

Distributional learning of L2 phonological categories by listeners with different language backgrounds

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Background

Question: How do people learn a second language?

- Generally hard, success depends on many factors
- Previous research:
 - ✦ focus on explaining *difficulties* in L2 learning
 - ✦ well-studied: L1 as a source of *interference*
- E.g., non-native speech perception & phonetic category learning
(Best, 1995; Best & Tyler, 2007; Flege, 1995; Kuhl & Iverson, 1995)

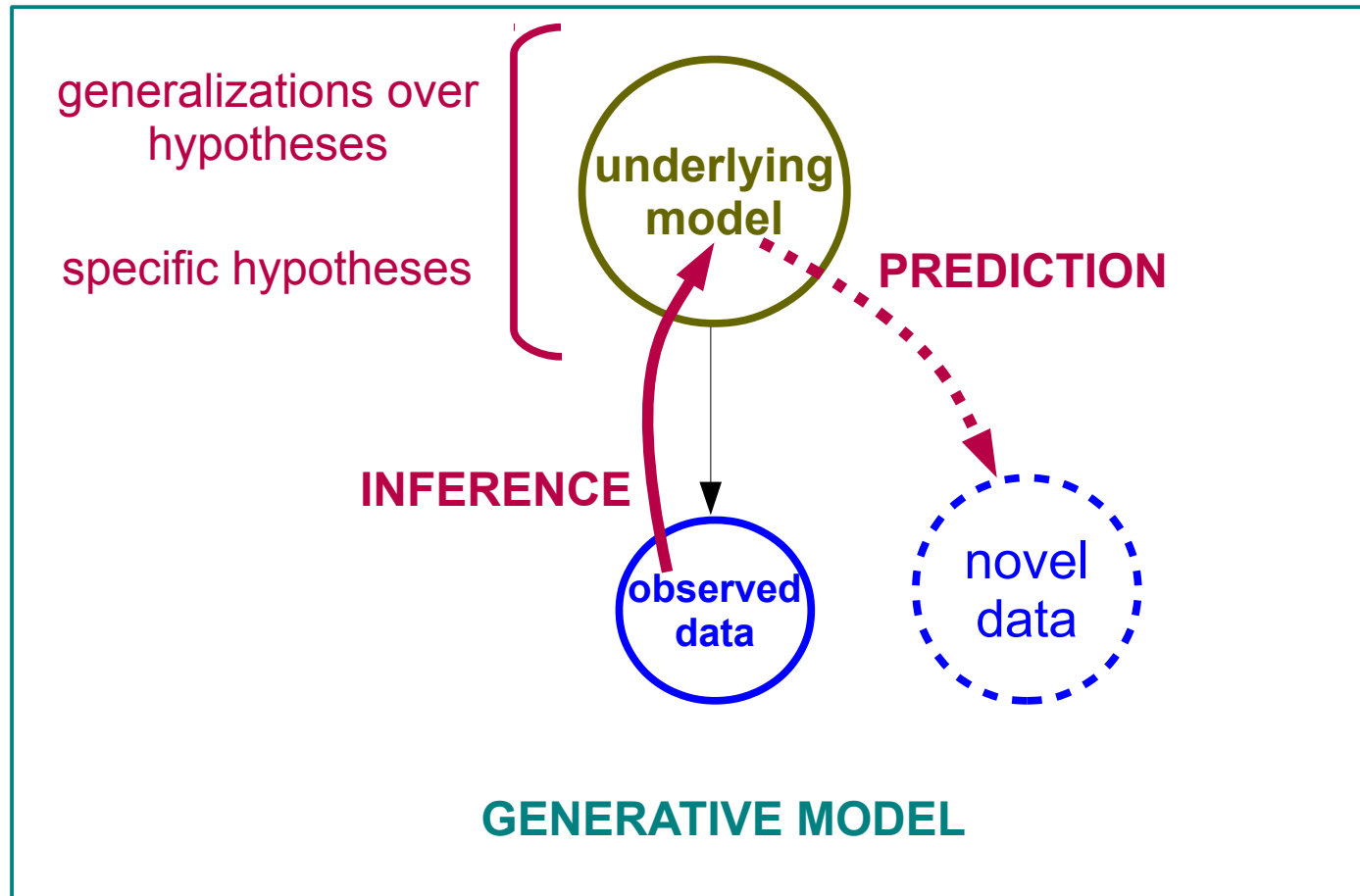


Questions

- **Less studied:** potential *benefits* of L1
 - ✦ known direct benefits: when L1 & L2 have very similar properties, e.g. cognates (e.g., Tréville, 1996)
- General learning literature:
 - ✦ people are very good at making generalizations and predictions based on current knowledge (e.g., Tenenbaum et al., 2011)
- **Our questions:**
 - ✦ Is there any evidence of *generalization* from L1 about L2 phonetic categories?
 - ✦ If so, how do these generalizations interact with the *distributional cues* from L2 input?

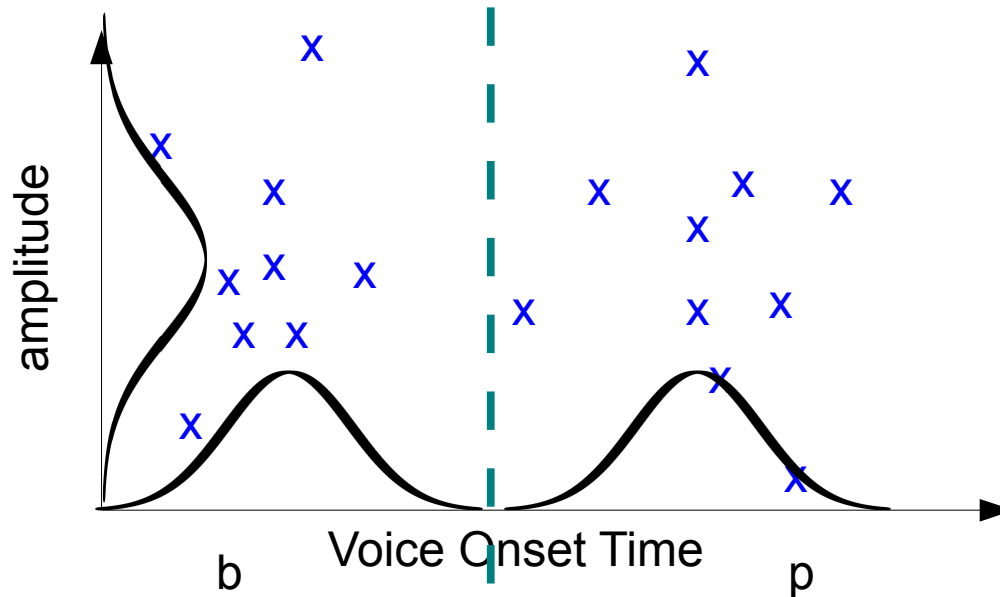
Framework

- Learning as hypothesis construction and testing
(e.g., Gerken 2010, Tenenbaum & Griffiths 2001, Tenenbaum et al. 2011, Xu & Tenenbaum 2007)



