

# INVESTIGATING CROSS-DOCUMENT EVENT COREFERENCE RESOLUTION FOR DUTCH

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# OVERVIEW

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- Event Coreference Resolution (ECR)
- Motivation

## **2. The ENCORE Corpus**

## **3. Experiments**

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- Results and Error Analysis

## **4. Conclusion and future work**



# EVENT COREFERENCE RESOLUTION (ECR)

- 1. A disciplined France team has beaten the Red Devils in the world cup semis earlier this week.*
- 2. Belgium lost the semi-final of the 2018 World cup against France with a score of 1-0 on Tuesday.*

# CONSIDERATIONS AND MOTIVATION

- Difficulties in ECR
  - Lack of data, sparsity and lack of uniformity
  - Coreference resolution across documents
- Why ECR research is valuable
  1. Move away from the paradigm of lexical semantics
  2. Focus on discourse-level relations to break down topic- and document barriers
  3. Practical multi-document applications such as summarization, content-based news recommendation and reading comprehension

# CROSS-DOCUMENT ECR FOR DUTCH

# ENCORE CORPUS

- ENCORE (*De Langhe, De Clercq, Hoste, 2022*)
  - Large-scale Dutch event coreference corpus
  - Data sourced from a variety of Dutch (Flemish) newspapers
  - Focus on unrestricted events
  - Coreference annotated within event clusters

Cluster id	Topic	# of documents
47	Tim Burton exposition in Genk	11
75	Royal Wedding Prince Harry	24
87	Election of Cuban president	12



# ENCORE CORPUS: EVENTS

- Events are annotated based on ECB+ guidelines *(Cybulska & Vossen, 2014)*

In **Guatemala**, the volcan de fuego has **erupted** again **today**

## Event Arguments

- Event Action (trigger)
- Event Time
- Event Location
- Event Participants
  - Human Participants
  - Non-Human Participants

## Event Properties

- Event Prominence (Main)
- Event Realis (Certain)
- Event Sentiment (Negative)



# ENCORE CORPUS: COREFERENCE

## 1. Identity Coreference between events

- Event time (1) , location (2) and participants (3) must match

*“The 2012 London Olympics were a succes. The games lasted from 27/07/12 to 13/08/12.”*

## 2. Part-whole Coreference between events

- One event is fully encompassed by the other

*“The oscar ceremony was about to begin, with the presenters preparing for the opening speech.”*

