Psych 156A/ Ling 150: Psychology of Language Learning

Lecture 10
Grammatical Categories

Αr	nnc	nır	nce	me	ents

Homework 3 will be returned on Tuesday

Homework 4 will be assigned today, and due next Thursday (5/8/08)

Quiz 4 will be on Tuesday (5/6/08)

Grammatical Categorization

Computational Problem: Identify grammatical categories
These will tell you how words are used in the language.

"This is a DAX."



"He is sibbing."

DAX = noun

SIB = verb

Categorization: How?

How might children initially learn what categories words belong to?

Deriving Categories from Semantic Information Semantic Bootstrapping Hypothesis (Pinker 1984)

Children can initially determine a word's category by observing what kind of entity in the world it refers to.

objects, substance = noun (goblins, glitter)

action = verb (steal, sing)



Word's semantic category (meaning) is then linked to innate grammatical category knowledge (noun, verb)

Categorization: How?

How might children initially learn what categories words belong to?

Deriving Categories from Semantic Information Semantic Bootstrapping Hypothesis (Pinker 1984)

Children can initially determine a word's category by observing what kind of entity in the world it refers to.

Slight problem: hard to identify the referent in the world for words sometimes (like verbs)

"Look! He's frepping!"

frep = climb, perch, glower, grab, yell, ...?



Categorization: How?

How might children initially learn what categories words belong to?

Deriving Categories from Semantic Information Semantic Bootstrapping Hypothesis (Pinker 1984)

Children can initially determine a word's category by observing what kind of entity in the world it refers to.

Another problem: mapping rules are not perfect Ex: not all action-like words are verbs

"active", "action" action-like meaning, but they're not verbs



Categorization: How? How might children initially learn what categories words belong to? Distributional Learning Children can initially determine a word's category by observing the linguistic environments in which words appear: relative location of words in an utterance: "He likes to SIB." phonological regularities within classes of words: the, a, an = short (monosyllabic) words, simple syllables co-occurrence relations between grammatical categories: Determiner Noun (the goblin) = Determiners (a, the, an, ...) precede Nouns (goblin)

Categorization: How?



How might children initially learn what categories words belong to?

Distributional Learning (Evidence)

Children are sensitive to the distributional properties of their native language when they're born (Shi, Werker, & Morgan 1999).

7 month olds can recognize and track specific functor words (a, the, to, will...) in fluent speech (Höhle & Weissenborn 2003)

15-16 month German infants can determine novel words are nouns, based on the distributional information around the novel words (Höhle et al. 2004)

18 month English infants can track distributional information like "is...-ing" to signal that a word is a verb (Santelmann & Jusczyk 1998)

Categorization: How?

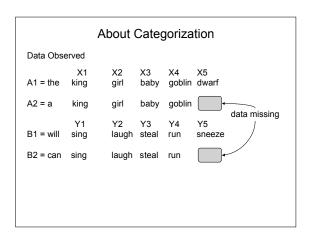


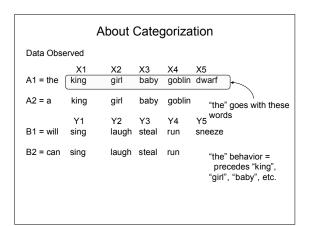
How might children initially learn what categories words belong to?

Idea (Gómez & Lakusta 2004)

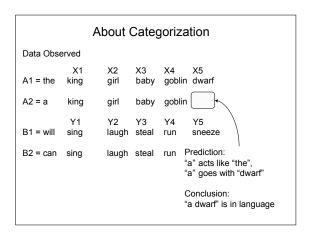
- (1) Sound properties of certain words can be tracked distributionally (monosyllabic, simple syllables = noticeable to infants).
- (2) Infants can group words together into categories based on these properties.

