

# Psych 156A/ Ling 150: Psychology of Language Learning

## Lecture 10 Grammatical Categories

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### Announcements

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)

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### Grammatical Categorization

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

"This is a DAX."



DAX = noun

"He is sibbing."

SIB = verb

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## 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)

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## 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!*"

*frepp* = climb, perch, glower, grab, yell, ...?



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## 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



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## 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)

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## 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 "s...-ing" to signal that a word is a verb (Santelmann & Jusczyk 1998)

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## 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.

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## About Categorization

Data Observed

|           | X1   | X2    | X3    | X4     | X5     |
|-----------|------|-------|-------|--------|--------|
| A1 = the  | king | girl  | baby  | goblin | dwarf  |
| A2 = a    | king | girl  | baby  | goblin |        |
|           | Y1   | Y2    | Y3    | Y4     | Y5     |
| B1 = will | sing | laugh | steal | run    | sneeze |
| B2 = can  | sing | laugh | steal | run    |        |

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## About Categorization

Data Observed

|           | X1   | X2    | X3    | X4     | X5     |
|-----------|------|-------|-------|--------|--------|
| A1 = the  | king | girl  | baby  | goblin | dwarf  |
| A2 = a    | king | girl  | baby  | goblin |        |
|           | Y1   | Y2    | Y3    | Y4     | Y5     |
| B1 = will | sing | laugh | steal | run    | sneeze |
| B2 = can  | sing | laugh | steal | run    |        |

data missing

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## About Categorization

Data Observed

|           | X1   | X2    | X3    | X4     | X5     |
|-----------|------|-------|-------|--------|--------|
| A1 = the  | king | girl  | baby  | goblin | dwarf  |
| A2 = a    | king | girl  | baby  | goblin |        |
|           | Y1   | Y2    | Y3    | Y4     | Y5     |
| B1 = will | sing | laugh | steal | run    | sneeze |
| B2 = can  | sing | laugh | steal | run    |        |

"the" goes with these words

"the" behavior = precedes "king", "girl", "baby", etc.

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### About Categorization

Data Observed

|           |            |             |             |              |              |
|-----------|------------|-------------|-------------|--------------|--------------|
| A1 = the  | X1<br>king | X2<br>girl  | X3<br>baby  | X4<br>goblin | X5<br>dwarf  |
| A2 = a    | king       | girl        | baby        | goblin       |              |
| B1 = will | Y1<br>sing | Y2<br>laugh | Y3<br>steal | Y4<br>run    | Y5<br>sneeze |
| B2 = can  | sing       | laugh       | steal       | run          |              |

Inference: "a" has almost the same distribution as "the", so "a" is the same category as "the"

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### About Categorization

Data Observed

|           |            |             |             |              |              |
|-----------|------------|-------------|-------------|--------------|--------------|
| A1 = the  | X1<br>king | X2<br>girl  | X3<br>baby  | X4<br>goblin | X5<br>dwarf  |
| A2 = a    | king       | girl        | baby        | goblin       |              |
| B1 = will | Y1<br>sing | Y2<br>laugh | Y3<br>steal | Y4<br>run    | Y5<br>sneeze |
| B2 = can  | sing       | laugh       | steal       | run          |              |

Prediction:  
"a" acts like "the",  
"a" goes with "dwarf"

Conclusion:  
"a dwarf" is in language

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### About Categorization

Data Observed

|           |            |             |             |              |              |
|-----------|------------|-------------|-------------|--------------|--------------|
| A1 = the  | X1<br>king | X2<br>girl  | X3<br>baby  | X4<br>goblin | X5<br>dwarf  |
| A2 = a    | king       | girl        | baby        | goblin       |              |
| B1 = will | Y1<br>sing | Y2<br>laugh | Y3<br>steal | Y4<br>run    | Y5<br>sneeze |
| B2 = can  | sing       | laugh       | steal       | run          |              |

"will" goes with these words

"will" behavior = precedes "sing", "laugh", "steal", etc.

