# Splitting Compounds By Semantic Analogy



Joachim Daiber, Lautaro Quiroz, Roger Wechsler and Stella Frank

Institute for Logic, Language and Computation University of Amsterdam





Lautaro Quiroz In Master of Al, UvA



Roger Wechsler In Master of AI, UvA



**Dr. Stella Frank** ILLC, UvA



## Introduction

### Compound words...

- ... make life hard for standard NLP applications, incl. MT
- ... are often modeled with shallow information (e.g. Moses frequency-based splitter)

**Question:** Can we use distributional semantics to do deeper processing of compounds in a simple way?



# Splitting compounds for SMT

- ► Koehn and Knight (2003) showed PBMT systems can better deal with compounds if they are split into their meaningful parts
- ▶ Difficulty: many possible splits, we need to choose the correct ones

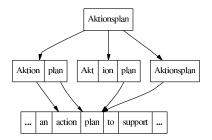


Figure: Compound splitting example from Koehn and Knight (2003).



### Semantic vector space

- ► Word embeddings saw surge of successful applications recently
- Basic idea: "You shall know a word by the company it keeps"
  - Words are mapped to vectors of real numbers in low dimensional space
  - These vectors are estimated on large amounts of text data using a neural network



#### Semantic vector space

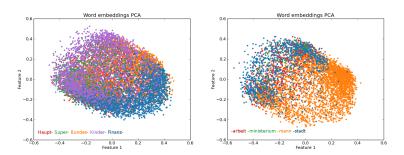
- ► Mikolov et al. (2013) showed that word embeddings capture some linguistic phenomena:
  - king is to man what queen is to woman  $v(\text{king}) v(\text{man}) + v(\text{woman}) \approx v(\text{queen})$
  - cars is to car what dogs is to dog  $v(cars) - v(car) + v(dog) \approx v(dogs)$



### Morphology induction from word embeddings

- Soricut and Och (2015) exploit these regularities to induce morphology from word embeddings
- ► Method:
  - Extract prefix and suffix replacement rules from the vocabulary
  - Keep 1000 examples of each rule
  - Judge how well each pair explains the other pairs: cars is to car what dogs is to dog?
  - Find most representative examples for each rule





(a) Compounds with same modifier.

(b) Compounds with the same head.



# The analogy test

- We model compounds based on their modifiers
- ► Potential compound splits are judged by how similar they are to a set of prototypical compounds for each modifier

**Analogy test:** Mauszeiger is to Zeiger what Mausklick is to Klick?

(mouse pointer)

(pointer)

(mouse click)

(click)



# Extracting potential compound splits

For all words in the vocabulary:

- ► Extract all possible string prefixes ≥ 4: Bundespräsident → Bund, Bunde, Bundes, ...
- ▶ Judge each Modifier+Compound pair by how well it explains others

The analogy test Computational considerations Prototypes Compound splitting algorithm



## Judging potential compound splits

### All potential compounds with prefix Maus

Maus kostüm Mauslzeiger

MausIstämme

Maus|klick

Maus|hirn

Maus|tasten

Mauslersatz

Mauslmutanten

Maus|knopf

Maus|steuerung

Maus|bewegung

Maus|gene

Mauslklicks

Mauslhirns Maus|zeiger

Maus|hirnen

Maus|bedienung

(up to 500)



### All potential compounds $\times$ All potential compounds

Maus|kostüm Mauslzeiger Mauslstämme Maus|klick Maus|hirn Maus|tasten Mauslersatz Mauslmutanten Maus|knopf Maus|steuerung Maus|bewegung Mauslaene Mauslklicks Mauslhirns Maus|zeiger Maus|hirnen Mauslbedienung

(up to 500)

Maus|bedienung ... (up to 500)

