

Nominalizations with the suffixes *-ee* and *-ation* A distributional semantic analysis

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Introduction: eventuality-related nominalizations

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- Research tends to focus on deverbal nominalizations (e.g., Barker 1998; Alexiadou 2010; Kawaletz & Plag 2015; Plag et al. 2018; Kawaletz 2021)
- Many nominalizing suffixes also attach to non-verbal bases (e.g., Plag 1999, 2004; Bauer et al. 2013, Schneider 2023)



Introduction: Distributional Semantics



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- **Distributional Semantics useful for analysis of nominalizations**
(e.g., Lapesa et al. 2018; Wauquier et al. 2018; Huyghe & Wauquier 2020)



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- Word vector: computed by list of words in context of target word
- Distance between vectors = semantic similarity
 - High distance → unsimilar
 - Low distance → similar



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Dress: *The **dress** was in the closet,
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Banana: *The **banana** tasted as delicious
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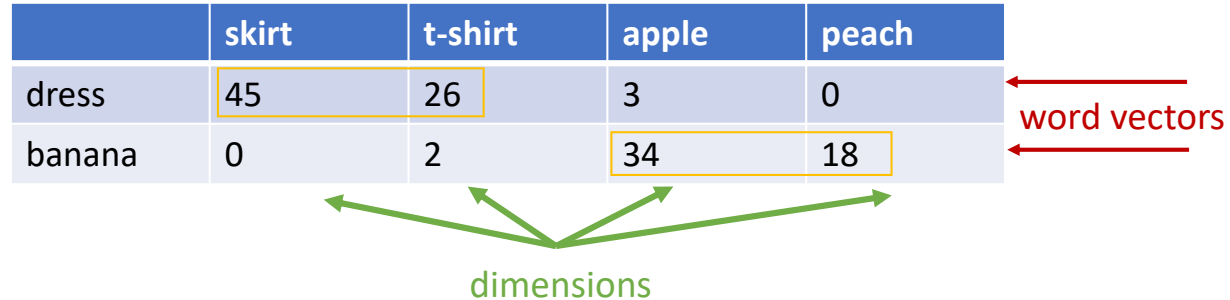
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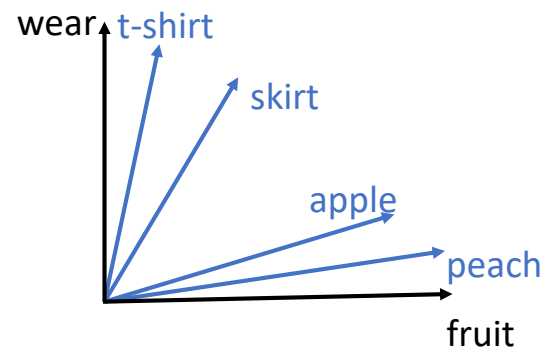
word vectors

dimensions



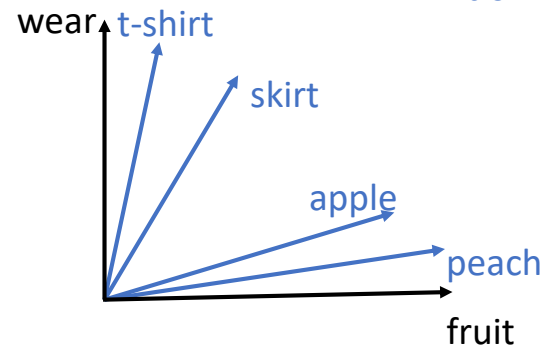
- Word vector represents a word's semantics
- Usually more dimensions: for example 300 dimensions

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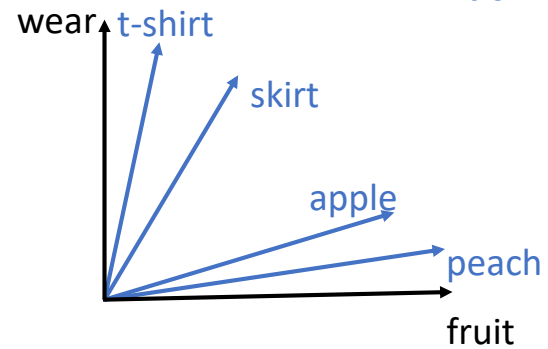
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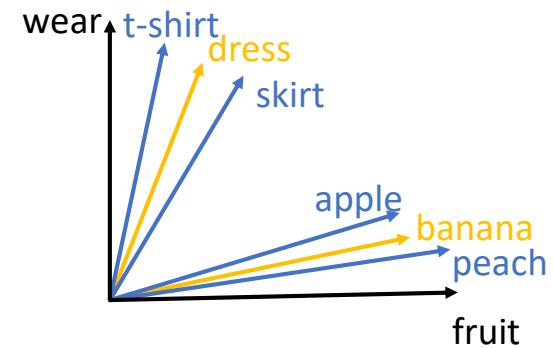
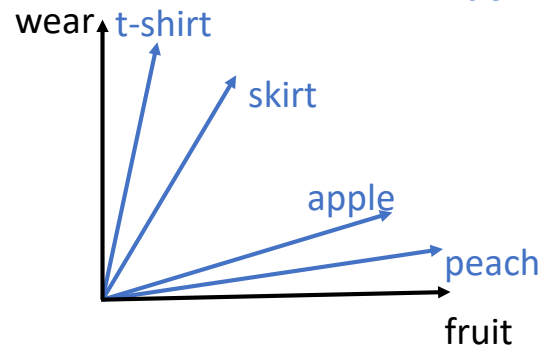
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- Are there differences in the semantics of deverbal and denominal nominalizations as the same suffix operates on different word classes?
 - use Distributional Semantics to find out
- Focus on *-ee* and *-ation*



Research questions



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 - How similar are the meanings of **deverbal** derivatives and their **verbal** base words?
- Which factors influence the similarity between base and derivative?
- Do we find differences regarding different suffixes, *-ee* and *-ation*?



Hypotheses



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- Base and derivative similar
 - Eventive elements for word formation process already in base
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 - Eventive elements for word formation process already in base
(e.g., Plag et al. 2018, Kawaletz 2021, Schneider 2023)
- Verbal bases more similar to their derivatives than nominal bases to their derivatives
 - Verbs clearly eventive (e.g., Van Valin & LaPolla 1997; Haspelmath 2001; Szabó 2015)
 - Word formation process more straightforward
 - Eventive elements more easily identifiable for word formation process



Method

- FastText *Common Crawl* subword model (Bojanowski et al. 2016, Mikolov et al. 2018)
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→ also for low frequent ones due to subword information

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- Compare cosine similarities of denominal/deverbal derivatives and their nominal/verbal bases



