from Yang (2010) that takes into account how frequently the child heard these words combine in the input.



...will dance...

The underlying representation whose **expected overlap best matches the observed overlap** is the most likely representation the child has.

Bates, Pearl & Braunwald, in prep.

observedAux
0.5Verb
0.6overlap0.50.6expectedAux
0.5Verb
0.6

VP \rightarrow **Aux** + dance

...will dance...

Underlying representations

$VP \rightarrow Aux + V$		VP \rightarrow will + dance	
Aux 0.7	Verb 0.8	Aux 0.8	Verb 0.7
$VP \rightarrow will + V$			
Aux 0.9	Verb 0.7		

Here, it looks like a semi-productive representation where the Aux word comes from a category while the verb doesn't is the best match.



Bates, Pearl & Braunwald, in prep.



Bates, Pearl & Braunwald, in prep.



Focus: categories in **Verb Phrases**...that appeared sufficiently frequently in this child's productions



Analyzing the utterances produced by a single American English child (L) between the ages of 20 and 24 months.

Bates, Pearl & Braunwald, in prep.



Focus: categories in **Verb Phrases**...that appeared sufficiently frequently in this child's productions



Analyzing the utterances produced by a single American English child (L) between the ages of 20 and 24 months.

Child output: 2154 verb phrases

Child input: 2184 verb phrases from her mother

Auxiliary

Negation

Bates, Pearl & Braunwald, in prep.



Focus: categories in **Verb Phrases**...that appeared sufficiently frequently in this child's productions

Verb

Noun

Child output: 2154 verb phrases

L's verb usage appears to be typical, compared against a group of <mark>93 children</mark> between 20 and 24 months from the American English CHILDES database











Bates, Pearl & Braunwald, in prep.



Utterances most compatible with having adult-like closed-class categories, but not adult-like open-class categories.



Bates, Pearl & Braunwald, in prep.



This is much earlier than when previous studies have thought children develop closed-class categories!





Focus: categories in Noun Phrases

NP → Det Noun …a penguin…

the glitter unicorn
Determiner Noun
a(n) penguin kitty



Yang 2010, 2011

Focus: categories in Noun Phrases

 $NP \rightarrow Det Noun$

...a penguin...





Data: Child-produced utterances from six American English corpora of the CHILDES database (age range 1;1 to 5;1).

First 100, 300, and 500 productions from all children to capture earliest stage of language production which should (presumably) be the least productive.







Yang 2010, 2011

 $NP \rightarrow$

Yang evaluated a representation where both words come from categories.

Focus: categories in **Noun Phrases**


The development of syntactic categories

Eacus: catagorias in Noun Phrases

Yang 2010, 2011

This contrasts with the representation where				a penguin	
neither word con category.	nes from a	De	the eterminer a(n)	idea glitter unicorn Noun penguin kitty	
$NP \rightarrow a pen$	guin	This mat	s doesn't tch as well		
	Observed Overlap	Expected Overlap	Expected Overlap		
First 100 utterances	21.8	19.6	17.2		
First 300 utterances	29.1	26.7	25.6		
First 500 utterances	34.2	32.3	30.2		

The development of syntactic categories

Yang 2010, 2011

Focus: categories in **Noun Phrases**

This suggests that for Noun Phrases, young children have created categories for both closed-class categories like Determiners and openclass categories like Nouns.



The development of syntactic categories

Yang 2010, 2011

Focus: categories in Noun Phrases

Though it's probably worth evaluating representations where one word comes from a category and the other doesn't just to make sure...



Recap: Syntactic categorization

Productivity, as measured by the lexical overlap of words for a syntactic category, is one way to assess whether children seem to have knowledge of a particular syntactic category.

Natural language use seems to have a Zipfian distribution, where many combinations are rarely (or never) heard. This can make it hard to learn, and it can also make it hard to figure out what knowledge children have.

There are formal metrics for figuring out exactly how much overlap words should have if children have particular representations in their minds, given that language use has a Zipfian distribution.

Based on these metrics, it seems like children may attain knowledge of closed-class categories like Auxiliary, Negation, and Determiner earlier than previously thought.

Questions?



You should be able to do all the review questions for syntactic categorization and all of HW5.