

# Verbessertes Entity Linking mit neuronalen Word Embeddings

Bachelorarbeit

Verfasser: Robert Dziuba  
Betreuer: Prof. Dr. habil. Alexander Löser  
Gutachter: Prof. Dr.-Ing. Joachim Schimkat

# Problemstellung

# 3

## Das Wort - Mention



## Der Kontext und Word Embedding

President **Trump** is in the White House.

card games USA Tower  
Hotel  
administration conglomerat  
engineer President transport company

The image shows a word cloud visualization of the sentence "President Trump is in the White House." The word "Trump" is highlighted in red. Other words in the cloud include "card games", "USA", "Tower", "Hotel", "administration", "conglomerat", "engineer", "President", and "transport company".

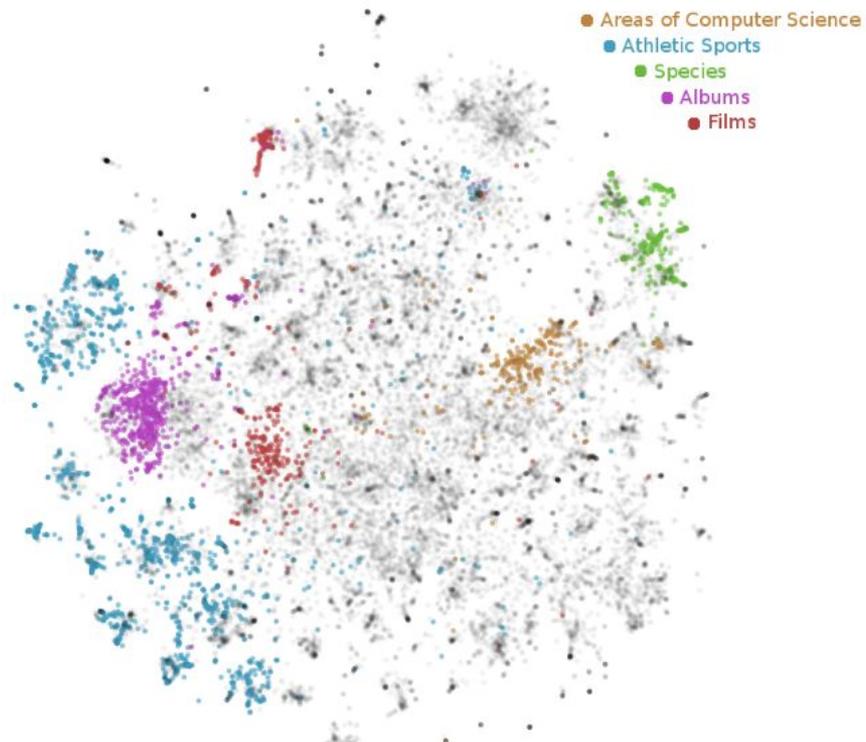
# Ziel der Arbeit

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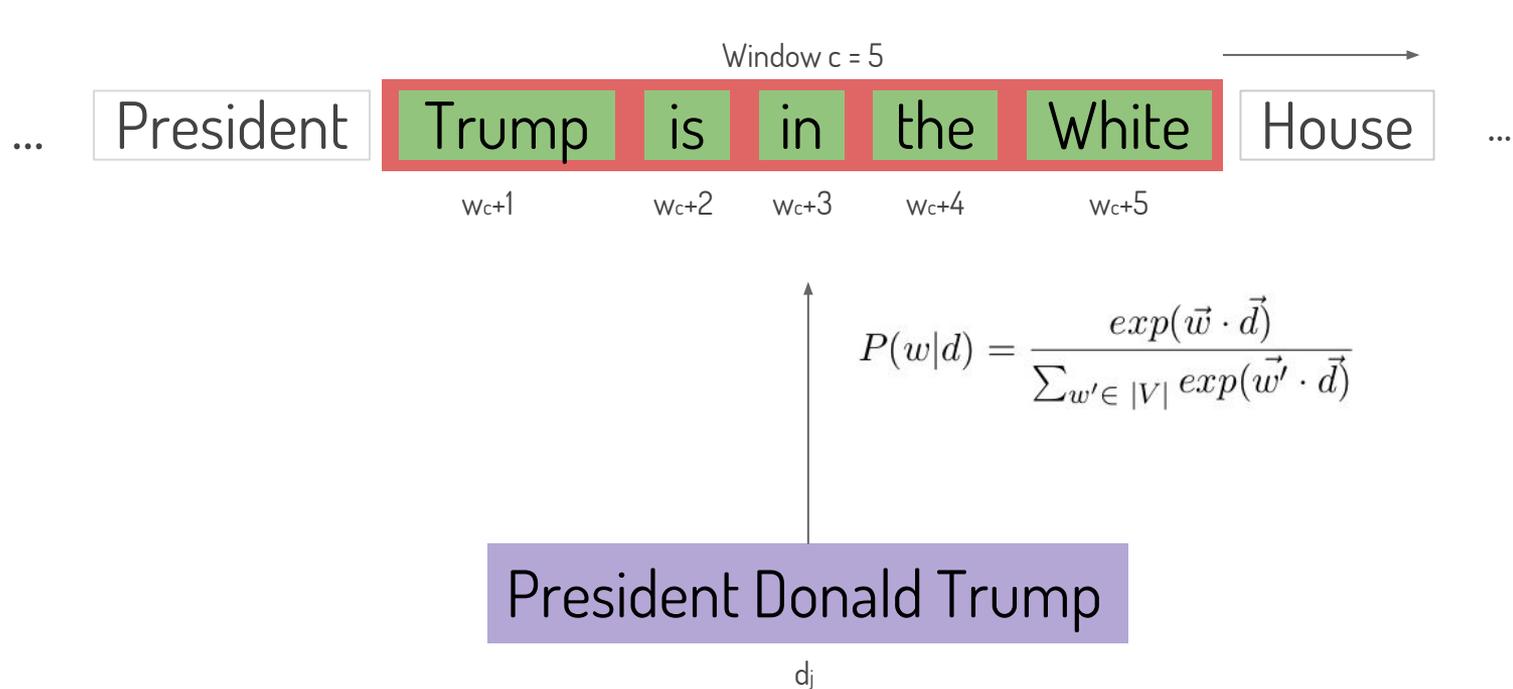
Das Ziel der Arbeit ist es, eine im Text genannten Mention, mit Hilfe ihres Kontextes, einer konkreten Entity zuzuordnen.

# Methodik

## Repräsentation von Kontext im Vektorraum - Entity Embeddings



# Paragraph Vector



# 10 Disambiguierung im Vektorraum

**Mention** = "Trump"

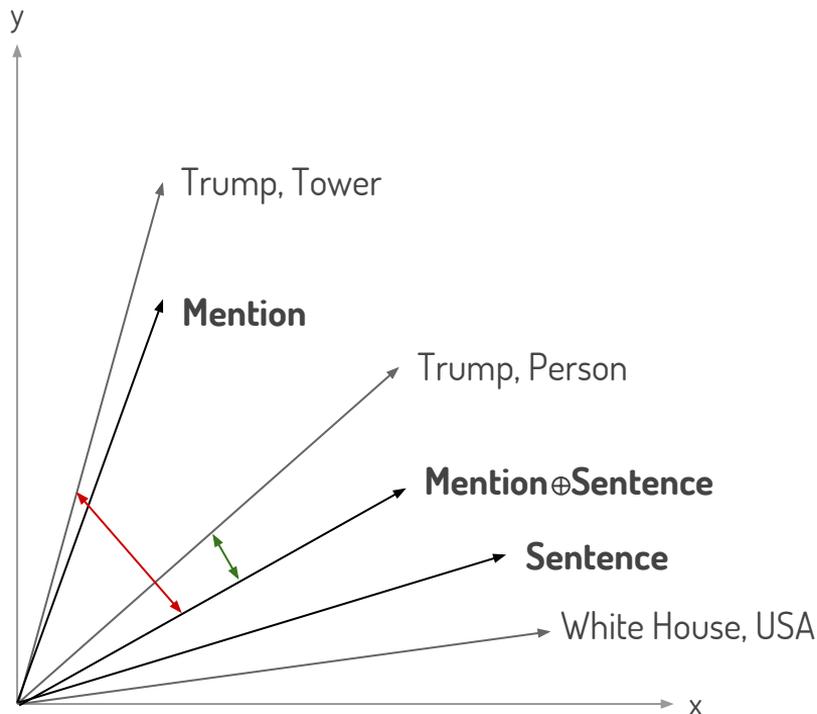
**Sentence** = "President Trump is in the White House."

