



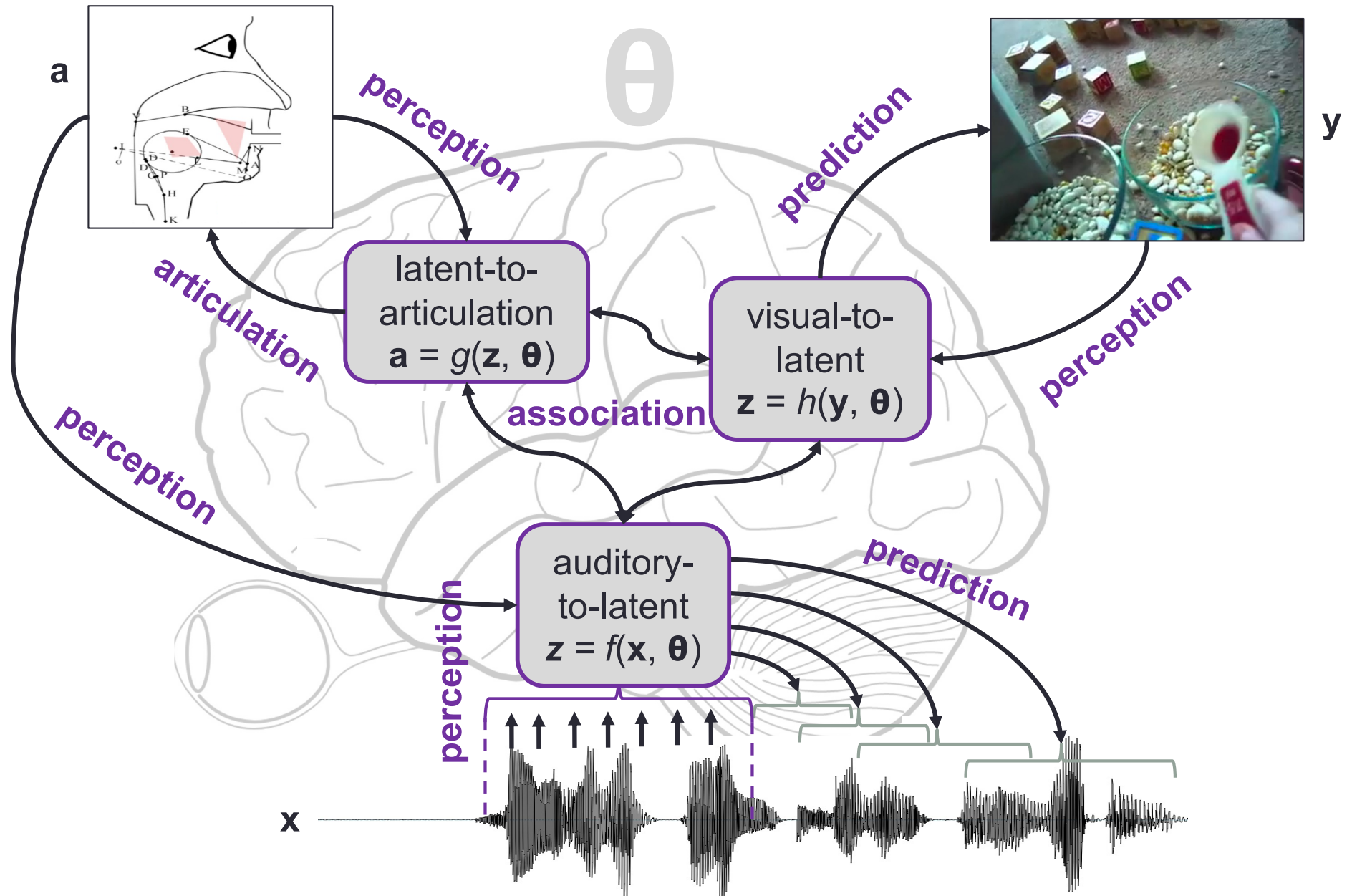
Speech and Cognition research group

COMPUTATIONAL MODELING OF INFANT LANGUAGE LEARNING FROM REALISTIC-SCALE SPEECH AND AUDIOVISUAL INPUT

Khazar Khorrami & Okko Räsänen