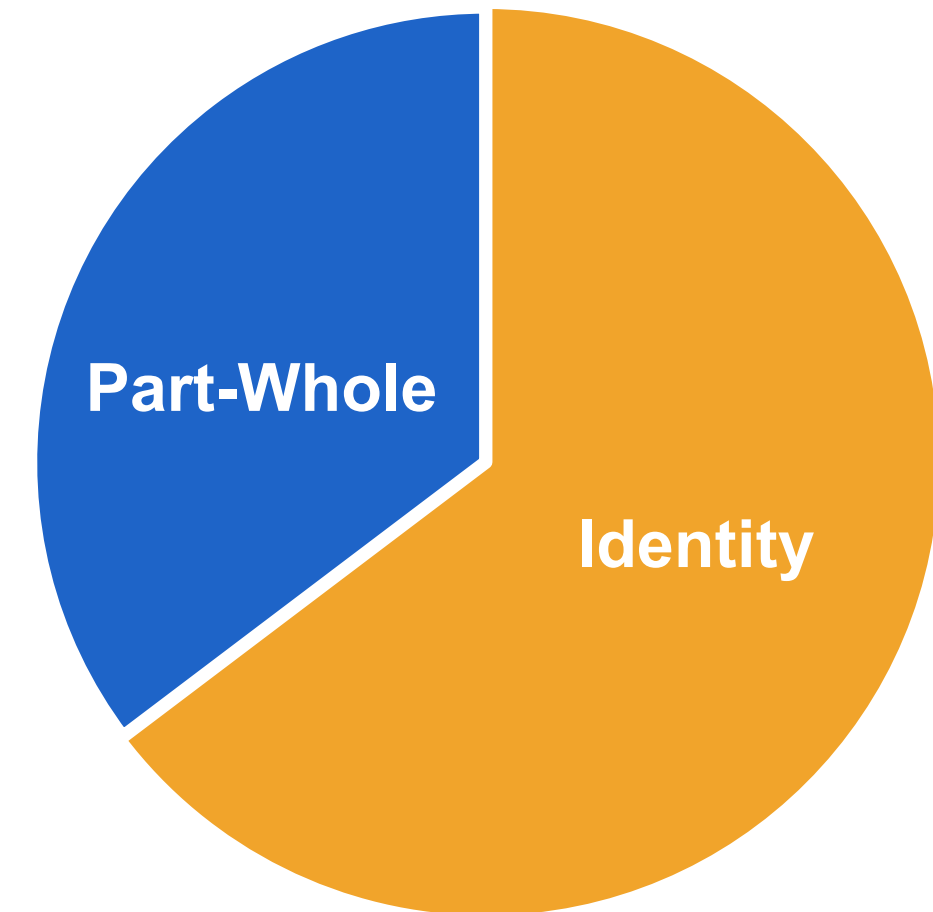




# ENCORE CORPUS: OVERVIEW

## Statistics

Description	Count
Documents	1115
Topics	91
Events	15407
Event Arguments	35315
Intra-document Event Coreference Chains	1018
Cross-Document Event Coreference Chains	1587



**Graph 2:** Distribution of Identity (28561) and Part-whole (15587) links between events

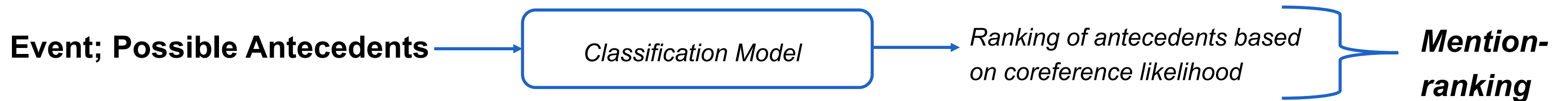
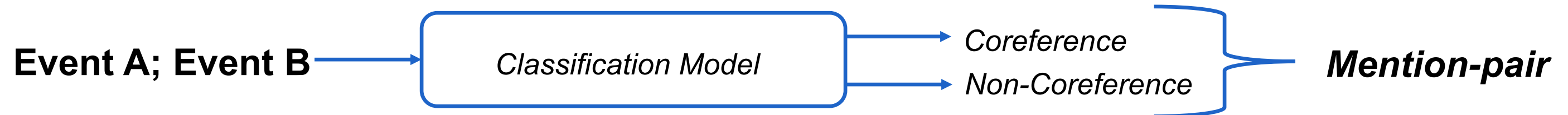


# EXPERIMENTS

# EXPERIMENTAL SETUP

## • Task Description

- Classification of coreference based on *gold-standard event* mentions
- For now, only identity links are considered
- *Within-document* and *cross-document* setting
- Experiments based on two existing paradigms in coreference studies:

