

Evidence for an encoding of morphological blocking effects within two English word embedding models

Background on morphological blocking

- Blocking is a process by which the existence of a word in the mental lexicon prevents regular affixation, if it would give rise to a complex word with the same meaning [1].
- In English, **-ity** and **-ness** affixation can apply to the same base: *monstros-ity*, *monstrous-ness*.
- Yet, only **-ity** affixation is subject to blocking: *glory* blocks **glorios-ity* but not *glorious-ness* [1]. Why?
 - Supposedly, **-ity** and **-ness** do not operate at the same “level”: **-ity** is **L1** (“word-creating”) while **-ness** is **L2** (“word-modifying”).
 - This division also has phonological manifestations: **-ity** shifts stress (*glóbal* → *globál-ity*), while **-ness** does not (*glóbal-ness*).
 - L1**-derived words may compete with **suppletive forms** from the lexicon, while **L2**-derived words may not.
- But this does not explain why **-ity** is **L1** and **-ness** **L2**!

Contribution

- We show that two recent word embedding models (GloVe [4] and fastText [2])...
 - Distinguish between **L1** and **L2** operations
 - Encode a stronger similarity between **suppletive operations** (such as *glorious* → *glory*) and **L1** operations, as opposed to **L2**.
- We therefore argue that embeddings encode some notion of semantic competition between **L1** vs **suppletive operations**, rather than **L2** vs **suppletive operations**.
- This semantic competition might then drive blocking effects.

Relevance of word embedding models

- Word embeddings are representations of words as vectors encoding their meanings, s.t. the words that are close in the vector space (in terms of cosine similarity) are semantically similar [3].
- Embeddings have been shown to encode morphological operations, such as comparative or superlative affixation, as stable geometrical translations defined as the difference between the affixed word-vector and the word-vector itself [4].
- We argue that the contrast between **-ity** and **-ness** affixation w.r.t. blocking is encoded as a configuration whereby **-ity** is on average closer to **sup.** (vector of the related suppletive transformation) than **-ness** is.

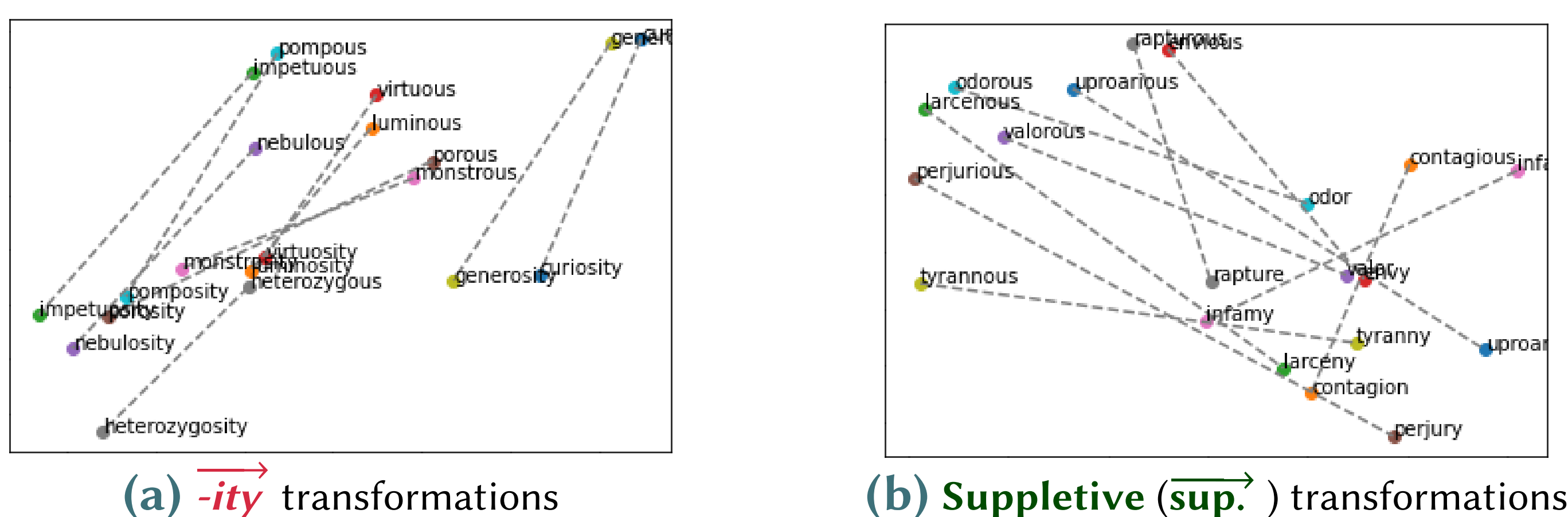


Figure 1: Morphological operations translate into systematic geometrical translations in a 2D-cosine-PCA-reduced space (GloVe embedding)