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Variables of interest	Expectation
Relative frequency of base/derivative	Higher relative frequency leads to higher segmentability (e.g., Hay & Baayen 2003) → more transparent → higher cosine similarity
Word class of base	Verbal bases more similar to derivatives due to clearer eventuality
Polysemy of base	Higher polysemy of base leads to decrease of cosine similarity

biographee

debtee

The suffix *-ee*

46 denominal 312 deverbial

mentee

covenantee

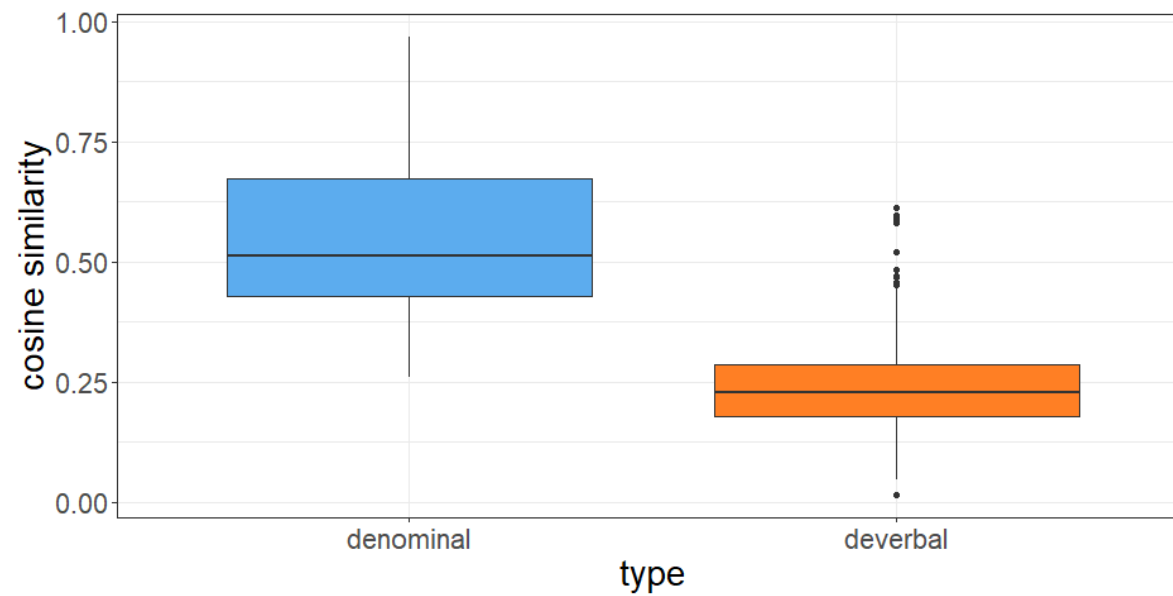
tutee

Data from COCA (Davies 2008) and BNC (Davies 2004)



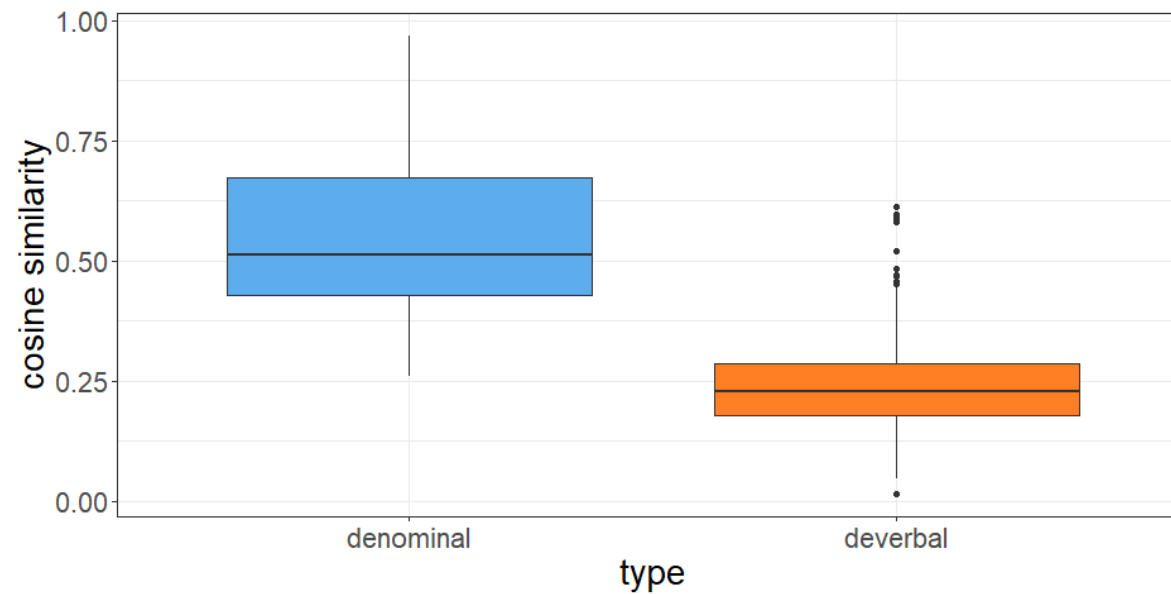
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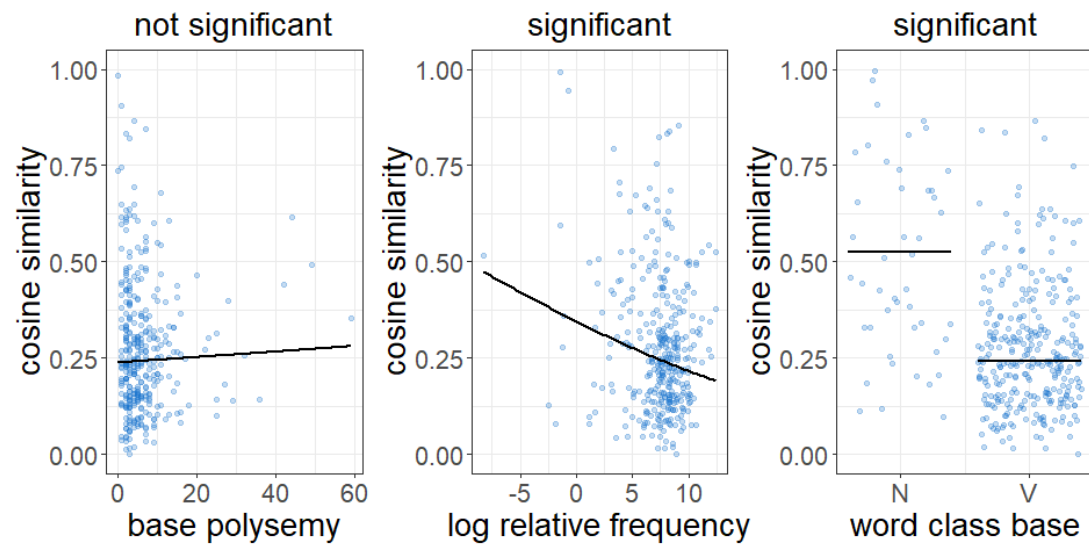
Similarity of nominal and verbal bases and derivatives for -ee

- Cosine similarity of denominal derivatives and nominal bases higher than that of deverbal derivatives and verbal bases
- Contra expectation that deverbal derivatives more similar to verbal bases

