

# **Joint part-of-speech and dependency projection from multiple sources**

Anders Johannsen\*

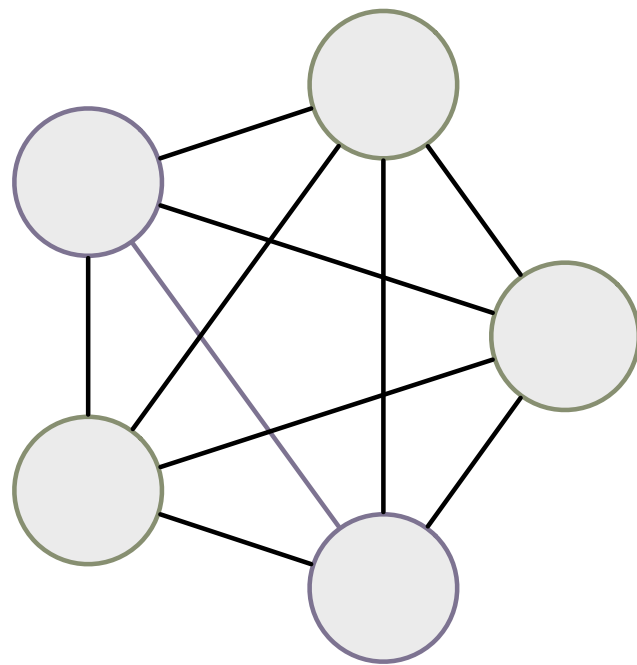
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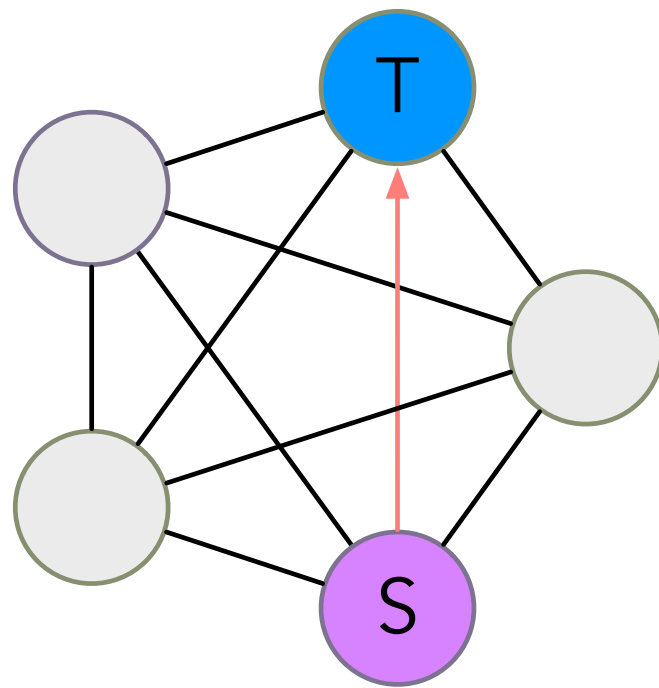
# Annotation projection

Parallel corpora



# Annotation projection

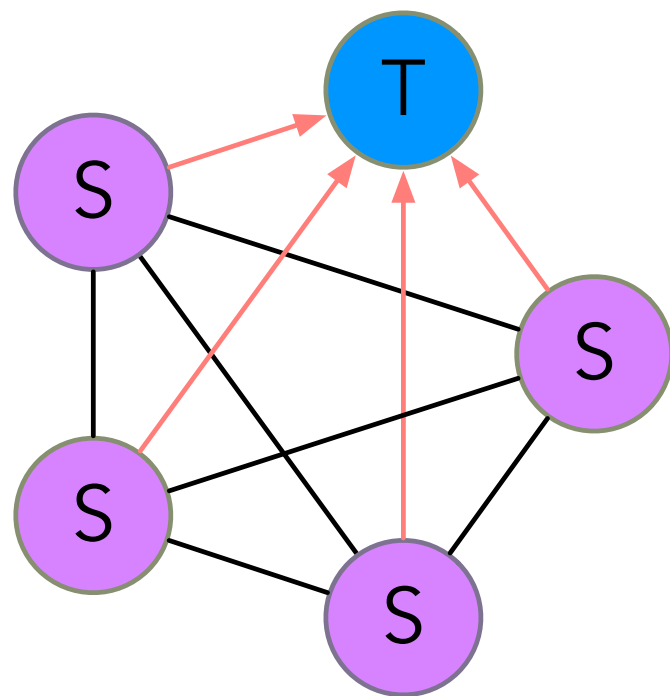
Parallel corpora



transfer annotation from source to target

# Annotation projection

Parallel corpora

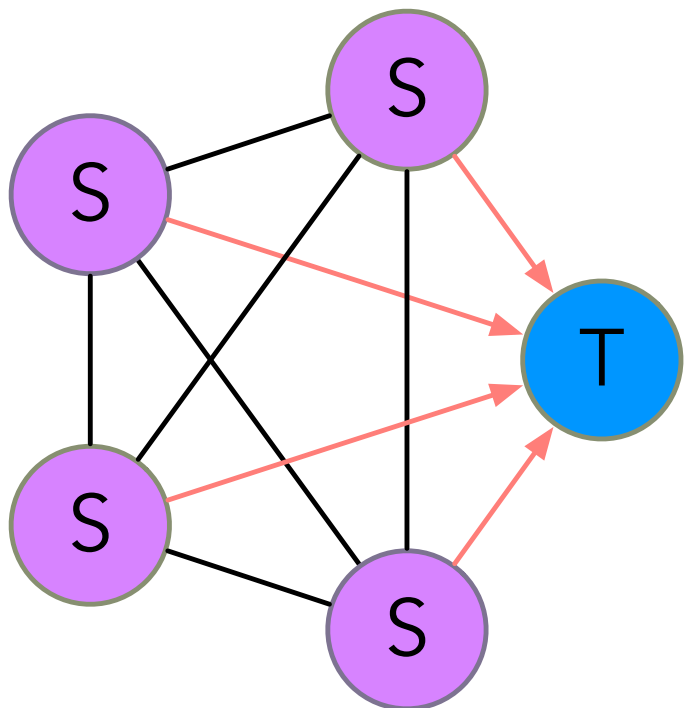


transfer annotation from source to target

may have multiple sources

# Annotation projection

Parallel corpora



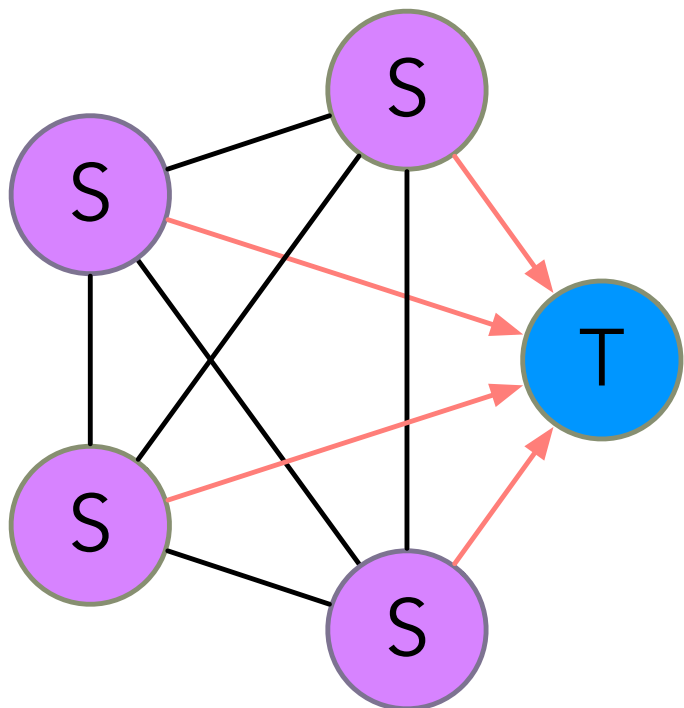
transfer annotation from source to target