Proposed model: assumptions

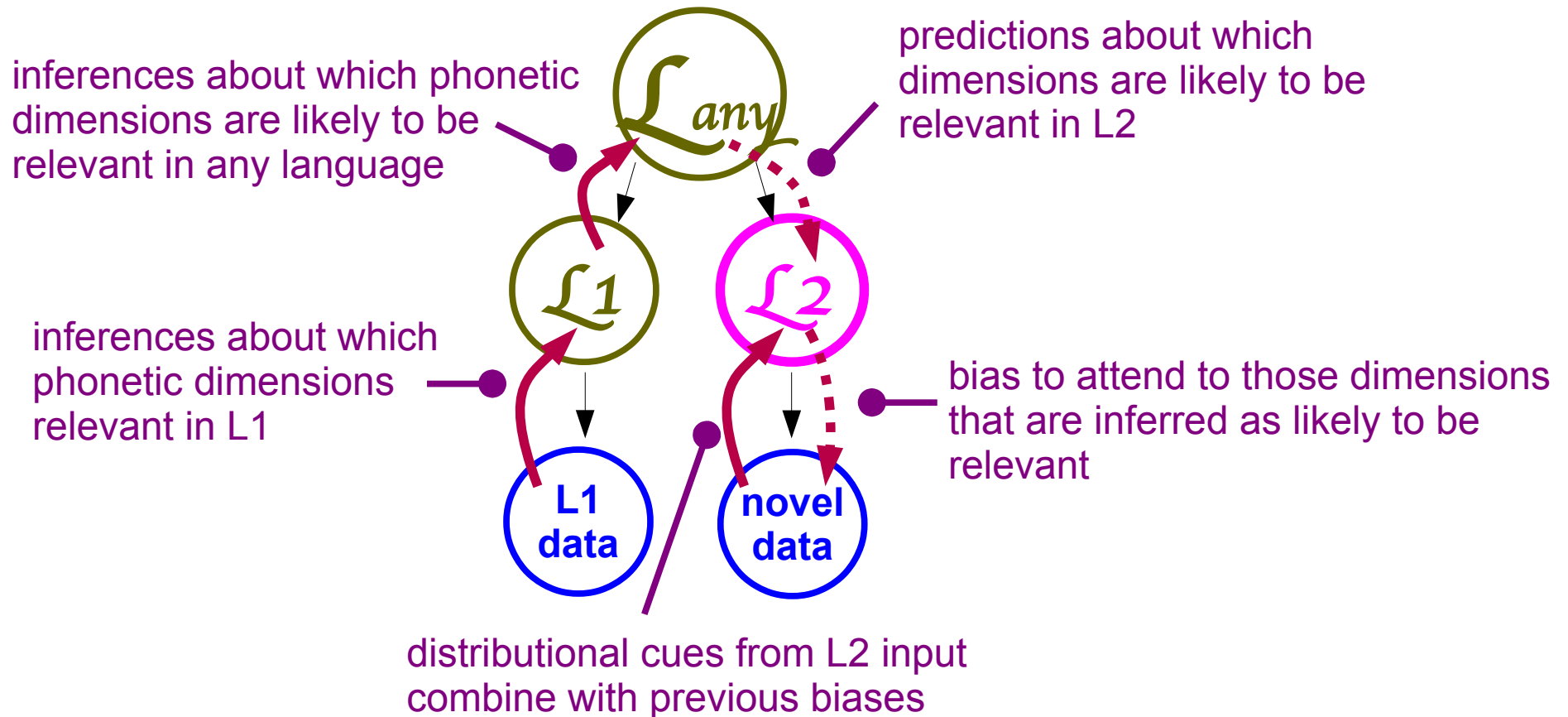
- Sounds vary along multiple acoustic-phonetic dimensions



- Categorization of sounds achieved through learning from distributional information to:
 - ◆ attend to relevant phonetic dimensions
 - ◆ disregard other dimensions

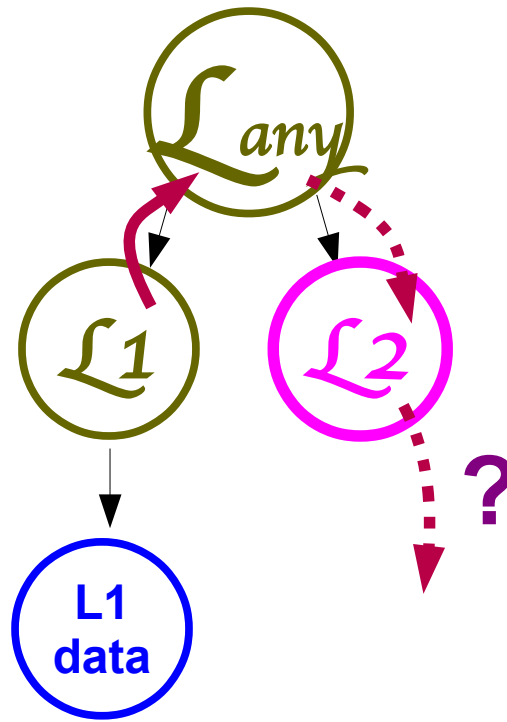
(Kuhl et al., 1992; Jusczyk, 1992; Maye, Werker, & Gerken, 2002; Nitttrouer & Miller 1997)

Proposed model: phonetic category learning



Pająk, 2010; in prep.

- Is there generalization from L1 in non-native speech perception for naïve listeners?



Pajał, 2010; in prep.

- Task: discriminating between words in a novel language
minimal pairs differing in:

- consonant **length**

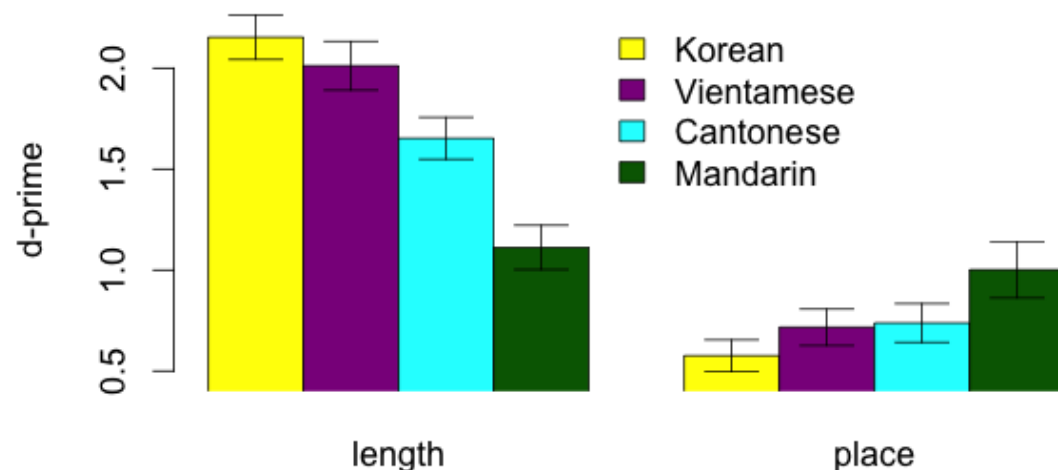
kena - kenna

- alveolo-palatal vs. retroflex **place** (w/o following vowel transition cue)

keəa - keša

Participants (n=96)	length short and long sounds	place alveolo-palatal and retroflex sounds
Korean speakers	✓	
Vietnamese speakers	✓	
Cantonese speakers	✓	
Mandarin speakers		✓

only for vowels!

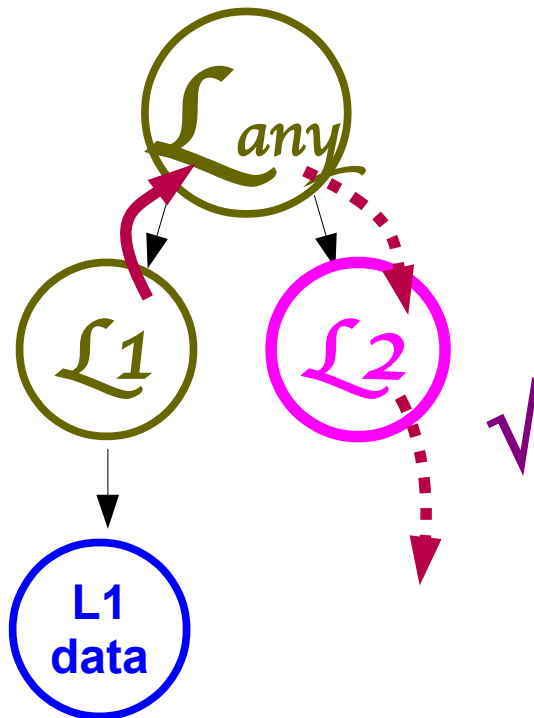


**There is generalization across segments
– even from vowels to consonants!**

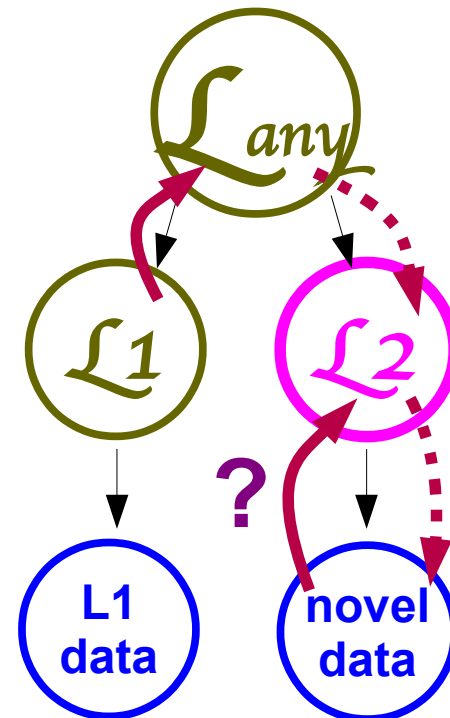
Experiment

- **Pajak, 2010; in prep.:** evidence of generalized L1 biases in perception of novel sounds
- **Current experiment:** do L1 biases change learners' interpretation of distributional cues from novel language input?

Pajak, 2010; in prep.