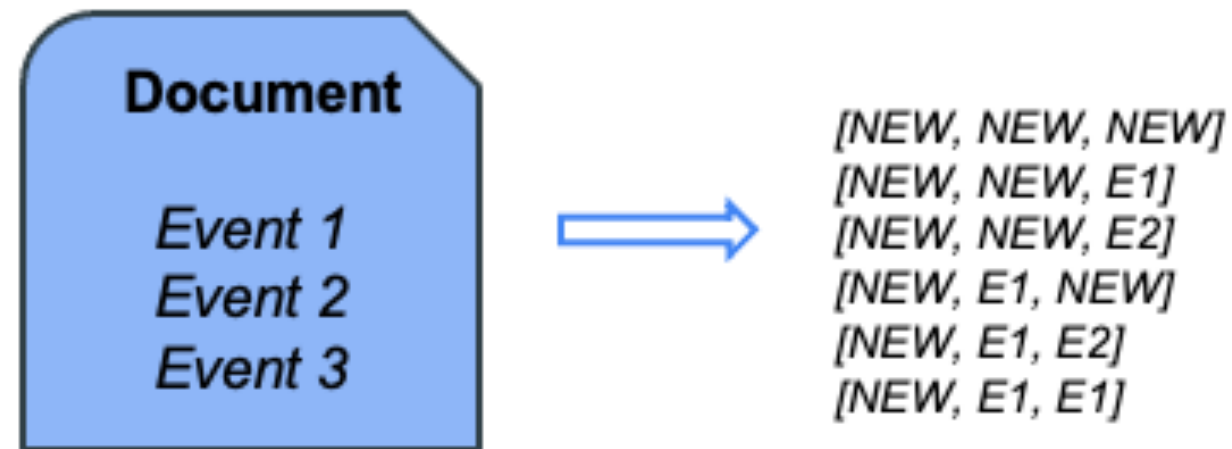


# EXPERIMENTAL SETUP: MENTION-PAIR MODELS

- Transformer-based language models
  - BERTje (*de Vries et al., 2019*)
  - RobBERT (*Delobelle et al., 2020*)
  - RobBERTje (*Delobelle et al., 2022*)
  - XLM-RobBERTa (*Lample and Conneau, 2019*)
  - mBERT (*Devlin et al., 2019*)
- Traditional feature-based learning
  - Gradient-boosted Tree algorithm (*Chen et al., 2015*)
  - Features based on those used in English-language ECR studies
    - Lexical Similarity
    - Discourse
    - Logical Constraining

# EXPERIMENTAL SETUP: MENTION-RANKING MODEL

- **Feature-based Mention-Ranking** (*Lu and Ng, 2017*)
  - Log-linear coreference model
  - Ranking of Document partitions
  - Task-specific loss based on type of error



$$L(\Theta) = \sum_{d=1}^t \log \sum_{a \in A(C_d^*)} p(a|d; w)' + \lambda \|\Theta\|_1$$

$$p(a|d; w)' \propto p(a|d; w) l(a, C_d^*)$$

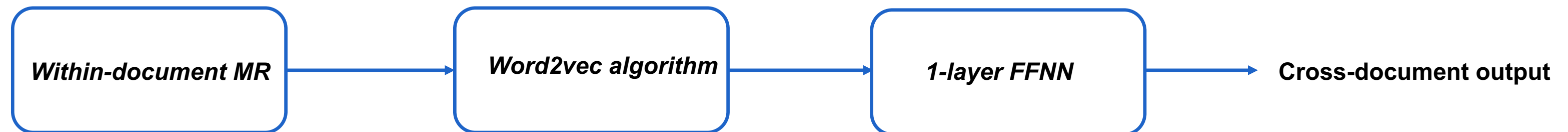
$$p(a|d; w) \propto \exp\left(\sum_{i=1}^n w \cdot f(i, a_i, d)\right)$$

# EXPERIMENTAL SETUP: MENTION-RANKING MODELS

- **Cross-Document Mention-Ranking**

1. Poses scalability issues in memory
2. Creates artificial sparsity problem in the data

➔ Pairwise classification of within-document coreference chains



# RESULTS AND DISCUSSION

# IDENTIFYING IDENTITY COREFERENCE RELATIONS

## ■ Results

	CONLL	LEA
MP XGBoost	0.36	0.21
MR <sub>base</sub>	0.39	0.25
MR <sub>task-specific</sub>	0.42	0.26
MR Embedding <sub>base</sub>	/	/
MR Embedding <sub>task-specific</sub>	/	/
MP BERTje	<b>0.52</b>	<b>0.33</b>
MP RobBERT	0.49	0.29
MP RobBERTje	0.48	0.29
MP XLM-RoBERTa	0.17	0.11
MP mBERT	0.14	0.08

