Emergent morphophonological representations in models of spoken word recognition

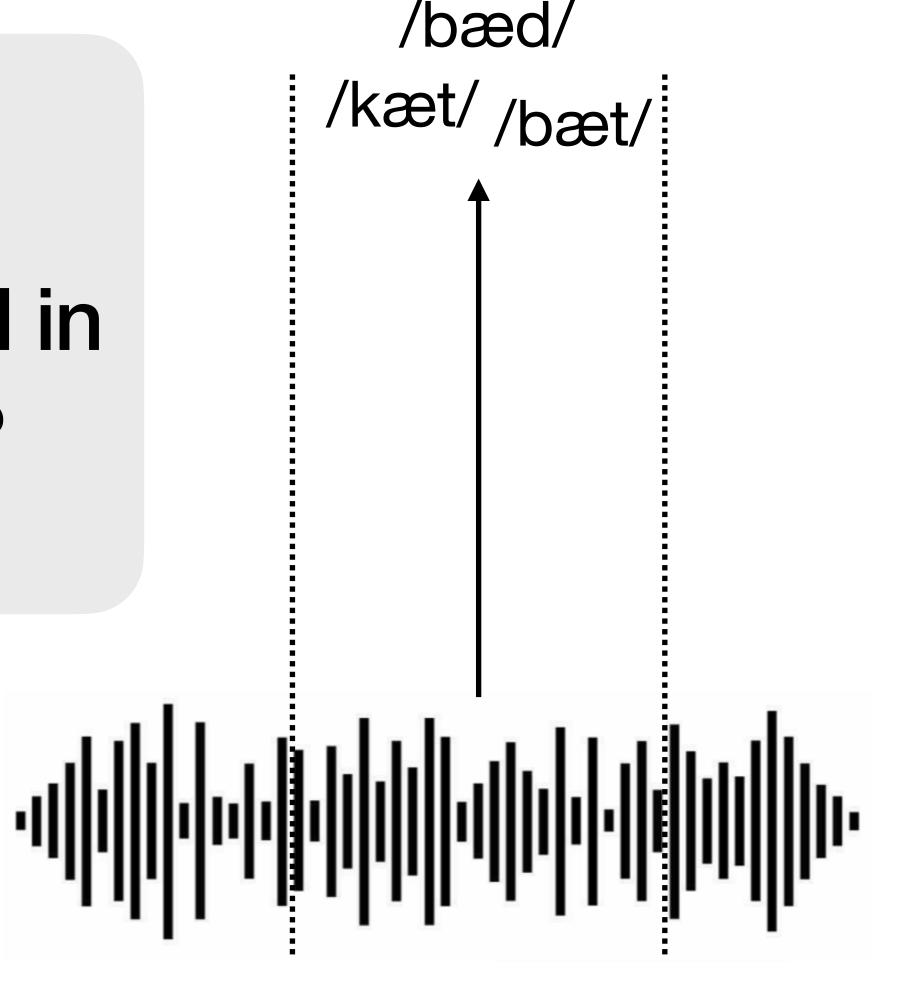
Jon Gauthier¹
Matthew Leonard¹

Canaan Breiss² Edward Chang¹

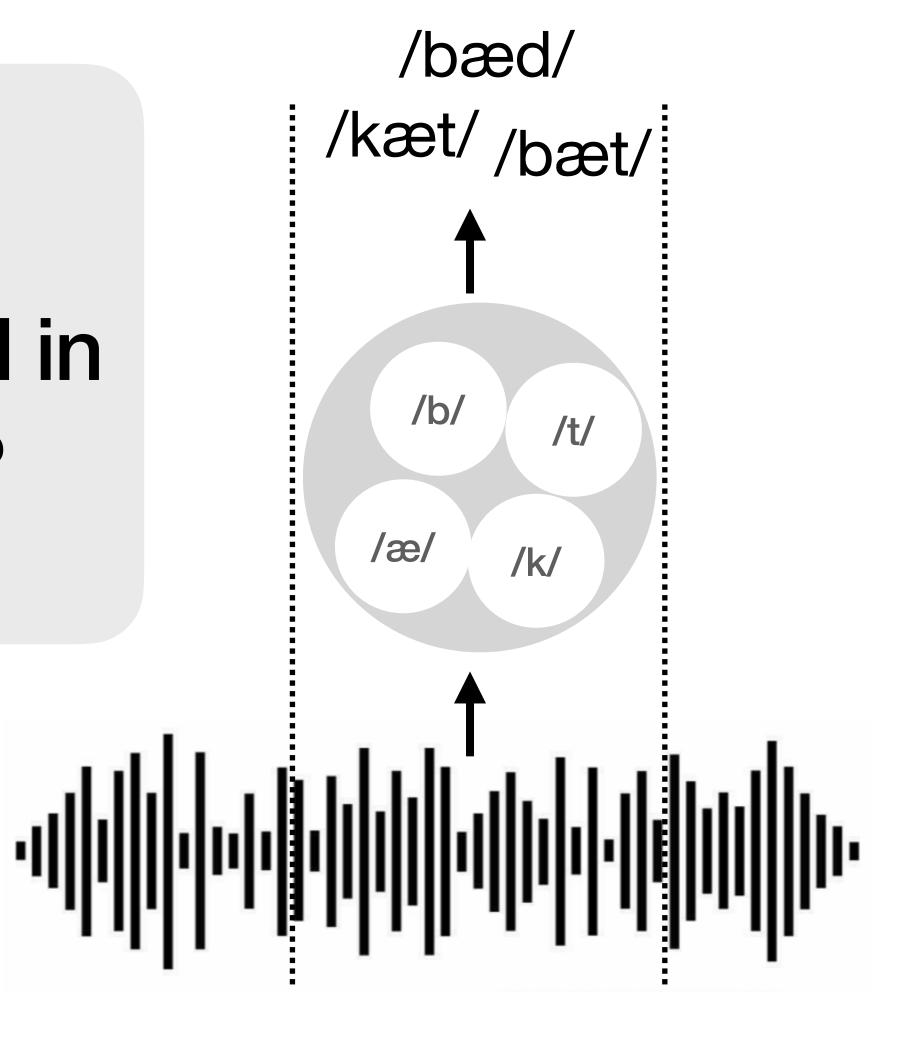
¹ UCSF ² USC

AMP 2025 UC Berkeley

What kinds of linguistic representations are recruited in spoken word recognition?



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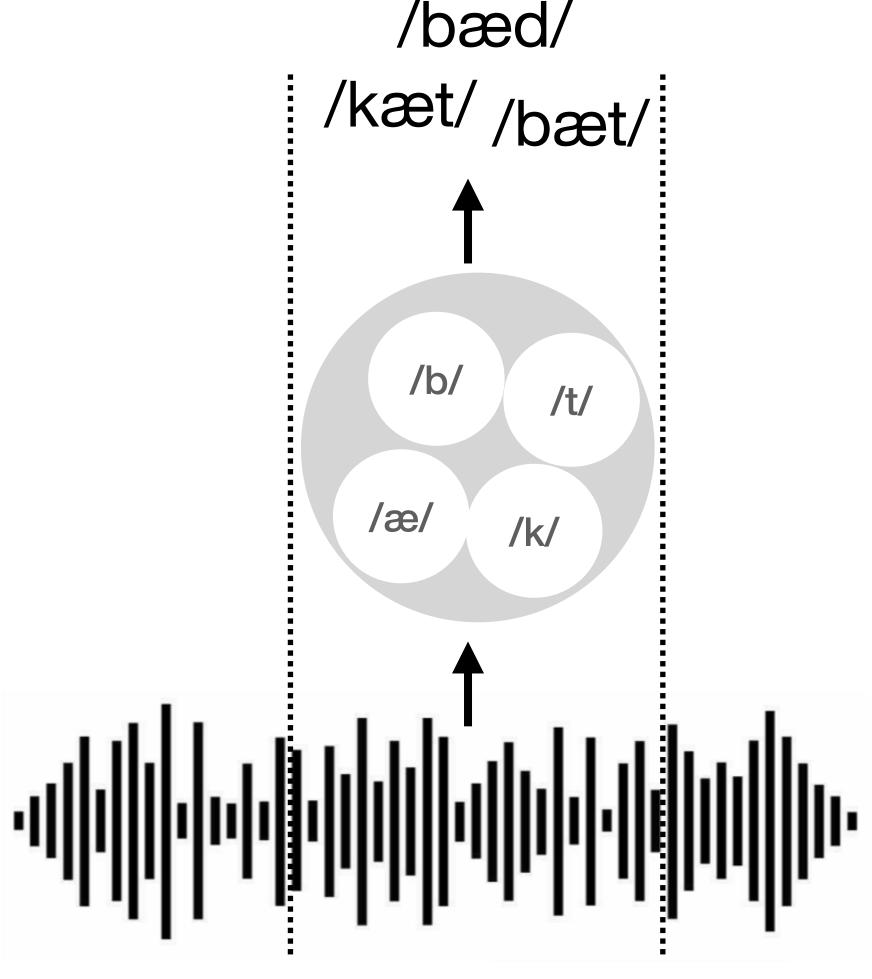


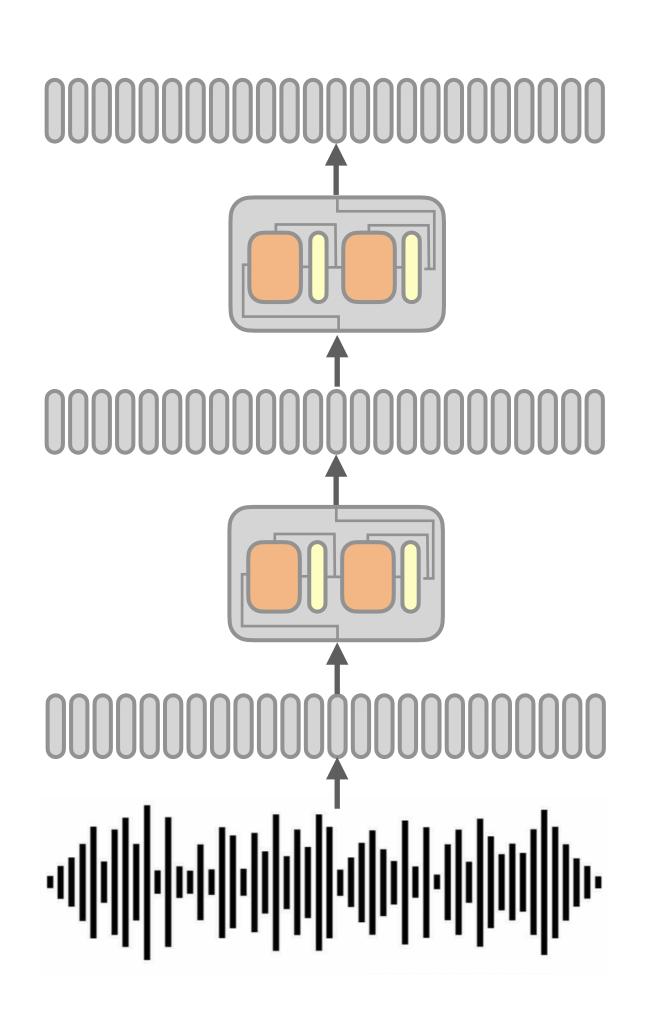
Classic computational theories

TRACE, Shortlist, Merge, DCM, ...