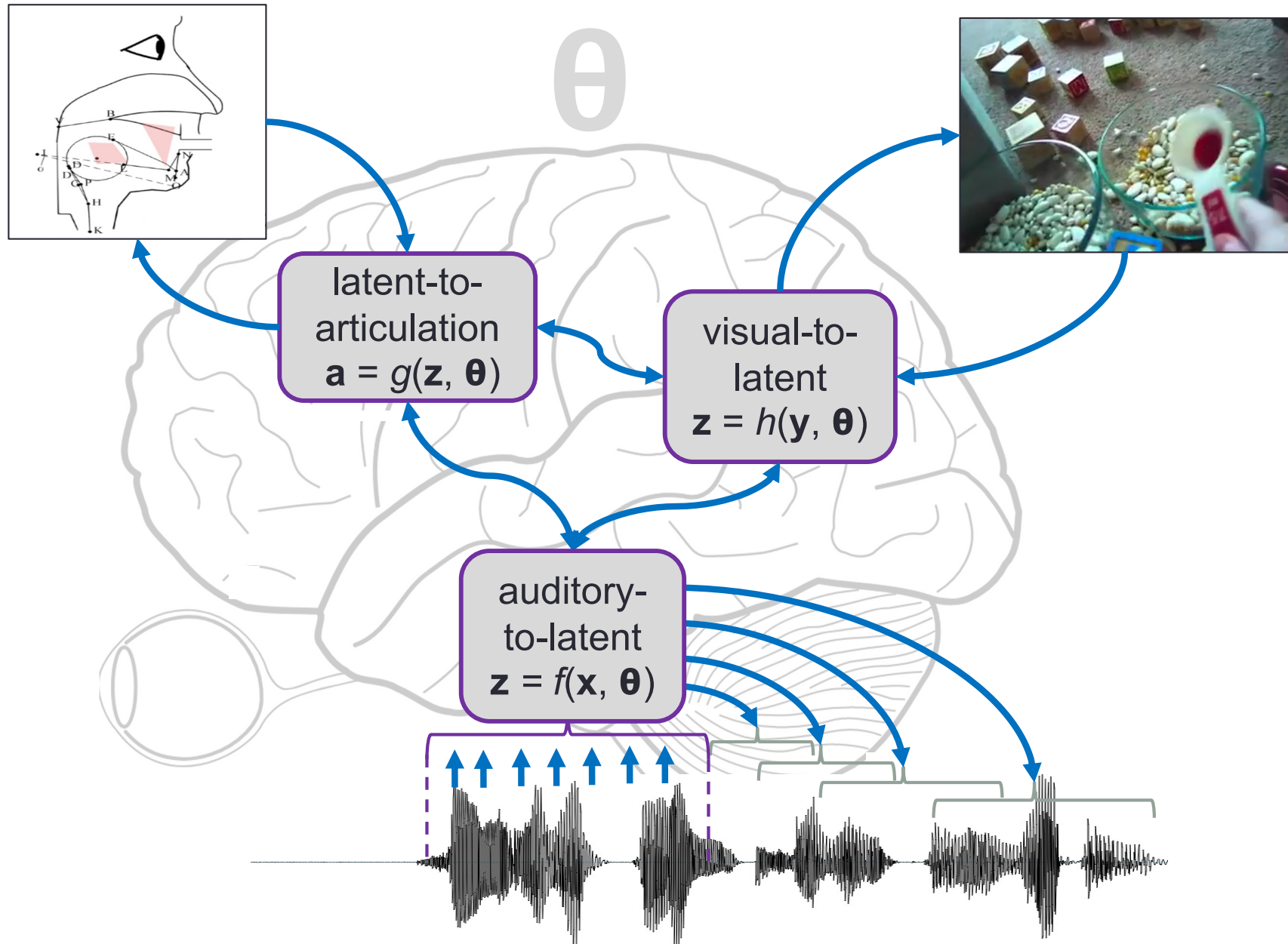
Speech and Cognition research group
Signal Processing Research Center
Tampere University, Finland

*Symposium on Computational Approaches to Early Language Development
ICIS-2024, Glasgow, UK*