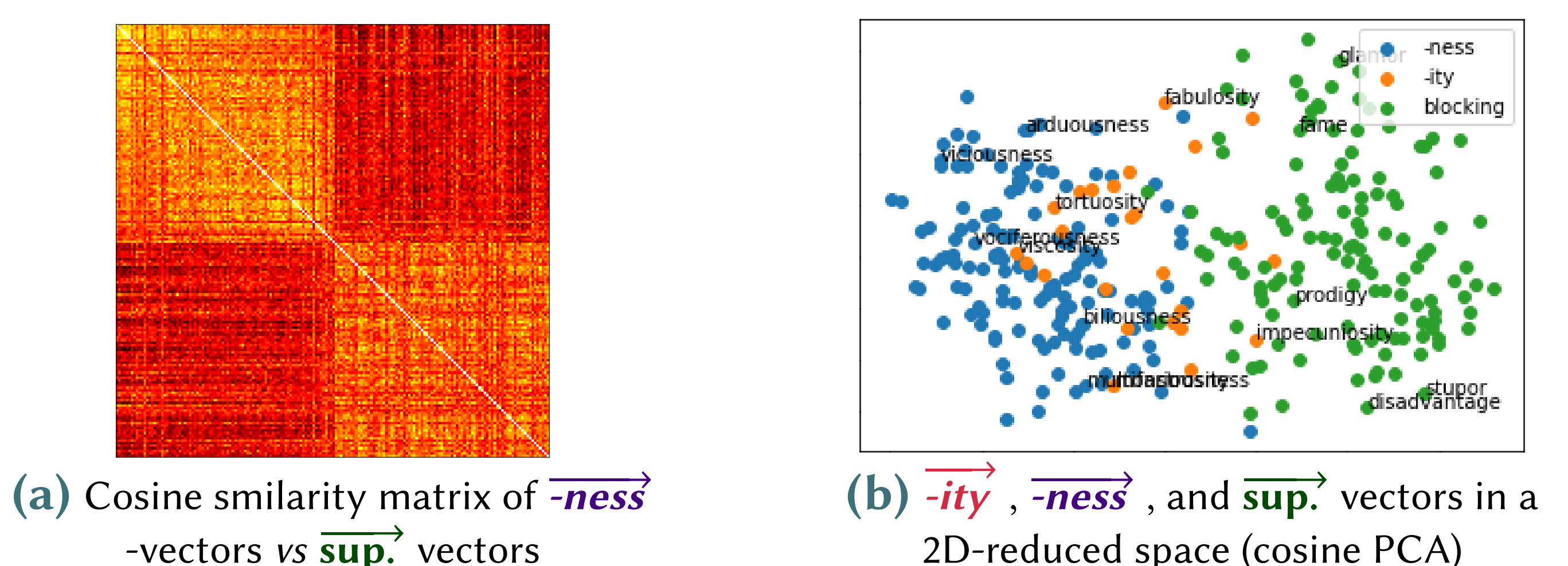


Figure 2: Evidence for a clustering of the vectors corresponding to the 3 kinds of transformations at stake (H_1), and for differences in cluster proximities (H_2)

Modeling of the problem

- We define \mathcal{B}_{-ity} , \mathcal{B}_{-ness} and $\mathcal{B}_{sup.}$ as the sets of base forms in **-ous** compatible with resp. **-ity**, **-ness**, and a **suppletive transformation** of the form *glorious* → *glory*.
- For instance, *monstrous* ∈ $\mathcal{B}_{-ity} \cap \mathcal{B}_{-ness}$, and *glorious* ∈ $\mathcal{B}_{-ness} \cap \mathcal{B}_{sup.}$, but also *glorious* ∉ \mathcal{B}_{-ity} .
- For a base $b \in \mathcal{B}_{-ity} \cup \mathcal{B}_{-ness}$ we define the suffixal vector(s):

$$\begin{aligned} \vec{-ity}_b &= \vec{b-ity} - \vec{b} && \text{if } b \text{ in } \mathcal{B}_{-ity} \\ \vec{-ness}_b &= \vec{b-ness} - \vec{b} && \text{if } b \text{ in } \mathcal{B}_{-ness} \end{aligned}$$