$$\text{cosine}(\vec{v}, \vec{w}) = \frac{\vec{v} \cdot \vec{w}}{|\vec{v}| |\vec{w}|} = \frac{\sum_{i=1}^N v_i w_i}{\sqrt{\sum_{i=1}^N v_i^2} \sqrt{\sum_{i=1}^N w_i^2}}$$

$$\hat{c} = \sum_{c \in |C_m|} \text{cosSim}(c)$$

mit  $\text{cosSim}(c) = \text{cosine}(\text{concat}(m, s), \text{concat}(c, c))$

und  $\text{concat}(a, b) = \text{vec}(a) \oplus \text{vec}(b)$



# Daten

## Wikipedia Korpus

Within American **political culture**, the **center-right** Republican Party is considered "conservative" and the **center-left** Democratic Party is considered "liberal".<sup>[375][376]</sup> The states of the **Northeast** and **West Coast** and some of the Great Lakes states, known as "**blue states**", are relatively liberal. The "**red states**" of the **South** and parts of the **Great Plains** and **Rocky Mountains** are relatively conservative.

Republican **Donald Trump**, the winner of the 2016 presidential election, is serving as the 45th President of the United States.<sup>[377]</sup> Leadership in the Senate includes Republican Vice President **Mike Pence**, Republican President Pro Tempore **Orrin Hatch**, **Majority Leader Mitch McConnell**, and **Minority Leader Chuck Schumer**.<sup>[378]</sup> Leadership in the House includes Speaker of the House **Paul Ryan**, **Majority Leader Kevin McCarthy**, and **Minority Leader Nancy Pelosi**.<sup>[379]</sup>

Wikipedia: Donald Trump - [https://en.wikipedia.org/wiki/United\\_States](https://en.wikipedia.org/wiki/United_States)



**Donald Trump**  
45th **President**  
since January 20,  
2017

**Mike Pence**  
48th **Vice President**  
since January 20, 2017

## Trainingsdaten

### Beispielsatz

Q22686    Republican

Donald Trump, the winner of the 2016 presidential election, is serving as the 45th President of the United States.

### Datensatz

2.725.363 Entities

63.812.227 Sätzen

11 GB Gesamtumfang

# Evaluation

## Ergebnisse der Vektorkonkatenation-Strategie

Layer/Epochen	MRR	PREC (%)	REC (%)	F1 (%)
Lucene	0.534	48.24	45.10	46.60
	CommonPreprocessor			
100 Layer/10 Epochen	0.592	56.73	53.10	54.84
	MinimalLowercasePreprocessor			
100 Layer/10 Epochen	0.594	57.20	53.52	55.28
	CommonPreprocessor			
300 Layer/10 Epochen	<b>0.630</b>	<b>62.01</b>	<b>57.96</b>	<b>59.90</b>

# Fazit und Ausblick

## Fazit

- Mention Embeddings sind besser als Lucene Exact Match
- Konkatenierte Entity Embeddings sind besser als Lucene
- Mehr Layers führen zu besseren Ergebnissen
  
- Trainingszeit steigt überproportional mit Iteration/ Epochen
- Lucene limitiert das Suchergebnis
- Entities die nicht in der Wikipedia verlinkt sind werden nicht gelernt

## Ausblick

- Minimierung der Modellgröße
- Lucene beim Entity Linking weglassen
- Mehr Kontext
- Einfluss benachbarter Entities im Text nutzen

# Live Präsentation

Vielen Dank