Type disambiguation of English -ment derivatives

Gabriella Lapesa¹, Lea Kawaletz², Marios Andreou² Max Kisselew¹, Sebastian Pado¹, Ingo Plag² University of Stuttgart¹, Heinrich-Heine University Düsseldorf²

Derivation and polysemy

Derived words tend to be **polysemous**

- We focus on *-ment* derivatives:
 - Eventive (assessment) vs. non eventive (pavement) reading
- **Context** often determines the reading of a derived word
 - Some nominalizations, however, remain ambiguous even in context

Question: to which extent does context determine the readings of derived words?

We try to answer this question for **newly derived**, nonlexicalized words, using manual annotation and corpus-based modeling

The *—ment* dataset

Selection of **sentences** containing low-frequency *—ment* nouns:

- Sources: COCA, GloWBE, WebCorp, BNC, Wikipedia, Google, ...
- Data: 401 tokens, 56 types
- **Base verbs:** change-of-state, force, psych, putting
- **Annotation:**
 - **EVENTIVE:** In many places, **emplacement** of granite plutons is synchronous to volcanic eruptions
 - **NON EVENTIVE:** I set down the scrap of doll's dress, a **bedragglement** of loose lace hem
 - **AMBIGUOUS**: [...] when it is evoked by a man who has suffered its most horrible **debauchments**

Corpus-based modeling

Distributional Semantics (DS):

