Psych 156A/ Ling 150: Psychology of Language Learning

Lecture 5 Sounds III

Announcements

Tayopa's office hours: T/Th 10:30-11:30am in SST 687

Homework 1 returned Average: 12.8 out of 16

Note on extra points: very good/funny answers will occasionally gain you an extra 1/2 point or so.

Homework 2 assigned (due next Tuesday: 4/22/08)

Speech Perception of Non-Native Sounds

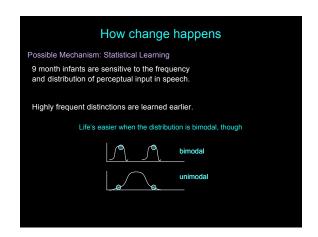
But when after 6-8 months is the ability to lost?

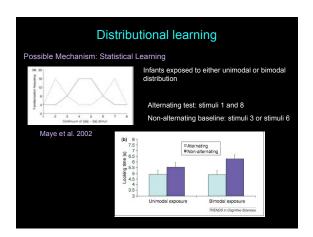
Werker & Tees (1984)

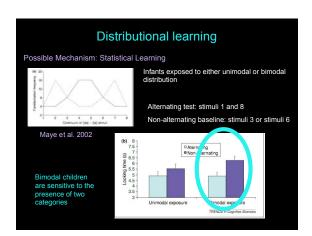
Salish & Hindi contrasts

omewhere and a sum of the sum of

Change happens somewhere around 8-10 months, depending on the sound



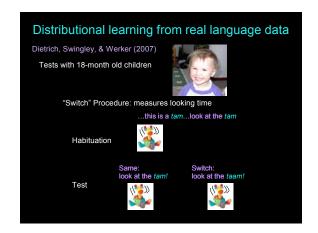


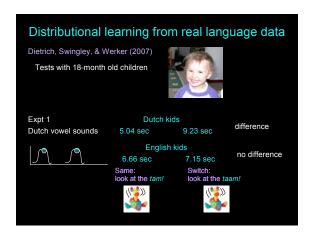


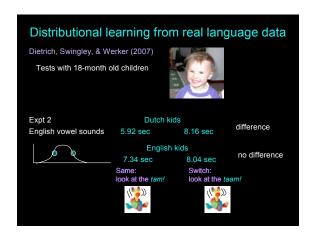
Distributional learning from real language data Dietrich, Swingley, & Werker (2007) Dutch and English vowel categories differ In English, the length of the vowel is not contrastive "cat" = "caat" In Dutch, the length of the vowel is contrastive "tam" is a different word from "taam" (Japanese also has this distinction)

Distributional learning from real language data Dietrich, Swingley, & Werker (2007)		
Dutch and English vowel sounds in the native language environment also seem to differ		
"studies suggest that differences between the long and short vowels of Dutch are larger than any analogous differences for English."		
frequency		Dutch = bimodal?
frequency		English = unimodal?
	range of sounds	

Distributional learning from real language data Dietrich, Swingley, & Werker (2007)			
Prediction if children are learning distributionally from the data:			
Dutch children interpret vowel duration as a meaningful contrast Implication: Change to vowel duration = new word			
English children should not Implication: Change to vowel duration = same word as before			
frequency			
frequency English = unimodal?			
range of sounds			







Distributional learning from real language data Dietrich, Swingley, & Werker (2007) Tests with 18-month old children Expt 3 Dutch kids difference Dutch contrastive 4.08 sec 5.72 sec vowel sounds for the Dutch kids English kids difference 6.31 sec 9.31 sec Switch: look at the tem! Same: look at the tam! ((**)) English contrastive sounds for the English Distributional learning from real language data Dietrich, Swingley, & Werker (2007) Tests with 18-month old children Expts 1, 2, & 3 Dutch kids recognize vowel durations as Dutch = bimodal contrastive English = unimodal English kids do not Native language influence Distributional learning from real language data

Distributional learning from real language data Dietrich, Swingley, & Werker (2007) Tests with 18-month old children A caveat about distributional learning "... preliminary investigation of Dutch child-directed speech indicated that the set of long and short instances formed largely overlapping distributions." Dutch = bimodal? Implication: Dutch children need other cues to help them out

Distributional learning from real language data	
Vallabha, McClelland, Pons, Werker, & Amano (2007)	
Tests with computational models (digital children)	
Distributional learning from real language data	
Vallabha, McClelland, Pons, Werker, &	
Amano (2007) Tests with computational models	
(digital children)	
Sounds: Vowel contrasts in English and Japanese English contrasts: contrast in quality (tense vs. lax) and a bit in duration	
/i/ vs. /i/ /ɛ/ vs. /e/ "ih" "ee" "eh" "ey"	
Japanese contrasts: contrast almost solely in duration (short vs. long) fi/ vs. /i:/ /e/ vs. /e:/ "ee" "eece" "ey" "eecy"	
Distributional learning from real language data	
Vallabha, McClelland, Pons, Werker, & Amano (2007)	
Tests with computational models (digital children)	
Data (input to model): Infant-directed speech of English and Japanese mothers	
Why? Idea = "motherese" makes differences more salient	
Learning algorithm: learns from a single data point at a time, trying to identify how many categories should be formed from the data points and how the categories should cover the acoustic sound space	

Distributional learning from real language data Vallabha, McClelland, Pons, Werker, & Amano (2007) Tests with computational models (digital children) Estimating how many categories from observation of the data points: probabilistic learning Hypotheses about how many categories exist are assigned probability based on how likely they are to have generated the observed data Distributional learning from real language data Vallabha, McClelland, Pons, Werker, & Amano (2007) Tests with computational models (digital children) Hypothesis probability raised is: 2 categories obability of generating data seen Hypothesis probability lowered Distributional learning from real language data Vallabha, McClelland, Pons, Werker, & Amano (2007) Tests with computational models (digital children) MM Trained on 50,000 data points Tested on 2,000 data points

Distributional learning from real language data Vallabha, McClelland, Pons, Werker, & Amano (2007) Tests with computational models (digital children) MM Results: About 92% successful categorization on average when learning from only a single speaker. 11/11 (0) Distributional learning from real language data Vallabha, McClelland, Pons, Werker, & Amano (2007) Tests with computational models (digital children) MM Results: One issue is that there is substantial variation even between speakers of the same language. 11/11 Testing on data from multiple English speakers and multiple Japanese speakers gave lower success rates (1988) OSEO. English = 69% Japanese = 77% Distributional learning from real language data Vallabha, McClelland, Pons, Werker, & Amano (2007) Tests with computational models (digital children) MM ZAN But speakers are able to categorize sounds from multiple speakers without Implication: Children need to learn from more than one speaker (not just their mother) in order to be able to generalize 11/11

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Announcements	
Quiz 2 on Thursday (4/17/08)	