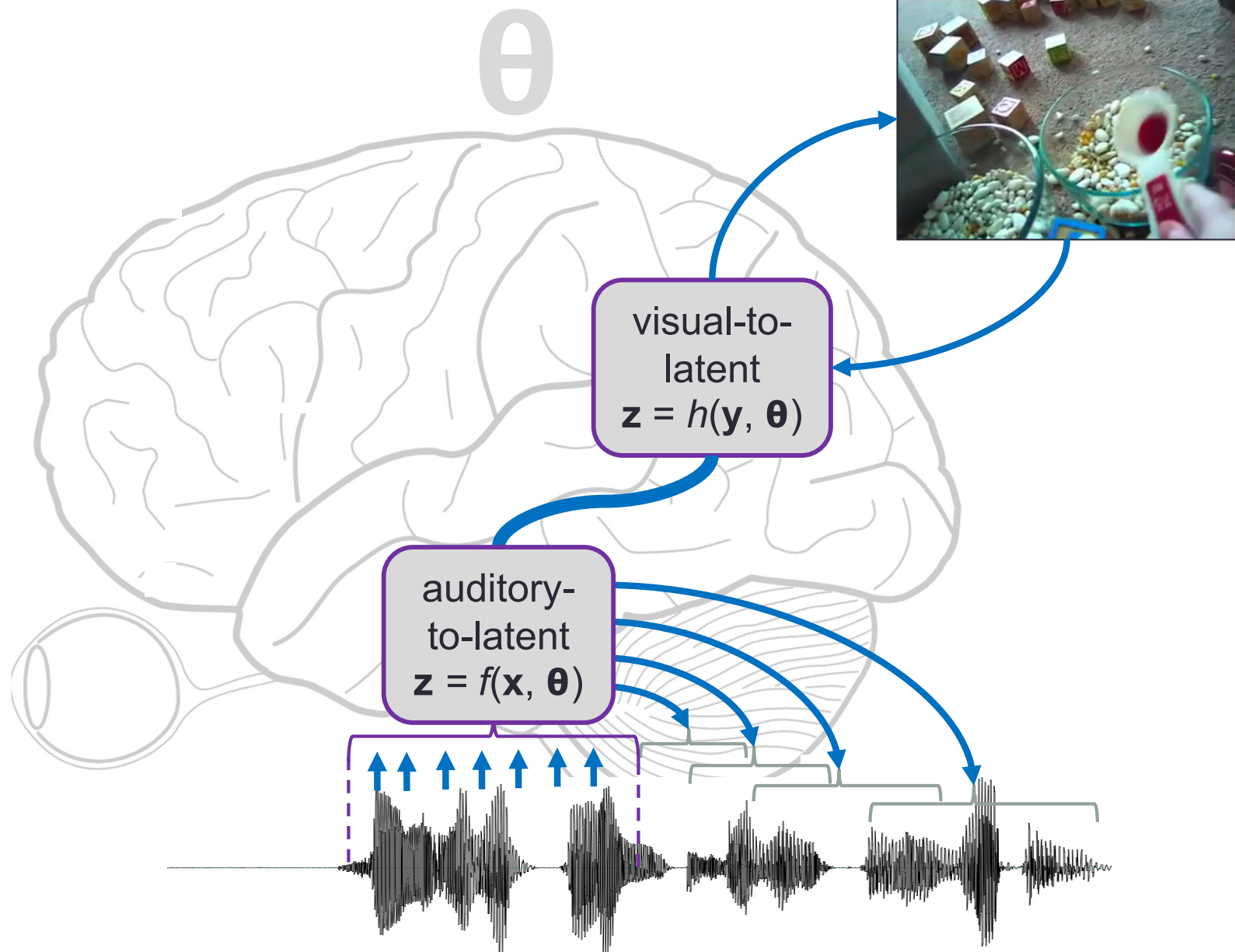
Basic framework for computational modeling: statistical learning via predictive processing

(e.g., Rao & Ballard, 1999; Friston, 2010; Clark, 2013)



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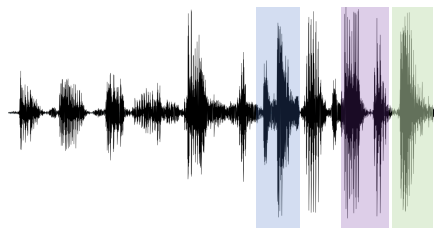
This work: auditory + audiovisual statistical learning

Background & research question

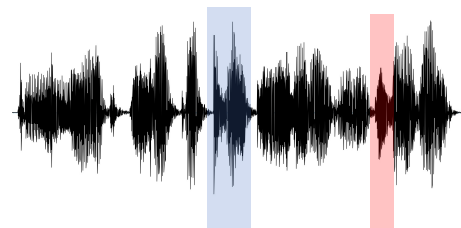
Previous research with audiovisual computational models:

