



# Analyzing the Surprising Variability in Word Embedding Stability Across Languages

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## Introduction

Does stability vary for different languages?  
Is stability associated with linguistic properties?

### Data

Wikipedia (40 languages)

Bible (97 languages)

World Atlas of Language Structures (WALS), phonological, lexical, and grammatical properties (>2,000 languages)

### Why word2vec and GloVe?

These algorithms continue to be used in many situations, including the computational humanities and low-resource languages!



## What is Stability?

Stability = percent overlap between ten nearest neighbors in an embedding space

$$\text{stability} = \frac{100}{|\text{words}|} \sum_{\text{words}} \frac{\text{neighbors}_0 \cap \text{neighbors}_1}{10}$$

neighbors<sub>0</sub> = ten words most similar to the word in embedding space 0

neighbors<sub>1</sub> = ten words most similar to the word in embedding space 1

Example: *international* in 2 embedding spaces

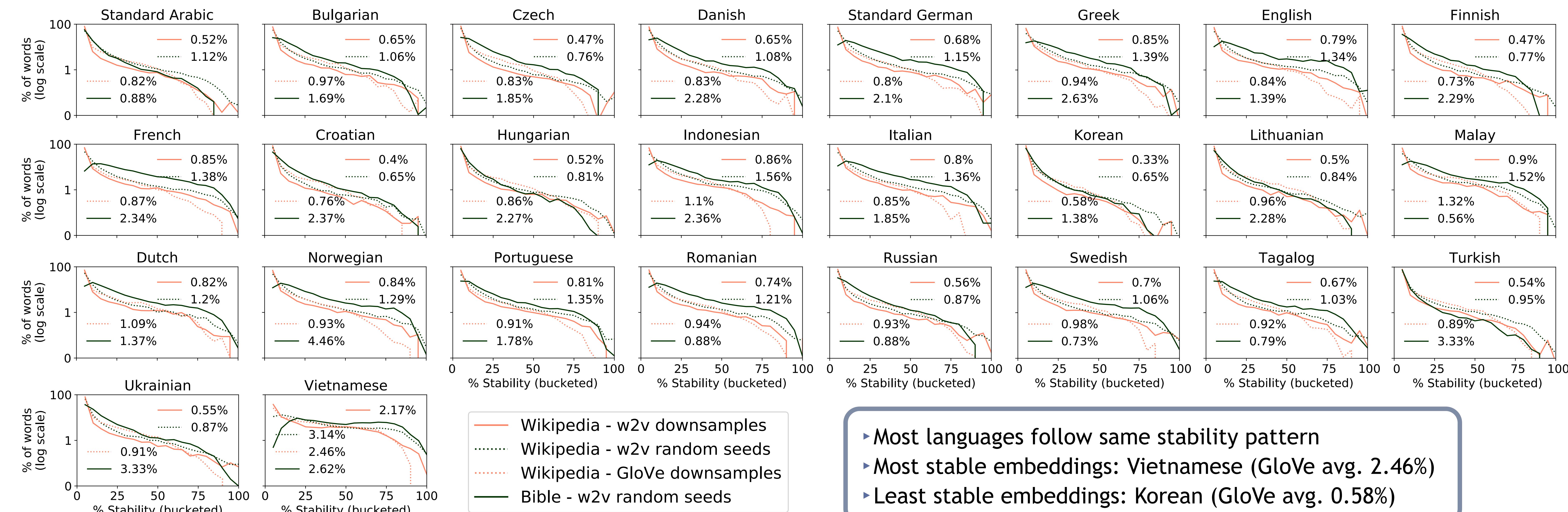
Stability = 40%

| Model 1        | Model 2      |
|----------------|--------------|
| metropolitan   | ballet       |
| national       | metropolitan |
| egyptian       | bard         |
| rhode          | chicago      |
| society        | national     |
| debut          | state        |
| folk           | exhibitions  |
| reinstallation | society      |
| chairwoman     | whitney      |
| philadelphia   | rhode        |

## Stability for Wikipedia and the Bible

We compare the stability of embeddings for 26 languages.