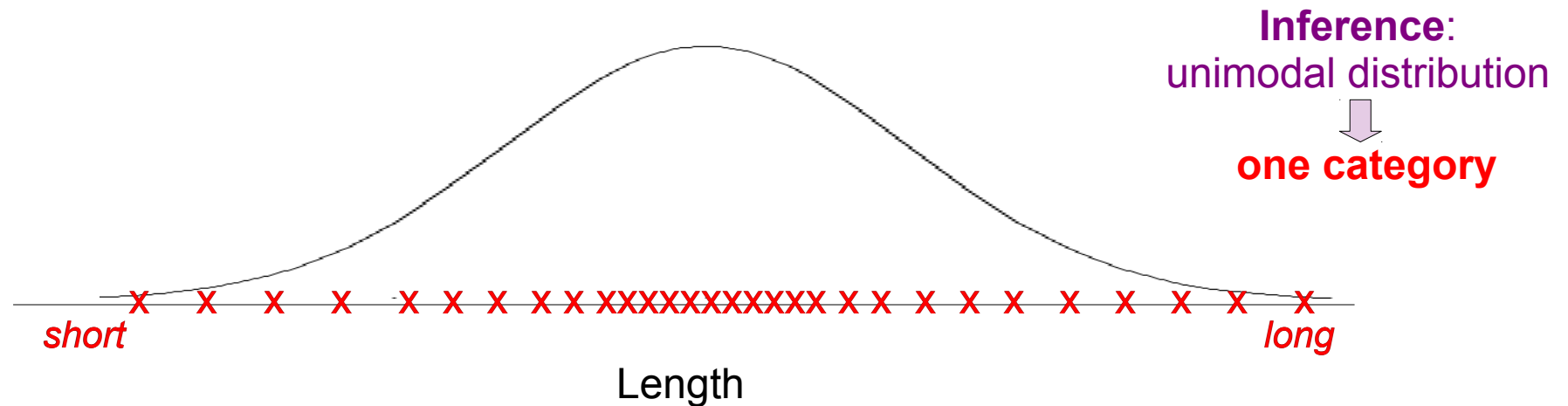
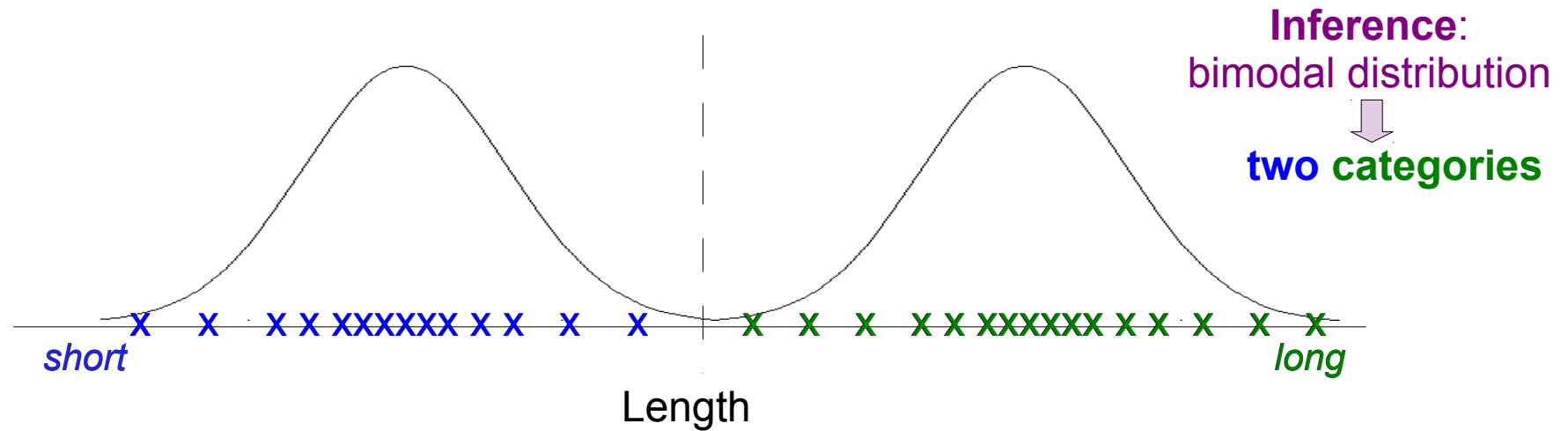


Current experiment



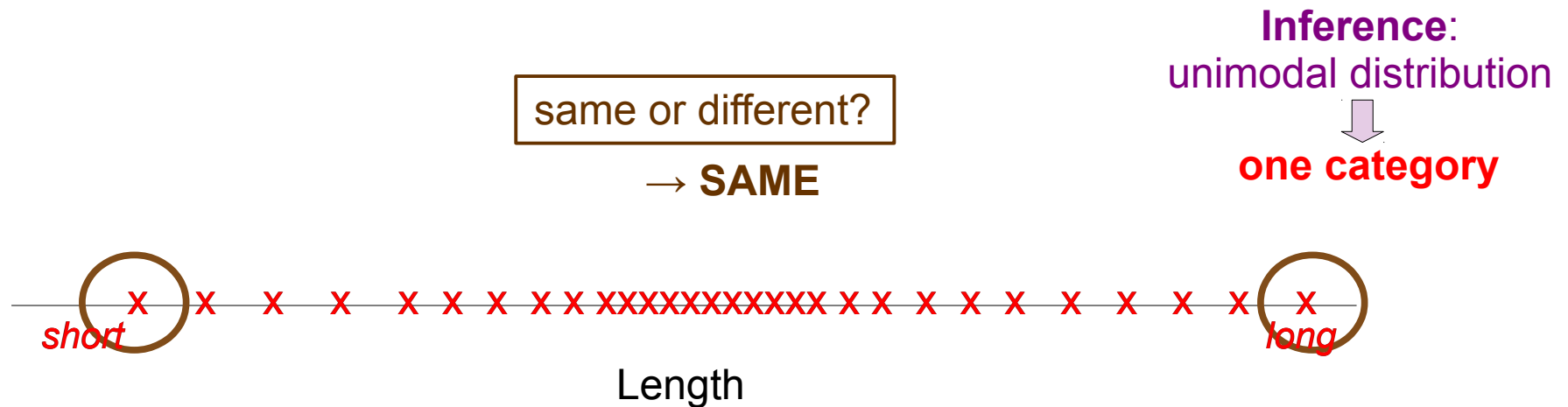
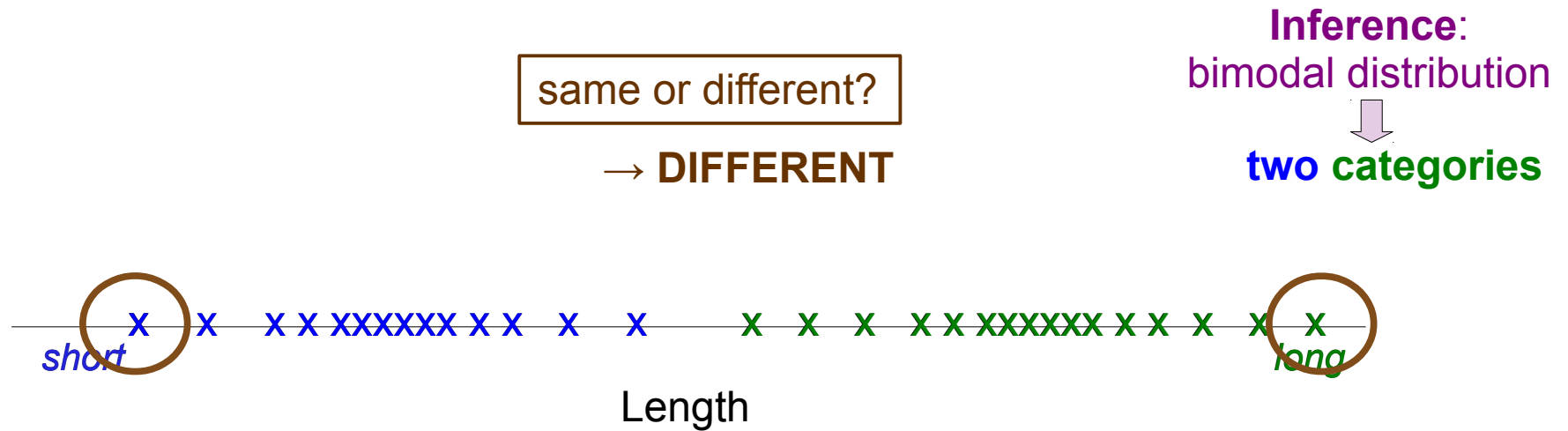
Experiment: paradigm

- **Distributional learning paradigm** (Maye & Gerken 2000, Maye, Werker, & Gerken 2002)
 - ◆ a tool to probe learners' interpretation of distributional cues in novel input



Experiment: paradigm

- Distributional learning paradigm (Maye & Gerken 2000, Maye, Werker, & Gerken 2002)
 - ◆ a tool to probe learners' interpretation of distributional cues in novel input



Experiment: instructions

- You'll be learning (sounds from) a new language.
- First, you'll listen to words from that language.
- Then, you'll hear pairs of words and, based on what you learned, decide whether these are two different words or the same word repeated twice.
- The same word can be pronounced a bit differently (e.g., with different intonation) – follow your intuition in deciding what counts as 'different' in this language.

