Psych156A/Ling150:
Psychology of Language Learning

Lecture 16
Learning Biases

## Announcements

Final, 6/12/08 in SH 134, 4pm-6pm:
Will be closed-note. Questions will come only from quizzes and homeworks. (No surprises.)

Final paper: Due by 6pm on 6/12/08. Hand in hard copy during final exam, or email me (Ipearl@uci.edu) in either .doc or .pdf format. Email me by next Thursday ( $5 / 29 / 08$ ) if you will be writing a final paper, and indicate which article(s) you will be writing a review of. If I do not receive email from you, I will assume you will be taking the final exam.

HW 6 assigned today, due next Thursday (5/29/08)
Quiz 6 on Tuesday (5/27/08)

Summary from last time:
Poverty of the Stimulus and Learning Strategies
Poverty of the stimulus: Children will often be faced with multiple generalizations that are compatible with the language data they encounter. In order to learn their native language, they must choose the correct generalizations.


Summary from last time:
Poverty of the Stimulus and Learning Strategies
Claim of prior (innate) knowledge: Children only seem to make the right generalization. This suggests something biases them to make that generalization over other possible generalizations. Importantly, that something isn't available in the data itself. It is knowledge they must already know to succeed at learning language.


## Summary from last time:

Poverty of the Stimulus and Learning Strategies
One Learning Bias: Experimental research on artificial languages
suggests that children prefer the more conservative generalization compatible with the data they encounter.

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Specificity of Innate Knowledge
"Innate capacities may take the form of biases or sensitivities toward particular types of information inherent in environmental events such as language, rather than a priori knowledge of grammar itself." - Seidenberg (1997)

Example: Children seem able to calculate transitional probabilities across syllables (Saffran, Aslin, \& Newport 1996)

Example: Adults seem able to calculate transitional probabilities across grammatical categories (Thompson \& Newport 2007)

## But is it always

just statistical information of some kind?

Gambell \& Yang (2006) found that tracking transitional probabilities across syllables yields very poor word segmentation on realistic English data.

Other learning strategies like the Unique Stress Constraint and algebraic learning did far better. These other learning strategies were not statistical in nature - they did not use probabilistic information.
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Peña et al. 2002: Experimental Study
Goal: examine the relation between statistical learning
mechanisms and non-statistical learning mechanisms like algebraic learning.

Adult learners' task on artificial language:
(1) word segmentation
(2) generalization about words (~categorization) $\qquad$
PURAKIBELIGATAFOOUPUFOKITALIDUBERAGA.. $\qquad$
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Peña et al. 2002: Experimental Study
The artificial language: "AXC language"
Syllables: A, X, C
Generalization:
A perfectly predicts C: A_C is a word in the language
pu_ki, be_ga, ta_du

Intervening syllable X: _ra_, _li_, _fo_
pu ra ki be li ga ta fo du pu fo ki ta li du be ra ga
PURAKIBELIGATAFODUPUFOKITALIDUBERAGA...

Peña et al. 2002: Experimental Study
The artificial language: "AXC language"
Note: transitional probability information is not informative.
Only non-adjacent syllables are informative about what words are in the language.
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pul ral ${ }^{\text {l }}$ ki be li ga ta fo du pu fo ki ta li ${ }^{\text {I }}$ du be ra ${ }^{\text {ra }}$ ga $\qquad$
PURAKIBELIGATAFODUPUFOKITALIDUBERAGA...
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Peña et al. 2002: Experimental Study
The artificial language: "AXC language" $\qquad$
Note: transitional probability information is not informative.

Only non-adjacent syllables are informative about what words are in the language. $\qquad$

pu ra ki be li galta fo du pu fo ki, ta li du be ra ga ...
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PURAKIBELIGATAFODUPUFOKITALIDUBERAGA...

Peña et al. 2002: Experimental Study
The artificial language: "AXC language" $\qquad$
Note: transitional probability information is not informative.

Only non-adjacent syllables are informative about what words are in the language.
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Prob $=1$
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First Question: Good word segmentation?

PURAKIBELIGATAFODUPUFOKITALIDUBERAGA...
10 minute familiarization period
Can adults recognize words from part-words?
Remember: transitional probability won't help - it'll bias them
the wrong way.
word: pu raki
$(\operatorname{TrProb}$ pura $)=1 / 3, \operatorname{TrProb}($ raki $)=1 / 3$,
$\operatorname{TrProb}\left(\right.$ puraki) $=\operatorname{TrPob}\left(\right.$ pura) ${ }^{*} \operatorname{TrProb}($ raki $)=1 / 3^{*} 1 / 3=1 / 9$
part-word: ra ki be
$\operatorname{TrProb}($ raki $)=1 / 3, \operatorname{TrProb}($ kibe $)=1 / 2$
$\operatorname{TrProb}($ rakibe $)=\operatorname{TrPob}(\text { raki })^{*} \operatorname{TrProb}($ kibe $)=1 / 3^{*} 1 / 2=1 / 6$
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First Question: Good word segmentation?
PURAKIBELIGATAFODUPUFOKITALIDUBERAGA... $\qquad$
$10 \mathrm{~m} . .$. PURAKIBELIGATAFODUPUFOKITALIDUBERAGA... 10 m
RAKIBE


Adults prefer real words to part-words that they actually heard. This means they can unconsciously track the non-adjacent $\qquad$ probabilities of the AXC language and identify the words.

Next Question:
Good generalization about words?

$10 \mathrm{~m} \ldots$...PURAKIBELIGATAFODUPUFOKITALIDUBERAGA $\ldots 10 \mathrm{~m}$


Adults prefer part-words that they actually heard over real words that follow the generalization about words in the language, but which they didn't actually hear. This means they can't use the non-adjacent probabilities of the AXC language to identify the words in general.