- Wikipedia (3 settings): Stability of...
  - GloVe embeddings across 5 downsampled corpora
  - word2vec (w2v) embeddings across 5 downsampled corpora
  - w2v using 5 random seeds on 1 downsampled corpus



- Most languages follow same stability pattern
- Most stable embeddings: Vietnamese (GloVe avg. 2.46%)
- Least stable embeddings: Korean (GloVe avg. 0.58%)

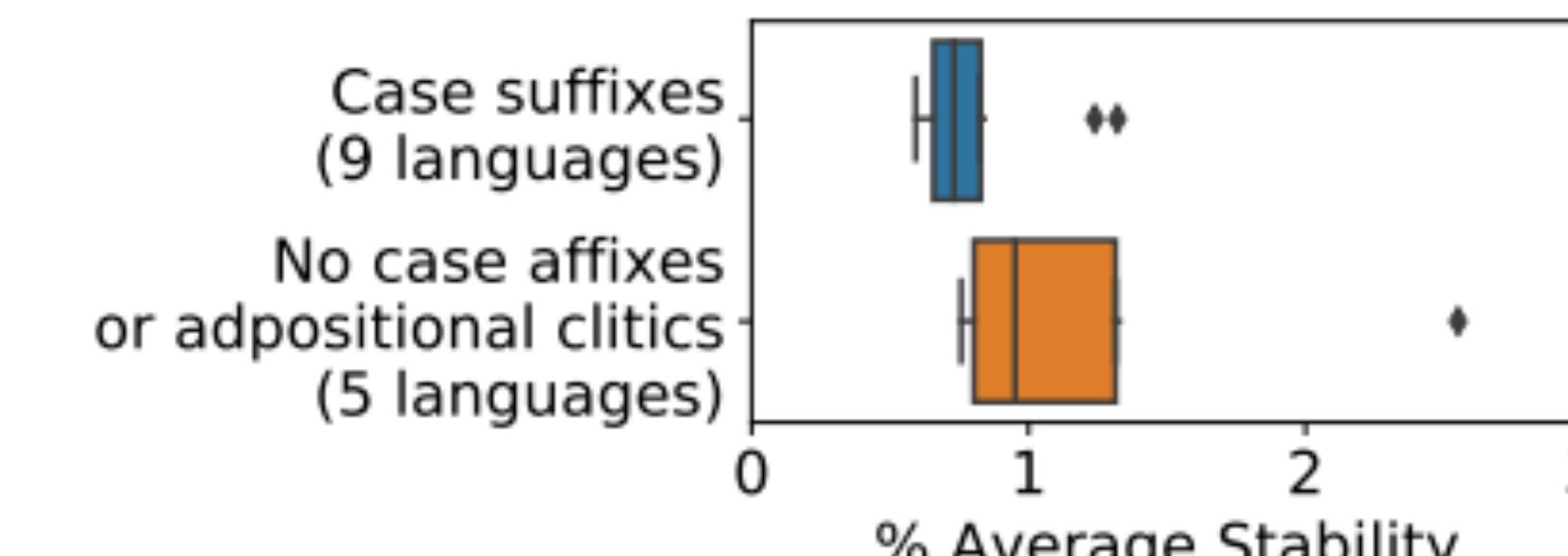
## Regression Modeling

We use a regression model to predict stability in a language using linguistic properties.