- Learning from photographs and their spoken descriptions
(e.g., Harwath & Glass, 2017; Alishahi et al., 2017; Merx et al., 2019; Khorrami & Räsänen, 2021; Peng et al., 2023)
 - Learning from infant head-mounted camera data + transcribed speech
(Vong et al., 2024)
 - Main findings: latent representations for phonemes, syllables, and words emerge as a side-product of audiovisual predictive optimization. ***No need for linguistic priors or proximal learning goals!***
 - Limitations: models trained on thousands of speech-image pairs (“naming events”) or with simplified speech representations (text).
- **Unclear if word learning succeeds from infant-scale sensory input with real speech**
- **This work: simulate auditory and audiovisual learning with realistic-scale input.**

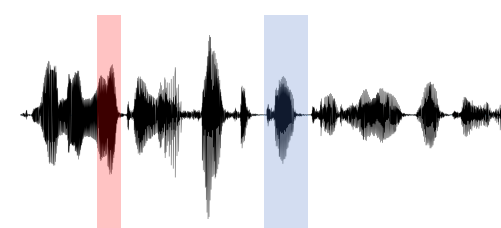
Referential ambiguity in audiovisual learning



“cake” “milk”
“bottle”



“cake” “cup”

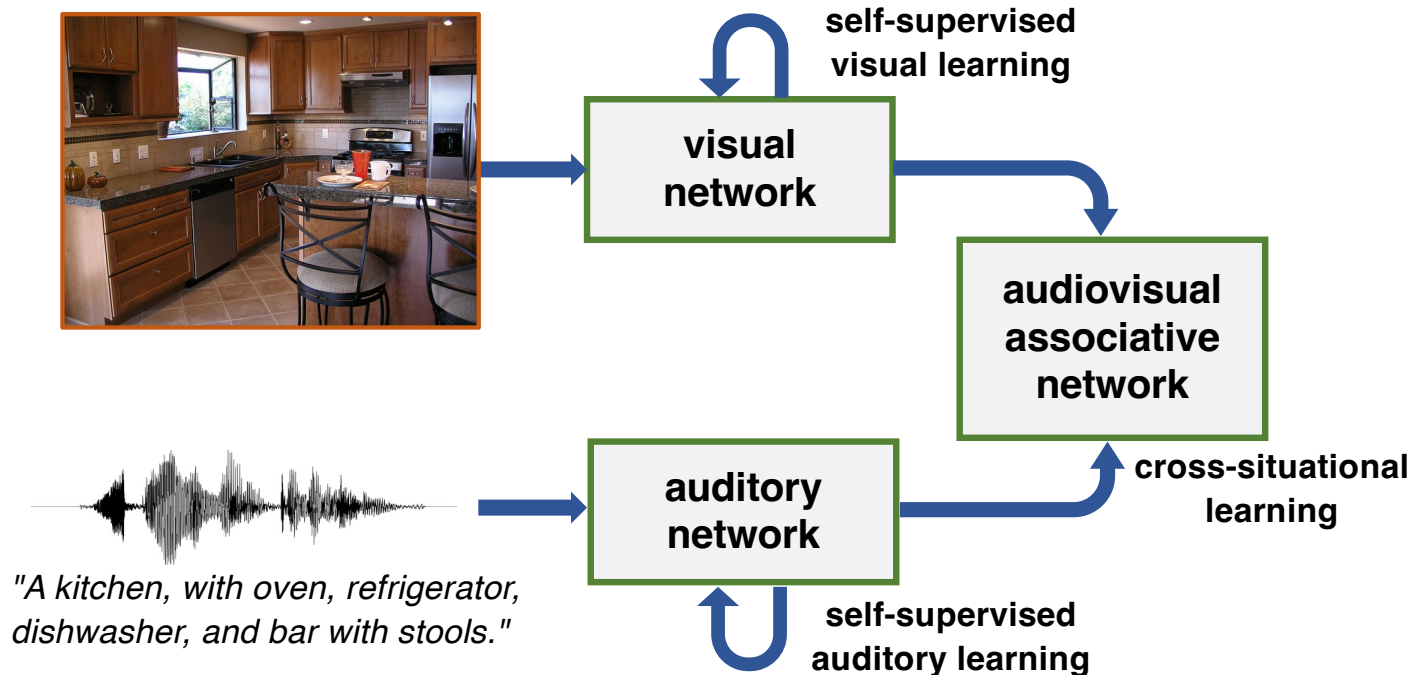


“cup” “cake”

Basic challenges:

- Segmentation problem in the auditory and visual domains (“where”)
- Recognition problem in both domains (“what”)
- Referential ambiguity across domains (e.g., Quine, 1960; Smith & Yu, 2008)