Experiment: ppts & overview

- 58 Korean speakers & 53 Mandarin speakers

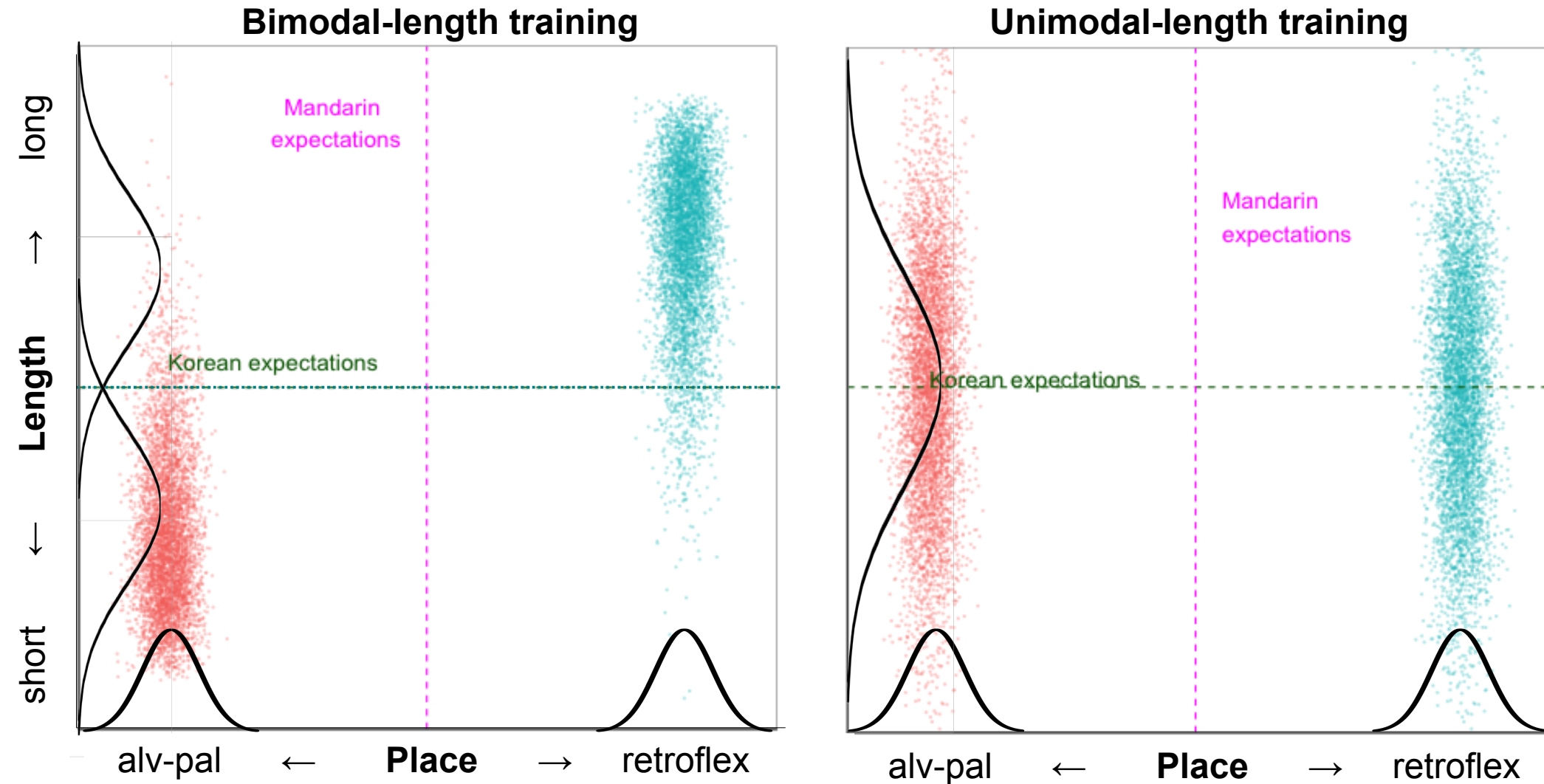
	length short and long sounds	place alveolo-palatal and retroflex sounds
Korean	✓	
Mandarin		✓

example → [n] - [nn] [ɕ] - [ʂ]

(Pinyin: x and sh)

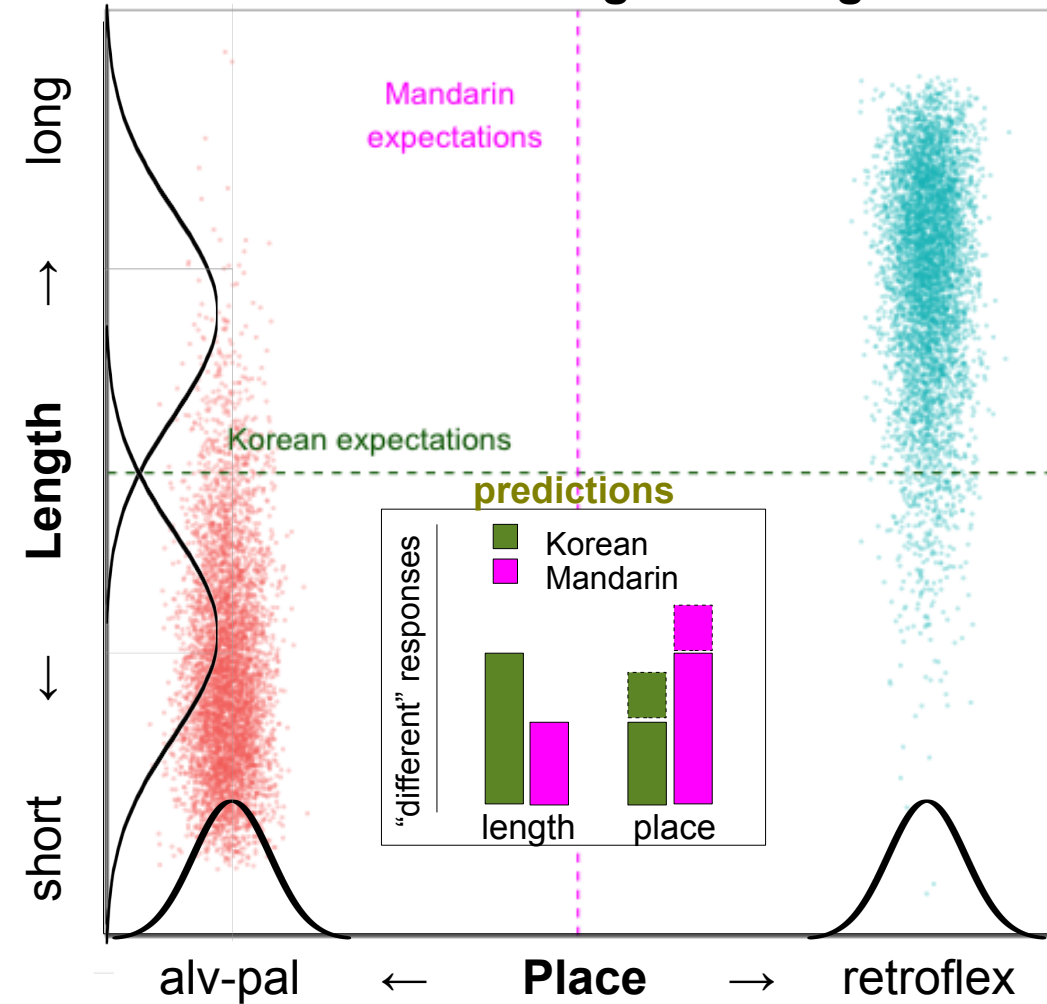
- Exposing participants to evidence suggesting a novel place contrast, with a less reliable length cue