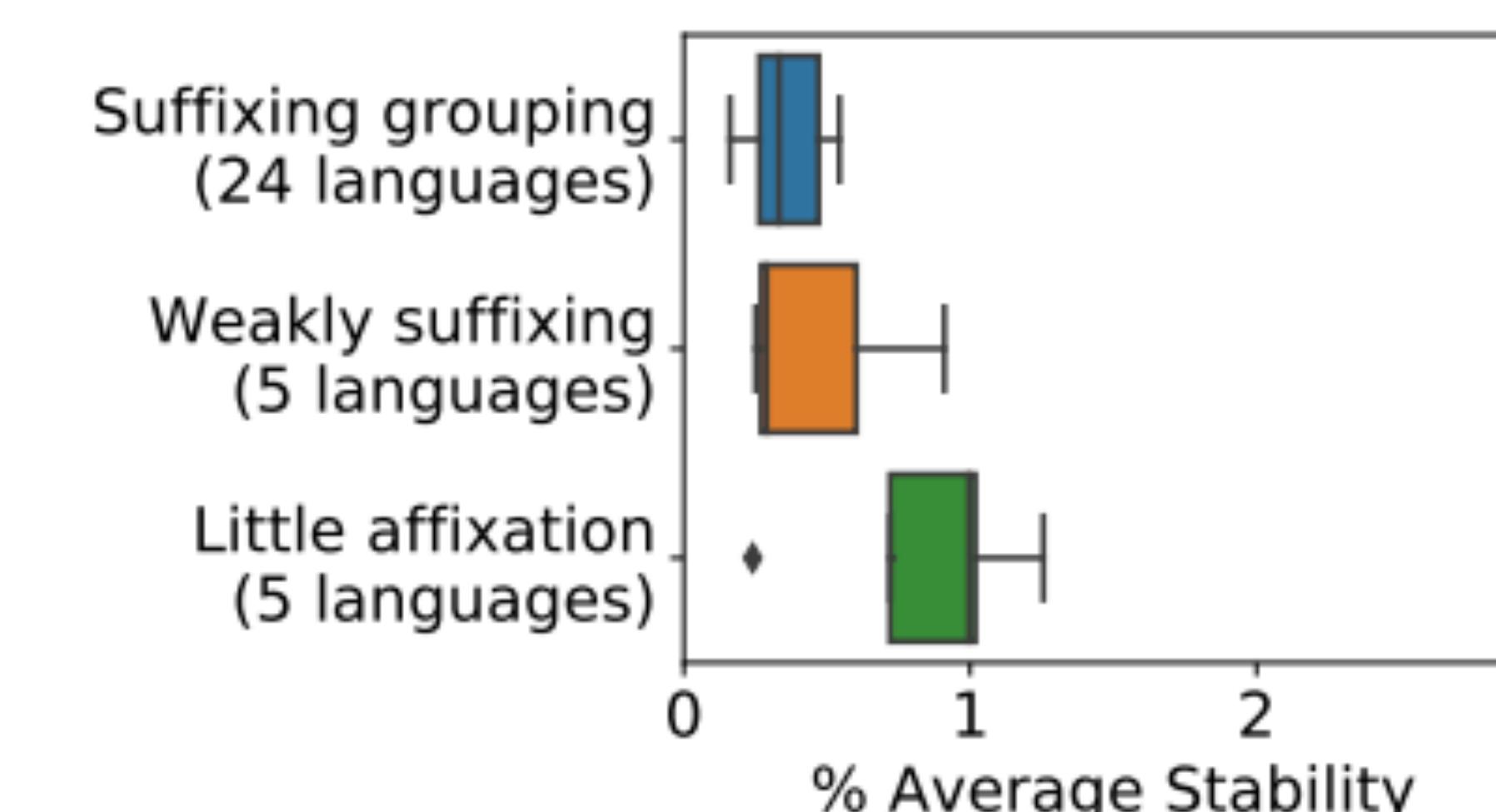
- Ridge regression
- 37 languages
- Input: 97 WALS properties
- Output: Average stability of all the words in a language
- High  $R^2$  score of  $0.96 \pm 0.00$

- More affixing (suffixing and prefixing) associated with lower stability
  - Affixes cause increased word variation
- Languages with no gender system associated with higher stability
  - Languages with gender systems have more word forms