Model architecture (adapted from Peng & Harwath, 2022)



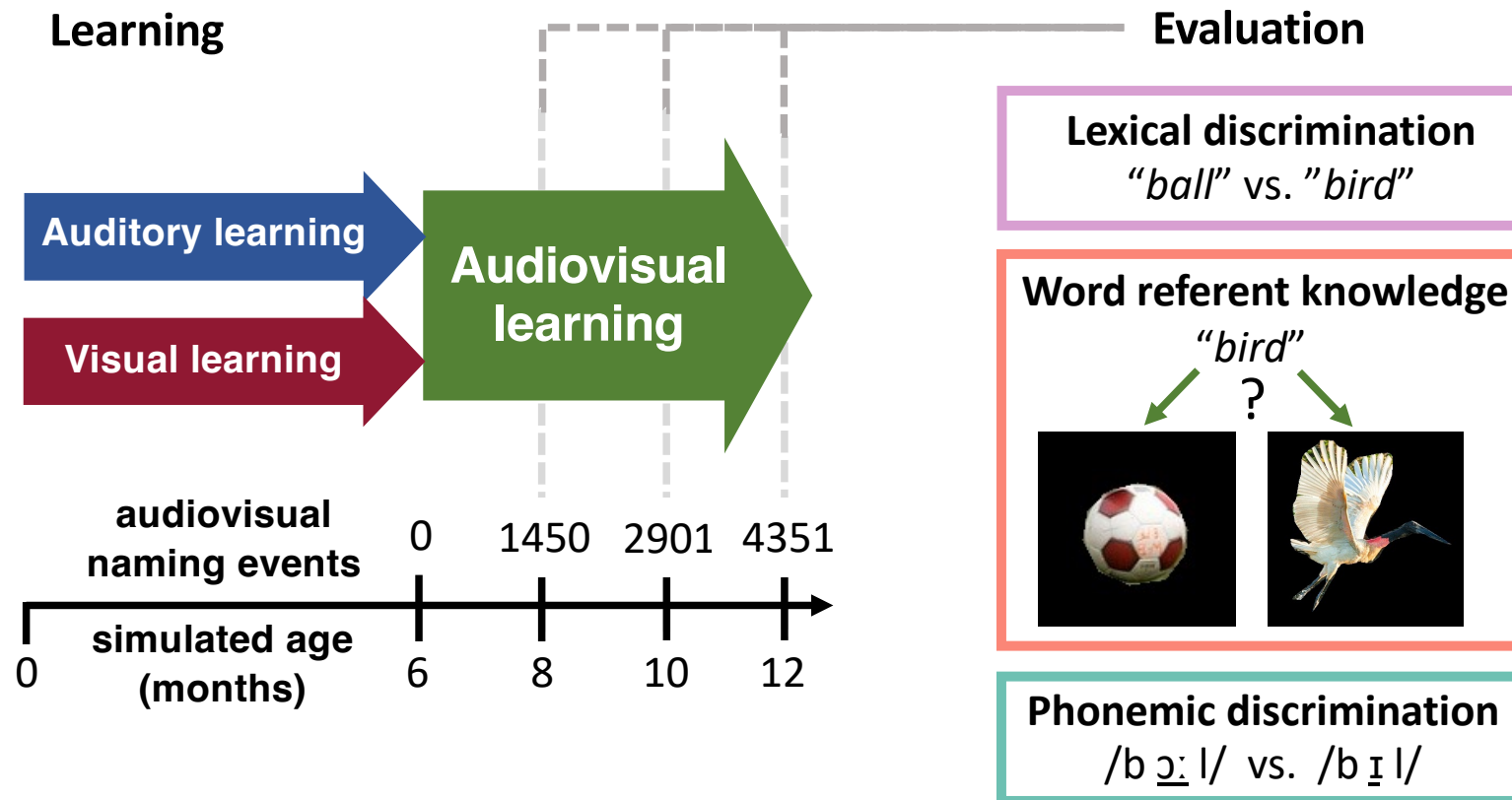
A deep neural network with three parts:

- 1) Visual encoder: DINO
- 2) Auditory encoder: Wav2Vec 2.0
- 3) Associative network: contrastive learning.

No supervision or data labels.

Only self-supervised (statistical) learning from sensory input.

Experimental setup



Training data design

Auditory learning:

1049 h of speech input to simulate 6 months of auditory learning (e.g., Cruz-Blandon et al., 2023; Coffey et al., 2024).