may have multiple sources

evaluate by leave-one-out

# Annotation projection

Parallel corpora



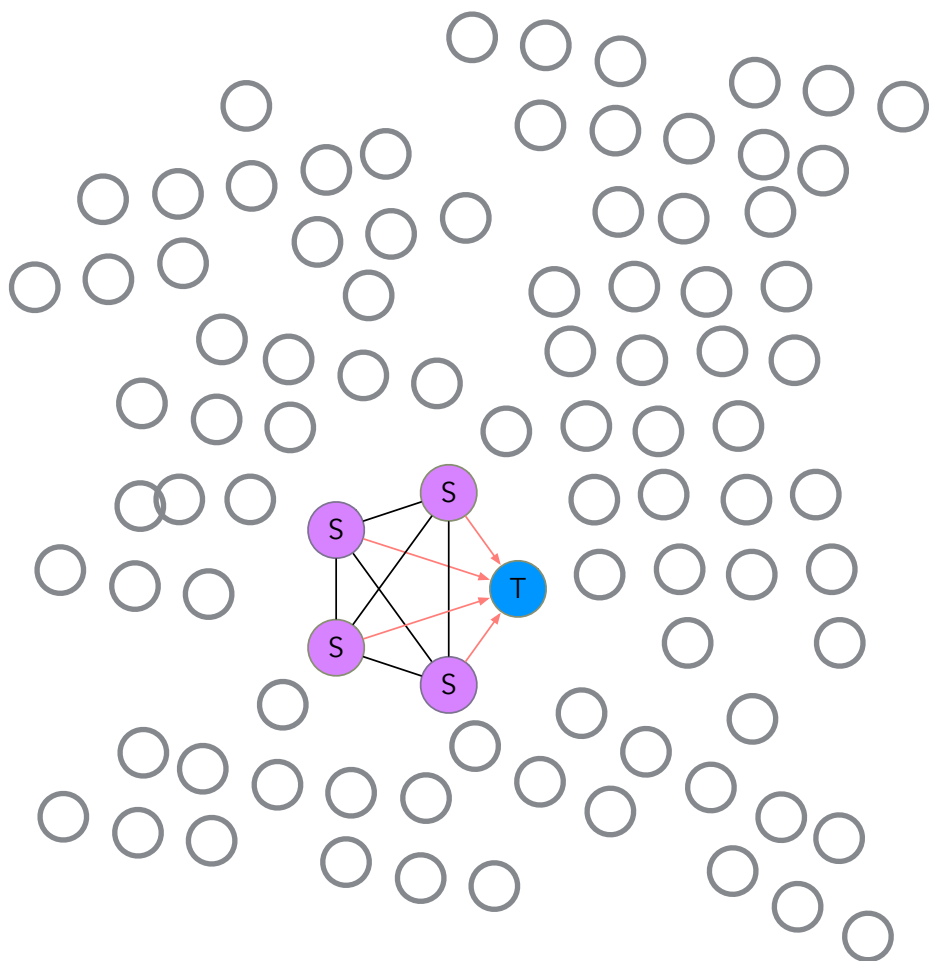
train parser



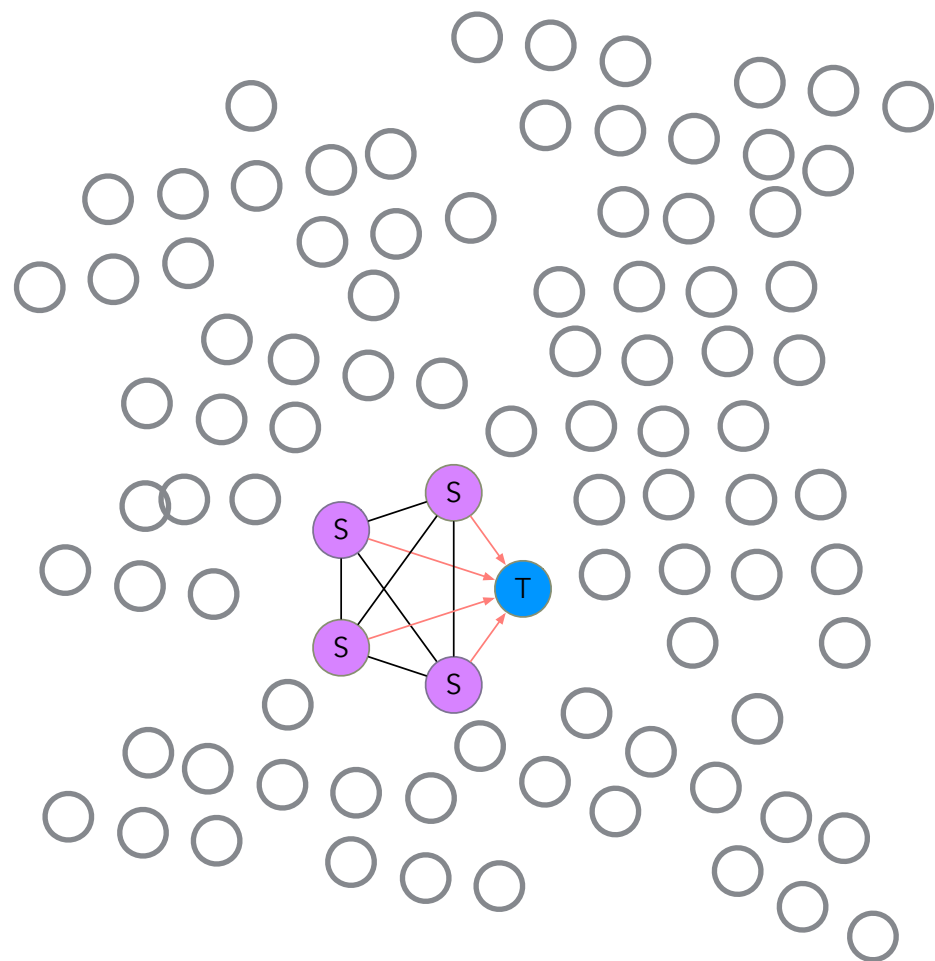
parse test set

# The many languages of the world

cross-lingual parsing suffers a little  
from EUROPARLalism



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cross-lingual parsing suffers a little  
from EUROPARLalism

This work extends Agić (2016):