Maus|kostüm

Mauslstämme

Mauslzeiger

Maus|klick

Maus|hirn

Maus|tasten

Mauslersatz

Maus|knopf

Mauslaene

Mauslklicks

Mauslhirns

Maus|zeiger

Maus|hirnen

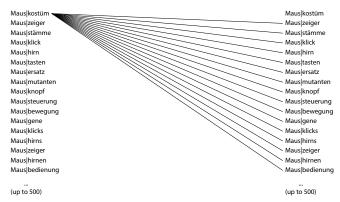
Mauslmutanten

Maus|steuerung

Maus|bewegung

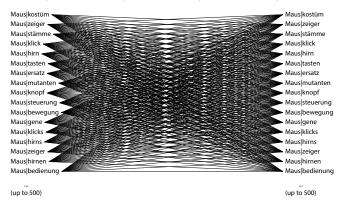


### All potential compounds $\times$ All potential compounds





### All potential compounds × All potential compounds





Maus zeiger —	
Maus stämme	Maus stämme
Maus klick	— Maus klick

**Perform analogy test:** Mauszeiger is to Zeiger what Mausklick is to Klick?

(mouse pointer) (pointer) (mouse click) (click)



## Computational considerations

- Analogy test is expensive!
- ► True and predicted vectors:
  - V<sub>Mausklick</sub>
  - $\hat{v}_{Mausklick} = Mauszeiger Zeiger + Klick$
- ► Two evaluation functions: RANK and COSINE



## Computational considerations

► Exact but slow implementation:

$$\operatorname{RANK}(v_{cmpd}, \hat{v}_{cmpd}) = \operatorname{RANK} \ \operatorname{OF} \ v_{cmpd} \ \operatorname{IN} \ \operatorname{arg} \ \operatorname{sort} \left[ \operatorname{Cosine} \left( v_w, \hat{v}_{cmpd} \right) \right]$$

- ► Approximate but fast implementation:
  - Approximate k-nearest neighbor search
  - We use the Spotify Annoy library (C++) to perform the search



## **Prototypes**

Compounds that are good examples of a compound modifier.

- ► These are best at explaining other similar modifier+compound pairs
- ► We call this set the modifier's *prototypes*



Mauslkostüm

Maus|zeiger

Maus|stämme Maus|klick

Maus|hirn

Mausitasten

Mauslersatz

Maus|mutanten

Maus|knopf

Maus|steuerung

Maus|bewegung

Maus|gene

Maus|klicks

Maus|hirns

Maus|zeiger

Maus|hirnen Maus|bedienung

Mauspediending

(up to 500)

Mauslkostüm

Maus|zeiger

Maus|stämme Maus|klick

Maus|hirn

Maus|tasten

Maus|ersatz

Maus|mutanten

Maus|knopf

Maus|steuerung

Maus|bewegung

Maus|gene

Maus|klicks

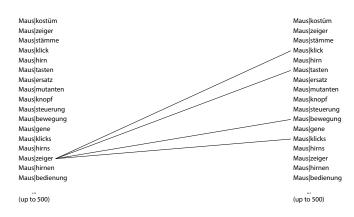
Maus|hirns

Maus|zeiger Maus|hirnen

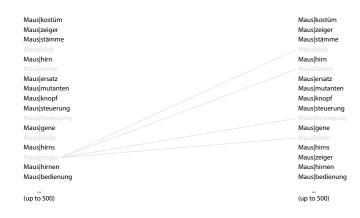
Maus|bedienung

(up to 500)

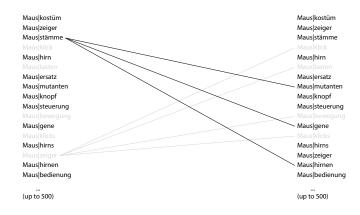




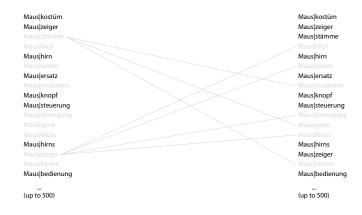














# Extracted prototypes for Maus-

Prototype	Evidence words
V-Zeiger	-Bewegung -Klicks -Klick -Tasten -Zeiger
V-Stämme	-Mutanten -Gene -Hirnen -Stämme
V-Kostüm	-Knopf -Hirn -Hirns -Kostüm
V-Steuerung	-Ersatz -Bedienung -Steuerung



### Mausmutation

▶ We start from the left...