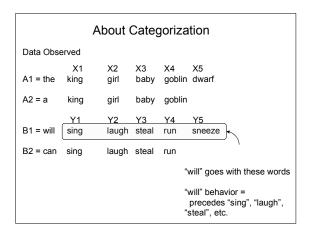
		Αŀ	oout (Cate	goriza	tion
Data Obse	erved					
A1 = the	X1 king		X2 girl	X3 baby		X5 dwarf
A2 = a	king		girl	baby	goblin	
B1 = will	Y1 sing		Y2 laugh	Y3 steal	Y4 run	Y5 sneeze
B2 = can	sing		laugh	steal	run	





		About (Cate	goriza	tion
Data Obs	erved				
A1 = the	X1 king	X2 girl	X3 baby	X4 goblin	X5 dwarf
A2 = a	king	girl	baby	goblin	"a" goes with almost all
B1 = will	Y1 sing	Y2 laugh	Y3 steal	Y4 run	Y5 the same words sneeze
B2 = can	sing	laugh	steal	run	Inference: "a" has almost the same distribution as "the", so "a" is the same category as "the"





About Categorization Data Observed A1 = theking girl baby goblin dwarf A2 = agoblin king girl baby Y2 Υ3 Y5 B1 = will sing laugh steal run sneeze "can" goes with almost sing B2 = can laugh steal run all the same words Inference: "can" has almost the same distribution as "will", so "can" is the same category as "will" **About Categorization** Data Observed A1 = the girl baby goblin dwarf A2 = aking girl baby goblin Y2 Y3 Y5 B1 = will sing laugh steal run sneeze B2 = can sing laugh steal run Prediction: "can" acts like "will" so "can" goes with "sneeze" Conclusion: "can sneeze" is in language Gómez & Lakusta 2004: Categorization Experiment Testing 12 month olds, using artificial language paradigm (so children couldn't have any experience with the categories beforehand) General procedure: Infants exposed to one of two training languages (L1 or L2). Used same set of vocabulary (all novel words). L1 generalization: a goes with X, b goes with Y (aX, bY language) L2 generalization: a goes with Y, b goes with X (aY, bX language)

L1

A1 = alt		X2 fengle				
A2 = ush	coomo	fengle	kicey	loga	paylig	wazil
	Y1	Y2	Y3	Y4	Y5	Y6
B1 = ong	deech	ghop	e jic	skige	vabe	tam
B2 = erd	deech	ghope	e jic	skige	vabe	tam



Gómez & Lakusta 2004: Categorization Experiment

LI	X1	X2	X3	X4	X5	X6	3
A1 = alt		fengle			paylig		Distribution
A2 = ush	coomo	fengle	kicey	loga	paylig	wazil	Disyllabic words
	Y1	Y2	Y3	Y4	Y5	Y6	
B1 = ong	deech	ghope	e jic	skige	vabe	tam	Monosyllabic words
B2 = erd	deech	ghope	jic	skige	vabe	tam	words

Gómez & Lakusta 2004: Categorization Experiment

L1

	X1	X2	Х3	X4	X5	X6	100
A1 = alt	coomo	fengle	kicey	loga	paylig	wazil	D: "
A2 = ush	coomo	fengle	kicey	loga	paylig	wazil	Disyllabic words
	Y1	Y2	Y3	Y4	Y5	Y6	
B1 = ong	deech	ghope	e jic	skige	vabe	tam	Monosyllabic words
B2 = erd	deech	ghope	e jic	skige	vabe	tam	Words

Association: alt/ush (a1,a2) go with these words (X1-X6) Abstraction: alt/ush (a1,a2) go with disyllabic words Categorization: alt/ush are a category whose behavior is to go with disyllabic words

L1 X2 Х3 X4 X5 X6 A1 = altcoomo fengle kicey loga paylig wazil Disvllabic A2 = ushcoomo fengle kicey loga paylig wazil Y3 Y4 Y5 Y6 B1 = ong deech ghope jic skige vabe tam Monosyllabic B2 = erd deech ghope jic skige vabe tam