## What's going on?

PURAKIBELIGATAFODUPUFOKITALIDUBERAGA... $\qquad$ +

X PURAKIBELIGATAFODUPUFOKITALIDUBERAGA...
"We conjecture that this reflects the fact that the discovery of components of a stream and the discovery of structural regularities require different sorts of computations...the process of projecting generalizations ...may not be statistical in nature." - Peña et al. (2002)
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Prediction for Different Types of Computation
PURAKIBELIGATAFODUPUFOKITALIDUBERAGA... $)^{\circ}$

X PURAKIBE-IGATAFODIPUFOKIIALIDUBERAGA... Y
"...it is the type of signal being processed rather than the amount of familiarization that determines the type of computation in which participants will engage...changing a signal even slightly may induce a change in computation." Peña et al. (2002)

Types of computation: statistical, algebraic

10 minute familiarization period with 25 ms
(subliminal) gaps after each word
If word segmentation is already accomplished, subjects wil be free to engage their algebraic computation. This should
allow them to succeed at identifying the properties of words in the artificial language (e.g. pu_ki, be_ga, ta_du), since this kind of structural regularity is hypothesized to be found by algebraic computation.

Question:
Good generalization about words?

$10 \mathrm{~m} . . \mathrm{PURAKIBELIGATAFODUPUFOKITALIDUBERAGA} . . .10 \mathrm{~m}$


Adults prefer real words that follow the generalization about words in the language, but which they didn't actually hear, over part-words they did hear. This means they can use the nonadjacent probabilities of the AXC language to identify the
words in general. They make the structural generalization.

## Prediction: Algebraic vs. Statistical

Idea: Subjects are really using a different kind of computation (algebraic) because of the nature of the input. Specifically, the input is already subliminally segmented for them, so they don't need to engage their statistical computation abilities to accomplish that. Instead, they are free to notice more abstract properties via algebraic computation.

Prediction 1: If the words are not segmented subliminally, statistical computation will be invoked. It doesn't matter if subjects hear a lot more data. Their performance on preferring a real word they didn't hear over a part-word they did hear will not improve.

Question:
Good generalization about words?
PURAKIBELIGATAFODIPEOKIIALIDUBERAGA...
$30 \mathrm{~m} \ldots$ PURAKIBELIGATAFODUPUFOKITALIDUBERAGA ... 30 m


If given 30 minutes of training on unsegmented artificial language, adults really prefer part-words that they actually heard over real words that follow the generalization about words in the language, but which they didn't actually hear. They can't make the generalization: prediction 1 seems true.

## Prediction: Algebraic vs. Statistical

dea: Subjects are really using a different kind of computation (algebraic) because of the nature of the input. Specifically, the input is already subliminally segmented for them, so they don't need to engage their statistical computation abilities to accomplish that. Instead, they are free to notice more abstract properties via algebraic computation.

Prediction 2: If the words are segmented subliminally, algebraic computation will be invoked. It doesn't matter if subjects hear a lot less data. They will still prefer a real word they didn't hear over a part-word they did hear.

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If given 2 minutes of training on segmented artificial language, adults really prefer real words that follow the generalization about words in the language, but which they didn't actually hear, over part-words that they actually heard. The still make the generalization: prediction 2 seems true. $\qquad$

Peña et al. (2002): Summary
While humans may be able to compute powerful statistical relationships among the language data they're exposed to, $\qquad$ this may not be enough to capture all the linguistic knowledge humans come to possess.

In particular, learning structural regularities (like structural properties of words) may require a non-statistical learning mechanism, perhaps algebraic computation.

Different kinds of computation can be cued in learners based on the data at hand. Statistical computation was cued by the need to group and cluster items together. Algebraic computation was cued once items were already identified, and generalizations had to be made among the items

What kind of things can statistical computation keep track of?

Idea: "Learners might be able to compute certain types of statistical regularities, but not others." - Newport \& Aslin (2004)

What kind of non-adjacent regularities do real language actually exhibit? Maybe only these non-adjacent regularities are the kinds that humans can compute using statistical computation.

Important: AXC-syllable language (statistical regularity between 1st and 3rd syllable of the word) does not naturally occur in real languages. $\qquad$
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Naturally occurring non-adjacent regularities

Example of non-adjacent dependency: between individual segments (sounds)

Semitic languages: words built from consonantal "stems", where vowels are inserted to make different words

Arabic: k-t-b = "write"
kataba $=$ "he wrote"
yaktubu = "he writes"
kitaab = "book" maktab = "office"

Non-adjacent segment regularities: consonants

Newport \& Aslin (2004): AXCXEX segment language
p_g_t, d_k_b filler vowels: a, i, æ, o, u, e

Subject exposure time to artificial language made up of these kinds of words: 20 minutes

Result 1: Subjects were able to segment words based on non-adjacent segment regularities. $\qquad$
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Non-adjacent segment regularities: vowels

Newport \& Aslin (2004): XBXDXF segment language _a_u_e, _o_i_æ filler consonants: p, g, t, d, k, b

Subject exposure time to artificial language made up of these kinds of words: 20 minutes

Result 2: Subjects were again able to segment words based on non-adjacent segment regularities.

## Newport \& Aslin (2004): Summary

When subjects are tested with artificial languages that reflect properties real languages have (such as statistical dependencies between non-adjacent segments), they are still able to track statistical regularities.

This suggests that statistical computation is likely to be something real people use to notice the statistical regularities (non-adjacent or otherwise) that real languages have. It is not just something that will only work for the regularities that have been created in a lab setting, such as those between non-adjacent syllables in artificial languages
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