- For a base $b \in \mathcal{B}_{sup.}$ (e.g., *glorious*) we have a suppletive form s_b (e.g., *glory*). We then define the suppletive vector:

$$\vec{sup.}_b = \vec{s_b} - \vec{b}$$

- For each possible label $X := -ity, -ness$ or $sup.$, we define the set of transformation vectors (suffixal or suppletive), and a measure of similarity between such sets:

$$\begin{aligned} \mathcal{V}_X &= \{\vec{X}_b \mid b \in \mathcal{B}_X\} \\ \mathcal{S}(X, Y) &= \text{Mean}(\{\text{CosSim}(\vec{x}, \vec{y}) \mid (\vec{x}, \vec{y}) \in \mathcal{V}_X \times \mathcal{V}_Y\}) \\ \mathcal{S}(X, X) &\triangleq \mathcal{S}(X) : \text{Intra-group similarity} \\ \mathcal{S}(X, Y), X \neq Y &: \text{Inter-group similarity} \end{aligned}$$

Hypotheses and testing

- H_1 : \mathcal{V}_{-ity} , \mathcal{V}_{-ness} and $\mathcal{V}_{sup.}$ form clusters, i.e.:

$$\forall X \neq Y \in \{-ity, -ness, sup.\}. \mathcal{S}(X) > \mathcal{S}(X, Y) \quad (H_1)$$
- H_2 : \mathcal{V}_{-ity} is closer to $\mathcal{V}_{sup.}$ than \mathcal{V}_{-ness} is, i.e.:

$$\mathcal{S}(-ity, sup.) > \mathcal{S}(-ness, sup.) \quad (H_2)$$
- Prior to testing, the embedding (GloVe or fastText) dimension was reduced twice using PCA:
 - Reduction based on the word vectors from the dataset (base words, plus their suffixed or suppletive counterparts).
 - Reduction based on the transformation vectors ($\vec{-ity}_b$, $\vec{-ness}_b$, $\vec{sup.}_b$) computed in the intermediate space.
- Dimension was reduced to 39 for GloVe and 28 for fastText.
- H_1 and H_2 yielded significant p -values (cf. Tab. 1).
- H_2 is corroborated by Tab. 2 which shows higher clustering scores (*-ed cells) between the **-ness** and **sup.** clusters as opposed to the **-ity** and **sup.** clusters.

	Similarities compared	fastText	GloVe	Score	fastText		GloVe	
					sup. / -ness	sup. / -ity	sup. / -ness	sup. / -ity
H_1	$\mathcal{S}(-ness) / \mathcal{S}(-ness, -ity)$	1.2e-16	2.1e-6					
	$\mathcal{S}(-ness) / \mathcal{S}(-ness, sup.)$	0.	0.					
	$\mathcal{S}(-ity) / \mathcal{S}(-ity, -ness)$	4.5e-129	2.3e-83	0.20*	0.12	0.23*	0.13	
	$\mathcal{S}(-ity) / \mathcal{S}(-ity, sup.)$	2.9e-21	6.7e-10	23.99*	3.77	32.31*	4.40	
	$\mathcal{S}(sup.) / \mathcal{S}(sup., -ness)$	3.4e-42	3.7e-13					
H_2	$\mathcal{S}(sup.) / \mathcal{S}(sup., -ity)$	4.2e-10	9.0e-4					
	$\mathcal{S}(-ity, sup.) / \mathcal{S}(-ness, sup.)$	3.9e-5	4.2e-3	3.08*	3.56	2.36*	3.14	

Table 1: p -values for H_1 and H_2 (independent t -tests, Holm-Bonferroni corrected for H_1)

Table 2: Clustering scores for the **-ness/sup.** and **-ity/sup.** cluster pairs.

Conclusion

- Word embeddings distinguish between **L1** (**-ity**), **L2** (**-ness**) and **suppletive operations**.
- L1** operations are closer to the latter than **L2** operations are.
- This implies that blocking is driven by semantic competition, and motivates the **L1 / L2** distinction.

[1] M. Aronoff. *Word Formation in Generative Grammar*. 1976. [2] Piotr Bojanowski et al. “Enriching Word Vectors with Subword Information”. In: *arXiv:1607.04606* (2016). [3] Daniel Jurafsky and Martin H. James. *Speech and language processing*. 2000. [4] Jeffrey Pennington, Richard Socher, and Christopher D Manning. “Global Vectors for Word Representation”. In: *EMNLP*. 2014.