Explicit levels of linguistic representation

Explain spoken word recognition at small scales



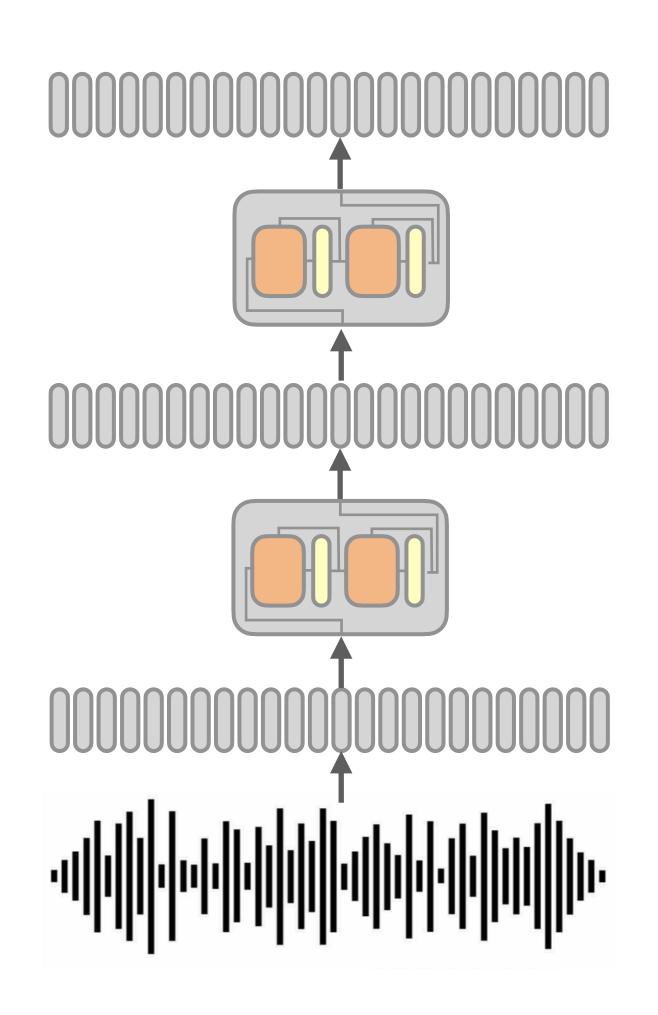


Self-supervised models

Word recognition at large scale, from raw audio

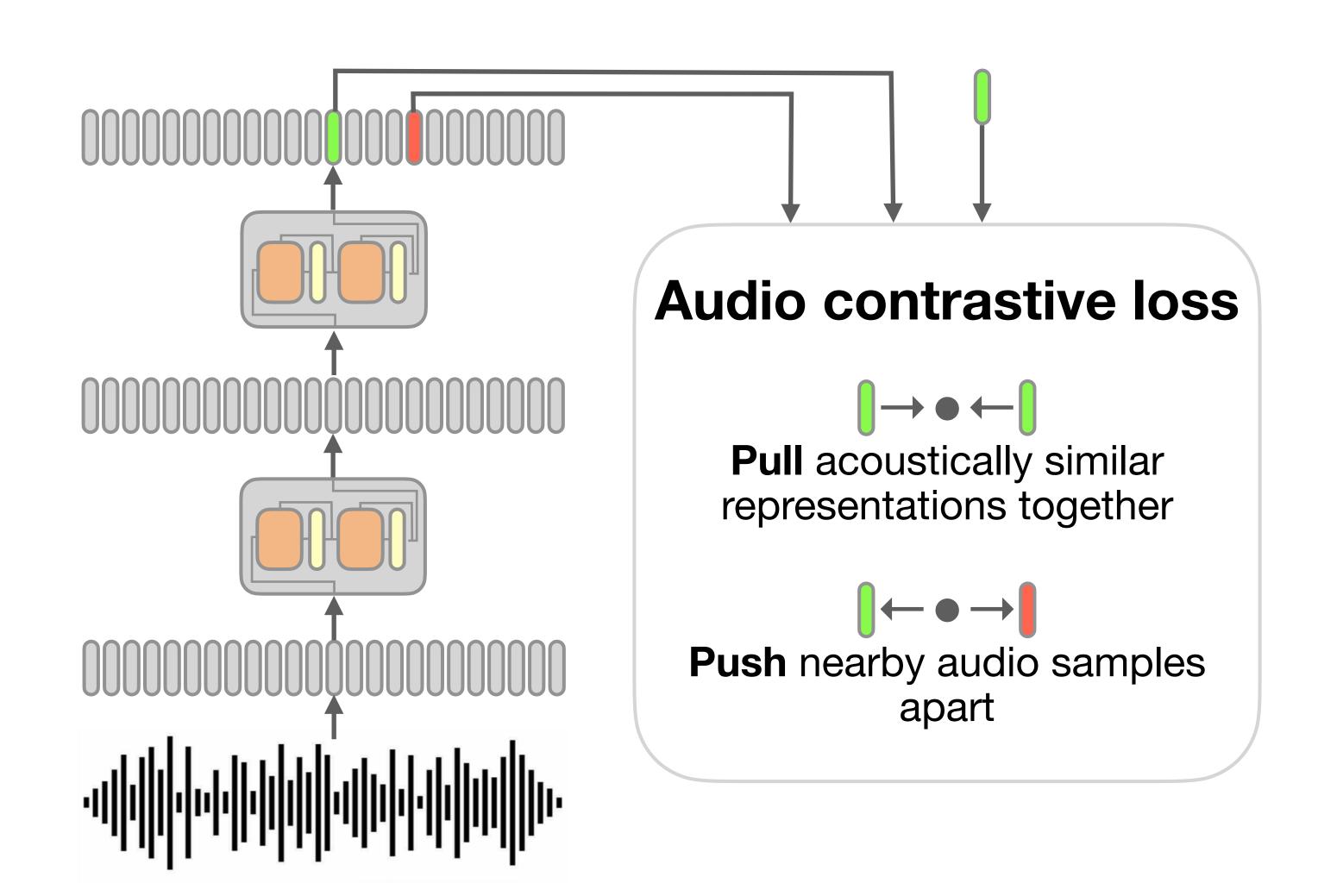
No explicit linguistic representation

Not (yet) interpretable as a cognitive theory

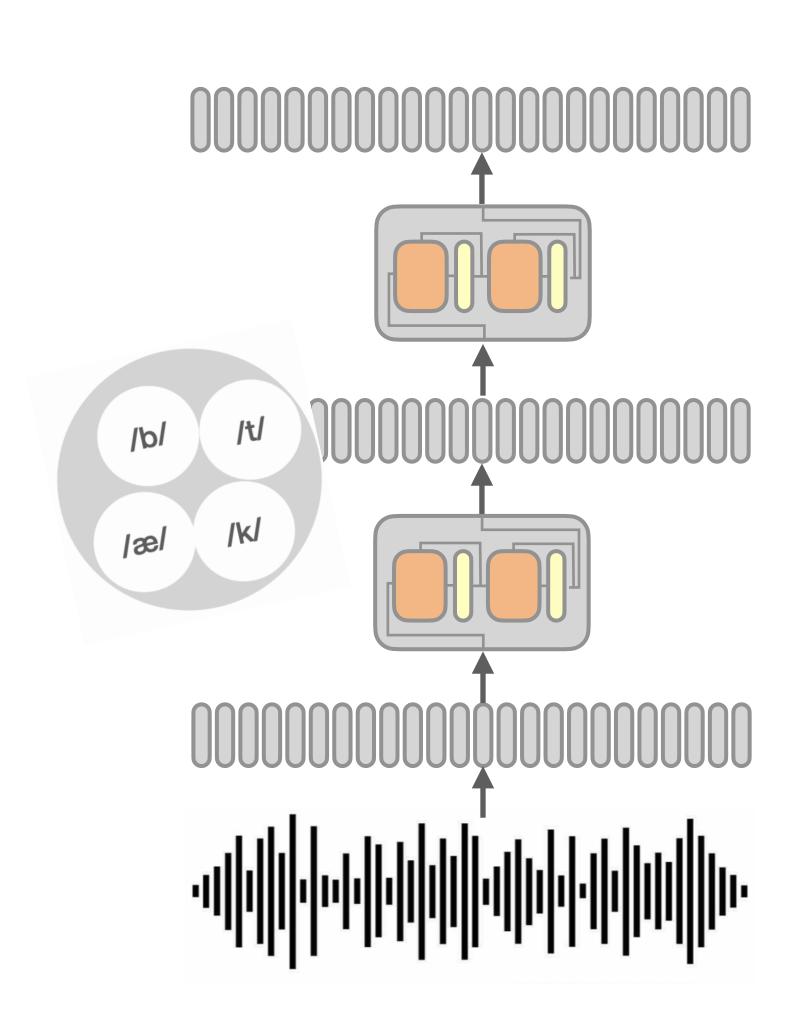


What kinds of linguistic representations are recruited in models of spoken word recognition?

A self-supervised model: wav2vec2



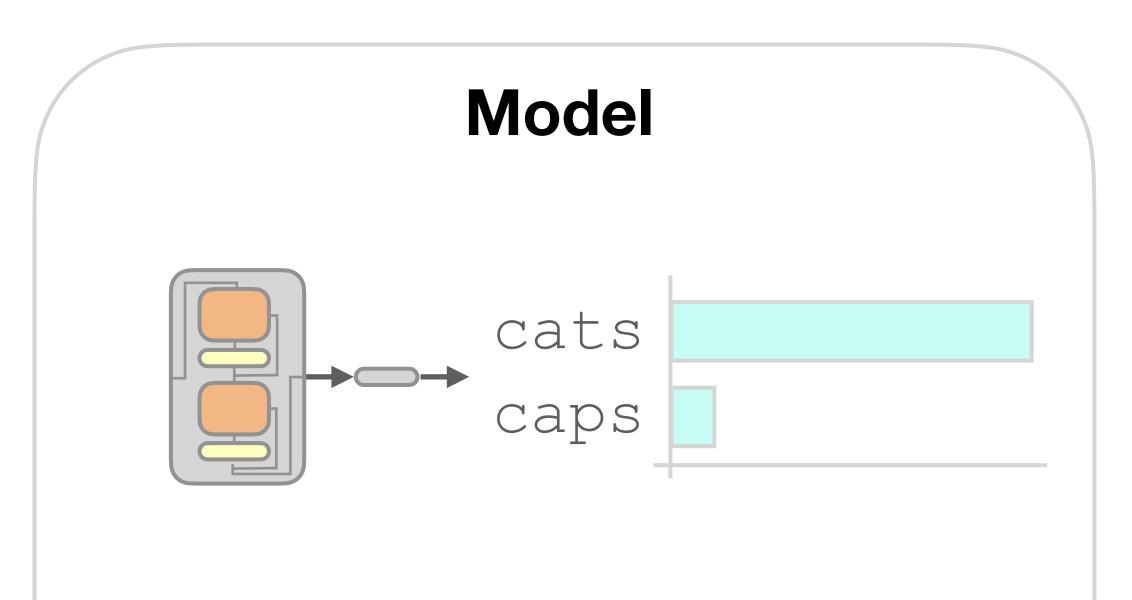
A self-supervised model: wav2vec2



Self-supervised models encode basic phonological categories
 but these may serve many functions beyond word recognition

(Pasad et al. 2021, 2023; Martin et al. 2023; Abdullah et al. 2023; Choi et al. 2024, 2025)