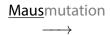






▶ Do I know the modifier Mau? No!





▶ Do I know the modifier Maus? Yes!



# <u>Maus</u>mutation

- ► Do I know the modifier *Maus*? Yes! Prototypes:
  - -Zeiger
  - -Stämme
  - -Kostüm
  - Steuerung

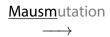




- ► Do I know the modifier *Maus*? Yes! Prototypes:
  - Zeiger
  - -Stämme √
- → Mausmutation is to Mutation what Mausstämme is to Stämme.

- -Kostüm
- Steuerung





▶ Do I know the modifier *Mausm*? No!



#### Mausmutation

► And so on...



## Maus mutation

- ► The prototype with the highest score will be our split!
- ► Recurse...



## Plantage

► Let's try another example...







▶ Do I know the modifier *Plan*? Yes!





- ▶ Do I know the modifier *Plan*? Yes! Prototypes:
  - Feststellung
  - -Wert
  - -Fertiger
  - .





- ▶ Do I know the modifier *Plan*? Yes! Prototypes:
  - Feststellung
  - -Wert
  - Fertiger
  - —



## Plantage

► No compound split!

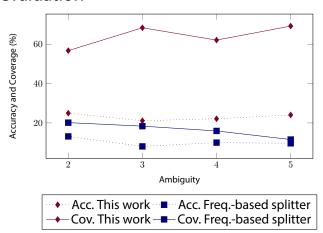


## Intrinsic evaluation

- ► Evaluation on human-annotated dataset (Henrich and Hinrichs, 2011)
  - ~50k compounds
  - only binary splits
- Baseline: Frequency-based Moses compound splitter (Koehn and Knight, 2003)
- ► We evaluate:
  - Accuracy: |correct splits| |compounds|
  - $\ \ \text{Coverage:} \ \tfrac{|\text{compounds split}|}{|\text{compounds}|}$



## Intrinsic evaluation





# Machine translation experiments (German to English)

	(a) No d	(a) No comp. splitting		(b) Ra	(b) Rare: $c(w) < 20$			(c) All words		
	Splits	BLEU	MTR	Splits	BLEU	MTR	Splits	BLEU	MTR	
Moses splitter This work	0	17.6	25.5	231 744	.,	25.7 <sup>C</sup> <b>26.1</b> <sup>ABC</sup>	244 1616	17.9 17.7	25.8 <sup>A</sup> 26.3 <sup>A</sup>	

 $<sup>^{\</sup>rm A}$  Stat. sign. against (a) at p<0.05  $^{\rm B}$  Stat. sign. against Moses splitter at same c(w) at p<0.05  $^{\rm C}$  Stat. sign. against best Moses splitter (c) at p<0.05



## Conclusion

- Regularities in semantic vector space can be used to model composition of compounds
- ▶ We can extract modifiers and prototypes (Soricut and Och, 2015)
- ► Compound splitting algorithm:
  - Good intrinsic performance on gold standard
  - Improved translation quality (standard PBMT setup)
  - Especially adept at splitting highly ambiguous compounds



Thank You!

Any questions?



## References

- Henrich, V. and Hinrichs, E. W. (2011). Determining immediate constituents of compounds in GermaNet. In *Proceedings of the International Conference on Recent Advances in Natural Language Processing 2011*.
- Koehn, P. and Knight, K. (2003). Empirical methods for compound splitting. In *Proceedings of the tenth conference on European chapter of the Association for Computational Linguistics-Volume 1*, pages 187--193. Association for Computational Linguistics.
- Mikolov, T., Yih, W.-t., and Zweig, G. (2013). Linguistic regularities in continuous space word representations. In Proceedings of the 2013 Conference of the North American Chapter of the Association for Computational Linguistics: Human Language Technologies, pages 746—751, Atlanta, Georgia. Association for Computational Linguistics.
- Soricut, R. and Och, F. (2015). Unsupervised morphology induction using word embeddings. In Proceedings of the 2015 Conference of the North American Chapter of the Association for Computational Linguistics: Human Language Technologies, pages 1627--1637, Denver, Colorado. Association for Computational Linguistics.