(a) Results for within-document ECR

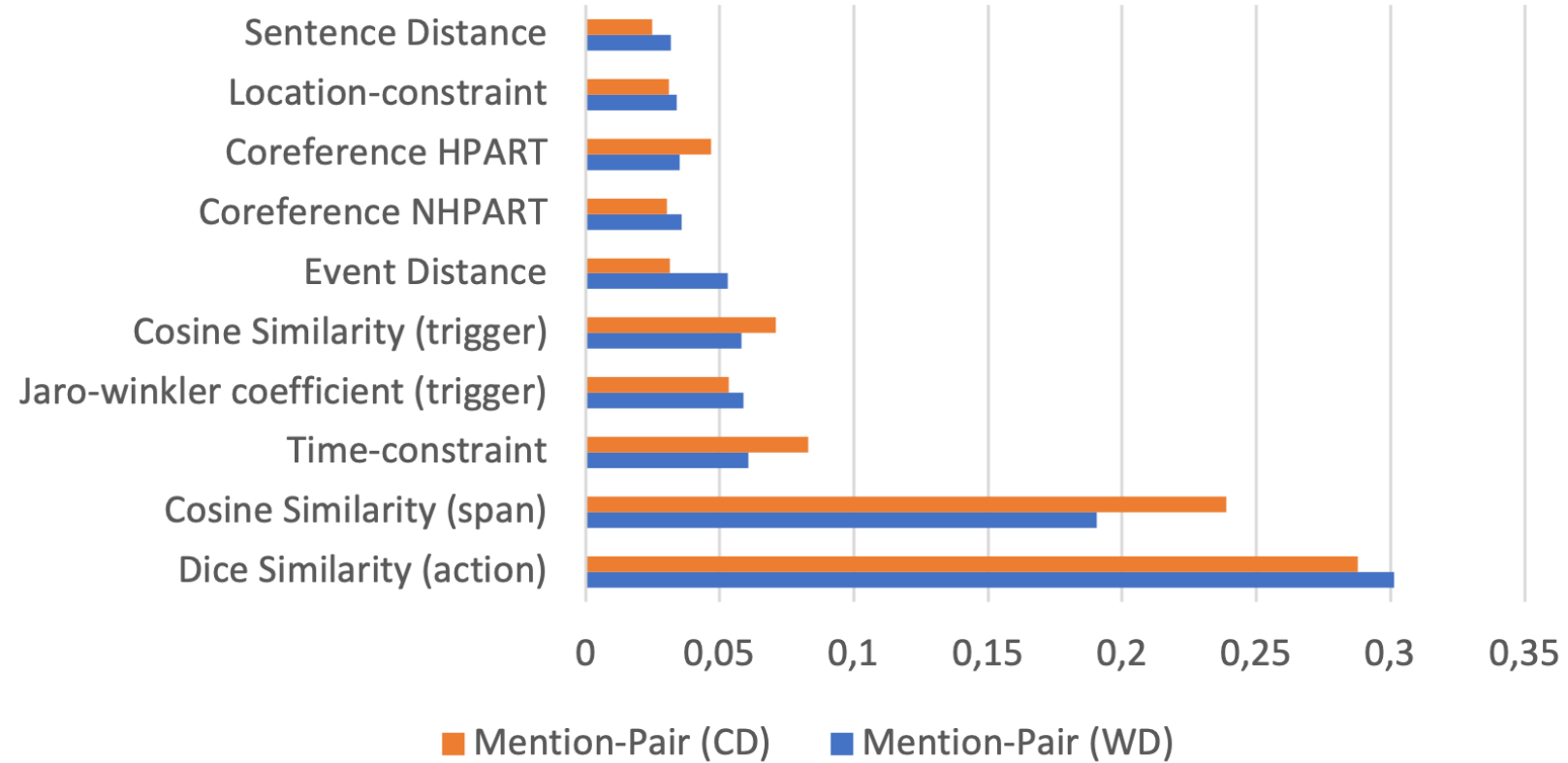
	CONLL	LEA
MP XGBOOST	0.37	0.23
MR <sub>base</sub>	0.35	0.22
MR <sub>task-specific</sub>	0.38	0.25
MR Embedding <sub>base</sub>	0.36	0.24
MR Embedding <sub>task-specific</sub>	0.40	0.28
MP BERTje	<b>0.59</b>	<b>0.39</b>
MP RobBERT	0.56	0.38
MP RobBERTje	0.54	0.35
MP XLM-RoBERTa	0.23	0.14
MP mBERT	0.19	0.10