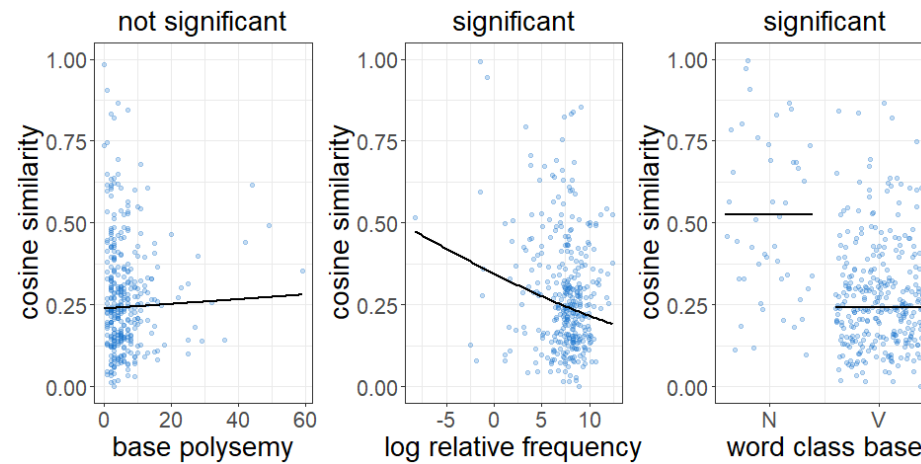


Beta regression model -ee

- Dependent variable cosine similarity

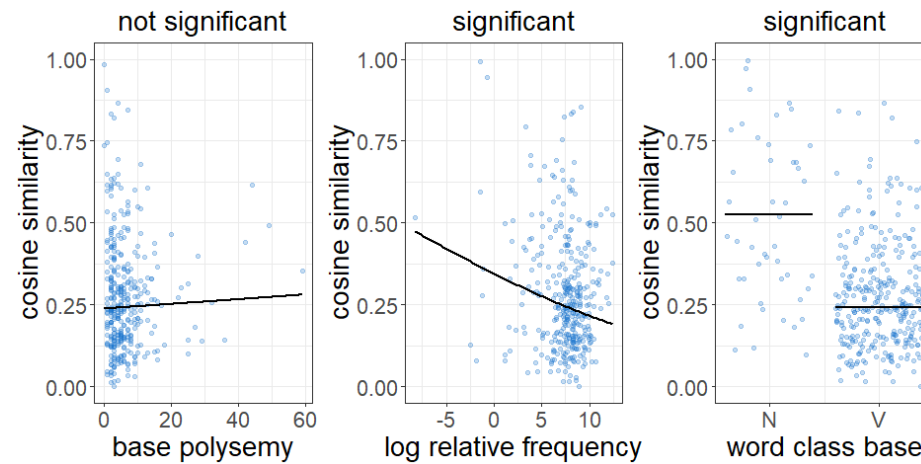


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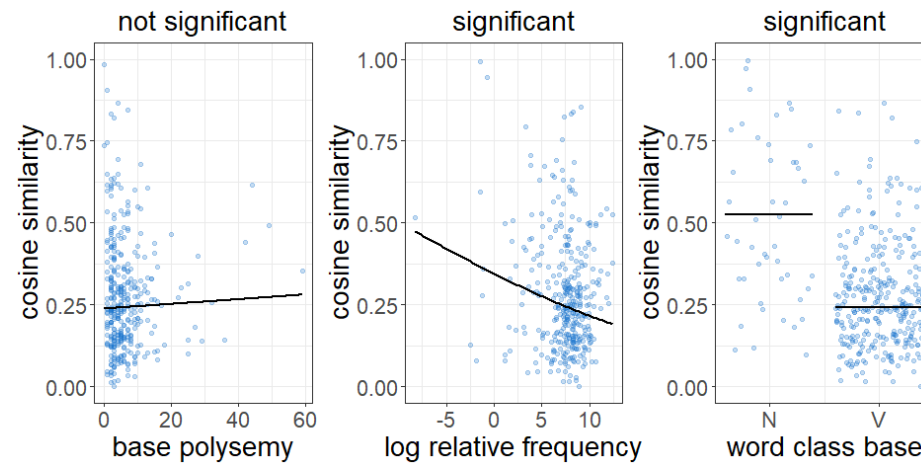
Beta regression model -ee

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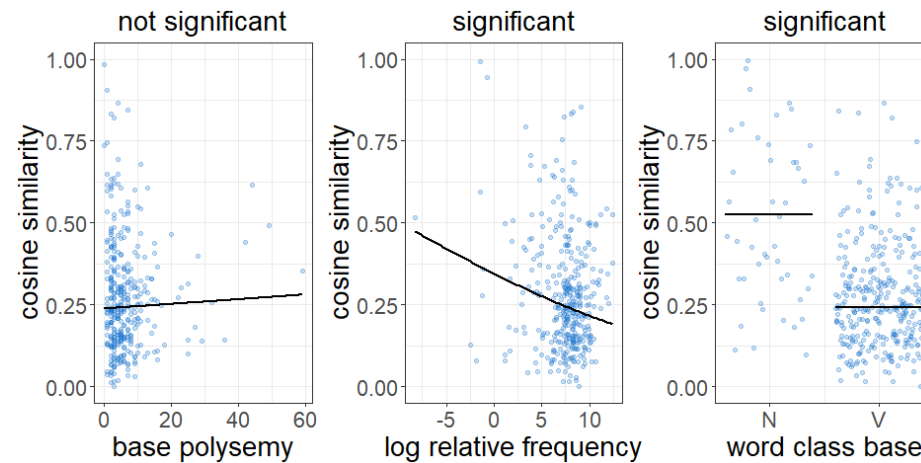
Beta regression model -ee

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Beta regression model -ee

- Polysemy of base
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- Relative frequency
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 - Not expected
- Word class base
 - Significant
 - Cosine similarity decreases if base is a verb
 - Not expected



concertation

pixelation

The suffix *-ation*

67 denominal 72 deverbal

instrumentation

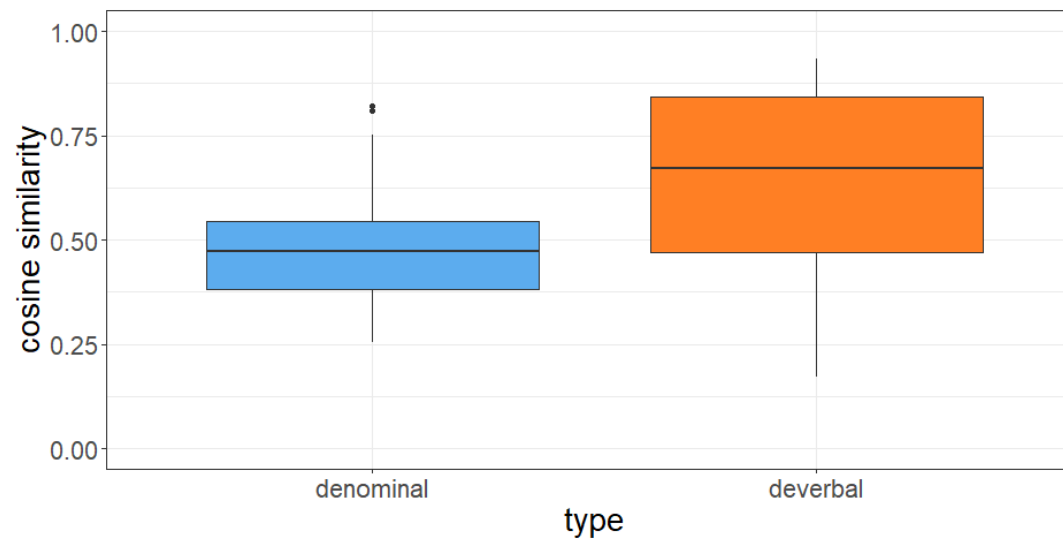
ozonation

impactation

Data from COCA (Davies 2008) and BNC (Davies 2004)