train models for hundreds of languages

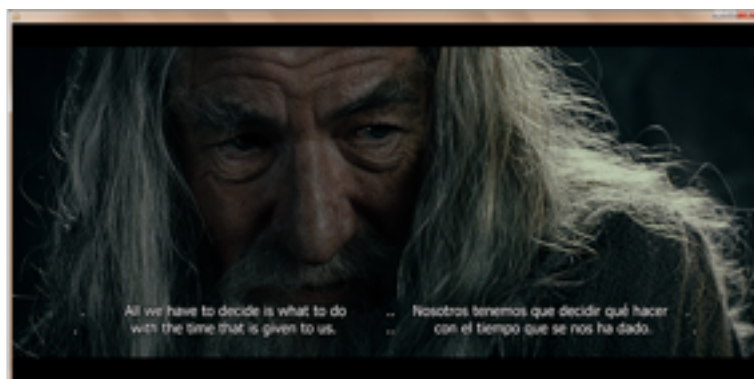
evaluate on 26 languages



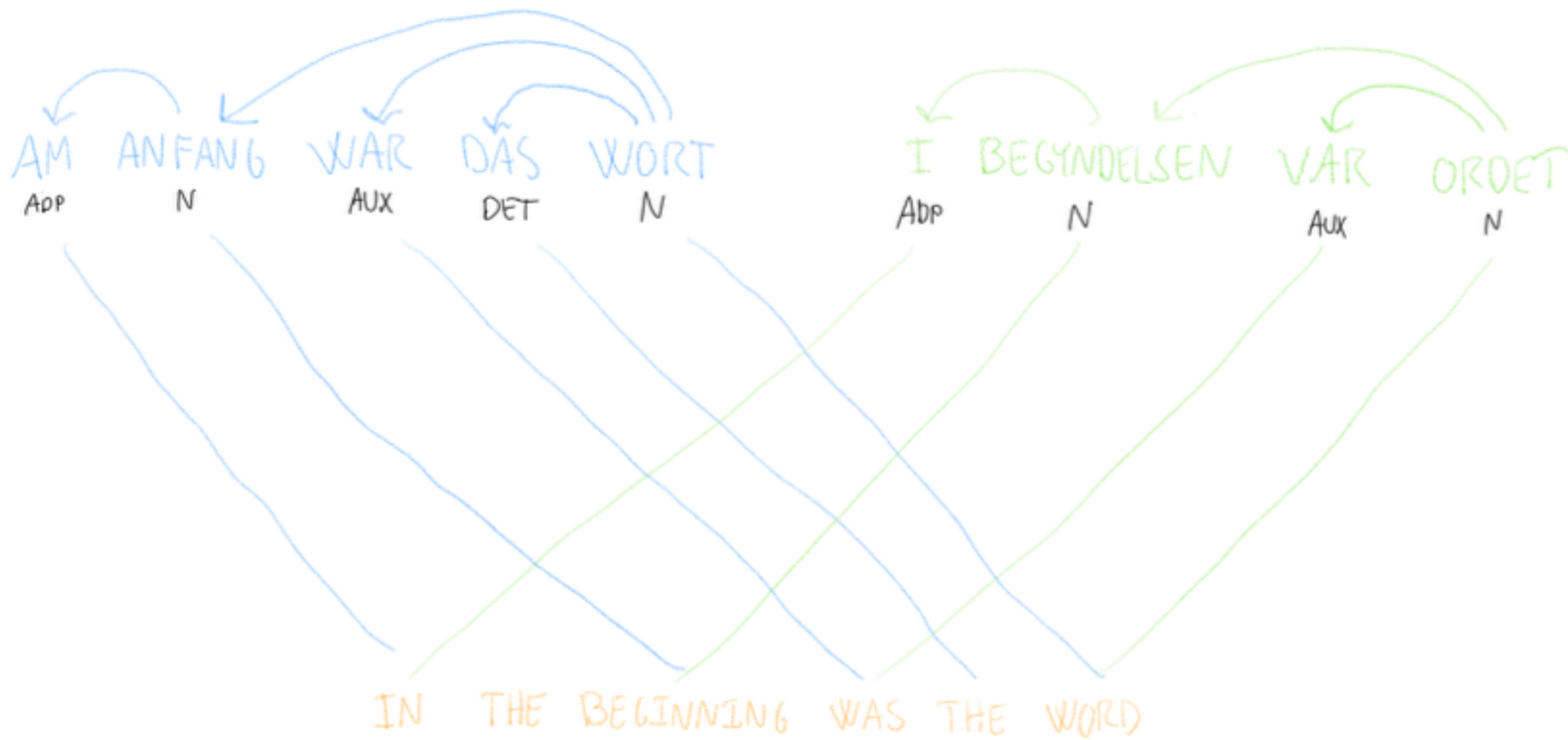
# Our corpora



# Our corpora

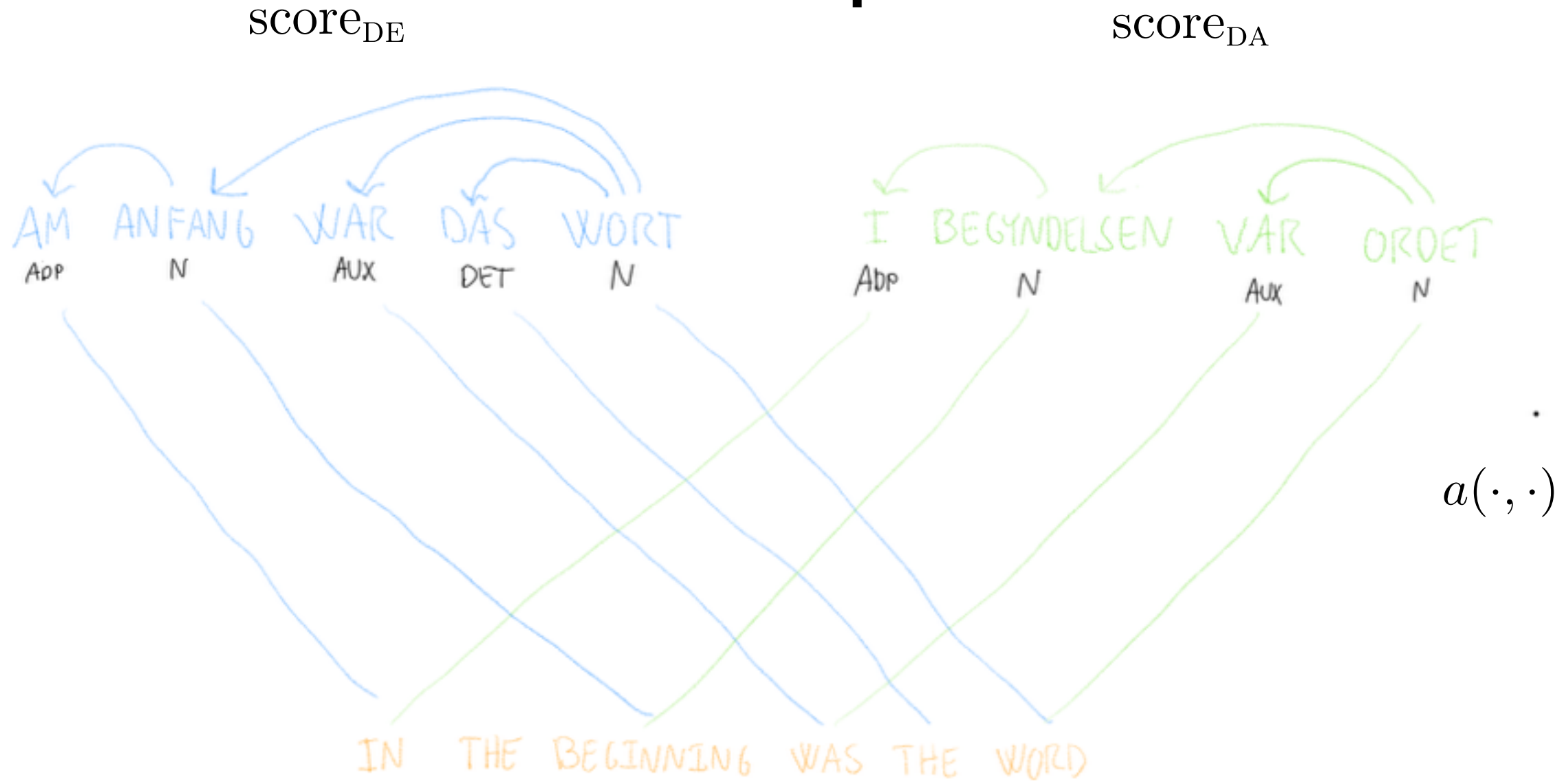