Plan



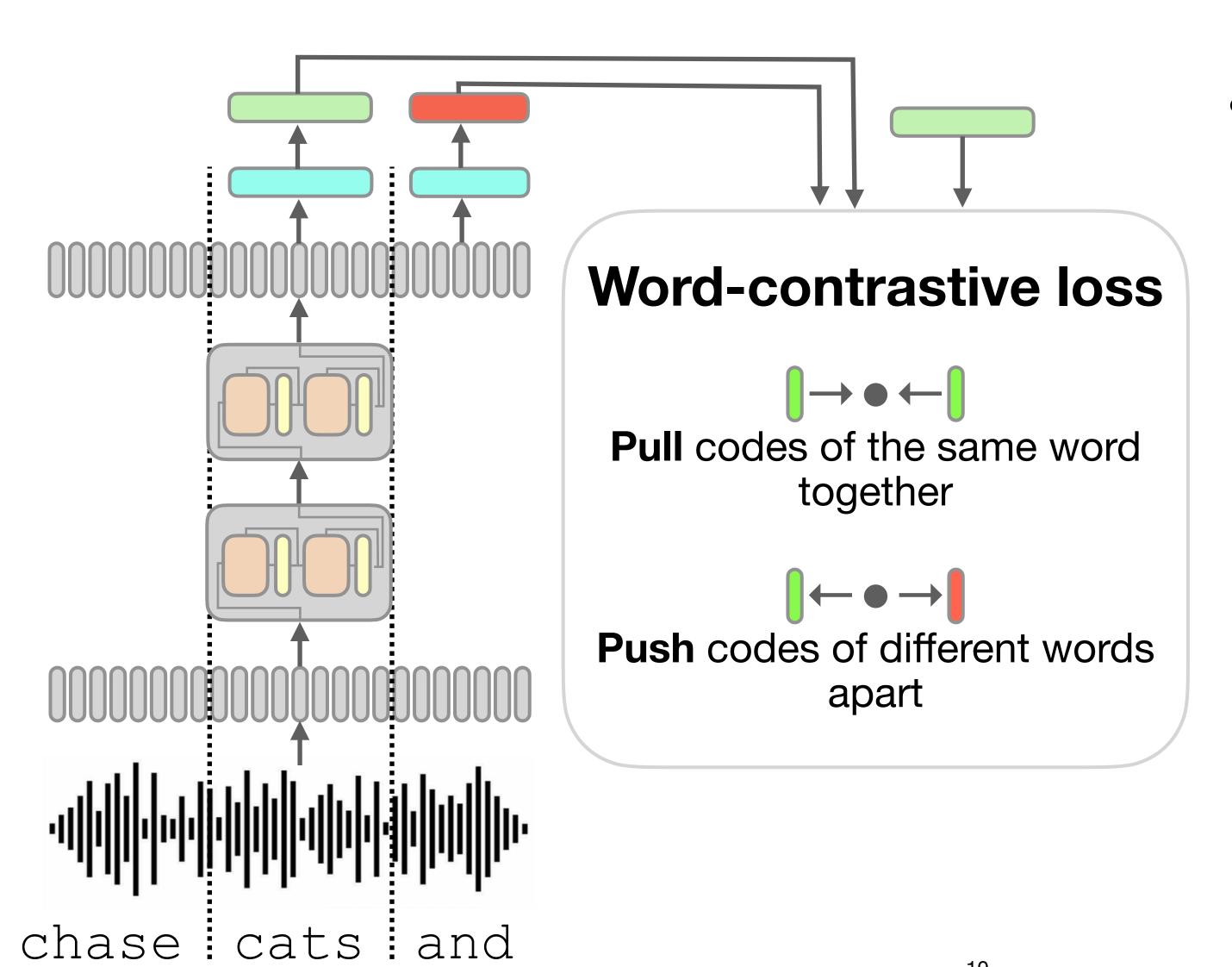
Derive a word recognition model from a self-supervised model

Experiment



Dissect its computations by treating it as an experimental subject

Word recognition model

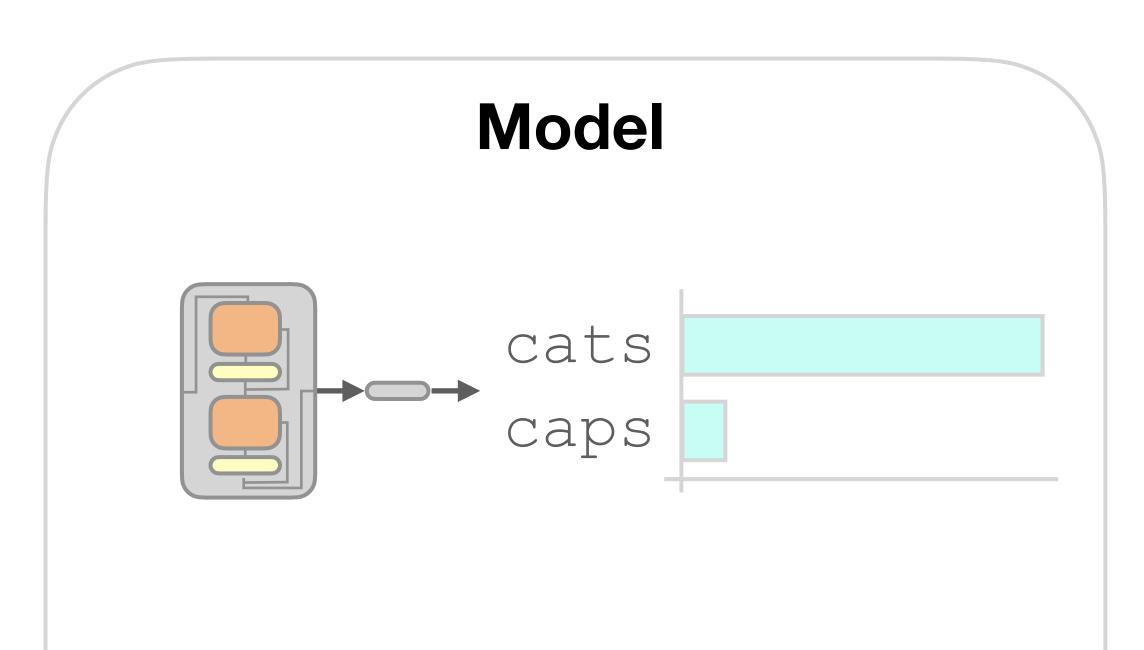


 We compute embeddings for every word token in a test corpus:

wav2vec
Audio-contrastive
embedding

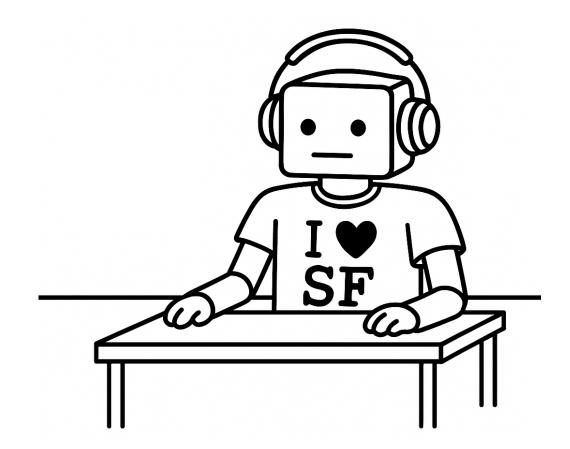
Word-Word-contrastive embedding

Plan



Derive a word recognition model from a self-supervised model

Experiment



Dissect its computations by treating it as an experimental subject

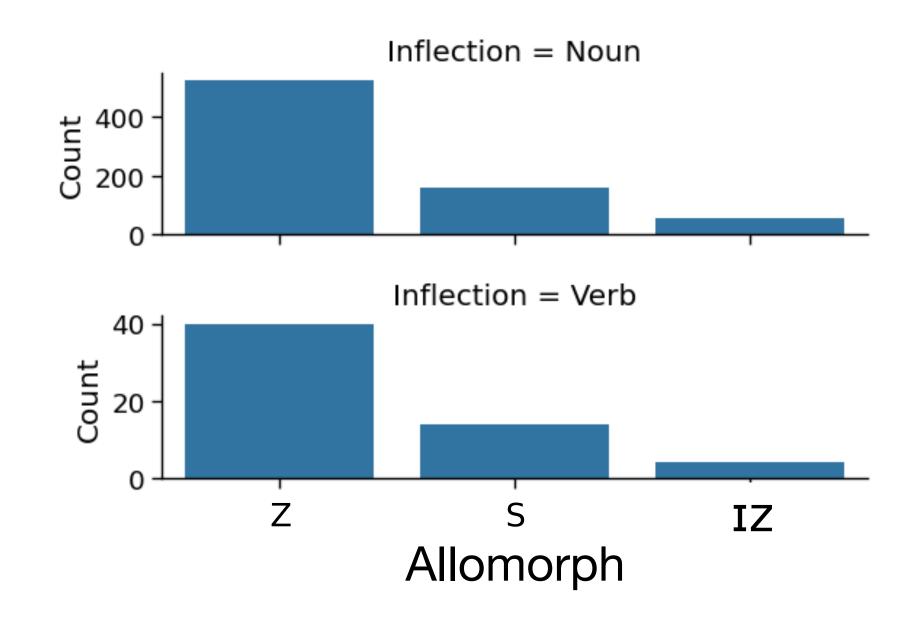
Phenomenon

- Word-final [z], [s], [IZ]
- Distributed by multiple morphological processes
- Governed by phonological rules:
 - [IZ] after sibilants
 - [z] after voiced segments
 - [s] after voiceless segments

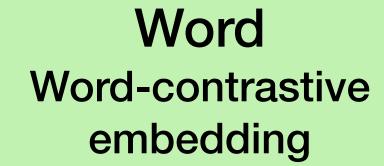
	Base	Inflected
	daughter	daughters
	lip	lips
	age	ages
	bring	brings
	speak	speaks
	please	pleases

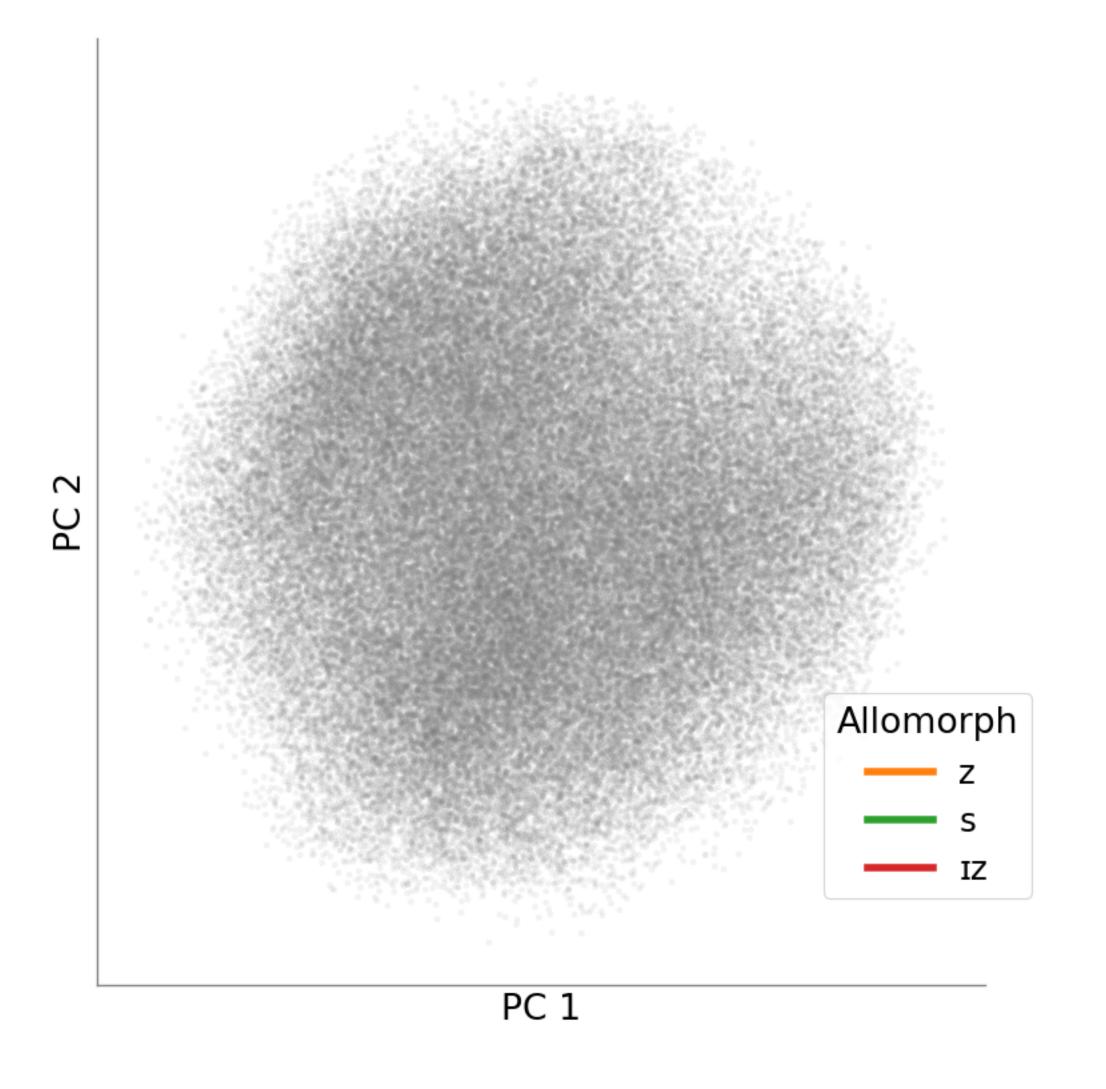
Corpus

- LibriSpeech corpus: 960 hours of amateur audiobook recordings (AmE, BrE)
- Source 786 regular nouns and 61 regular verbs whose inflected forms are **unambiguous**, e.g.
 - belongs is only a 3SG verb and not a plural noun
 - currents is only a plural noun and not a 3SG verb