Similarity of nominal and verbal bases and derivatives for *-ation*

- Denominal derivatives and nominal bases show lower cosine similarity than deverbal pairs → opposite picture than for *-ee*





Beta regression model – Principal component analysis (PCA)

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 - Correlations of relative frequency, base polysemy, word class
 - Collinearity in model

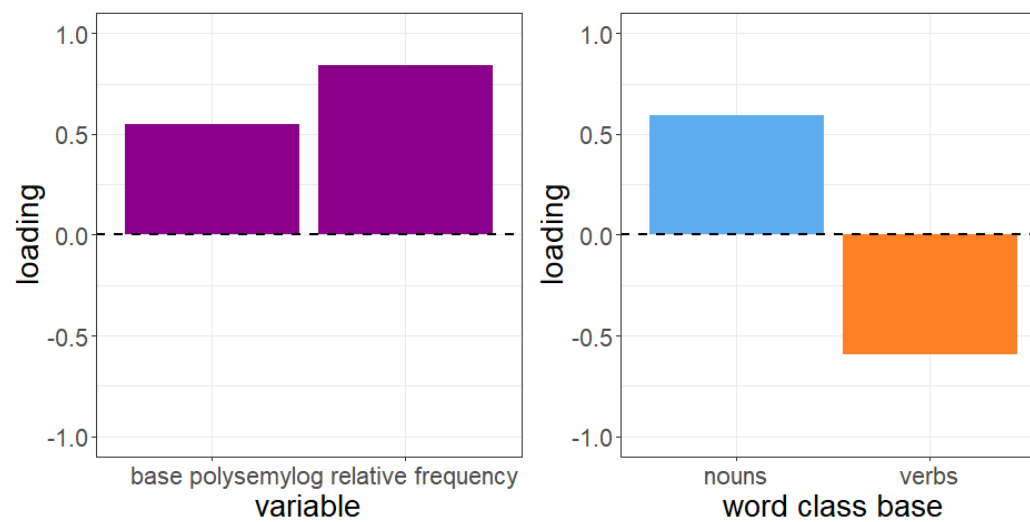
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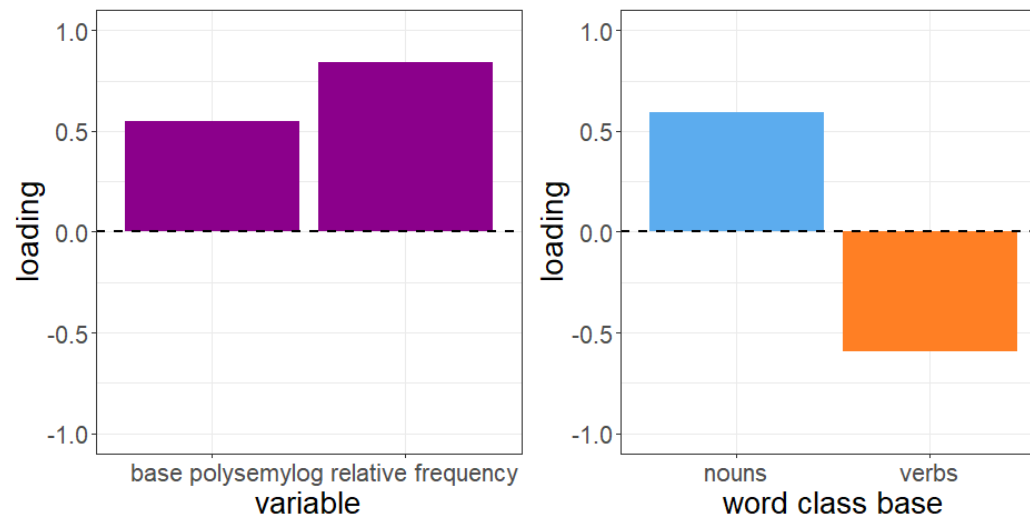
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- First principal component is retained for analysis as fulfills common criteria (e.g., O'Rourke et al. 2005; Baayen 2008; Schmitz et al. 2021, Schmitz 2022)

Beta regression with PC for *-ation*



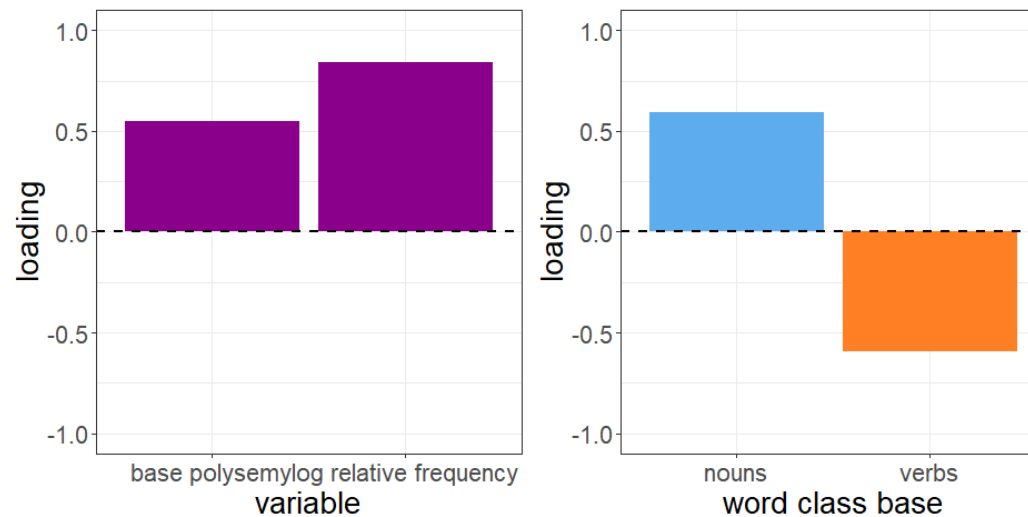
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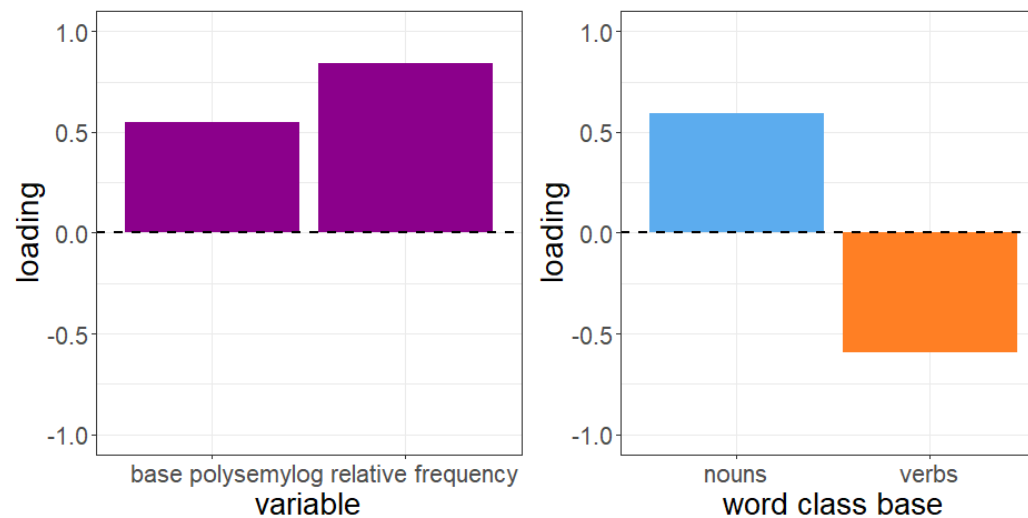
Beta regression with PC for *-ation*