Experiment: novel language input

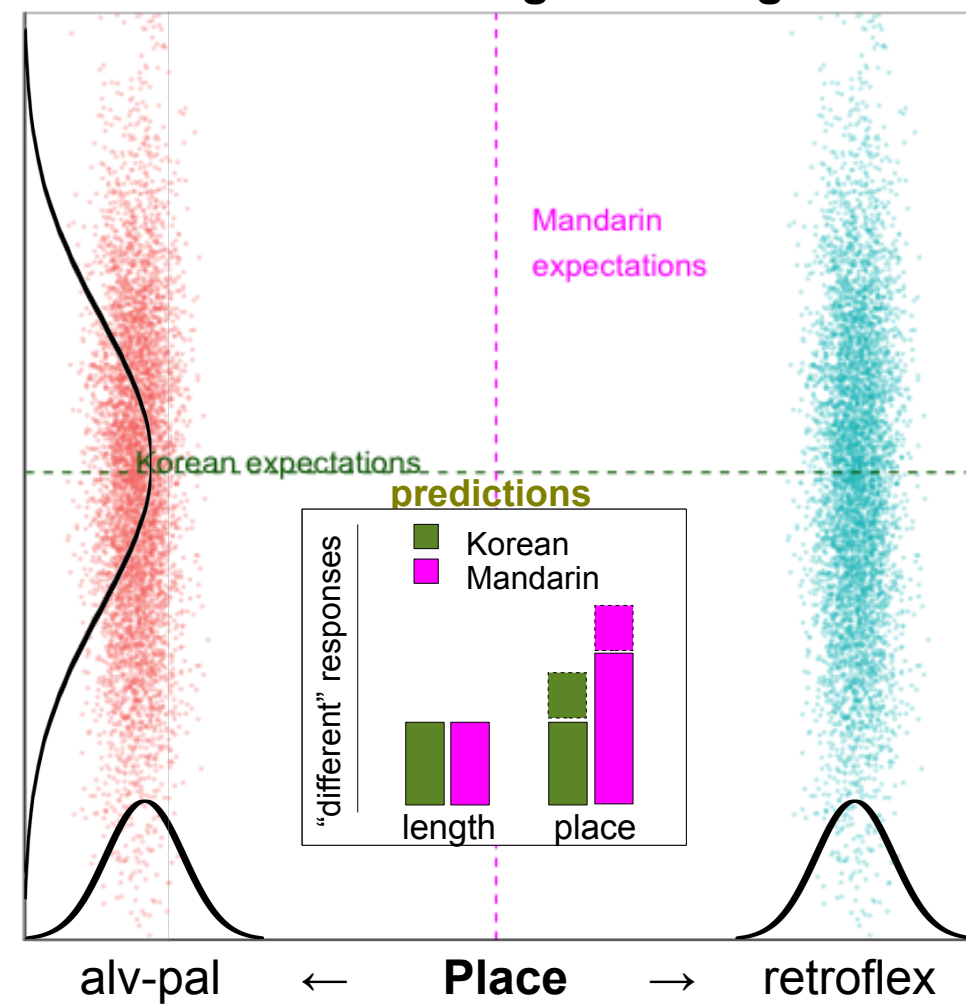


Experiment: novel language input

Bimodal-length training



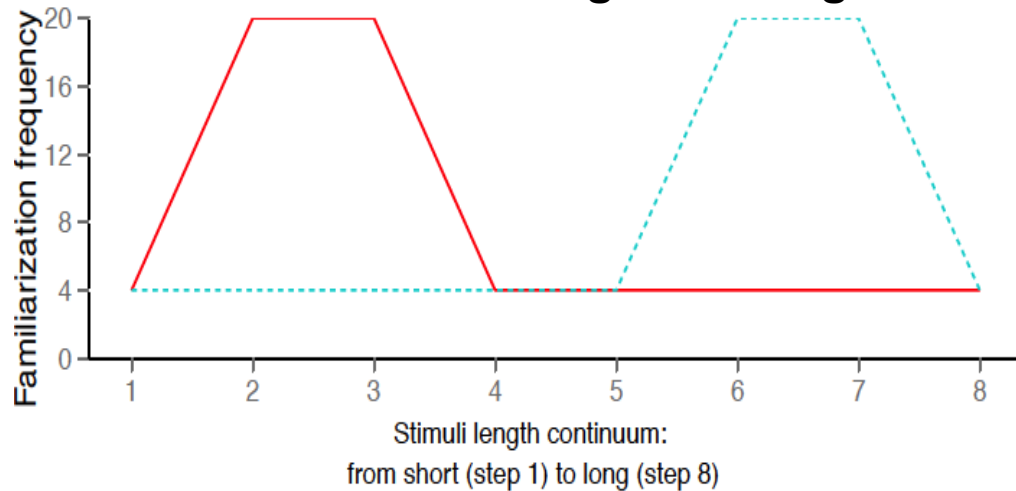
Unimodal-length training



Experiment: materials & design

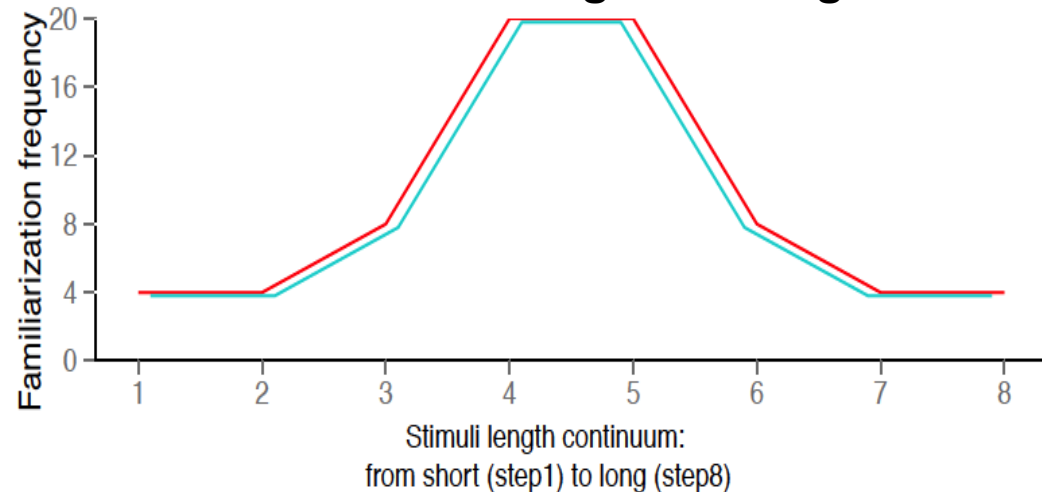
— alveolo–palatal
- - - retroflex

Bimodal-length training



— alveolo–palatal
— retroflex

Unimodal-length training



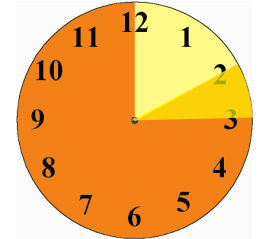
Training	x	Language	x	Contrast
bimodal-length / unimodal-length		Korean / Mandarin		length / place

+ baseline condition
(training on fillers with no
variability in length)