- From Librispeech + SpokenCOCO corpora

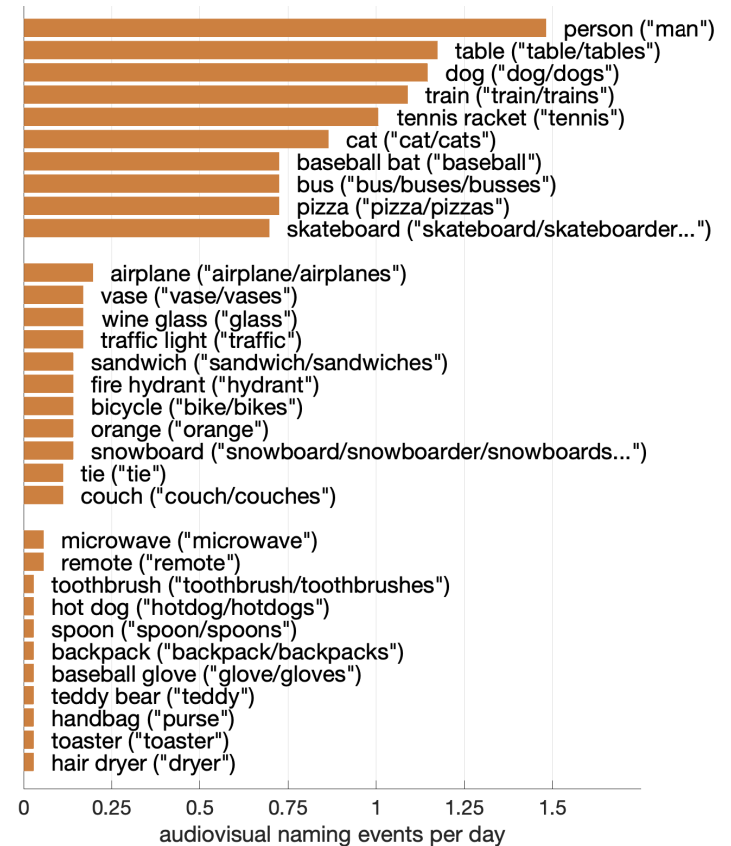
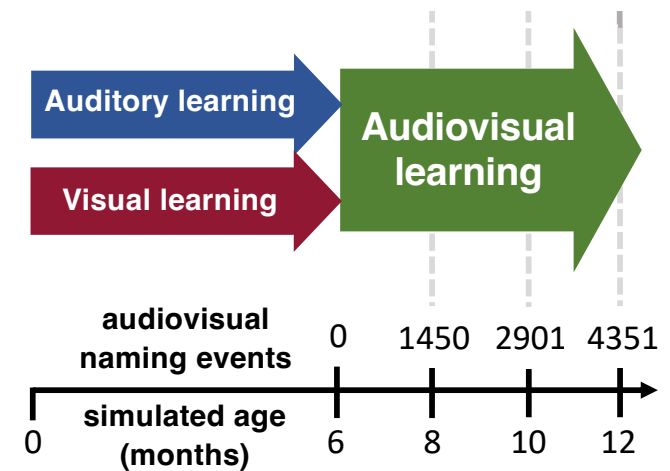
Audiovisual learning:

Photographs and their spoken descriptions from SpokenCOCO dataset.

Empirical estimates of daily object naming rates for the 80 most frequent word-object pairs (from Clerkin & Smith, 2019; 2022).

- Extrapolate counts to 2, 4 or 6 months
- Select images + utterances that satisfy the statistics.

- Words per utterance: 11.3 ± 2.59
- Content words per utterance: 5.87 ± 1.47
- Visual targets per image: 2.9 ± 1.84



Model evaluation

Evaluate model at 6, 8, 10, and 12 months for:

- **Phonemic discrimination** (ABX-test; Schatz et al., 2023)
- **Auditory word-form discrimination** (CDI-Lextest, Khorrami et al., 2023).
- **Word referent knowledge** for the 80 audiovisual concepts in SpokenCOCO (an audiovisual forced-choice task)

Lexical discrimination

"ball" vs. "bird"

Word referent knowledge

"bird"

?