- Loadings of retained PC
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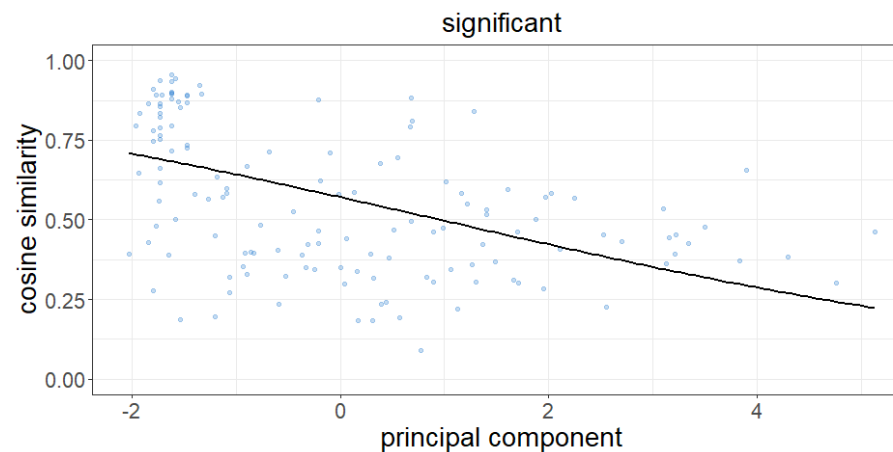
Beta regression with PC for *-ation*

- Loadings of retained PC
 - Polysemy of base and relative frequency same direction, relative frequency stronger represented in PC
 - Word classes point to different directions



Beta regression with PC for *-ation*

- Effect of retained PC
 - Higher polysemy of base word decreases cosine similarity (expected)
 - Higher relative frequency decreases cosine similarity (unexpected)
 - Word class of base influences cosine similarity (verbs higher cosine similarity, expected)