Experiment: training



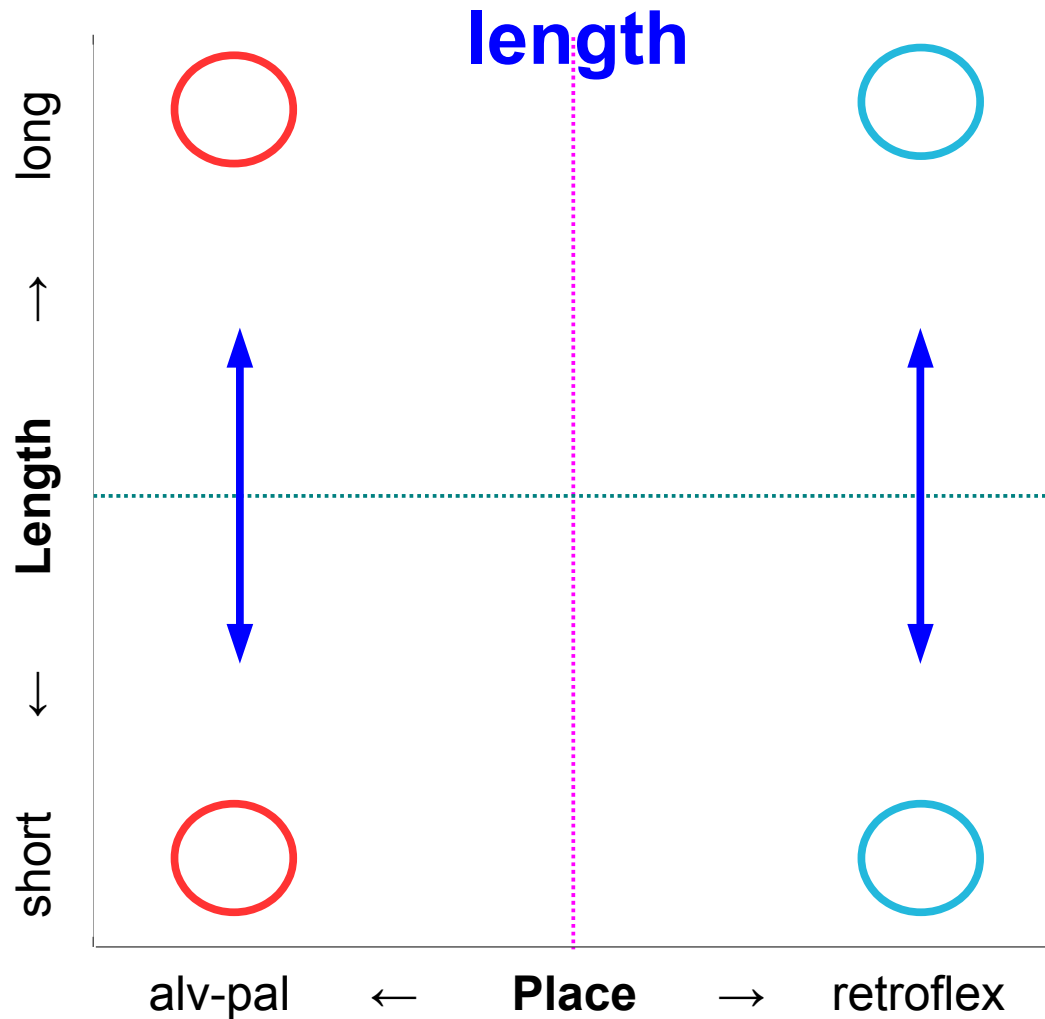
[aʁa]_{step1}
[aʁa]_{step4}
[aʁa]_{step7}
[aka]
...



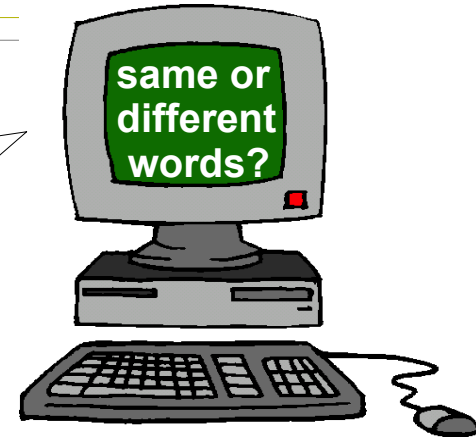
Experiment: testing

Identical testing for all participants:

→ length & place contrasts



word1 word2
word1 word2
...



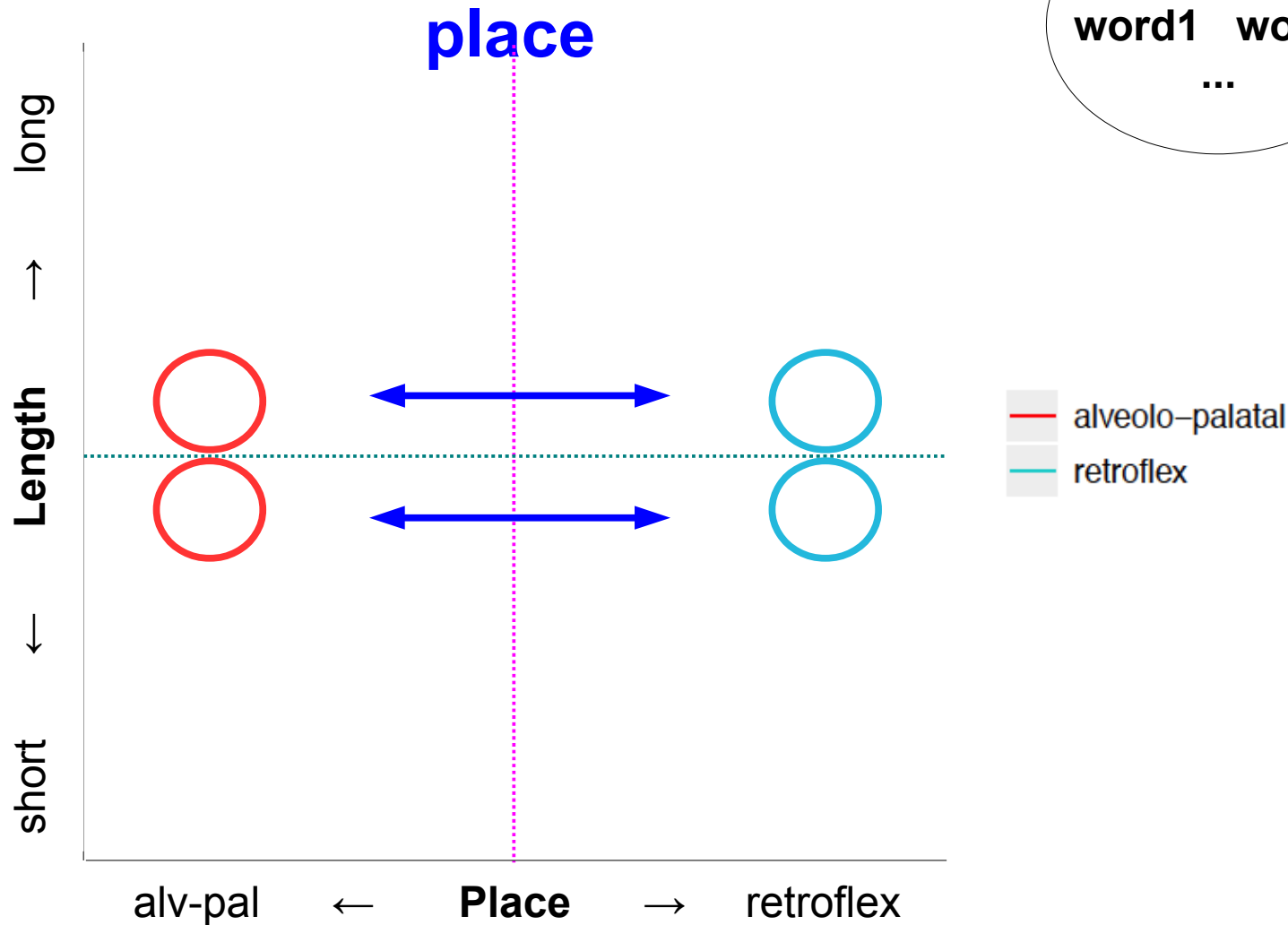
— alveolo-palatal
— retroflex



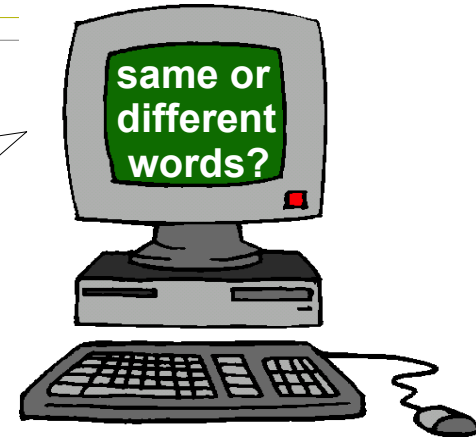
Experiment: testing

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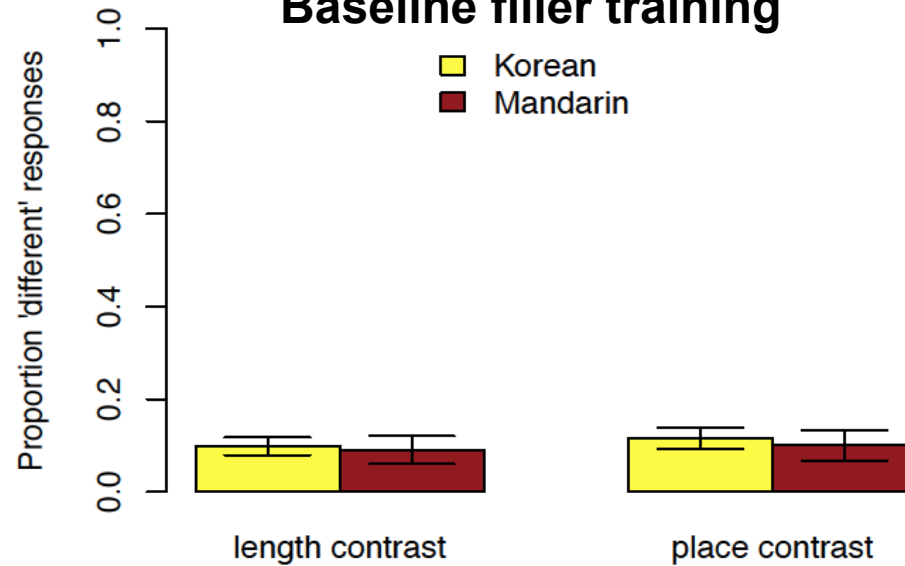
word1 word2
word1 word2
...