# Example

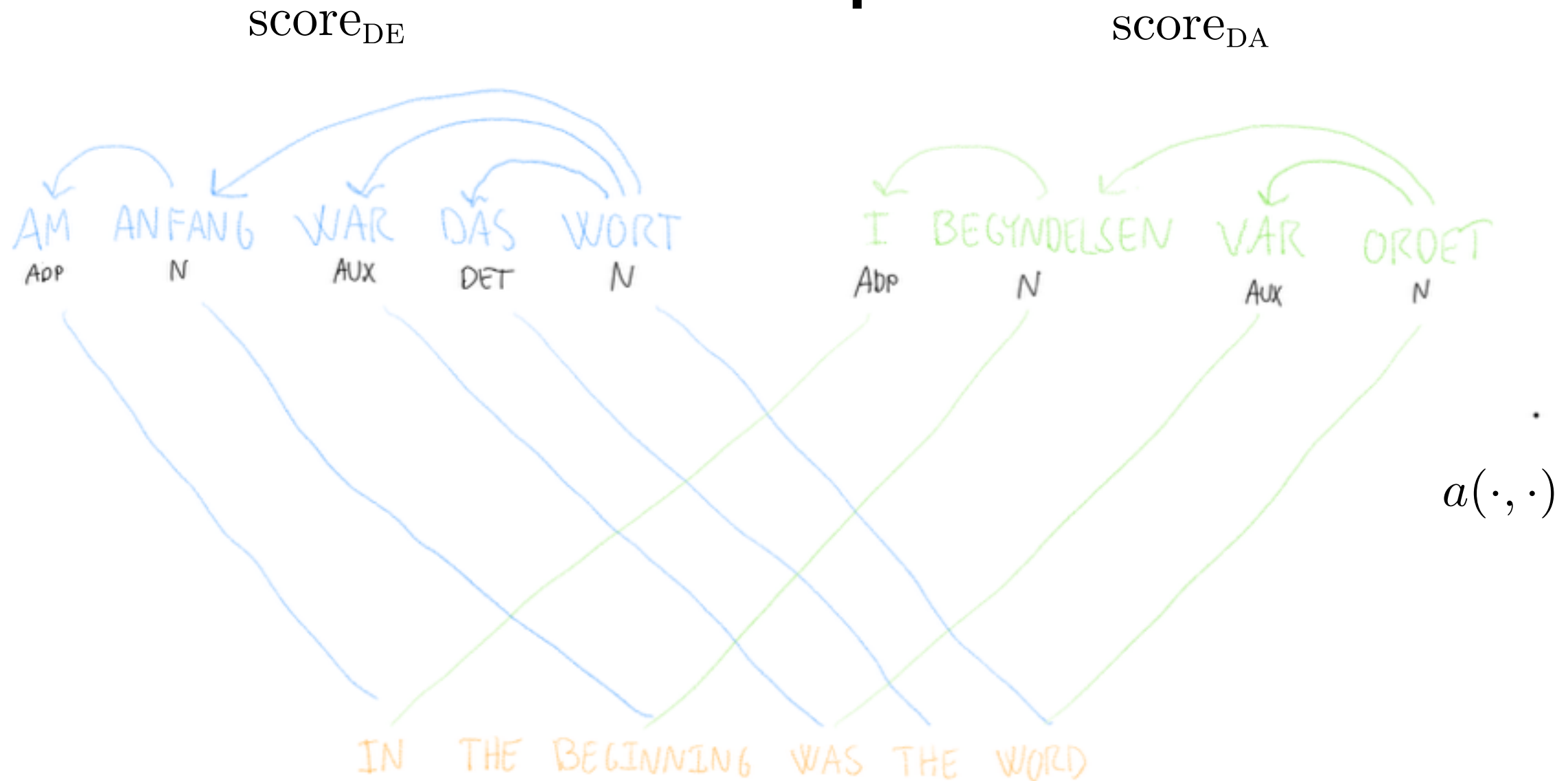




# Example

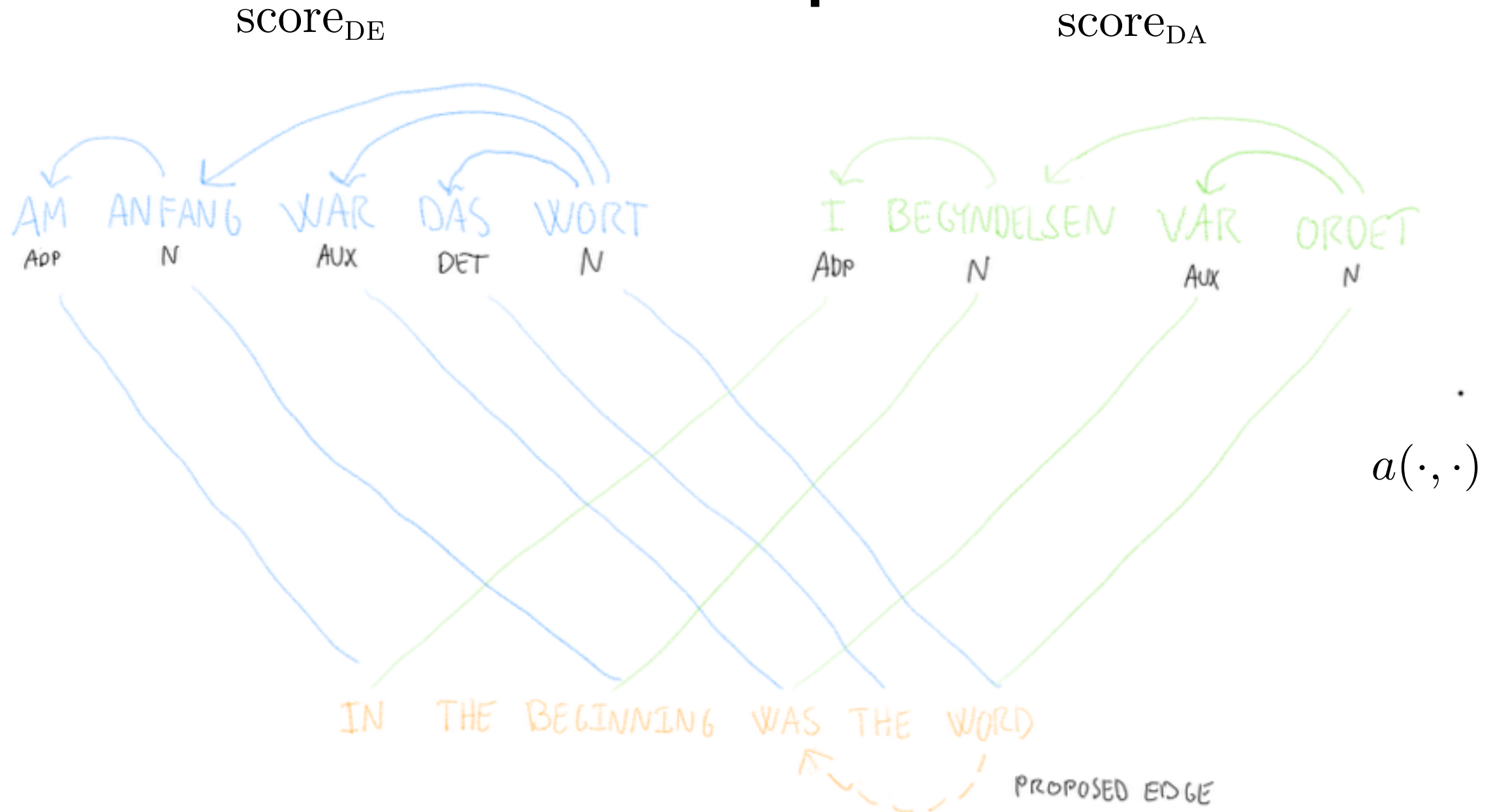


# Example



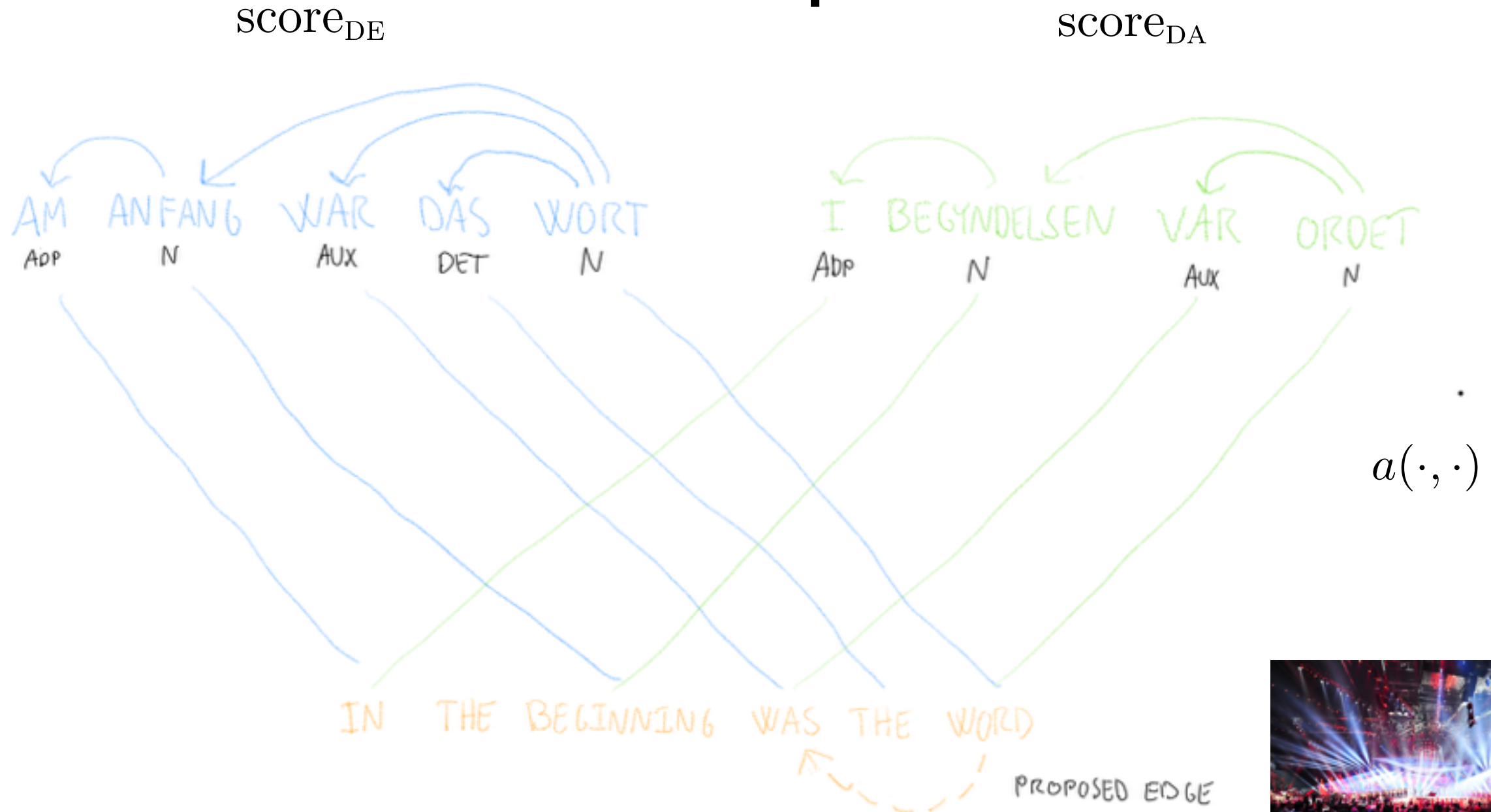
$$\arg \max_y \sum_{(i,j) \in y} \text{score}_T(i,j) \quad \text{s.t. } y \text{ is a tree}$$