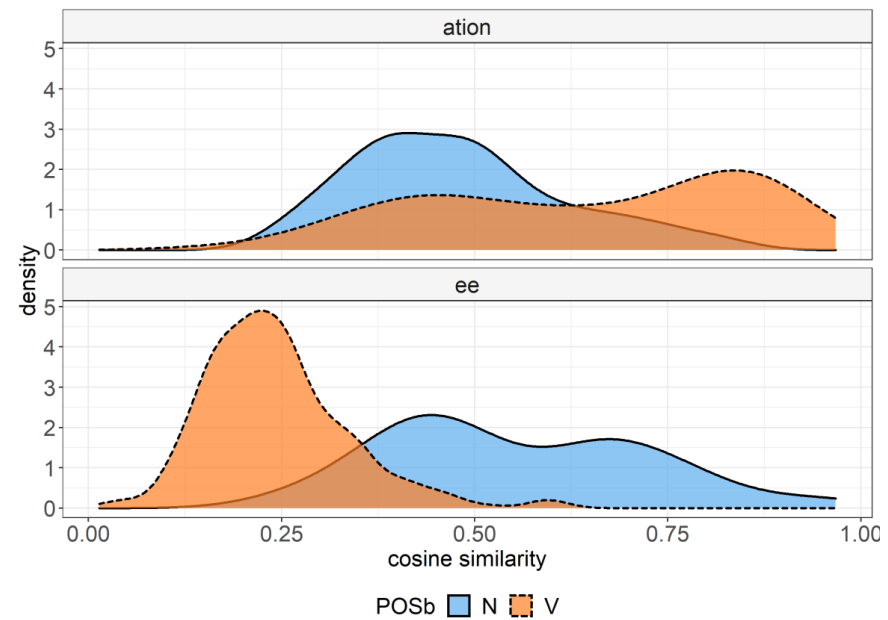


Summary *-ation* and *-ee*

Cosine similarity

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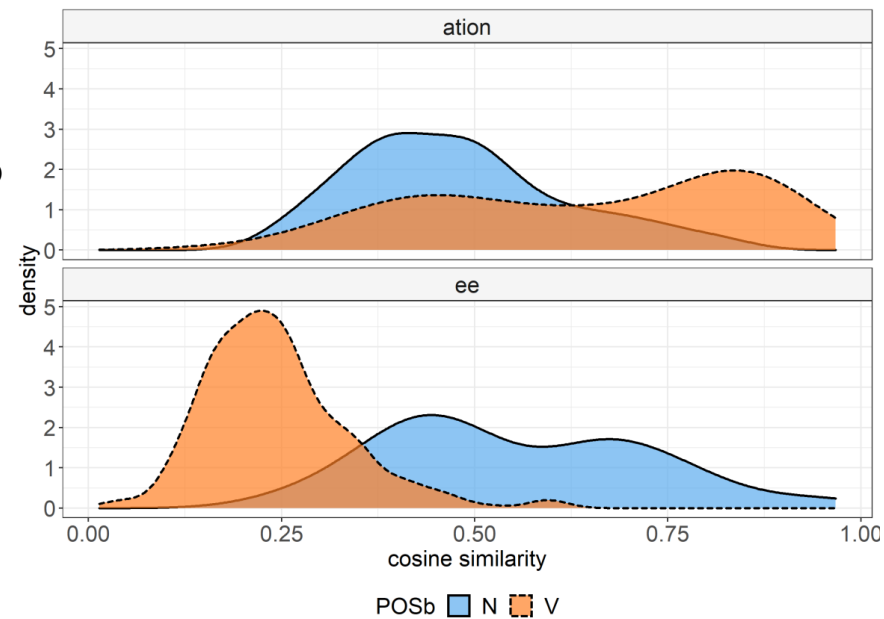
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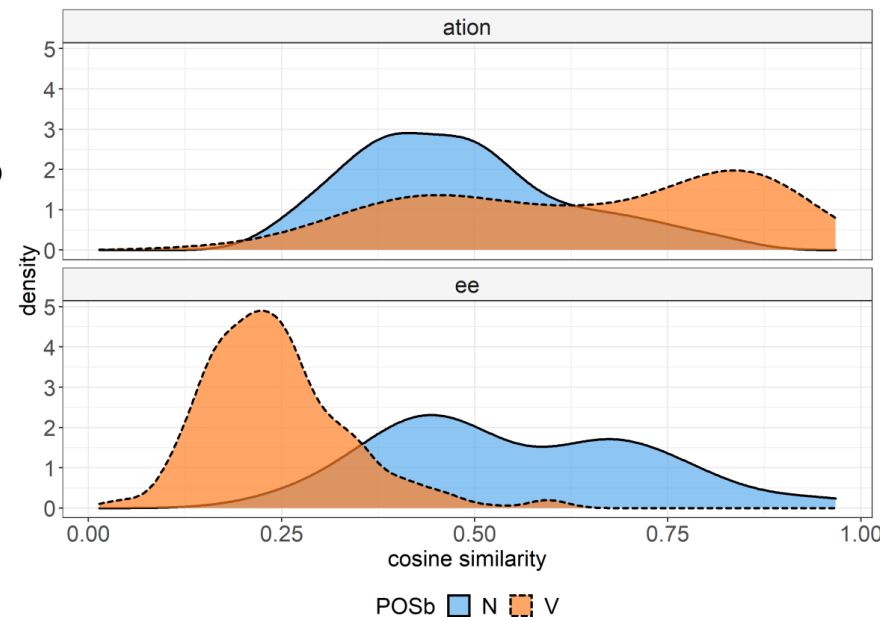
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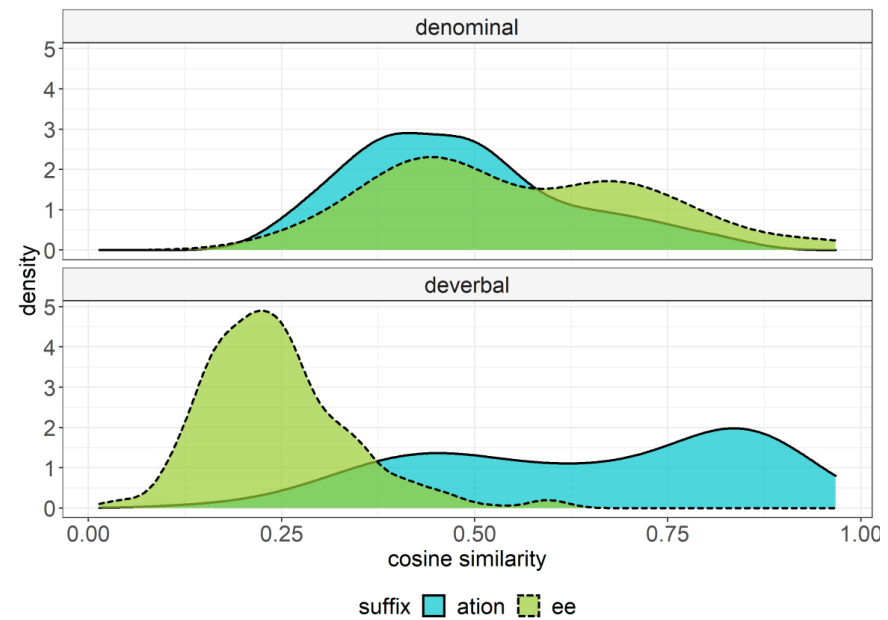
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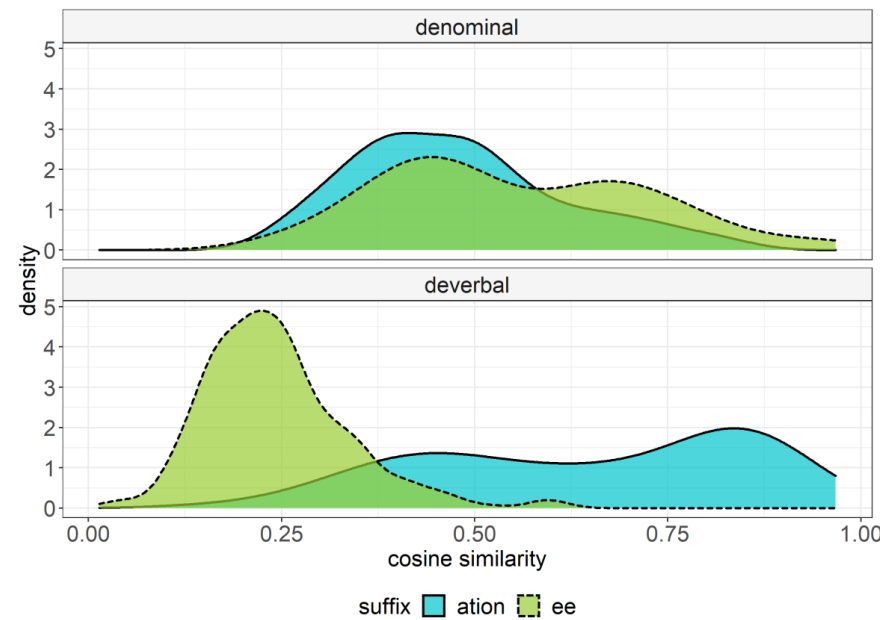
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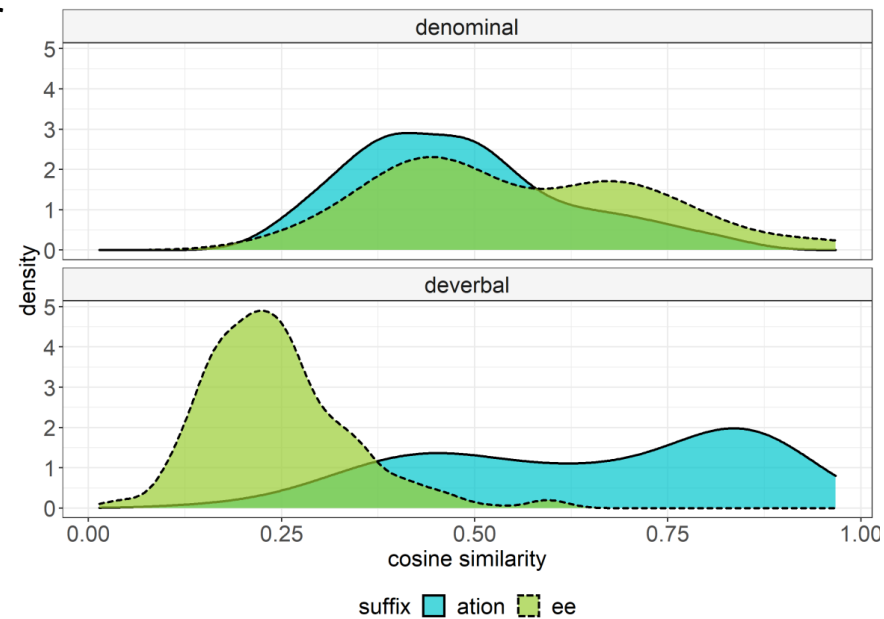
Denominal *-ation*
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Differences denominal and deverbal