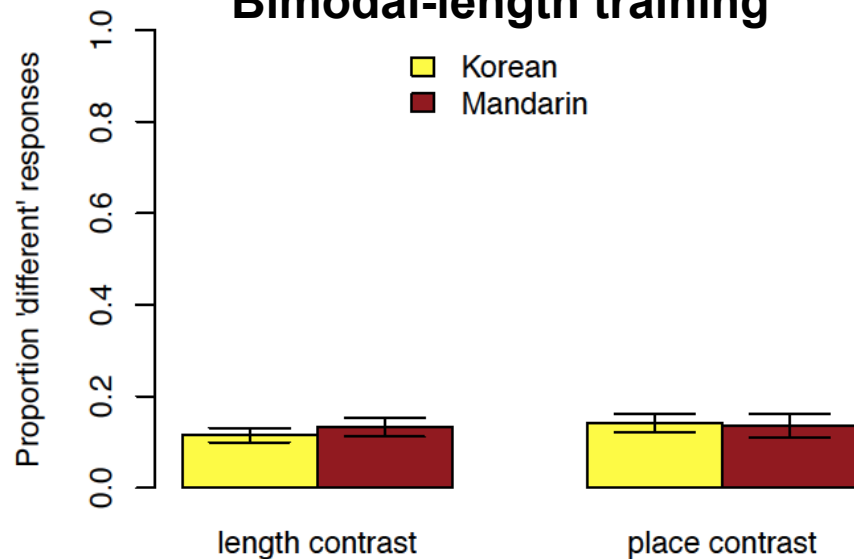
Experiment: results

SAME trials

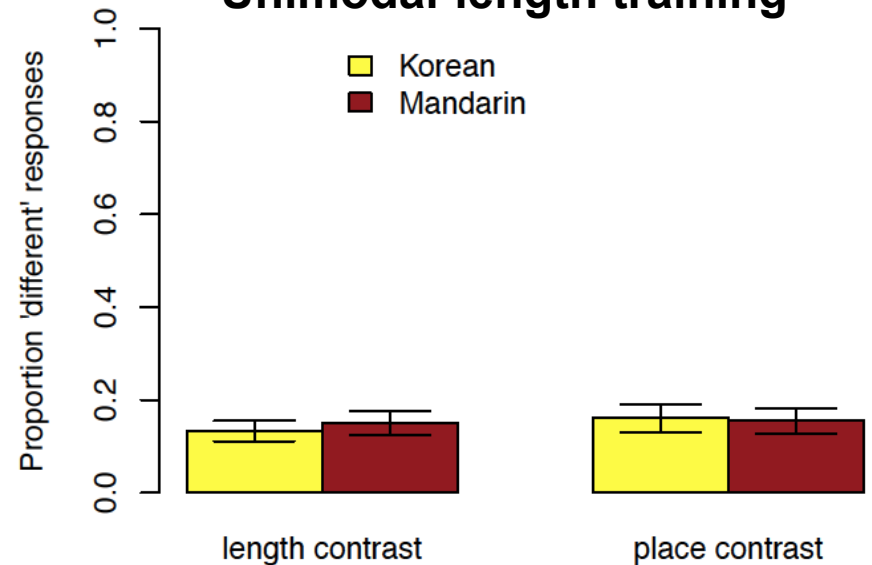
Baseline filler training



Bimodal-length training

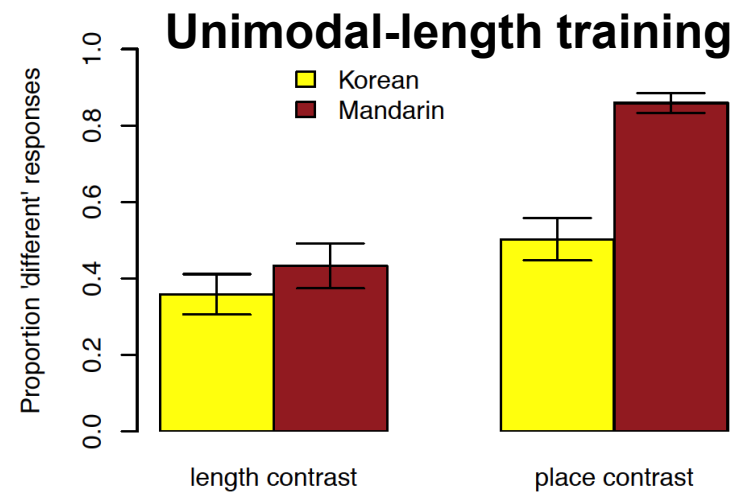
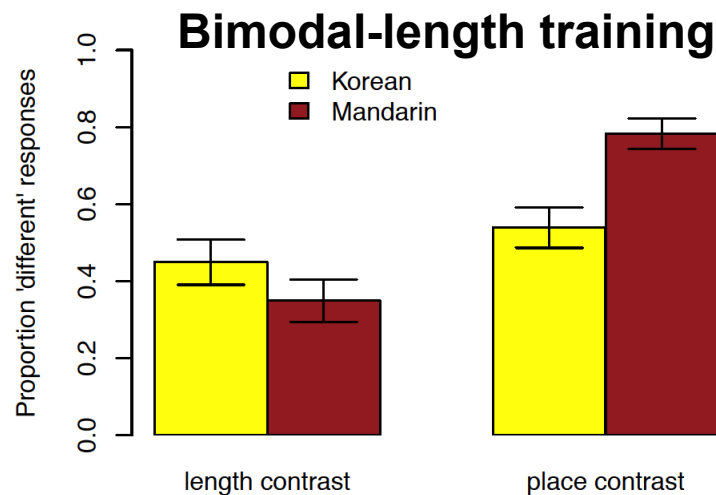
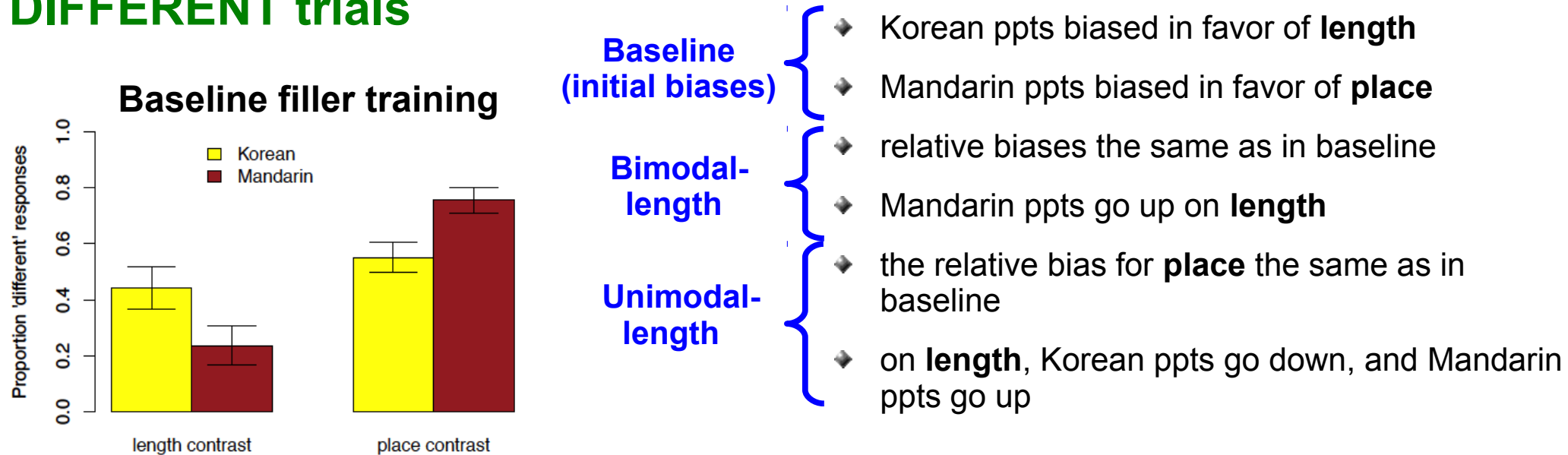


Unimodal-length training



Experiment: results

DIFFERENT trials



Experiment: results

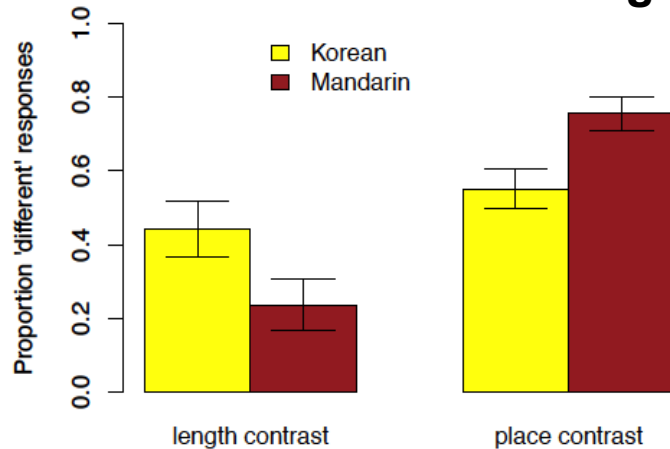
DIFFERENT trials

What does it mean?