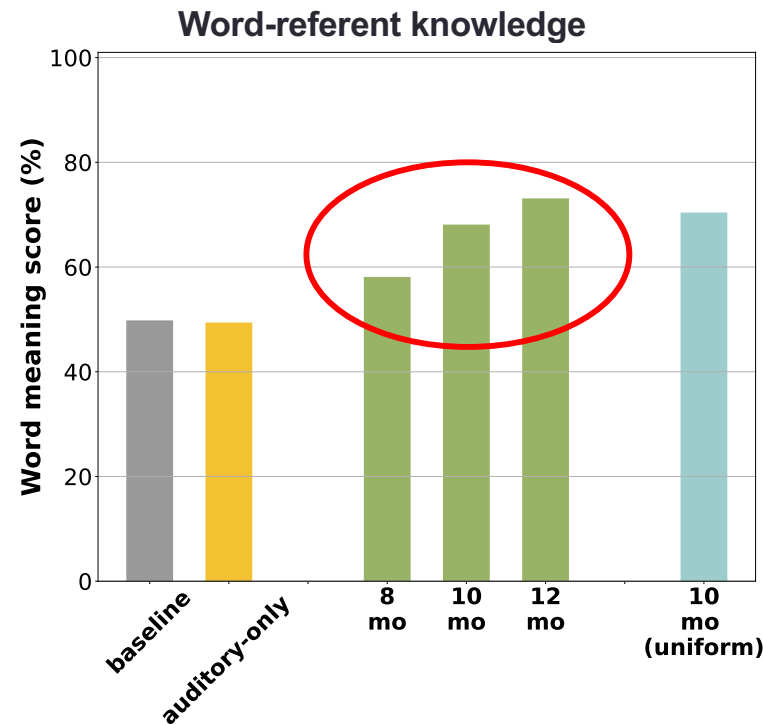
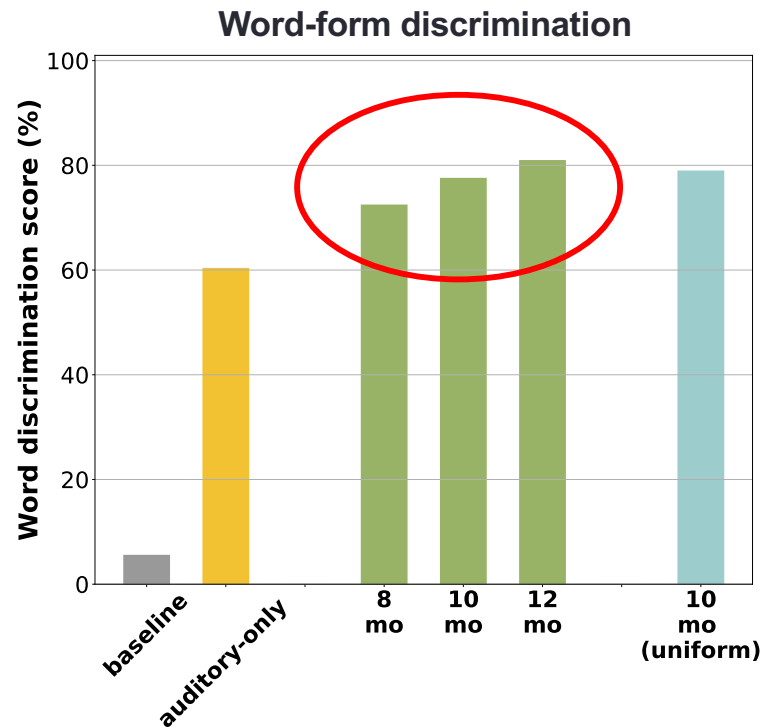