(b) Results for cross-document ECR

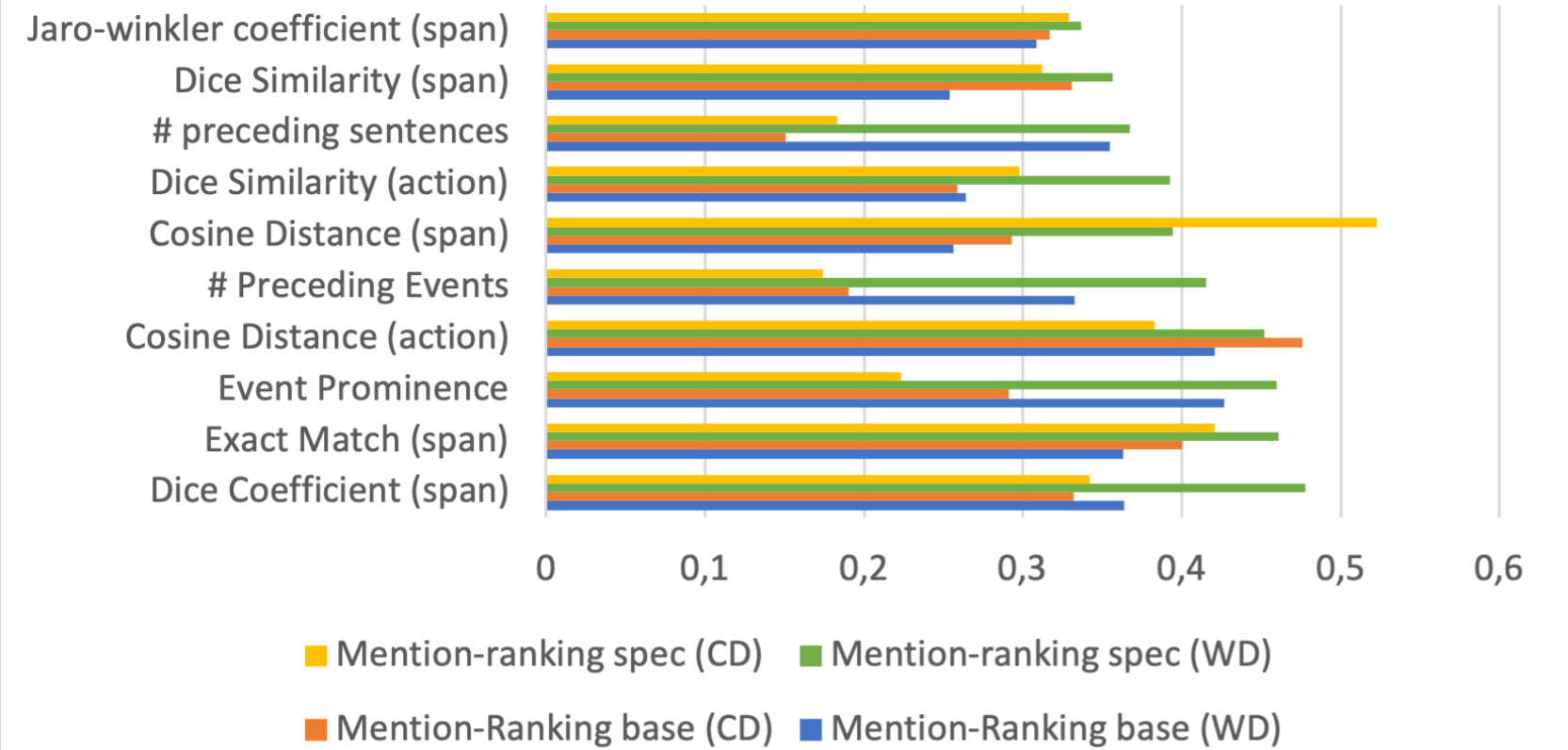


# DISCUSSION: FEATURE-BASED MODELS

Feature importance XGBOOST

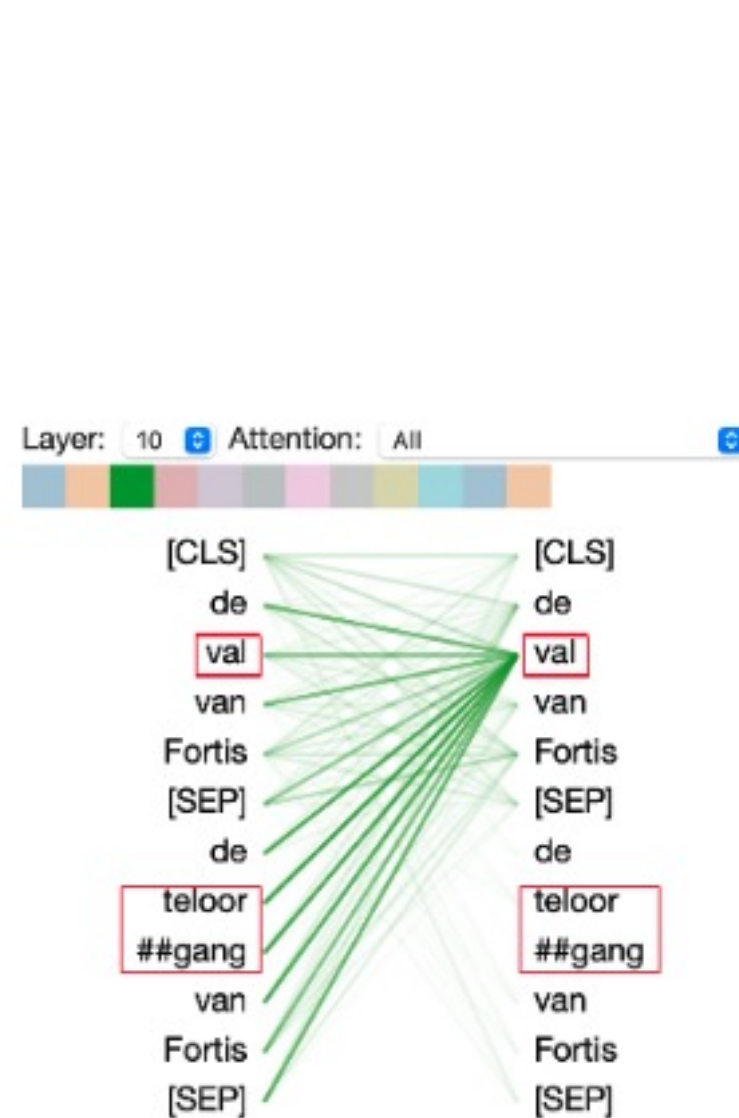


Feature Importance Mention-Ranking

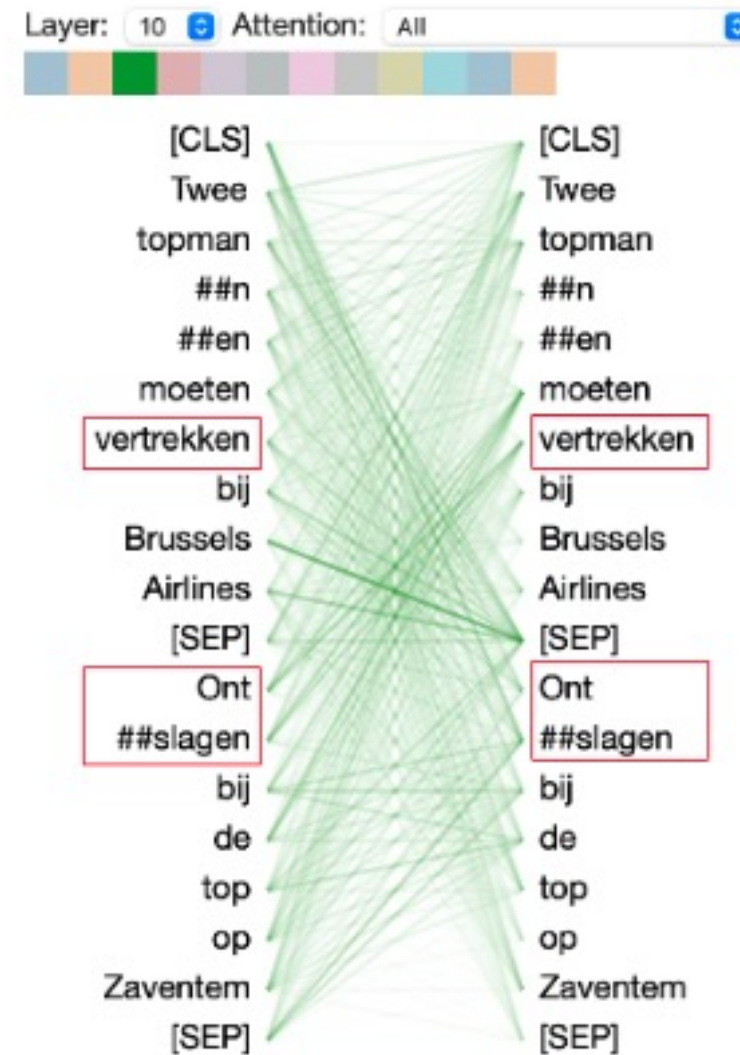


# DISCUSSION: TRANSFORMER-BASED MODELS

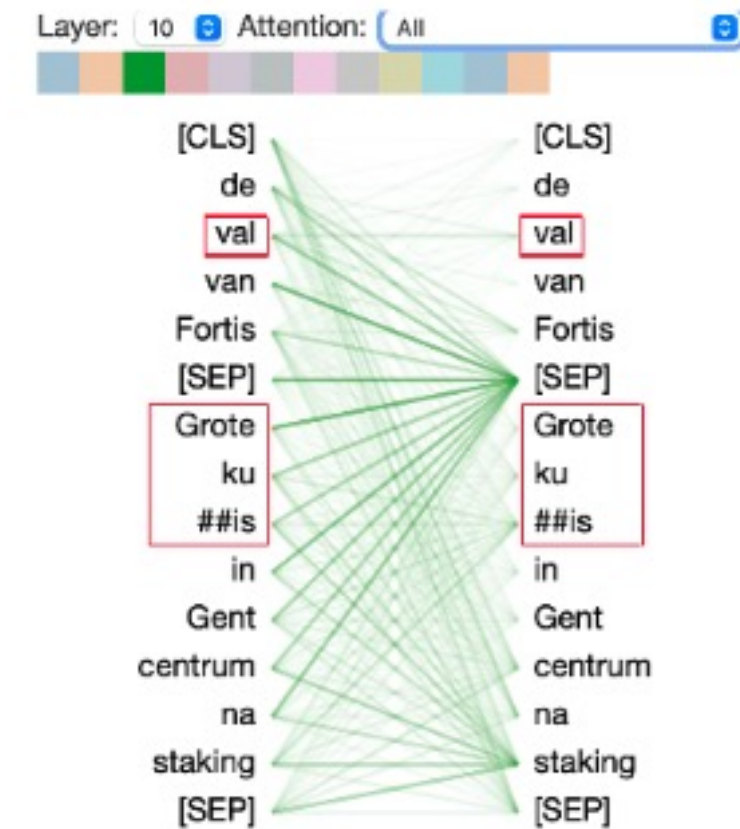
- Visualisation of transformer attention heads (Vig, 2019)



(a) The [downfall] of Fortis vs. The [decline] of Fortis



(b) Two executives have to [leave] at Brussels Airlines vs. [Dismissals] at the top of Zaventem



(c) The [downfall] of Fortis vs. [Large cleanup] in Gent city center after strikes

# CONCLUSION AND FUTURE RESEARCH

# CONCLUSION

- **Conclusion**

- ENCORE corpus allows exploration of Dutch event coreference resolution
- Large number of baseline experiments performed using both transformer-based and feature-based methods
- Analysis of the baseline (feature-based) experiments show trends similar to ECR studies in English

- **Future work**

- Integration of the baseline models into existing event mention detection systems (pipeline architecture)
- Exploratory studies regarding event-subevent relationships
- Development of joint extraction-coreference systems for Dutch ECR using SpanBERT architectures, Graph Neural Networks (GNN), ...