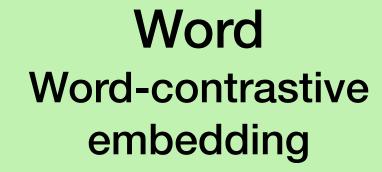


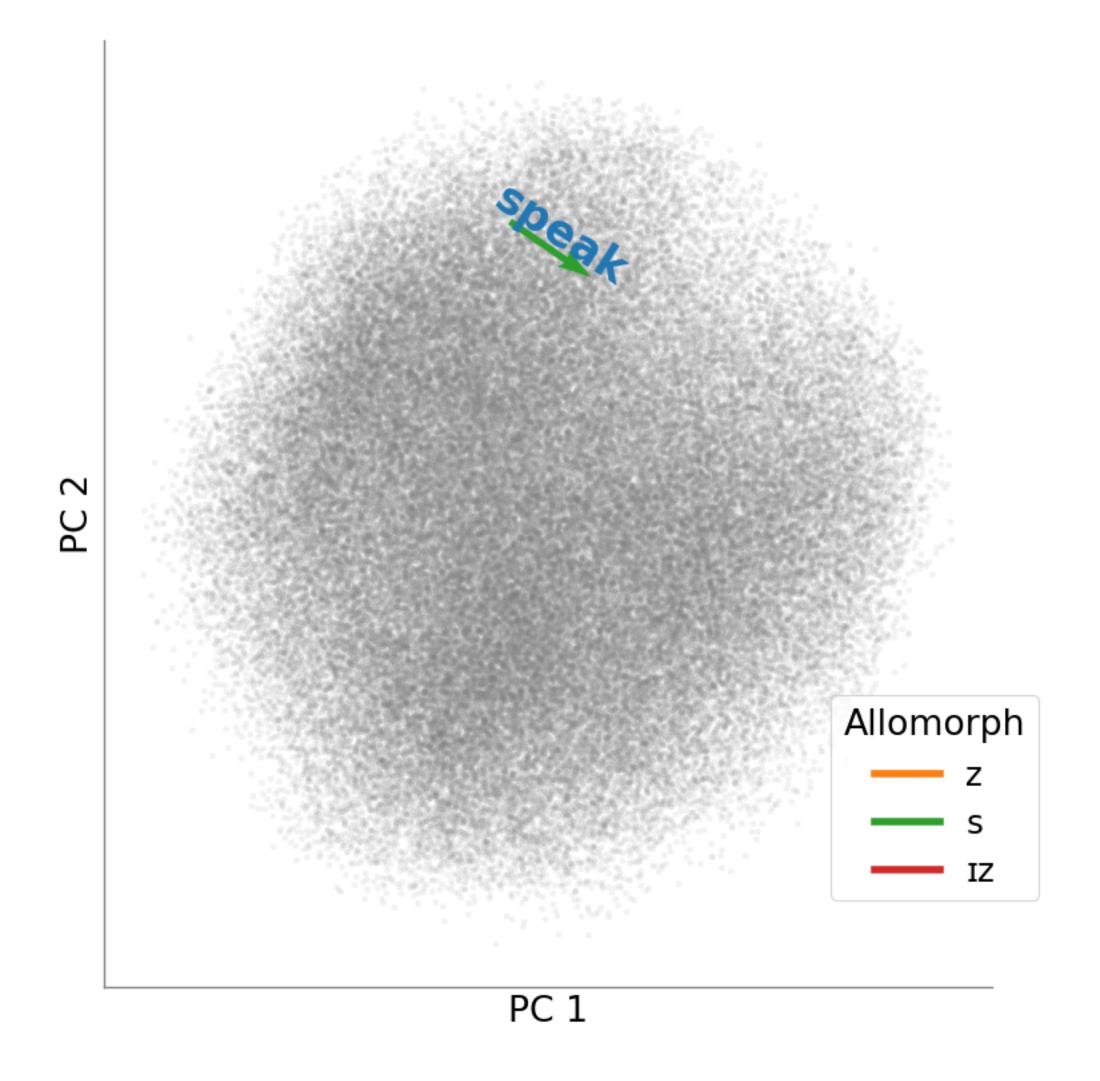
Global linear geometry



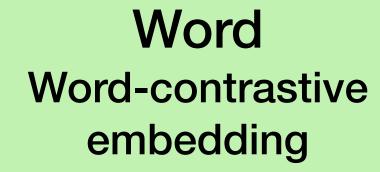


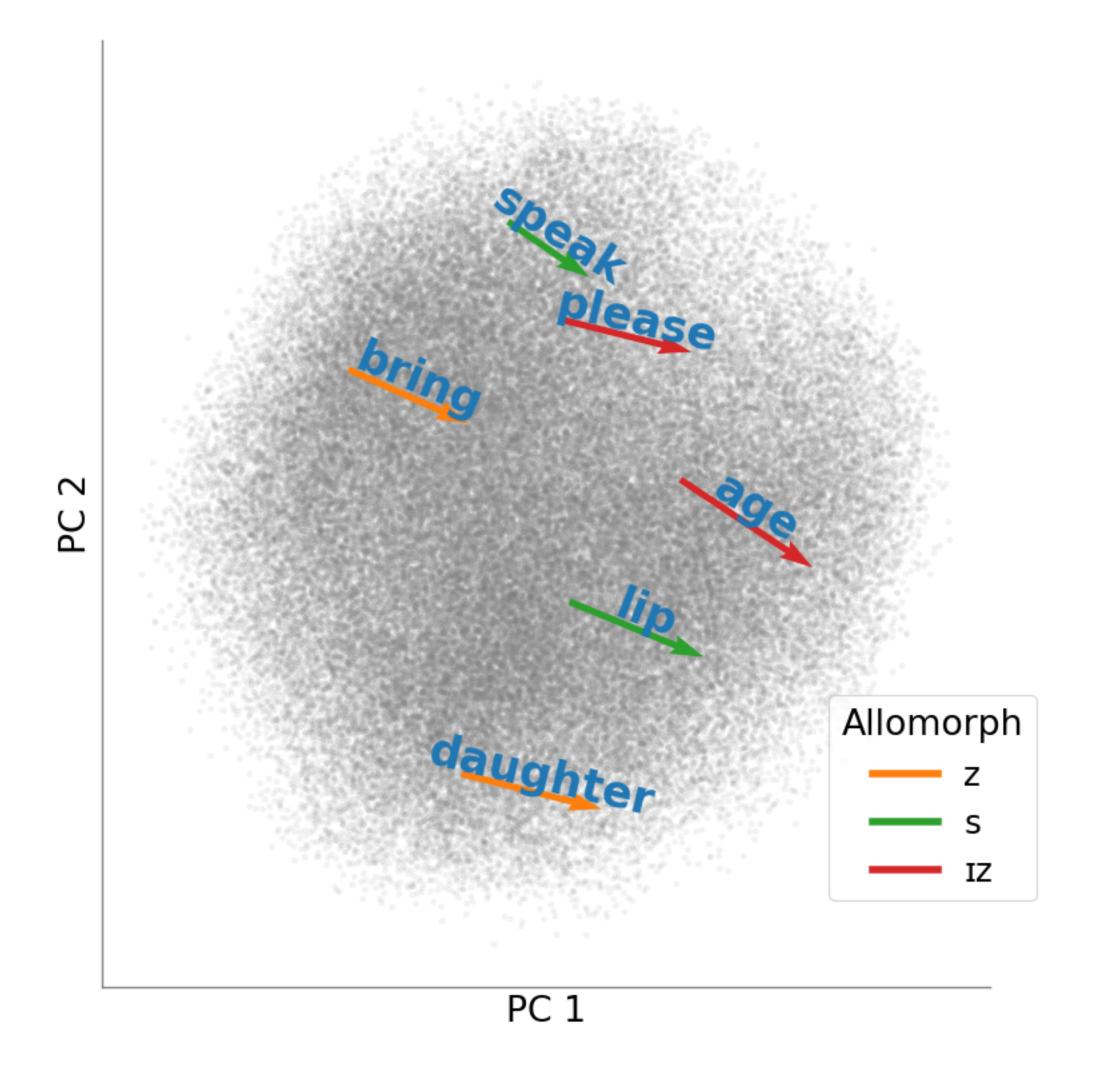
Global linear geometry





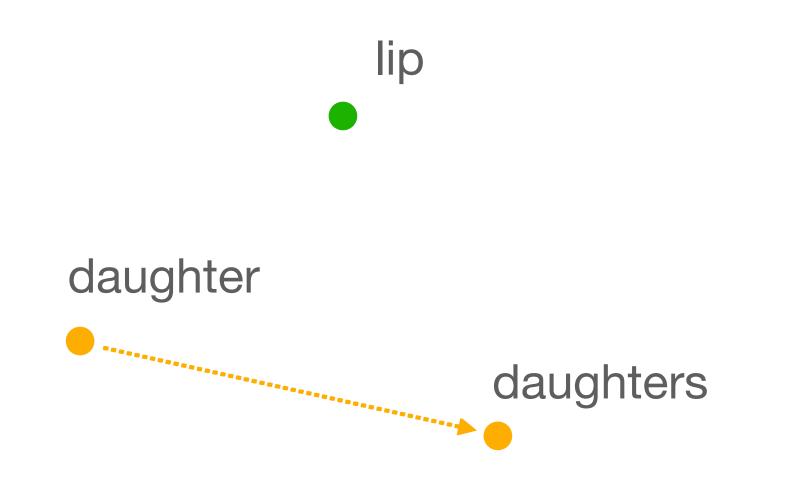
Global linear geometry

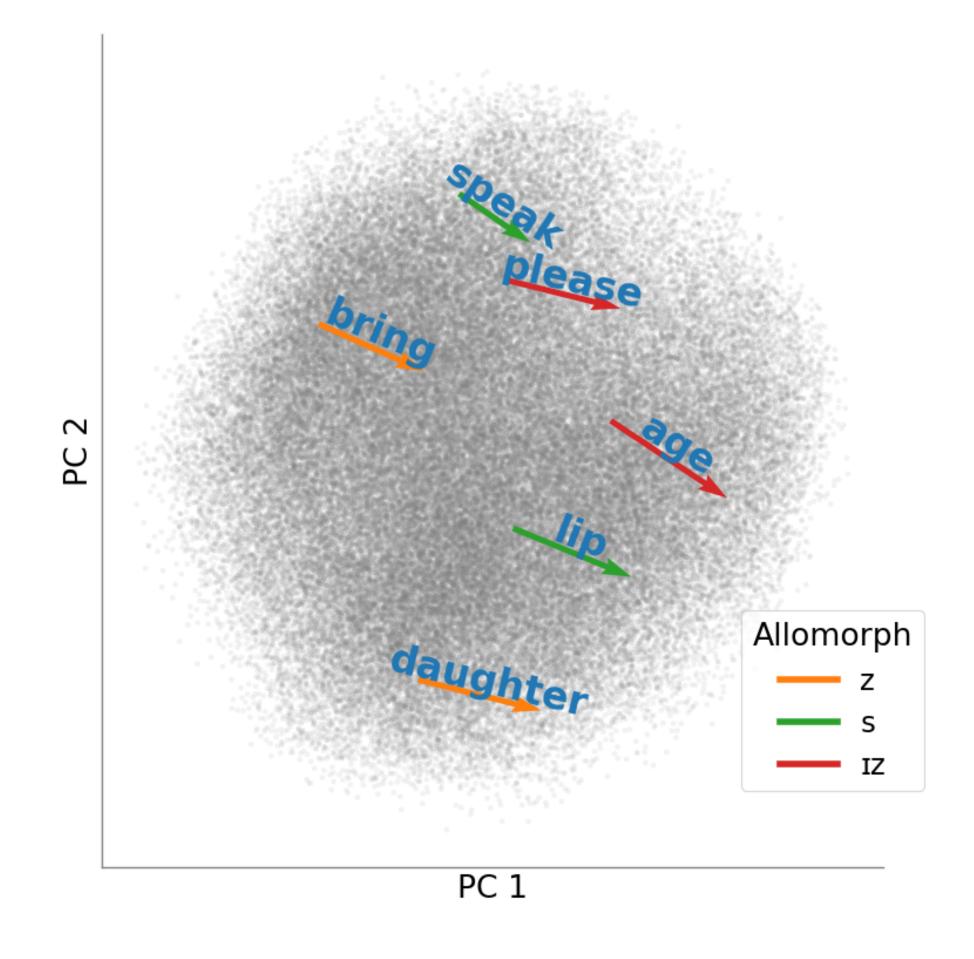




Hypothesis

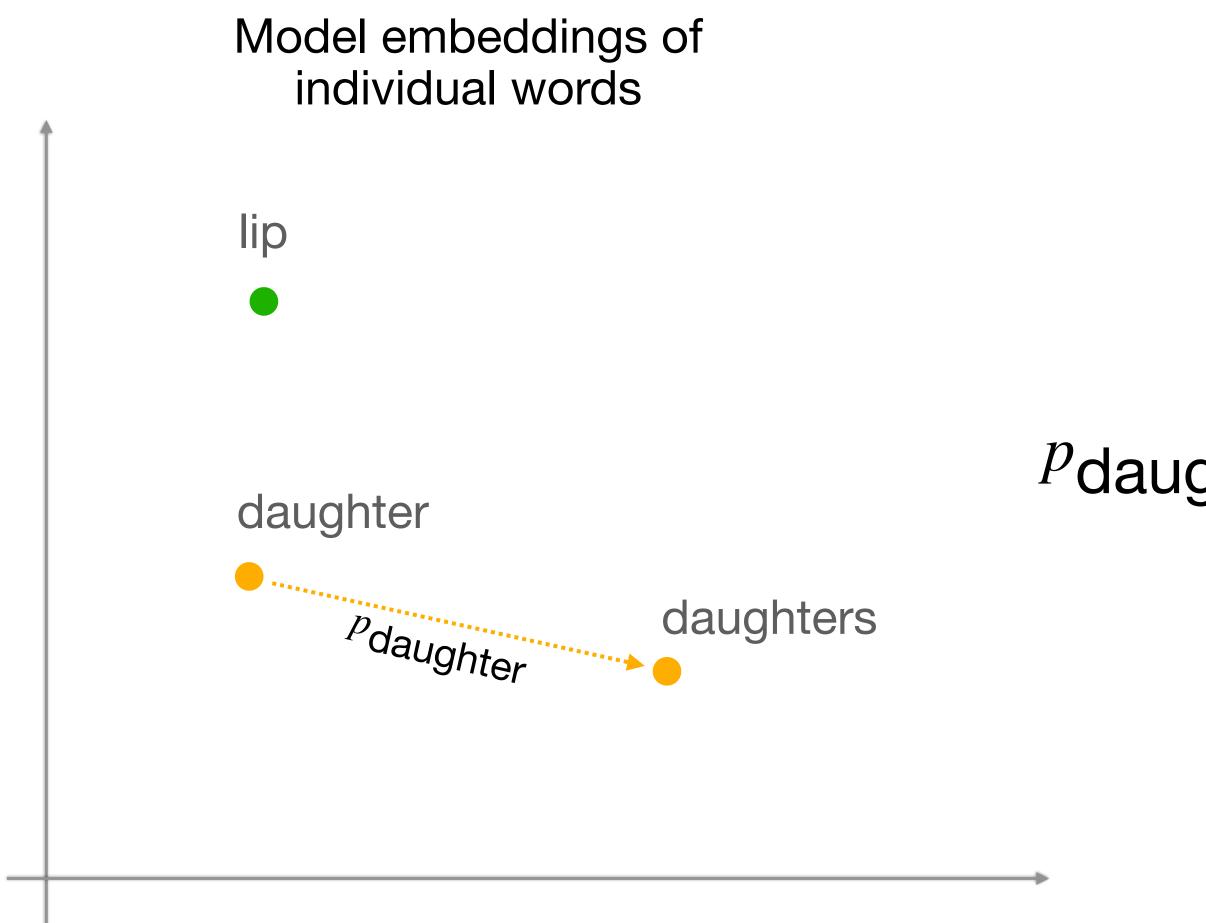
A global **linear** translation links the representations of **base** and **inflected** forms