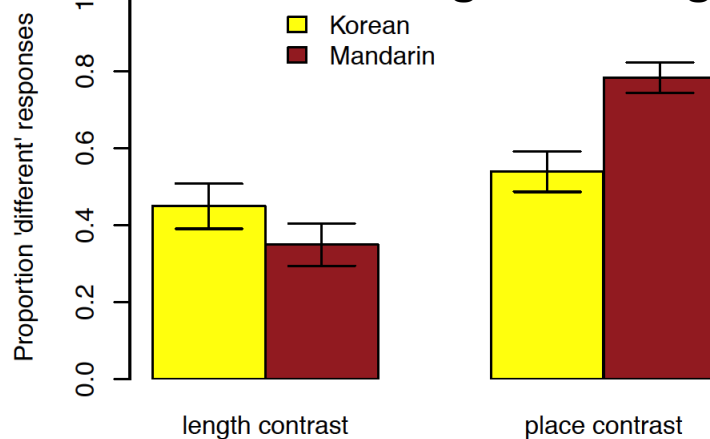
Data not entirely clear, but:

- ◆ initial biases – as predicted
- ◆ training didn't override these biases for **place**
- ◆ Korean pts were able to properly interpret distributional cues for **length**
- ◆ for Mandarin pts, mere exposure to variability along the length dimension (in either training) increased sensitivity to **length** ← not able to use distributional cues?

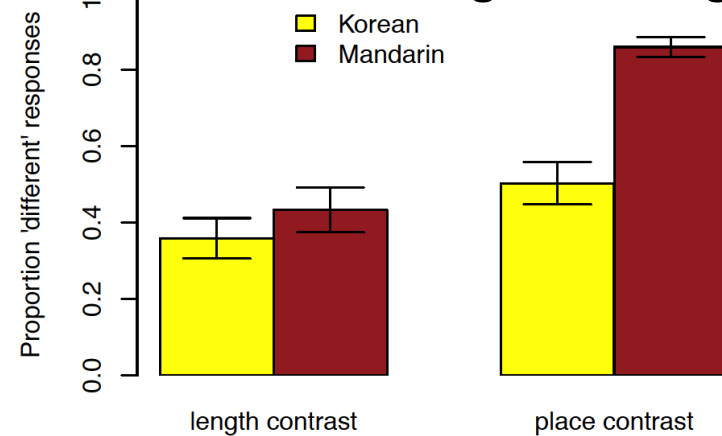
Baseline filler training



Bimodal-length training

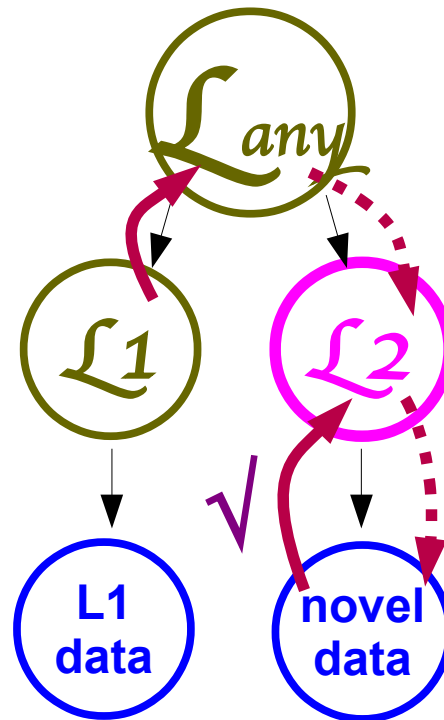


Unimodal-length training



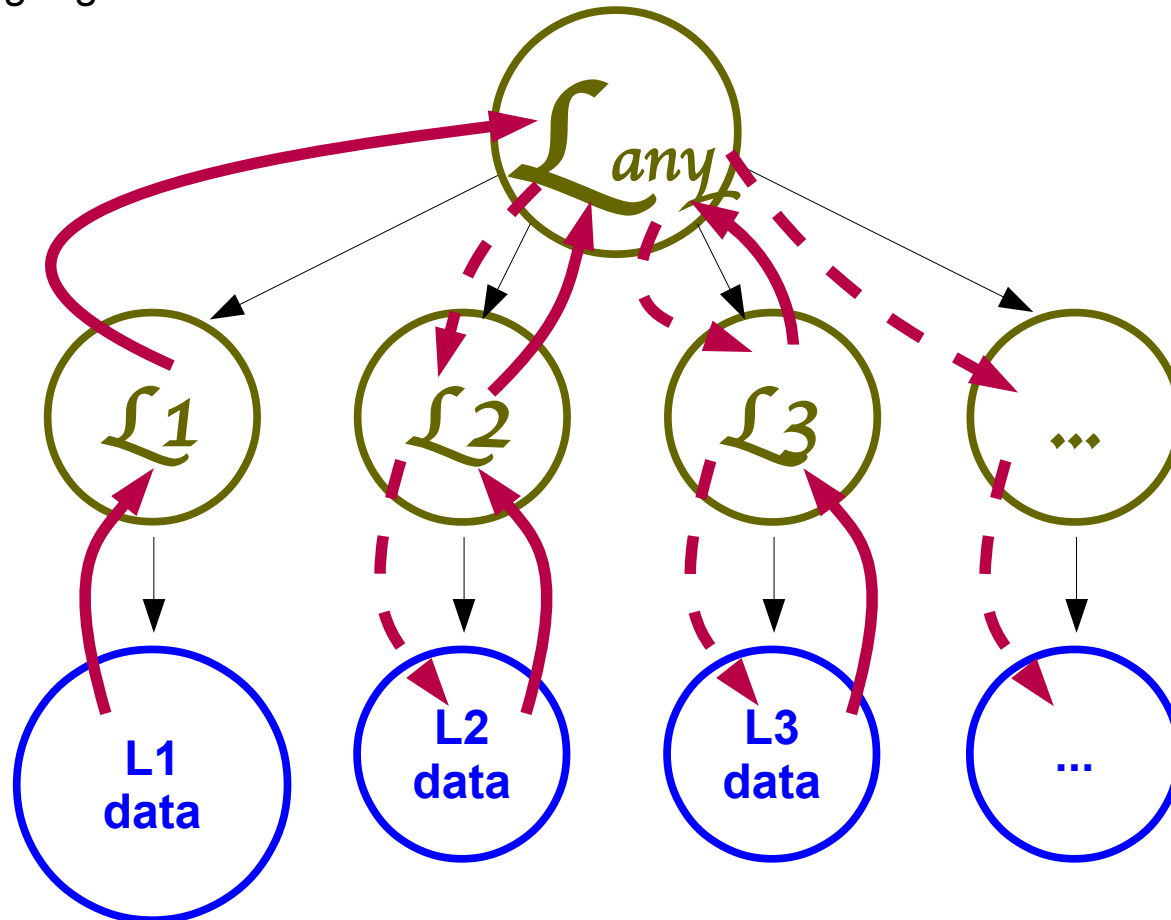
Experiment: summary

- L1 biases were reflected in categorization of novel sounds
- But categorization was also affected by short exposure to distributional cues in a novel language
- Distinct L1 biases led to different interpretations of distributional cues



Extending the model beyond L2

- The model provides a straightforward way of combining inferences from multiple languages
- This captures the intuition that learning each additional language becomes easier – because learners' \mathcal{L}_{any} inferences become sharper
- The model can make predictions about how learners' inferences change after learning each additional language



Thank you

Discussion:

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