# Example



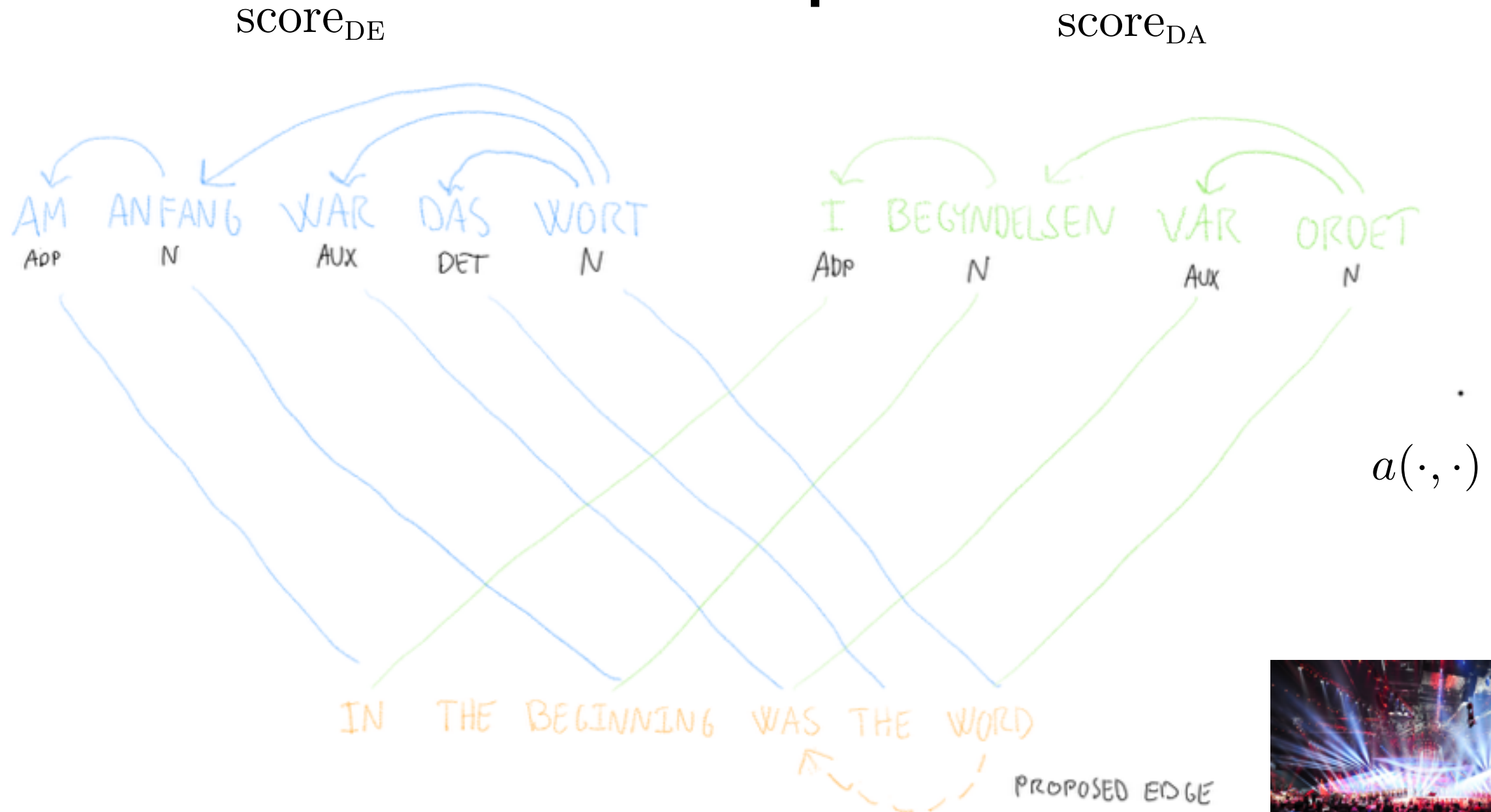
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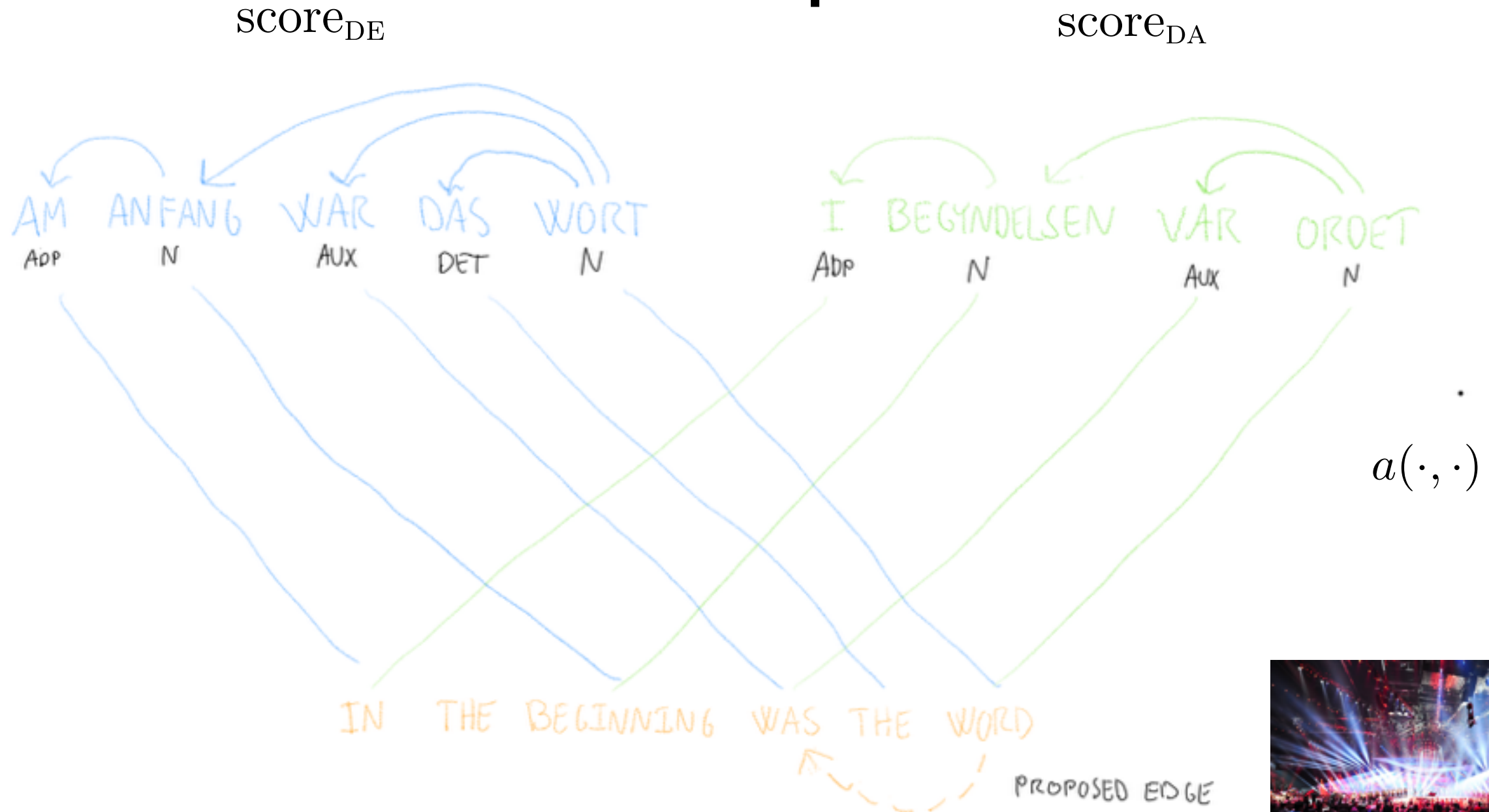


$$\arg \max_y \sum_{(i,j) \in y} \text{score}_T(i, j) \quad \text{s.t. } y \text{ is a tree}$$

$$\text{score}_T(\text{word}, \text{was}) = \text{score}_{\text{DA}}(\text{ordet}, \text{var})$$