Prediction

daughter : daughters :: lip :

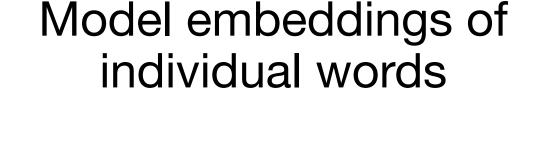


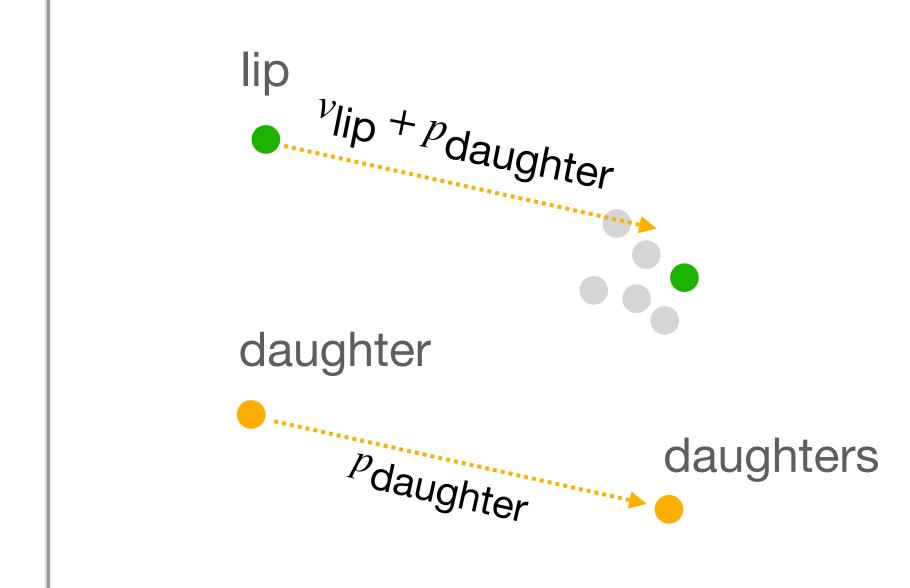
Compute analogy by vector algebra:

 $p_{\text{daughter}} = v_{\text{daughters}} - v_{\text{daughter}}$

Prediction

daughter : daughters :: lip : _____



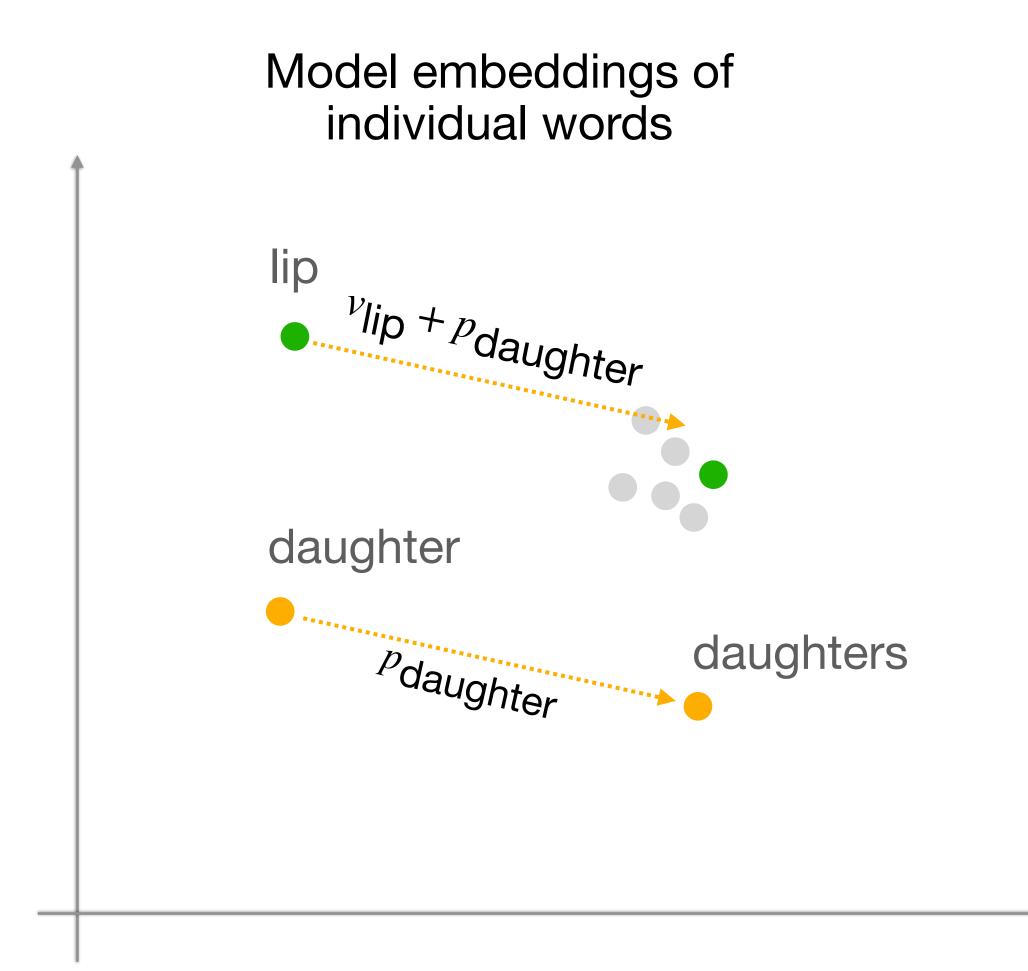


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daughter : daughters :: lip : _____