Denominal *-ation* derivatives

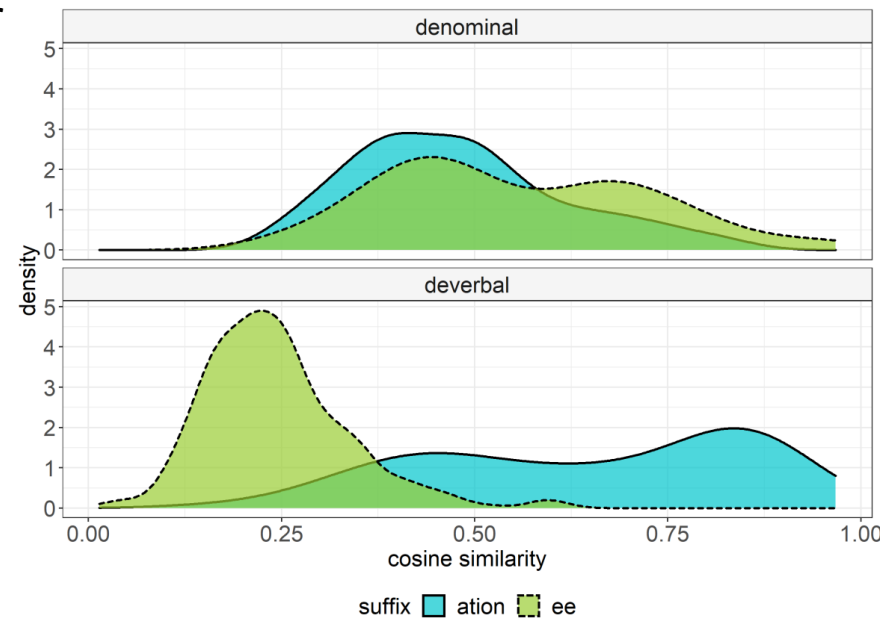
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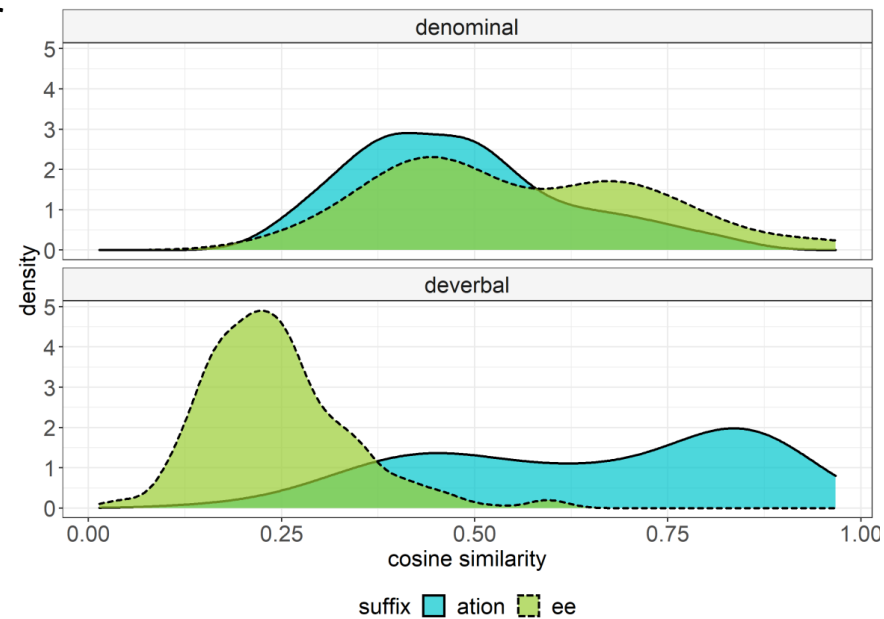


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Deverbal *-ee* derivatives



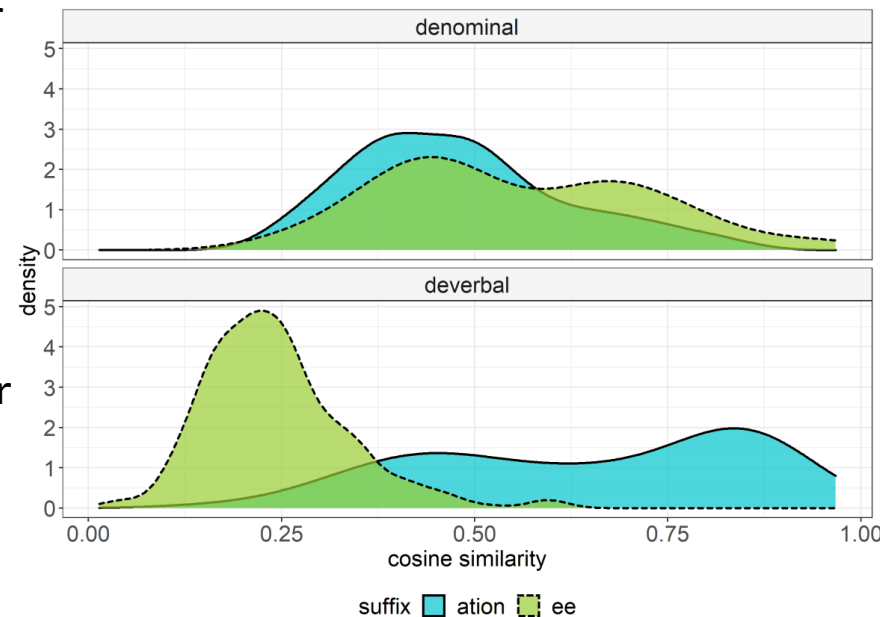
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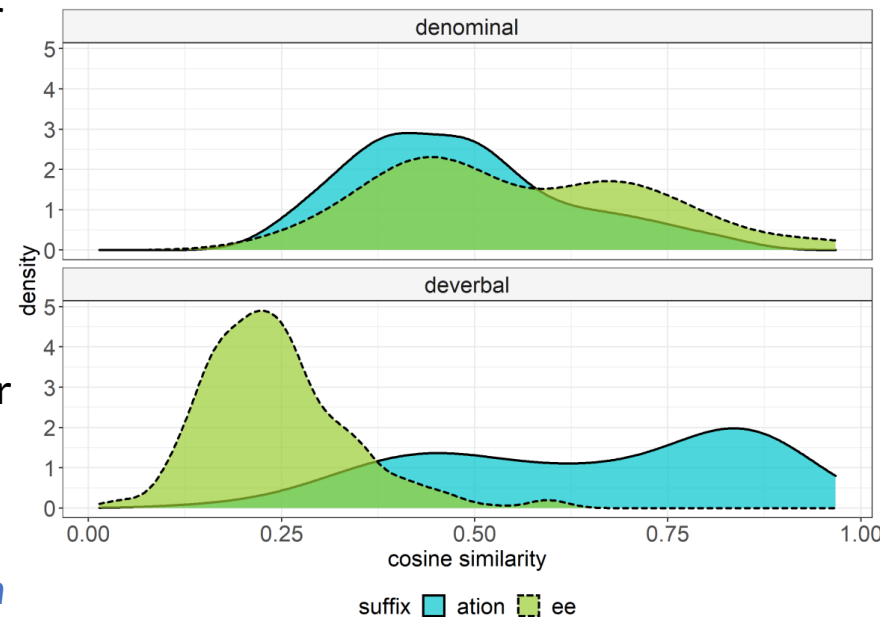
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Deverbal *-ee* derivatives