## Selected WALS Properties Associated with Affixing

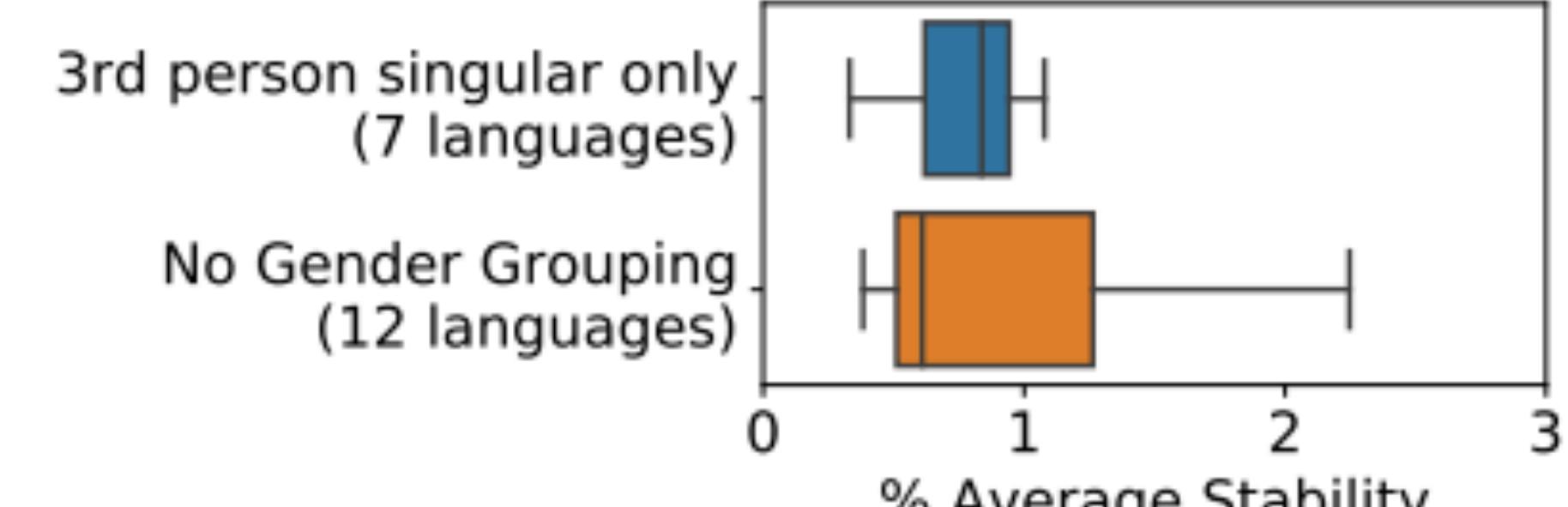


### Position of Case Affixes

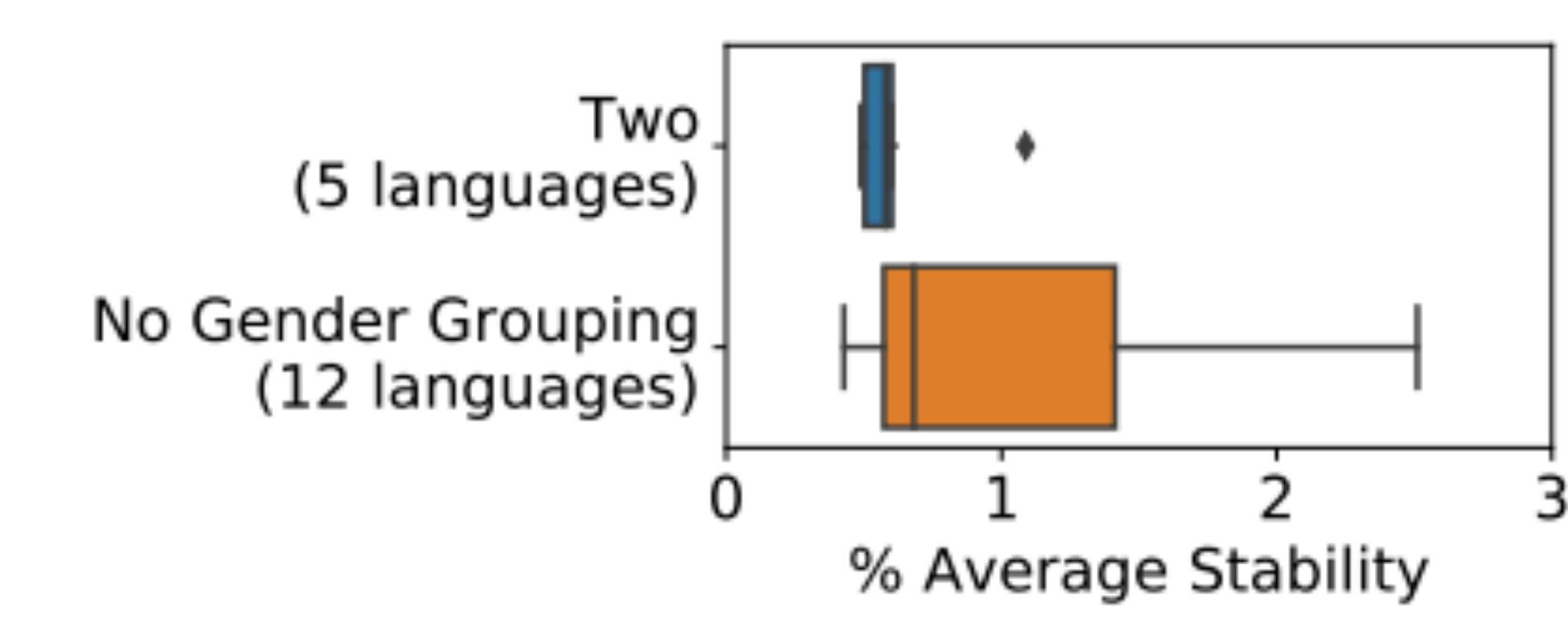


Prefixing v. Suffixing in Inflectional Languages

## Selected WALS Properties Associated with Gender



### Gender Distinctions in Independent Personal Pronouns



Number of Genders