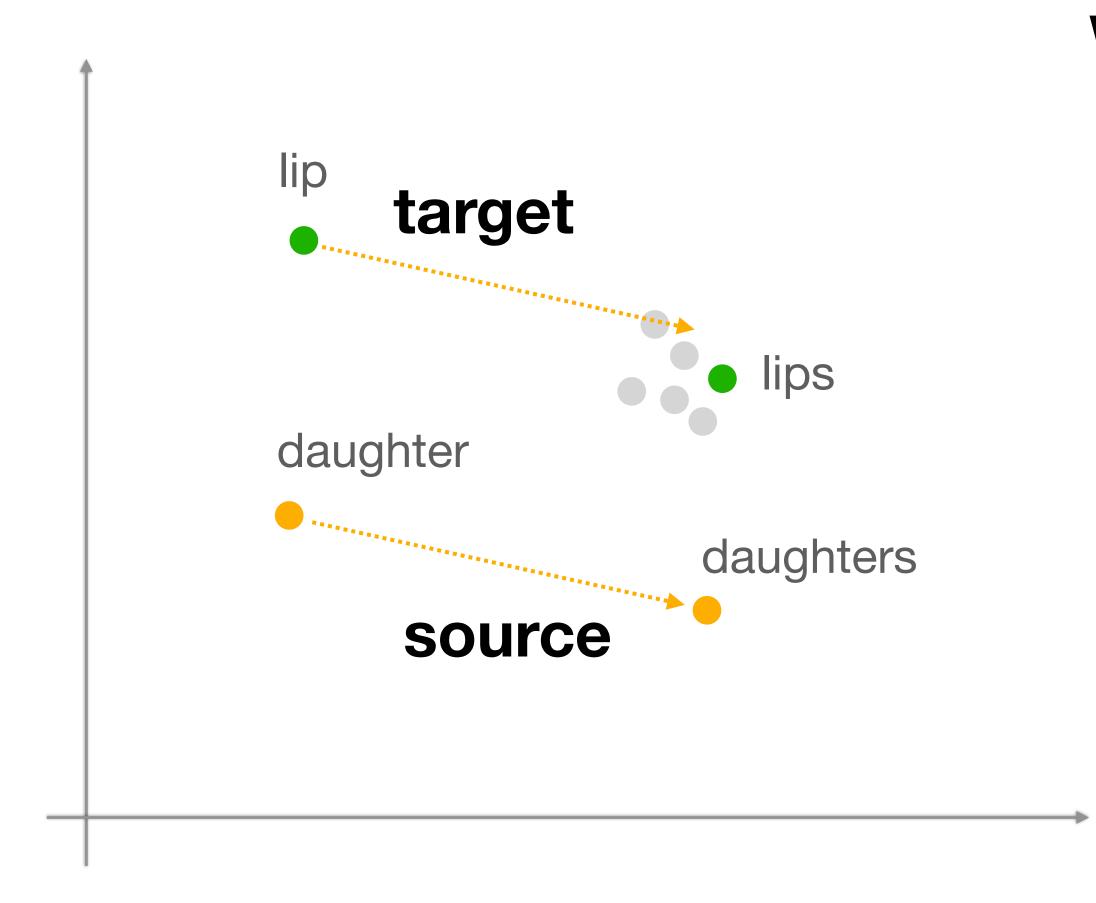
Compute analogy by vector algebra:

 $p_{\text{daughter}} = v_{\text{daughters}} - v_{\text{daughter}}$

Rank evaluation:

Rank	Word	
0	list	
1	less	
2	lips	
3	lend	

Experimental questions

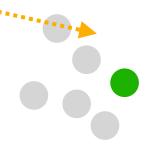


What is encoded in this translation?

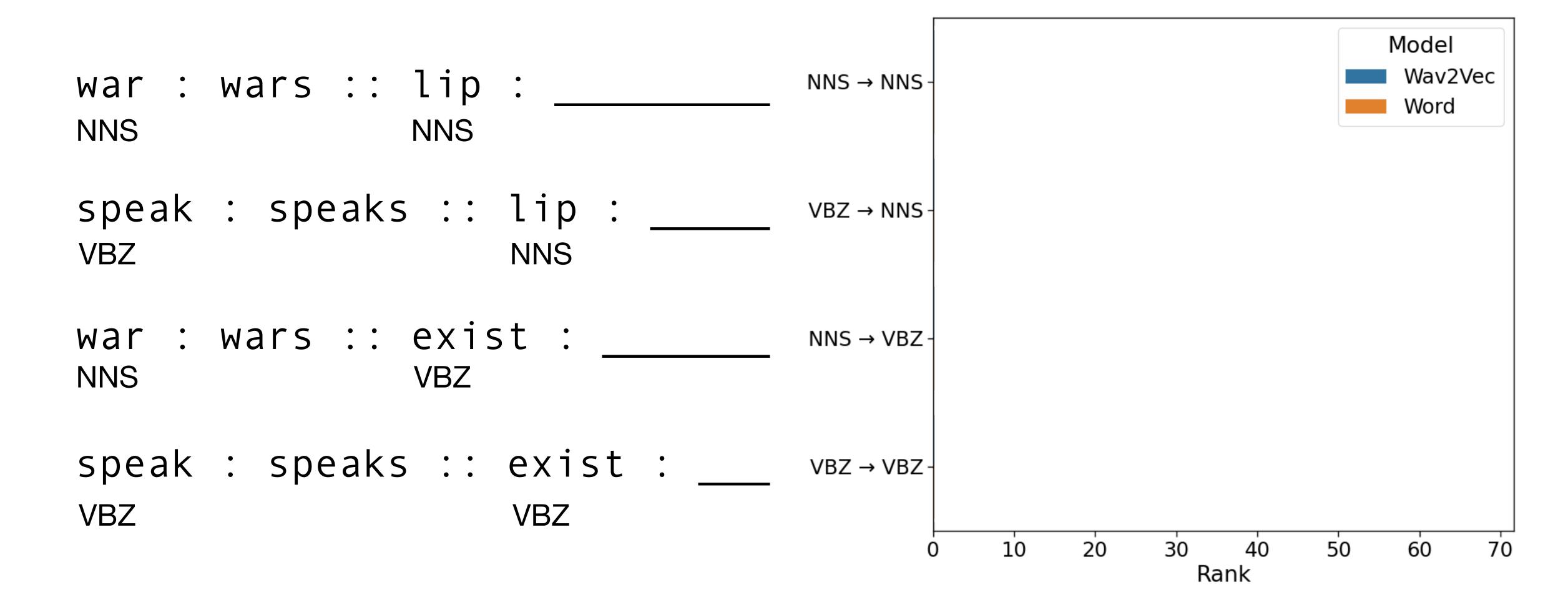
- Is it a morphological transformation?
- Is it a **phonological** transformation?
- How does this vary in a model trained for word recognition?

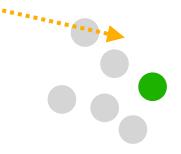
wav2vec
Audio-contrastive
embedding

Word-Word-contrastive embedding

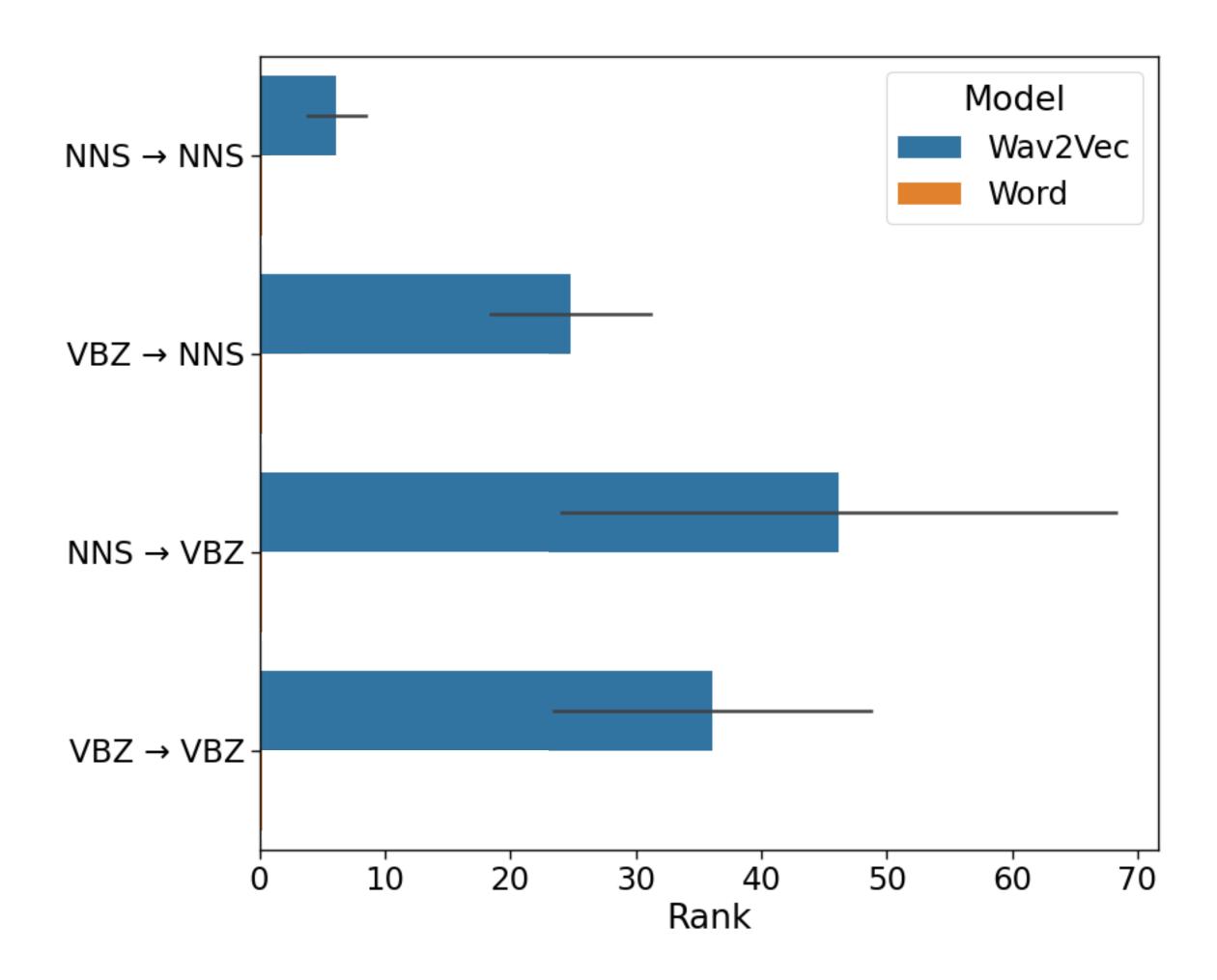


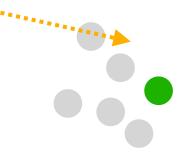
Is this a morphological transformation?