- are more uniform in their similarity than *-ation* derivatives
- overall much less similar to their bases than *-ation* derivatives





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(e.g., Van Valin & LaPolla 1997; Haspelmath 2001; Szabó 2015)
 - *-ee* creates participant readings → participants usually denoted by nouns
 - *-ation* refers to eventualities → eventualities usually denoted by verbs



Open questions & future directions



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- Do we find similar results with a different methodology, e.g., Linear Discriminative Learning (LDL)?
 - Error-driven learning → different vectors
 - Different measures → explanation for findings?

References

- Alexiadou, Artemis. 2010. Nominalizations: A probe into the architecture of grammar part i: The nominalization puzzle. *Language and Linguistics Compass* 4(7), 496-511.
- Baayen, R. Harald. (2008). *Analyzing linguistic data: A practical introduction to statistics using R*. Cambridge University Press. doi:10.1017/CBO9780511801686.
- Barker, Chris. 1998. Episodic -ee in English: A thematic role constraint on new word formation. *Language* 74 (4), pp. 695-727.
- Bauer, Laurie, Rochelle Lieber & Ingo Plag. 2013. *The Oxford reference guide to English morphology*. Oxford: Oxford University Press.
- Bojanowski, Piotr, Edouard Grave, Armand Joulin & Tomas Mikolov. 2016. Enriching word vectors with subword information. arXiv preprint arXiv:1607.04606.
- Davies, Mark. 2004. *British National Corpus* (from Oxford University Press). Available online at <https://www.english-corpora.org/bnc/>.
- Davies, Mark. 2008. *The Corpus of Contemporary American English: 400+ million words, 1990-present*. Available online at <https://www.english-corpora.org/coca/>.
- Haspelmath, Martin. 2001. Word classes and parts of speech. In Neil J. Smelser & Paul B. Baltes (eds.), *International encyclopedia of the social & behavioral sciences*, 16538-16545. Amsterdam: Elsevier.
- Hay, Jennifer & Harald Baayen. 2003. Phonotactics, parsing and productivity. *Italian Journal of Linguistics* 1. 99-130. doi:10.1.1.171.705.
- Kawaletz, Lea. 2021. *The semantics of english -ment nominalizations*. PhD Dissertation, Heinrich Heine-Universität Düsseldorf.
- Kawaletz, Lea & Ingo Plag. 2015. Predicting the semantics of English nominalizations: A frame-based analysis of -ment Suffixation. In: *Semantics of complex words*. Bauer, Laurie, Livia Körtvélyessy, Pavol Stekauer (Eds.), pp. 289-319.
- Lapeša, Gabriella, Lea Kawaletz, Ingo Plag, Marios Andreou, Max Kisselew & Sebastian Padó. 2018. Disambiguation of newly derived nominalizations in context: A distributional semantics approach. *Word Structure* 11(3). 277-312. doi: 10.3366/word.2018.0131.
- Mikolov, Tomas, Edouard Grave, Piotr Bojanowski, Christian Puhersch & Armand Joulin. 2018. *Advances in Pre-Training Distributed Word Representations*.
- O'Rourke, Norm; Hatcher, Larry; Stepanski, E. J. (2005). *Using SAS for Univariate & Multivariate Statistics*. SAS Institute Inc.
- Plag, Ingo. 1999. *Morphological productivity: Structural constraints in English derivation*. Berlin: Mouton de Gruyter.
- Plag, Ingo. 2004. Syntactic category information and the semantics of derivational morphological rules. *Folia Linguistica* 38(3-4). 193-225.
- Plag, Ingo, Marios Andreou & Lea Kawaletz. 2018. A frame-semantic approach to polysemy in affixation. In Olivier Bonami, Gilles Boyé, Georgette Dal, Hélène Giraudo & Fiammetta Namer (eds.), *The lexeme in descriptive and theoretical morphology*, 467-486. Berlin: Language Science Press.
- Schmitz, Dominic. 2022. Production, perception, and comprehension of subphonemic detail. Language Science Press. <https://langsci-press.org/catalog/book/365>.
- Schmitz, D., Plag, I., Baer-Henney, D., & Stein, S. D. (2021). Durational Differences of Word-Final /s/ Emerge From the Lexicon: Modelling Morpho-Phonetic Effects in Pseudowords With Linear Discriminative Learning. *Frontiers in Psychology*, 12, 2983.
- Schneider, Viktoria. 2023. Eventualities in the semantics of denominal nominalizations. In Sven Kotowski & Ingo Plag (eds.), *The semantics of derivational morphology: Theory, methods, evidence*, de Gruyter. In press.
- Szabó, Zoltán Gendler. 2015. Major parts of speech. *Erkenntnis* 80(S1), 3-29.
- Van Valin, Robert D. & Randy J. LaPolla. 1997. *Syntax: Structure, meaning and function*. Cambridge textbooks in linguistics. Cambridge: Cambridge Univ. Press.

Thank you!



Method: Data

- Corpus search on BNC (Davies 2004) and COCA (Davies 2008) for nominalizations with suffixes *-ee* and *-ation*
- Identification of bases manually
 - *biography* → *biographee*
 - *ozone* → *ozonation*
- Determination of word class of base on basis of a frequency criterion
 - As soon as the verbal base represents over 30% of all tokens of the base forms → base deverbal

<i>charge</i>	Tokens	Percentage
V	8265	31
N	26469	69



Beta regression with principal component for *-ation*

Beta regression with principal component for *-ation*

- Common criteria PCA

Beta regression with principal component for *-ation*

- Common criteria PCA
 - Eigenvalue higher than 1

Beta regression with principal component for *-ation*

- Common criteria PCA
 - Eigenvalue higher than 1
 - Cumulative percentage explained higher than 80%

Beta regression with principal component for *-ation*

- Common criteria PCA
 - Eigenvalue higher than 1
 - Cumulative percentage explained higher than 80%
 - PC has to make sense in their loadings

Beta regression with principal component for *-ation*

- Common criteria PCA
 - Eigenvalue higher than 1
 - Cumulative percentage explained higher than 80%
 - PC has to make sense in their loadings
 - Here it decreases cosine similarity