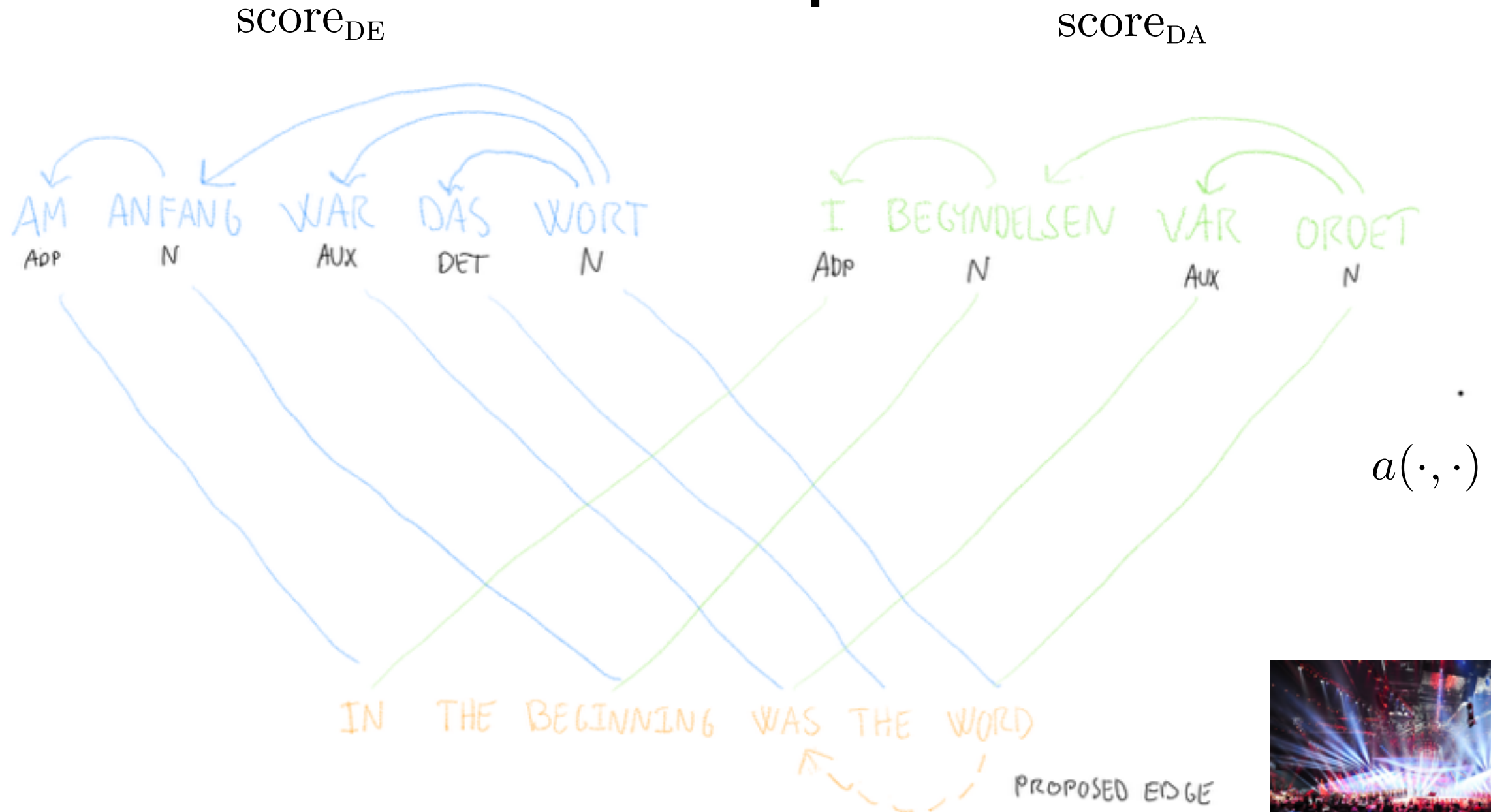
# Example



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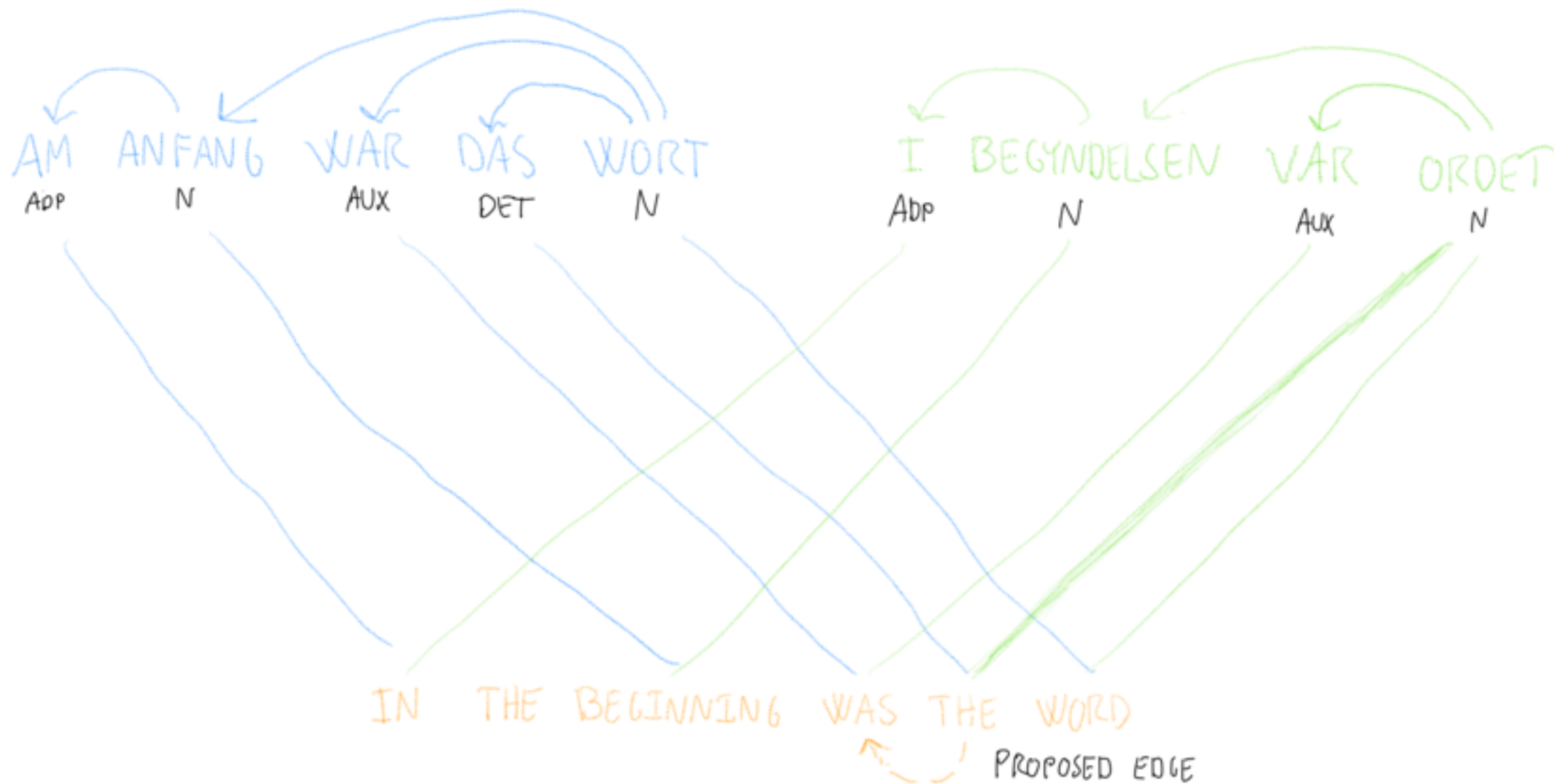
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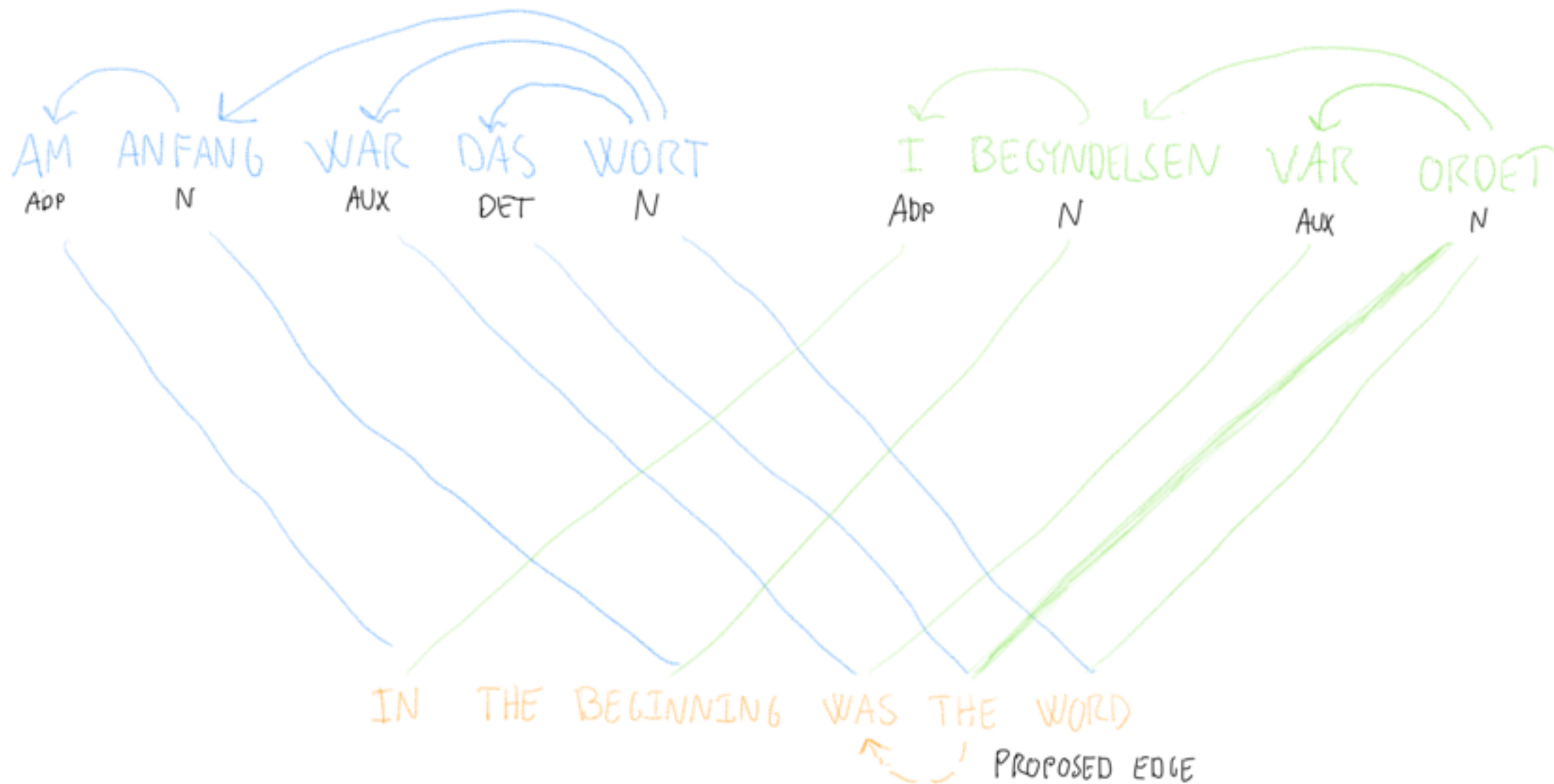
$$\begin{aligned} \text{score}_T(\text{word}, \text{was}) = & \text{score}_{\text{DA}}(\text{ordet}, \text{var}) a(\text{var}, \text{was}) a(\text{ordet}, \text{word}) \\ & + \text{score}_{\text{DE}}(\text{wort}, \text{war}) a(\text{war}, \text{was}) a(\text{wort}, \text{word}) \end{aligned}$$