Wav2vec
(audio-contrastive)
model shows sensitivity to
morphological distinctions

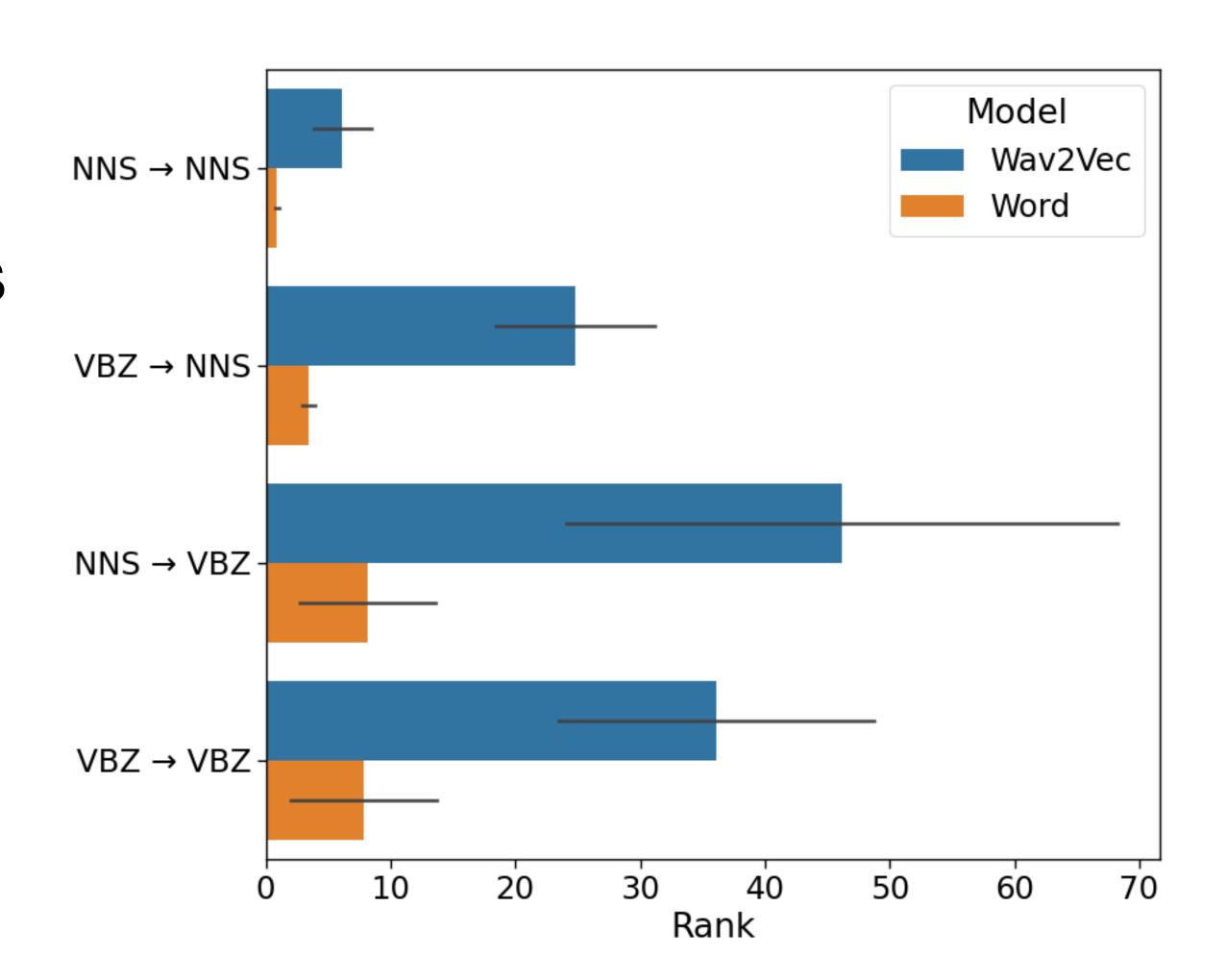




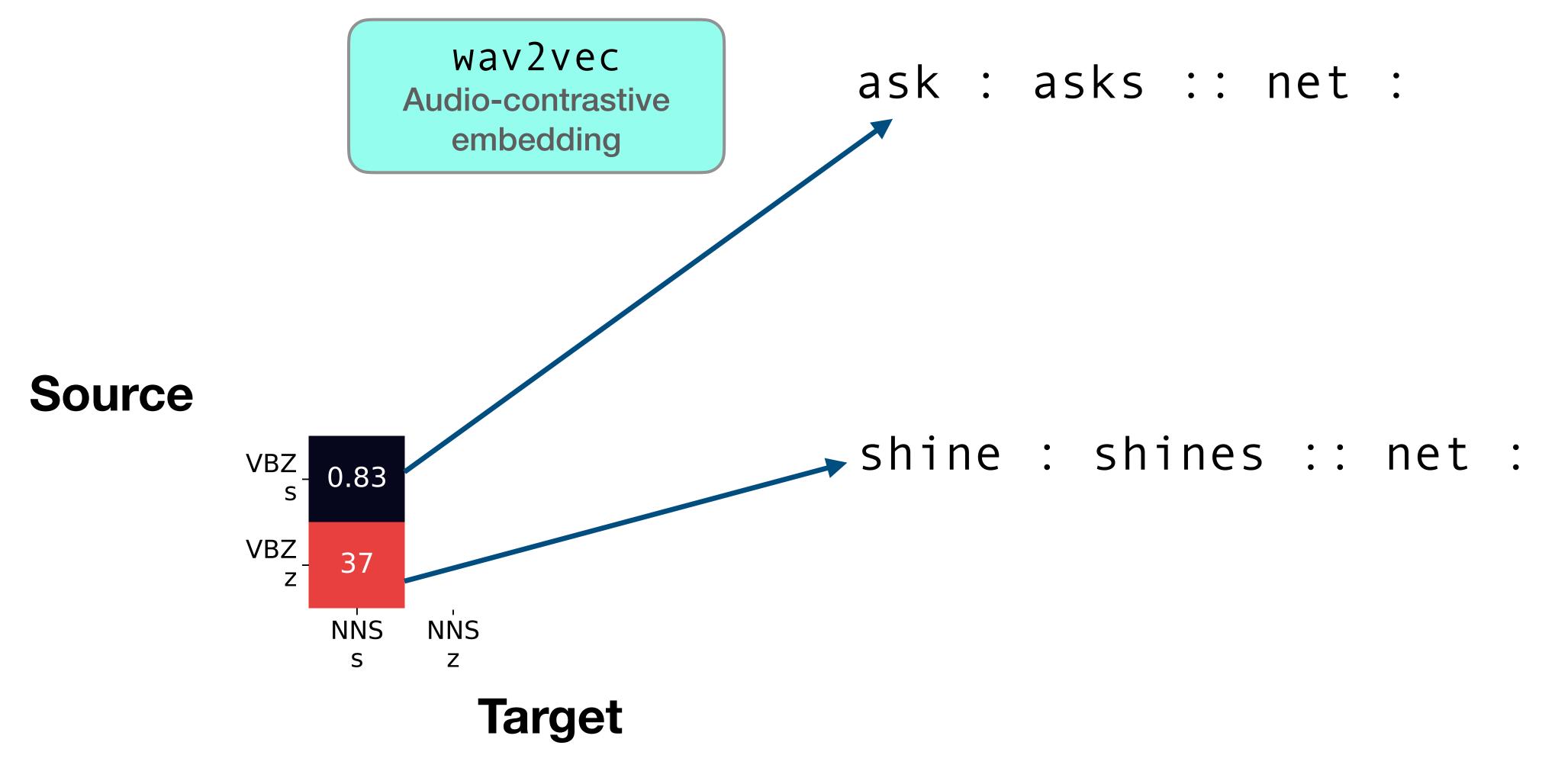
Word-contrastive model shows

reduced sensitivity to

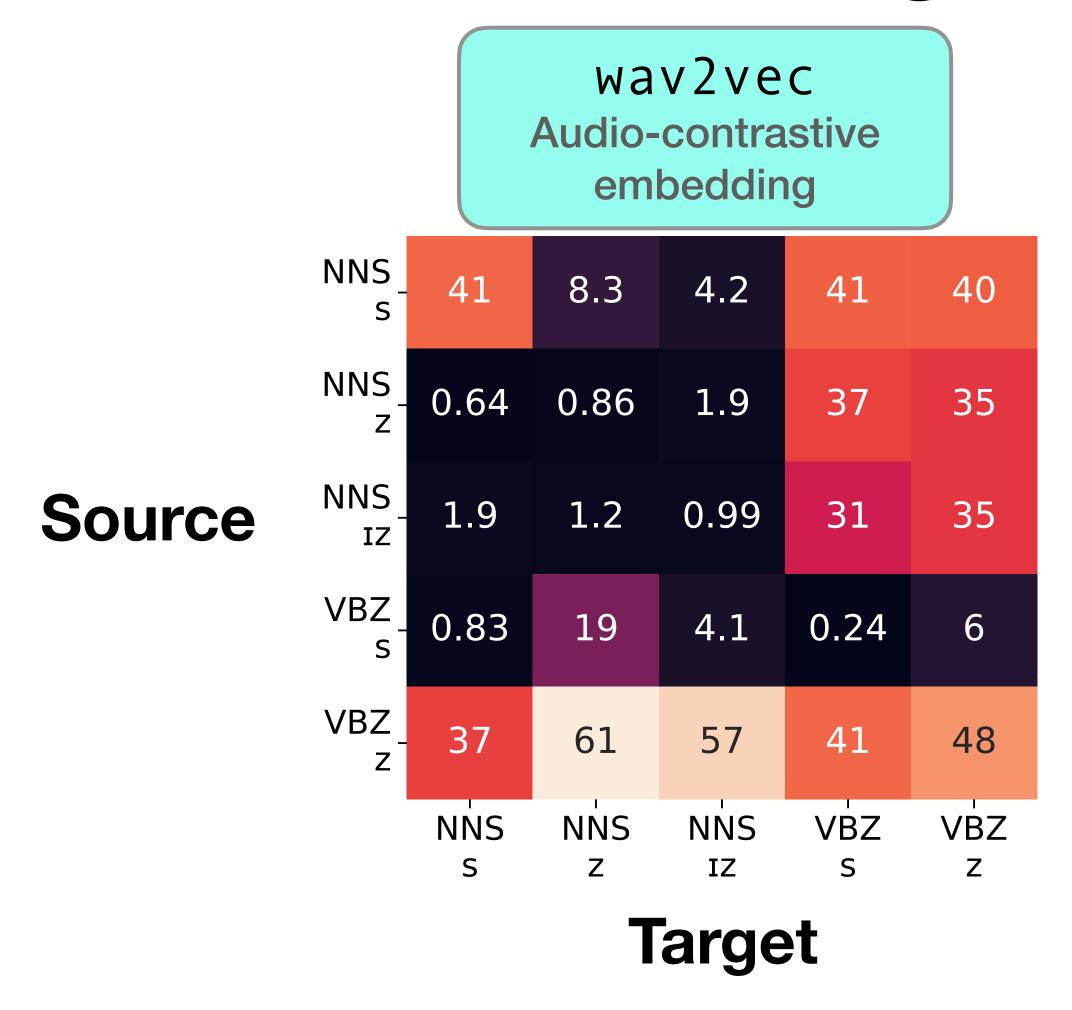
morphological distinctions



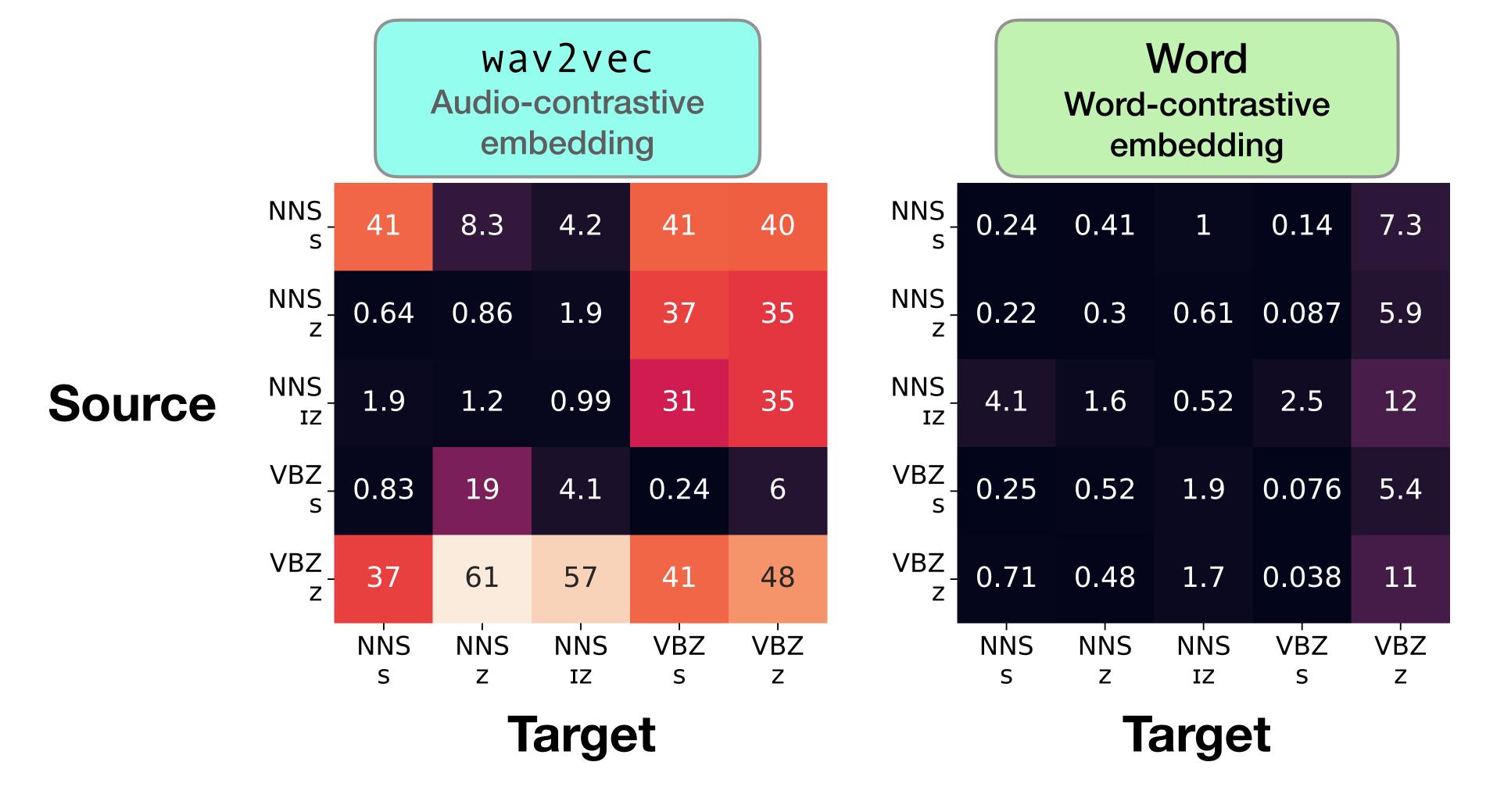
Is this a phonological transformation?

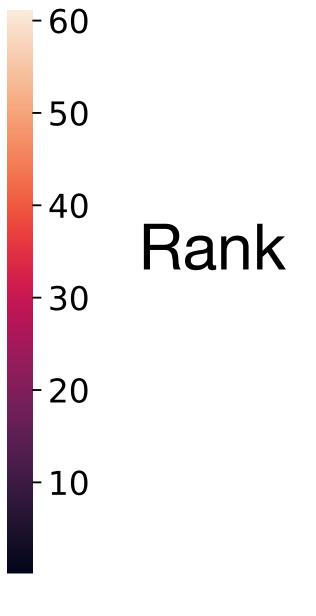


Is this a phonological transformation?



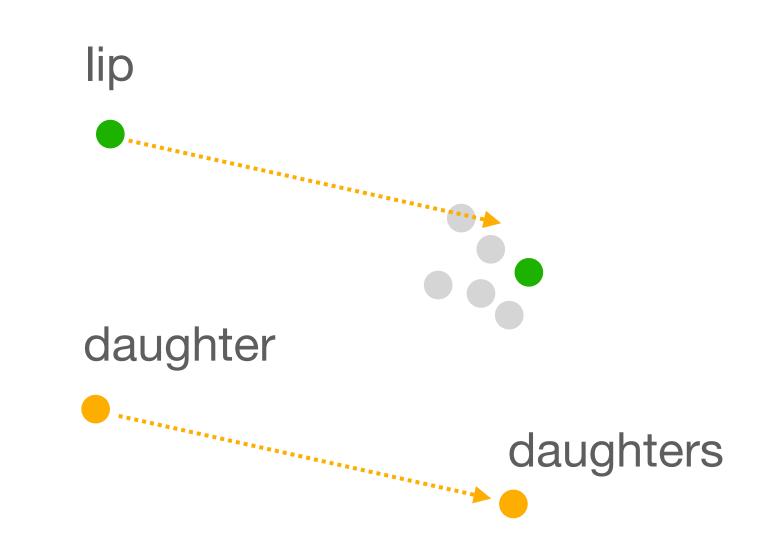
Is this a phonological transformation?





Interim summary

 wav2vec's representations are sensitive to both morphological (noun plurals vs. verbs) and phonological ([z], [s], [ɪz]) distinctions



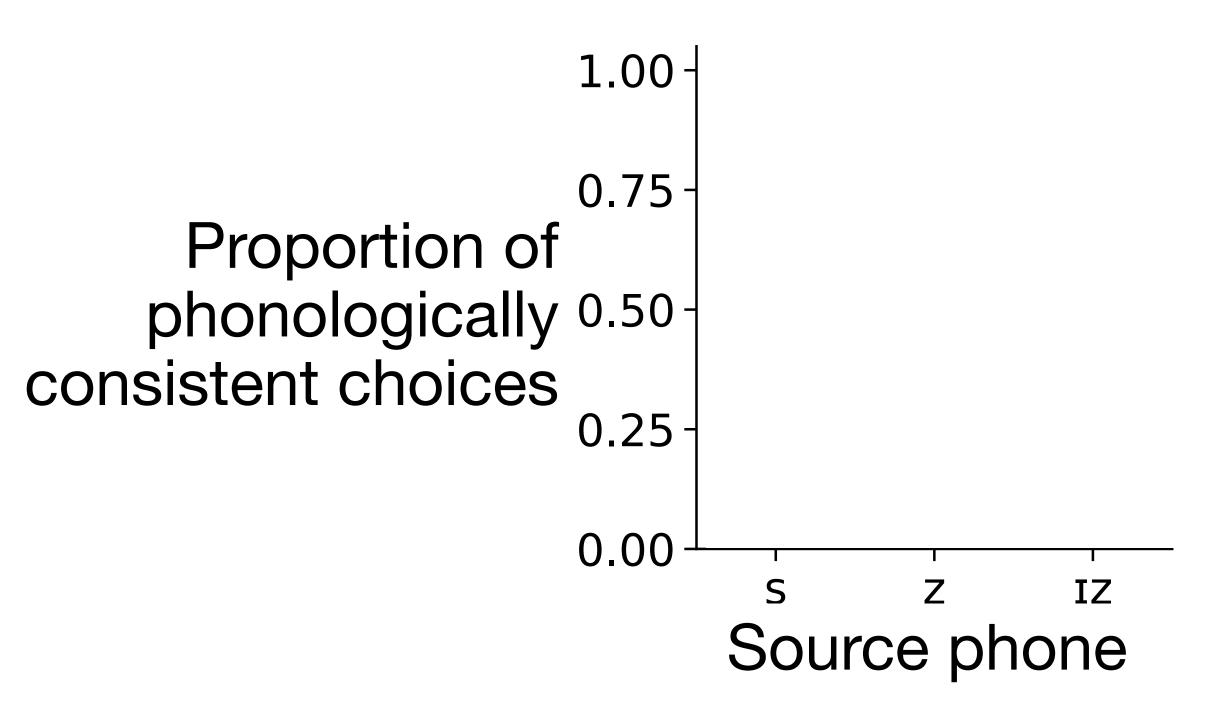
- Optimizing for word recognition minimizes these distinctions
- What about cases where phonological distinctions matter?

Hypothesis: analogy maps to the phonologically consistent item

```
own : owns :: bay : { bays (consistent) base (inconsistent) }

[s]
