

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



Anders Johannsen*

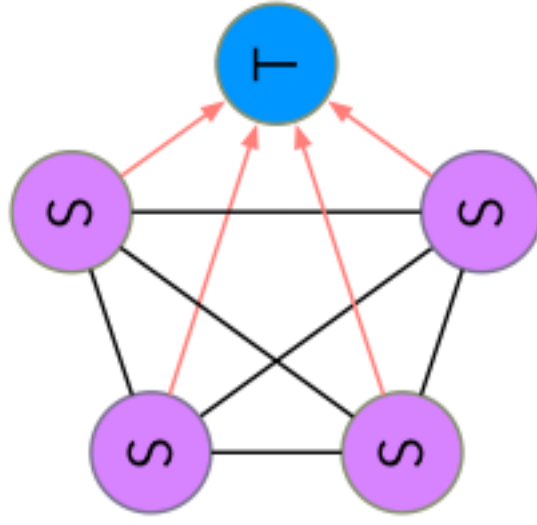
Željko Agić*

Anders Søgaard

(formerly)* University of Copenhagen

Annotation projection

Parallel corpora

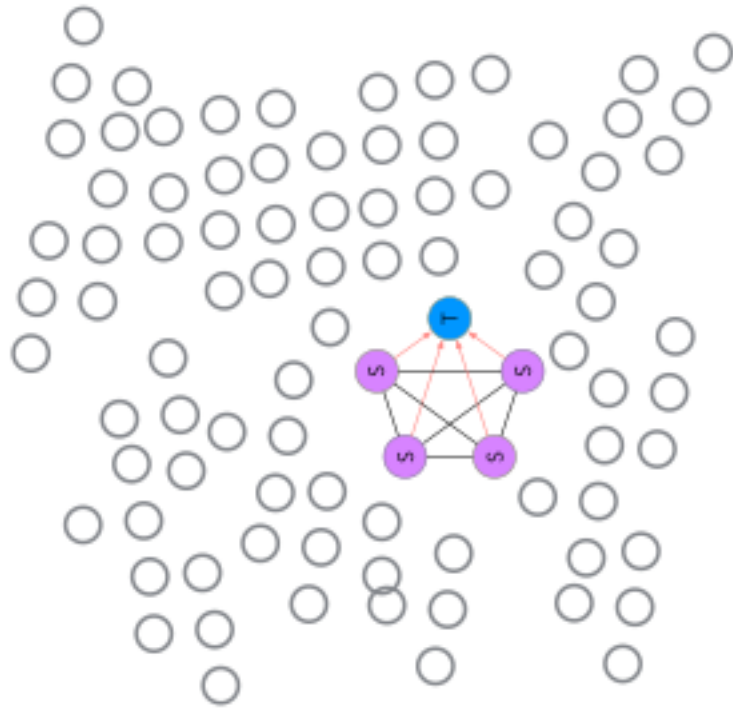


transfer annotation from source to target

~~may have parallel source~~ sparse test set

evaluate by leave-one-out

The many languages of the world



cross-lingual parsing suffers a little from
EUROPARLalism

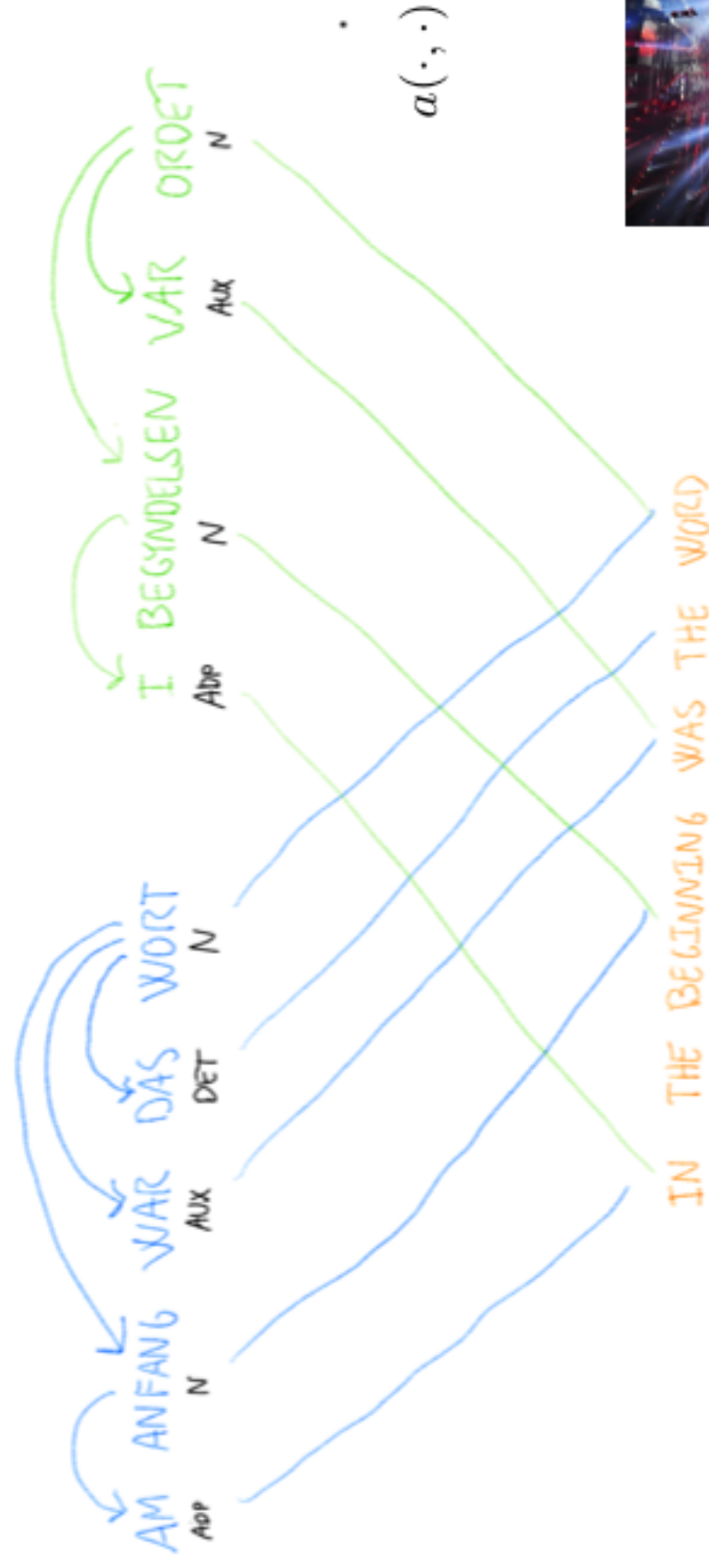
This work extends Agić et al. (2016):