Workflow: DS & Machine Learning

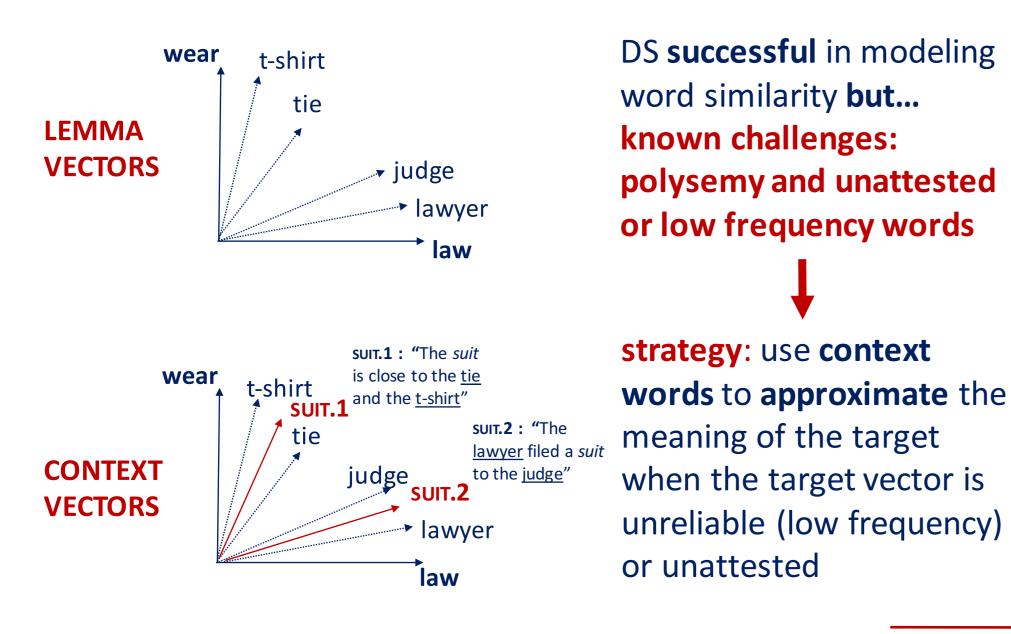
TRAINING NOUNS

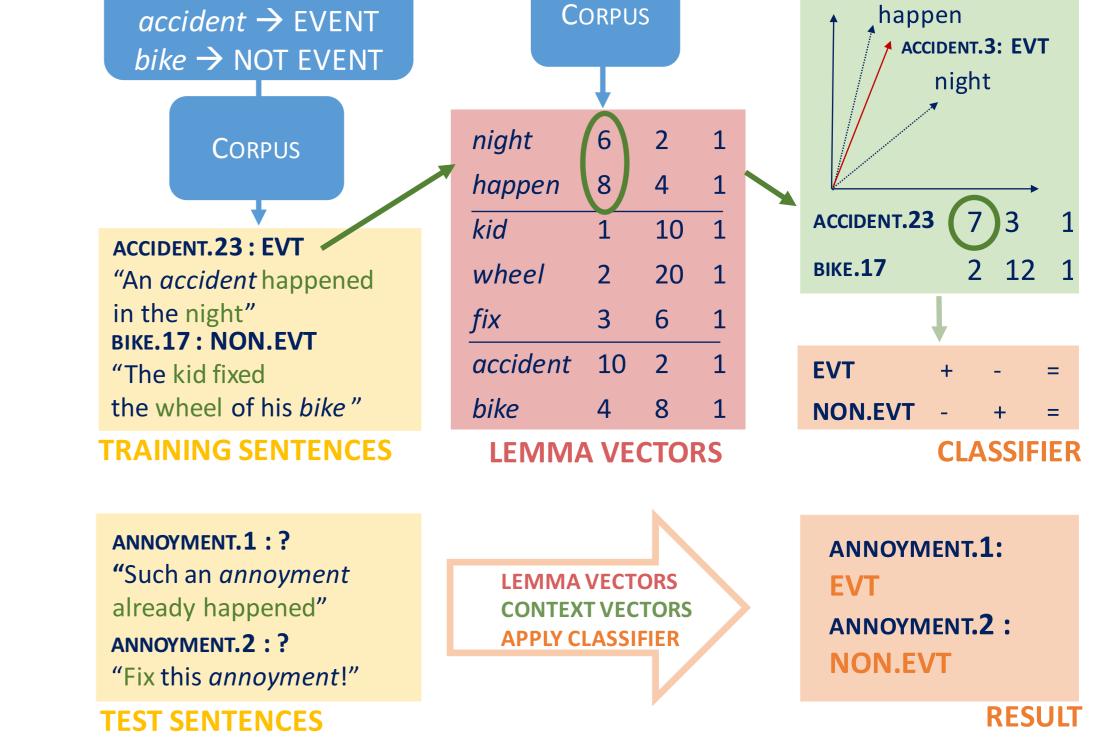
CONTEXT VECTORS

difference in meaning = difference in distribution

DS meaning \rightarrow word vector

list of words which occur in the context of a target





Experiments & Results

- **Training nouns: WordNet**
 - **EVENTIVE** •
 - state, feeling, process, phenomenon, event, act
 - **NON EVENTIVE**
 - **STRICT OBJECT:** object, substance, food, location, artefact, body
 - LAX OBJECT: communication, quantity, relation,

ANALYSIS 1: can we correctly classify training nouns and their

1.0

0.6

contexts as EVT vs. NON EVT? **Result:** yes, very well

- All classifiers are above baseline
- Performance matches linguistic intuitions
- 8.0 F-score Context vectors: lower performance than lemma vectors (task more difficult: target not used)

social relation, possession LIVING: person, animal, plant

- **Corpus:** BNC + Ukwac + Wikipedia
- Lemma vectors: state-of-the art DS model
- **Context vectors:** window size **2** (most immediate context) \checkmark vs. 8 (wider context: more info, more noise?)
- **Classifier:** support vector classifier (tested in many configurations) \checkmark **F-score:** high if the classifier captures all instances of class A and not too many non-instances of A

The authors gratefully acknowledge financial support by the Deutsche Forschungsgemeinschaft Collaborative Research Centers 732 (project B9: Lapesa, Kisselew, Pado) and 991 (project C08: Kawaletz, Andreou, Plag) Contacts: gabriella.lapesa@ims.uni-stuttgart.de, lea.kawaletz@hhu.de

Slight preference for larger windows

EVT/ EVT/ NON-EVT STRICT LIVING

EMMA VECTORS

ONTEXT VECTORS (WIN=2

ITEXT VECTORS (WIN=8

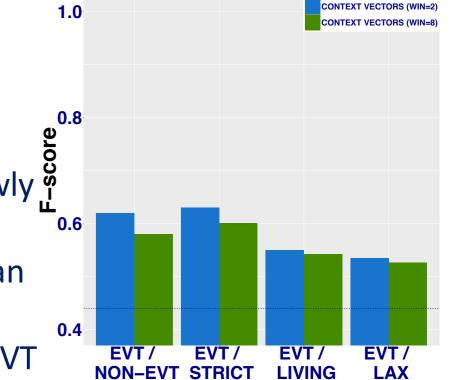
ANALYSIS 2: based on context, can we correctly classify *-ment*

instances?

Result: yes, but the task is difficult

(lemma vectors are not available)

- All classifiers above baseline Small window the best choice: newly derived words have more informative immediate context than training nouns
- AMBIGUOUS mostly classified as EVT



DS vectors can successfully disambiguate newly derived -ment nominalizations. Future work:

- qualitative evaluation of the predictions
- quantitative investigation of the factors which determine the predictions of the classifier (e.g., frequency)

References

- Firth, J. (1957). A synopsis of linguistic theory 1930-1955. Studies in Linguistic Analysis (special volume of the Philological Society), 1952-59, 1-32.
- Harris, Z. (1954). Distributional structure. Word, 10(23), 146–162.
- Kipper-Schuler, K. 2005. VerbNet: A broad-coverage, comprehensive verb lexicon. PhD. Thesis. Computer and Information Science Dept., University of Pennsylvania. Philadelphia, PA.
- Levin, B. (1993): English verb classes and alternations. A preliminary investigation. Chicago: University of Chicago Press.
- Omer Levy and Yoav Goldberg (2014) Dependency-based word embeddings. Proceedings of the 52nd Annual Meeting of the Association for Computational Linguistics (Short Papers), pages 302–308.
- Lieber, R. (2016). English nouns: The ecology of nominalization. Cambridge: Cambridge University Press.
- Plag, I., Andreou, M., & Kawaletz, L. (forthcoming). A frame-semantic approach to polysemy in affixation. In The lexeme in descriptive and theoretical morphology.
- Rainer, Franz (2014): Polysemy in derivation. In Rochelle Lieber, Pavol Štekauer (Eds.): The Oxford Handbook of Derivational Morphology. 1. ed. Oxford: Oxford Univ. Press (Oxford handbooks in linguistics), pp. 338–353.
- Schütze, H. (1998). Automatic word sense discrimination. Computational Linguistics, 27(1), 97–123.

Contacts: gabriella.lapesa@ims.uni-stuttgart.de, lea.kawaletz@hhu.de, Marios.Andreou@uniduesseldorf.de, max.kisselew@ims.uni-stuttgart.de, pado@ims.uni-stuttgart.de, Ingo.Plag@uniduesseldorf.de