# Example gone bad



$$\text{score}_T(\text{the}, \text{was}) = \text{score}_{DA}(\text{ordet}, \text{var}) a(\text{var}, \text{was}) a(\text{ordet}, \text{word})$$

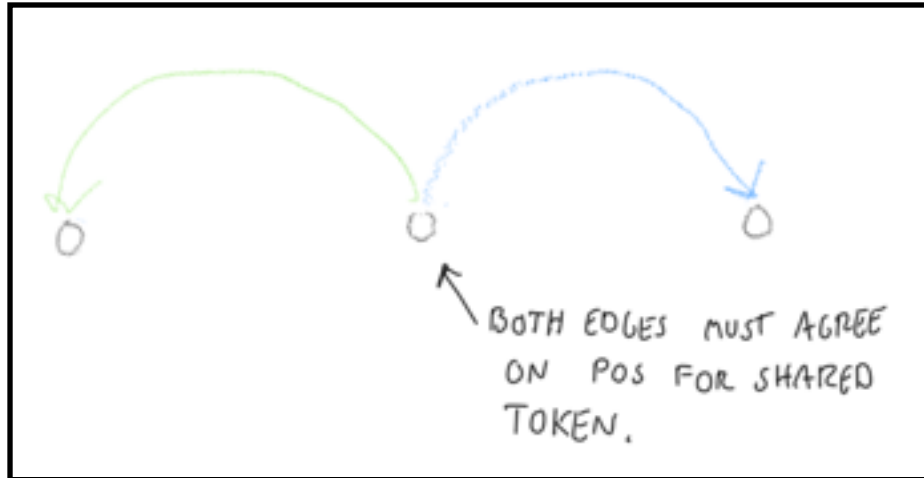
# Example gone bad



$$\text{score}_T(\text{the}, \text{was}) = \cancel{\text{score}_{DA}(\text{ordet}, \text{var})} \alpha(\text{var}, \text{was}) \alpha(\text{ordet}, \text{word})$$

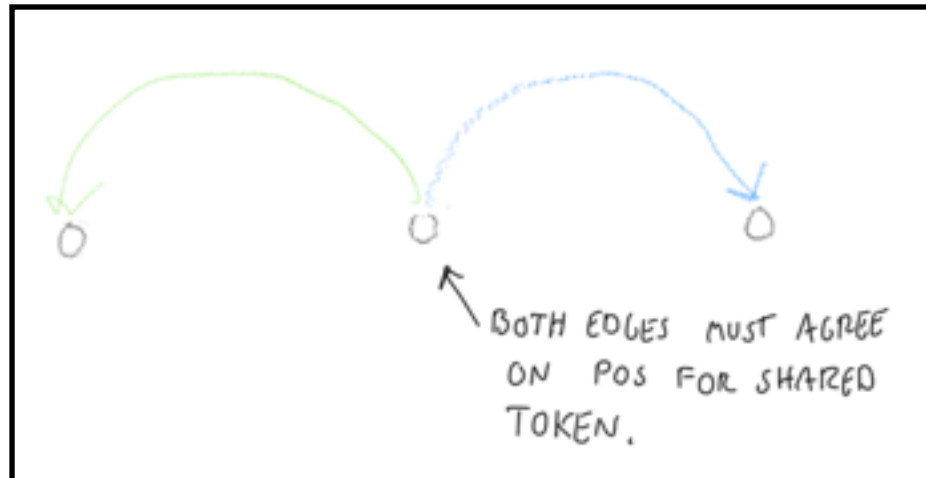
Yes, but only if "was" is AUX and "the" is N

# Projecting layers of annotation



$$\arg \max_y \sum_{(i,k,j,l) \in y} \text{score}_T(i, k, j, l) \quad \text{s.t. } y \text{ is a tree}$$

# Projecting layers of annotation

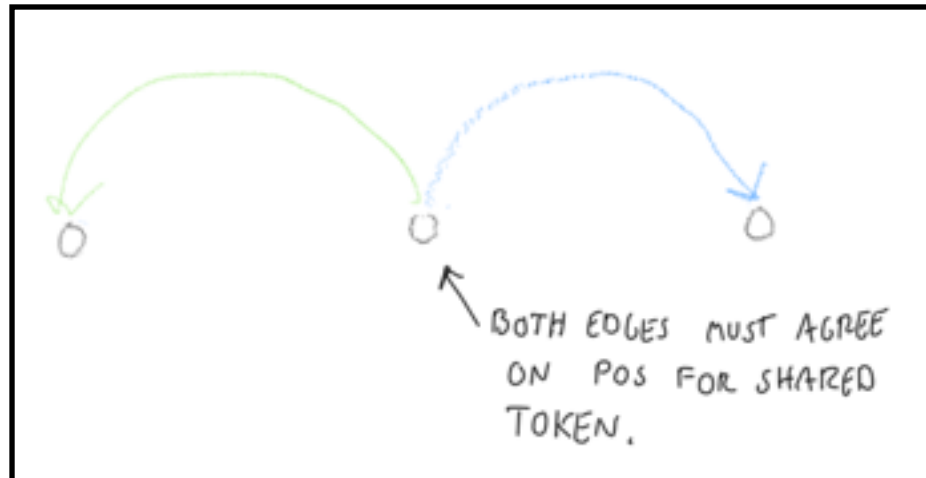


the edge  $(i, j)$

$$\arg \max_y \sum_{(i,k,j,l) \in y} \text{score}_T(i, k, j, l) \quad \text{s.t. } y \text{ is a tree}$$

Two red arrows point from the text "the edge (i, j)" to the variables  $i$  and  $j$  in the score function of the equation above.