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### About Categorization

Data Observed

A1 = the X1 king X2 girl X3 baby X4 goblin 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

"can" goes with almost all the same words

Inference: "can" has almost the same distribution as "will", so "can" is the same category as "will"

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### About Categorization

Data Observed

A1 = the X1 king X2 girl X3 baby X4 goblin 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

Prediction: "can" acts like "will" so "can" goes with "sneeze"

Conclusion: "can sneeze" is in language

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### 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)

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### Gómez & Lakusta 2004: Categorization Experiment



L1

|          | X1    | X2     | X3    | X4    | X5     | X6    |
|----------|-------|--------|-------|-------|--------|-------|
| A1 = alt | coomo | fengle | kicey | loga  | paylig | wazil |
| A2 = ush | coomo | fengle | kicey | loga  | paylig | wazil |
|          | Y1    | Y2     | Y3    | Y4    | Y5     | Y6    |
| B1 = ong | deech | ghope  | jic   | skige | vabe   | tam   |
| B2 = erd | deech | ghope  | jic   | skige | vabe   | tam   |

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### Gómez & Lakusta 2004: Categorization Experiment



L1

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

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### Gómez & Lakusta 2004: Categorization Experiment



L1

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

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

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### Gómez & Lakusta 2004: Categorization Experiment



L1

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

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### Gómez & Lakusta 2004: Categorization Experiment



L2

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

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### 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)

**Test phase:**

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

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### Gómez & Lakusta 2004: Categorization Experiment



L1 test

|          |        |        |        |       |        |        |                       |
|----------|--------|--------|--------|-------|--------|--------|-----------------------|
|          | X1     | X2     | X3     | X4    | X5     | X6     |                       |
| A1 = alt | beevit | meeper | gackle | roosa | nawlup | binnow | Disyllabic<br>words   |
| A2 = ush | beevit | meeper | gackle | roosa | nawlup | binnow |                       |
|          | Y1     | Y2     | Y3     | Y4    | Y5     | Y6     |                       |
| B1 = ong | vot    | pel    | tood   | rud   | biff   | foge   | Monosyllabic<br>words |
| B2 = erd | vot    | pel    | tood   | rud   | biff   | 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.

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### Gómez & Lakusta 2004: Categorization Experiment



L1 process

|          |       |        |            |
|----------|-------|--------|------------|
|          | X1    | 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     |

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### Gómez & Lakusta 2004: Categorization Experiment



L1 process

|          |       |        |            |             |
|----------|-------|--------|------------|-------------|
|          | X1    | X2     | ... X6     |             |
| A1 = alt | coomo | fengle | .... wazil | Association |
| A2 = ush | coomo | fengle | .... wazil | Association |
|          | Y1    | Y2     | ...Y6      |             |
| B1 = ong | deech | ghope  | ...tam     | Association |
| B2 = erd | deech | ghope  | ...tam     | Association |

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### Gómez & Lakusta 2004: Categorization Experiment



**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 before.

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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### Mintz 2003: Digital Children & Categorization



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 non-adjacent in the speech stream (Santelmann & Jusczyk 1998, Gómez 2002)

he *is* running

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### Mintz 2003: Digital Children & Categorization



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

Frequent 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                    can \_\_ him  
                 the king is...            can trick him...  
                 the goblin is...        can help him...  
                 the girl is...                can hug him...

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## Mintz 2003: Digital Children & Categorization



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

CHILDES 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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## Mintz 2003: Digital Children & Categorization



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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## Mintz 2003: Digital Children & Categorization



Determining success with frequent frames:

Precision =  $\frac{\# \text{ of words identified correctly as Category within frame}}{\# \text{ of words identified as Category within frame}}$

Recall =  $\frac{\# \text{ of words identified correctly as Category within frame}}{\# \text{ of words that should have been identified as Category}}$

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## Mintz 2003: Digital Children & Categorization



Determining success with frequent frames:

**Precision** =  $\frac{\text{\# of words identified correctly as Category within frame}}{\text{\# of words identified as Category within frame}}$

**Recall** =  $\frac{\text{\# of words identified correctly as Category within frame}}{\text{\# of words that should have been identified as Category}}$

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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## Mintz 2003: Digital Children & Categorization



Determining success with frequent frames:

**Precision** =  $\frac{\text{\# of words identified correctly as Category within frame}}{\text{\# of words identified as Category within frame}}$

**Recall** =  $\frac{\text{\# of words identified correctly as Category within frame}}{\text{\# of words that should have been identified as Category}}$

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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## Mintz 2003: Digital Children & Categorization



Some actual results of frequent frames

Frame: you \_\_\_ it  
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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## Mintz 2003: Digital Children & Categorization



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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## Mintz 2003: Digital Children & Categorization



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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## Mintz 2003: Digital Children & Categorization



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

Many frames overlap in the words they identify.

| <u>the ___ is</u> | <u>the ___ was</u> | <u>a ___ is</u> | <u>that ___ is ...</u> |
|-------------------|--------------------|-----------------|------------------------|
| dog               | dog                | dog             | cat                    |
| cat               | cat                | goblin          | goblin                 |
| king              | king               | king            | king                   |
| girl              | teddy              | girl            | teddy                  |

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

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## 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)

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## 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.

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Questions?

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