lip : lips :: bay : { bays (consistent) base (inconsistent) base (inconsistent) }
```

Phonological consistency



A direction in model space

encodes a phonological rule:

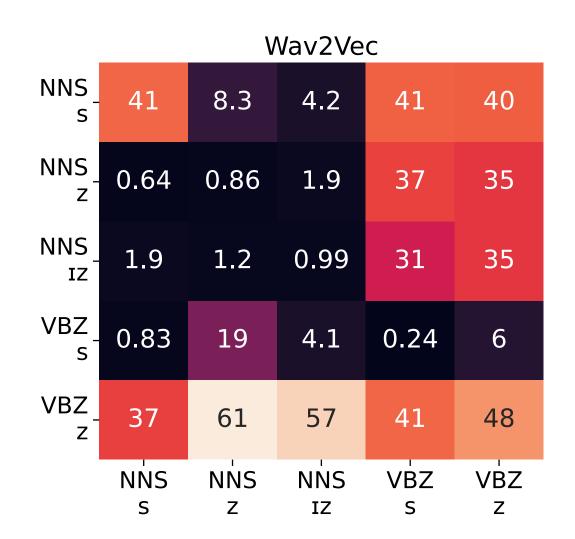
Add the phonologically consistent choice of [z], [s], [ɪz], as in noun plurals and verb inflections

```
[z]
own: owns: bay: bays (consistent)
base (inconsistent)
```

Conclusion: for modelers

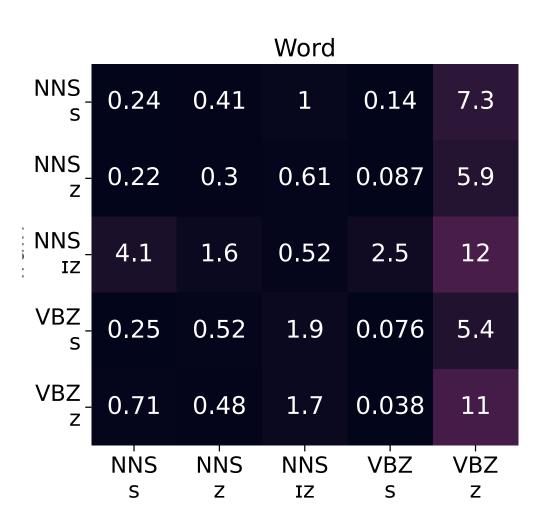
Unconstrained task

"What representations do speech models use?"

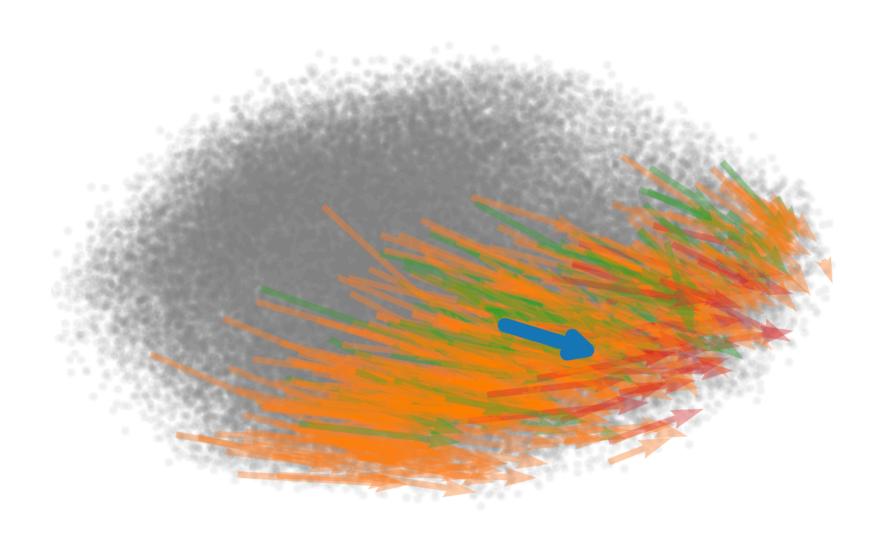


Specific task

"What representations do speech models use for spoken word recognition?"



- An optimal word recognition model tracks the phonological rules involved in noun and verb inflections using a simple geometric relationship
- This is an abstract computation, bridging phonology and morphology
- Next: use these findings to design predictions about the neural implementation of speech comprehension

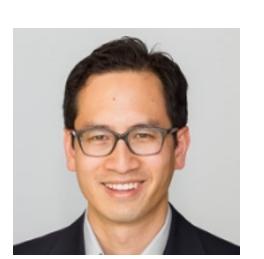








Matt Leonard



Edward Chang