Association: ong/erd (b1,b2) go with these words (Y1-Y6)
Abstraction: ong/erd (b1,b2) go with monosyllabic words
Categorization: ong/erd are a category whose behavior is to go with monosyllabic words

Gómez & Lakusta 2004: Categorization Experiment

L2

B2 = erd

Х3 X5 skige vabe tam Monosyllabic A1 = alt ghope jic A2 = ushdeech ghope jic skige vabe tam Y1 Y2 Y3 Y5 Y6 Y4 B1 = ongcoomo fengle kicey paylig wazil loga DisvIlabic

paylig wazil

coomo fengle kicey loga

Gómez & Lakusta 2004: Categorization Experiment



General procedure: Infants exposed to one of two training languages (L1 or L2). Used same set of vocabulary (all novel words).

L1 generalization: a goes with X, b goes with Y (aX, bY language) L2 generalization: a goes with Y, b goes with X (aY, bX language)

Infants exposed to new phrases from their training language L1 children: new aX, bY examples L2 children: new aY, bX examples

L1 test

X2 X3 X4 X5 A1 = alt

beevit meeper gackle roosa nawlup binnow

A2 = ush beevit meeper gackle roosa nawlup binnow

Disyllabic

Υ3 Y4 Y5 Y6 Y2

B1 = ong biff pel vot tood rud foge Monosyllabic

B2 = erd biff pel tood rud foge

The point: Children needed to complete association, abstraction, and categorization in order to realize that these new instances of aX and bY were part of the artificial language L1.

Gómez & Lakusta 2004: Categorization Experiment



L1 process

X2 ... X6 A1 = alt coomo fengle wazil

A2 = ush coomo fengle wazil

Y1 Y2 ...Y6

B1 = ong deech ghope ...tam

B2 = erd deech ghope ...tam

Gómez & Lakusta 2004: Categorization Experiment



L1 process

X1 X2 ... X6 coomo fengle wazil A1 = alt

Association A2 = ushcoomo fengle wazil

> Y1 Y2 ...Y6

Association B1 = ong deech ghope ...tam

B2 = erd Association deech ghope ...tam

L1 process

X1 X2 ... X6 coomo fengle wazil Abstraction: disyllabic words A1 = alt A2 = ush coomo fengle wazil Abstraction: disyllabic words

> Y1 Y2 ...Y6

Abstraction: monosyllabic words B1 = ong deech ghope ...tam B2 = erd deech ghope ...tam Abstraction: monosyllabic words

Gómez & Lakusta 2004: Categorization Experiment

L1 process

X2 ... X6 coomo fengle wazil A1 = alt

Categorization based on similar distribution: disyllabic words

A2 = ushcoomo fengle wazil

> ...Y6 Y1 Y2

B1 = ong deech ghope ...tam

B2 = erd deech ghope ...tam Categorization based on similar distribution: monosyllabic words

Gómez & Lakusta 2004: Categorization Experiment

L1 process

X1 X2 ... X6 coomo fengle wazil A1 = alt

Extension to new examples: alt beevit

A2 = ush coomo fengle wazil

> Y1 Y2 ...Y6

B1 = ong deech ghope ...tam

ong pel

B2 = erd deech ghope ...tam Extension to new examples:





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Results:

12 month olds listened longer to the test items that obeyed the categorizations of the language they were trained on, even though the words in the test items were ones they had never heard

This suggests that 12 month olds were able to complete association, abstraction, and categorization for this artificial language - based only on the distributional information available.

Specifically, the distributional information was the occurrence of one item next to another one in the training phase (L1: aX, bY).

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Idea: Children may be attending to other kinds of distributional information available in the linguistic environment

There is evidence that children can track information that is nonadjacent in the speech stream (Santelmann & Jusczyk 1998, Gómez 2002)

he is running

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Idea: What categorization information is available if children track frequent frames?

Frequent frame: X_

ent frame: X___Y where X and Y are words that frame another word and appear frequently in the child's linguistic environment

Examples:

the__is the king is... can___him can trick him...

the goblin is... the girl is...

can help him... can hug him...