# Projecting layers of annotation

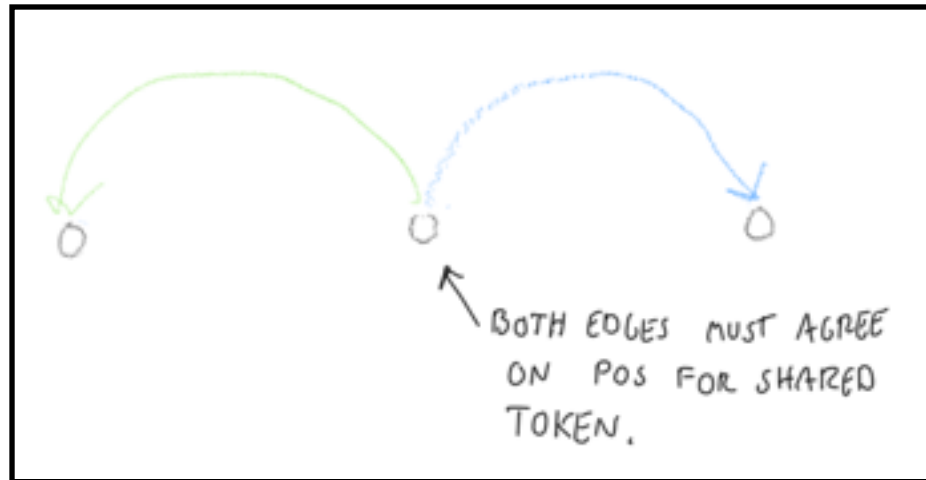


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tags for  $i$  and  $j$

# Projecting layers of annotation



the edge (i, j)

$$\arg \max_y \sum_{(i,k,j,l) \in y} \text{score}_T(i, k, j, l) \quad \text{s.t. } y \text{ is a tree}$$

tags for i and j

more difficult



## ILP model

$$\begin{array}{lll} \text{Edges} & e_{i,k,j,l} & \in \{0, 1\} \\ \text{Vertices} & v_{i,k} & \in \{0, 1\} \\ \text{Flow} & \phi_{i,k,j,l} & \in \mathbb{R}^+ \end{array}$$

$$\text{Maximize} \quad \sum_{i,k,j,l} e_{i,k,j,l} w_{i,k,j,l}$$

One parent per token

$$\sum_{i,k,l} e_{i,k,j,l} = 1 \quad \forall j \neq 0$$

The root token (index 0) sends  $n$  flow

$$\sum_{j,l} \phi_{0,0,j,l} = n$$

Each token consumes one unit of flow

$$\sum_{i,k,l} \phi_{i,k,x,l} - \sum_{k,j,l} \phi_{x,k,j,l} = 1 \quad \forall x \neq 0$$

One POS per token

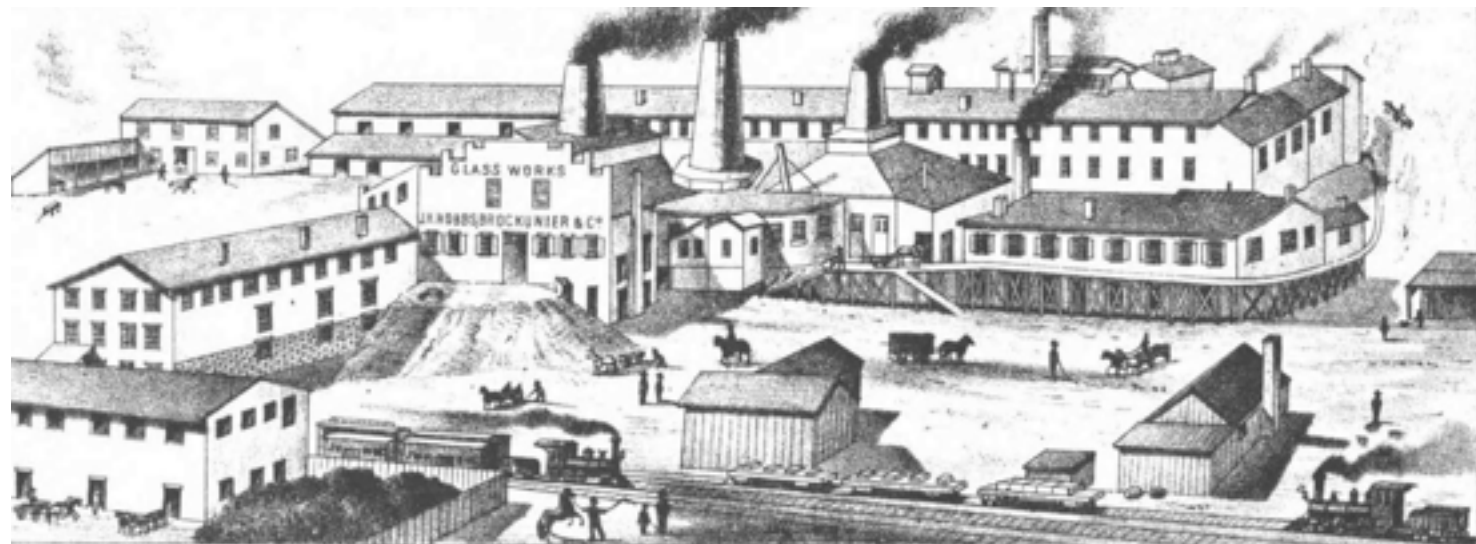
$$\sum_k v_{i,k} = 1 \quad \forall i \neq 0$$

Active edges choose token POS

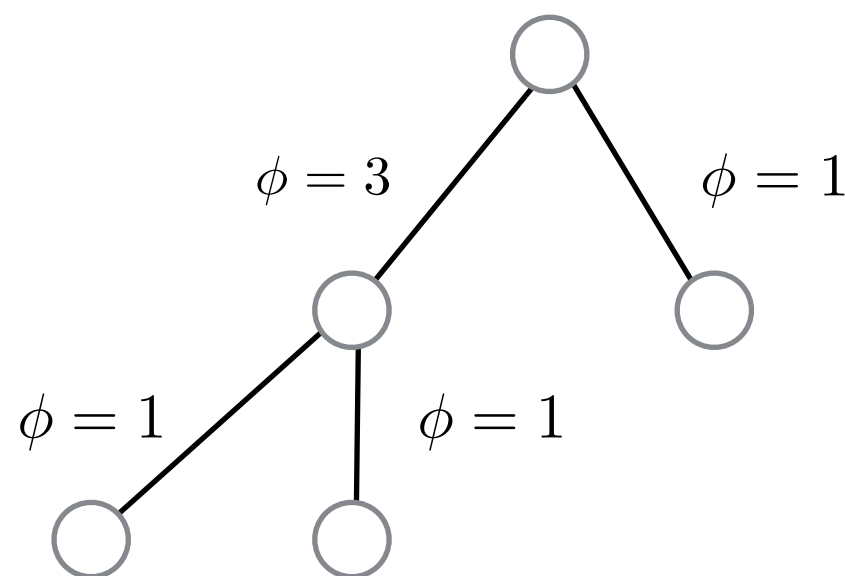
$$v_{i,k} \geq e_{i,k,j,l} \quad \forall i \neq 0, j, k, l$$

$$v_{i,l} \geq e_{i,k,j,l} \quad \forall i, j, k, l$$

Above,  $i, j$ , and  $x$  are token indices, while  $k$  and  $l$  refer to POS. Quantification over these symbols in the equations are always with respect to a given target graph.



Root produces  $n$  flow



Each node consumes one flow

# Results

		<i>Approach</i>		
<i>Predicted POS</i>		ILP	DCA	DELEX
	EBC	<b>51.62</b> (18)	48.39 (8)	42.44 (1)
	WTC	<b>53.58</b> (20)	48.40 (0)	47.35 (3)
<i>Gold POS</i>				
	EBC	<b>65.43</b> (25)	59.94 (2)	64.13 (–)
	WTC	<b>66.51</b> (23)	55.73 (0)	66.68 (–)

<i>POS tagging</i>	
EBC	WTC
69.40	73.05

# Conclusion

We extended Agić (2016) to project multiple layers of annotation jointly.

Approach stays simple and heuristics-free.

The initial experiments show promising results.

## Future work

Project higher/lower layers of annotation, or larger tree parts.

Penalise inconsistent structures instead of disallowing.

**Questions?**