Phonemic discrimination

/b ɔ l/ vs. /b ɪ l/

Results

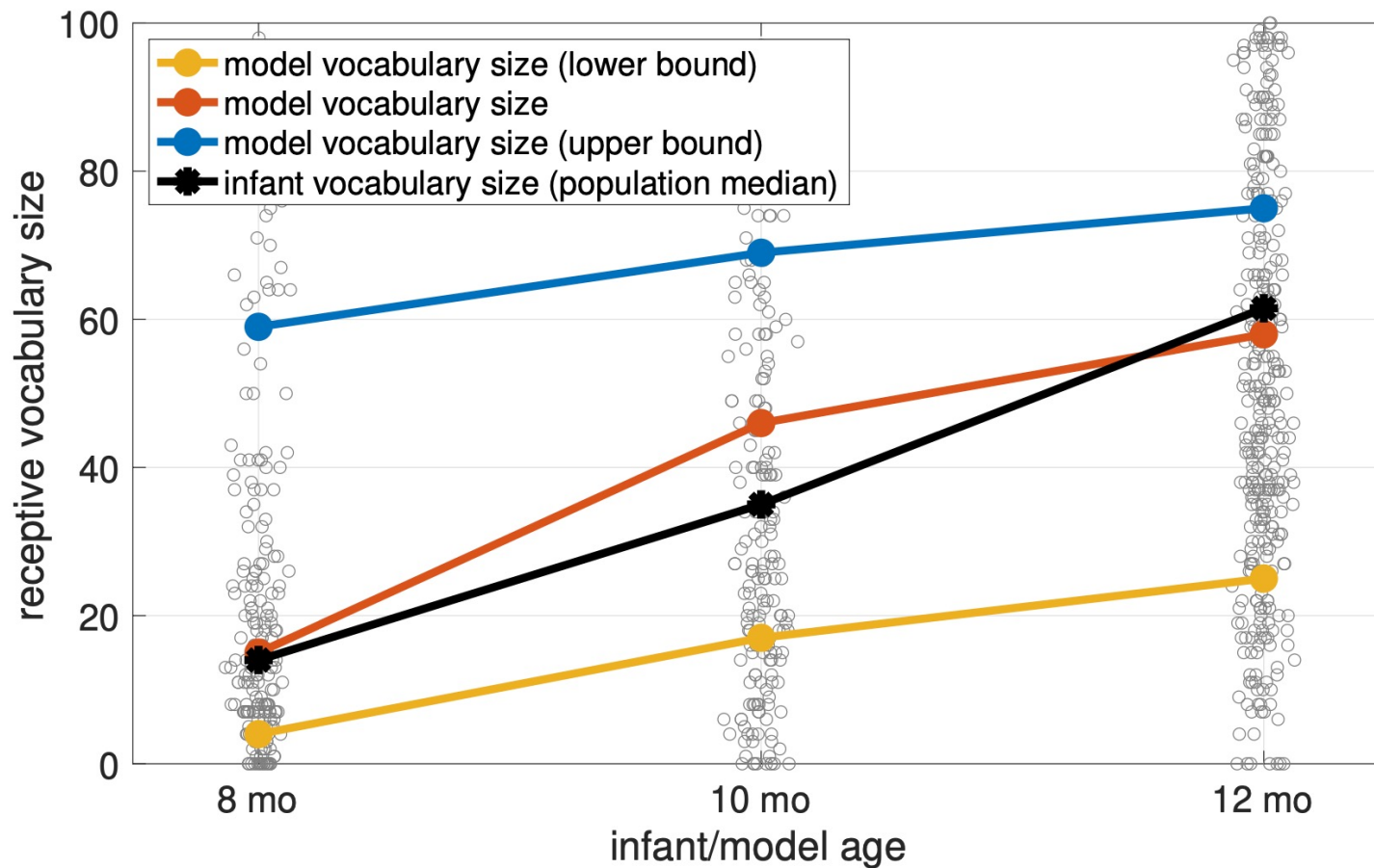
Word referent learning didn't work without the auditory learning stage.



Phoneme error rate: 7.1% after 6 mo auditory learning (chance: 50%). No change during audiovisual stage.

Phoneme and word comprehension skills emerge from plausible-scale data!

Vocabulary growth: model vs. CDI-norms



CDI data: North-American infants, receptive lexicon (from Wordbank; Frank et al., 2017)

Conclusions

The model succeeds in learning proto-lexical (and phonemic) representations from infant-scale input.

Learning operates on *real speech* and images, and *without linguistic priors*, data labels, or other strong constraints.

Supports the idea of statistical learning as a means to bootstrap early language acquisition.

Supports the “Latent Language Hypothesis”, according to which linguistic structures are not proximal targets of learning, but side products of predictive optimization (e.g., Khorrami & Räsänen, 2021, *Lang. Dev. Res*).

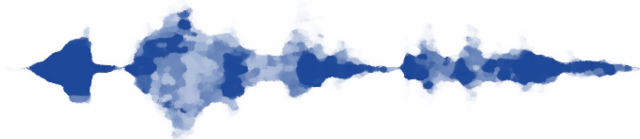
- No need to “cluster” phone(me)s or segment words as intermediate stages. Only prediction within and across sensory modalities.

The end

Paper pre-print:
<https://arxiv.org/abs/2406.05259>



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khazar.khorrami@tuni.fi

okko.rasanen@tuni.fi

<https://webpages.tuni.fi/specog/>