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Data representing child's linguistic environment:
6 corpora of child-directed speech from the CHILDES database

Child Language Data Exchange System



Definition of "frequent" for frequent frames: Frames appearing a certain number of times in a give corpus (ex: 45 times).

Meant to represent the idea that the child will encounter these frames often enough to recognize them and use them for categorization.

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Trying out frequent frames on a corpus of child-directed speech.

Frame: the ___ is

"the radio is in the way...but the doll is...and the teddy is..."

radio, doll, teddy = Category1 (similar to Noun)

Frame: you ____ it

"you draw it so that he can see it... you dropped it on purpose!...so he hit you with it..."

draw, dropped, with = Category 2 (similar-ish to Verb)

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Determining success with frequent frames:

Precision = # of words identified correctly as Category within frame
of words identified as Category within frame

Recall = # of words identified correctly as Category within frame # of words that should have been identified as Category

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Determining success with frequent frames:

Precision = # of words identified correctly as Category within frame # of words identified as Category within frame

Recall = # of words identified correctly as Category within frame # of words that should have been identified as Category

Frame: you ____ it draw, dropped, with = Category 2 (similar-ish to Verb)

of words correctly identified as Verb = 2 # of words identified as Verb = 3 Precision = 2/3

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Determining success with frequent frames:

Precision = # of words identified correctly as Category within frame # of words identified as Category within frame

Recall = # of words identified correctly as Category within frame # of words that should have been identified as Category

Frame: you ____ it draw, dropped, with = Category 2 (similar-ish to Verb)

of words correctly identified as Verb = 2 # of words should be identified as Verb = many (all verbs in corpus) Recall = 2/many = small number

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Some actual results of frequent frames

put, want, do, see, take, turn, taking, said, sure, lost, like, leave, got, find, throw, threw, think, sing, reach, picked, get, dropped, seen, lose, know, knocked, hold, help, had, gave, found, fit, enjoy, eat, chose, catch, with, wind, wear, use, took, told, throwing, stick, share, sang, roll, ride, recognize, reading, ran, pulled, pull, press, pouring, pick, on, need, move, manage, make, load, liked, lift, licking, let, left, hit, hear, give, flapped, fix, finished, drop, driving, done, did, cut, crashed, change, calling, bring, break, because, banged

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Some actual results of frequent frames

Frame: the ___ is moon, sun, truck, smoke, kitty, fish, dog, baby, tray, radio, powder, paper, man, lock, lipstick, lamb, kangaroo, juice, ice, flower, elbow, egg, door, donkey, doggie, crumb, cord, clip, chicken, bug, brush, book, blanket, Mommy

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Precision & Recall of frequent frames across corpora

Precision: Above 90% for all corpora (high)

Interpretation: When a frequent frame clustered words together into category, they often did belong together. (Nouns together, verbs together, etc.)

Recall: Around 10% for all corpora (very low)

Interpretation: A frequent frame made lots of little clusters, rather than being able to cluster all the verbs together and all the nouns together.

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Getting better recall (forming one category of Verb, Noun, etc.)

Many frames overlap in the words they identify.

the is the was that is ... dog dog dog cat cat cat goblin goblin king king king teddy

What about putting clusters together that have a certain number of words in common?

Mintz 2003: Digital Children & Categorization Getting better recall (forming one category of Verb, Noun, etc.) Many frames overlap in the words they identify. the/a/that__is/was dog teddy cat goblin king girl Recall goes up to 91% (very high). Precision stays above 90% (very high) Mintz 2003: Digital Children & Categorization Summary Frequent frames are non-adjacent co-occurring words with one word in between them. They are likely to be information young children are able to track, based on experimental evidence. When tested on realistic child-directed speech, frequent frames do very well at grouping words into clusters which are very similar to actual grammatical categories like Noun and Verb. Frequent frames could be a very good strategy for children to use.

Questions?