train models for hundreds of languages

evaluate on 26 languages

Our corpora

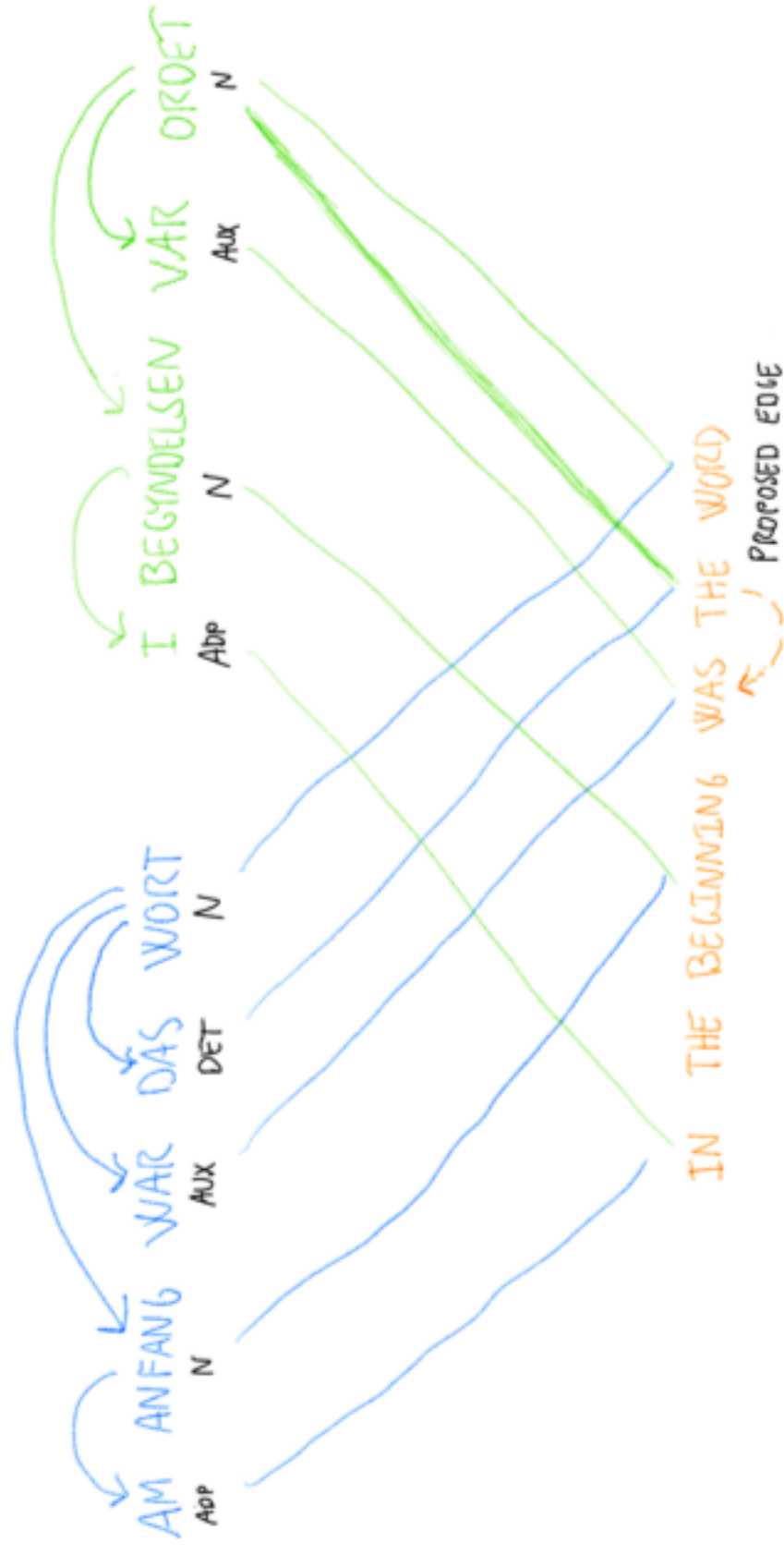


SCORE_{DE}SCORE_{DA}

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

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

Agic et al. gone bad

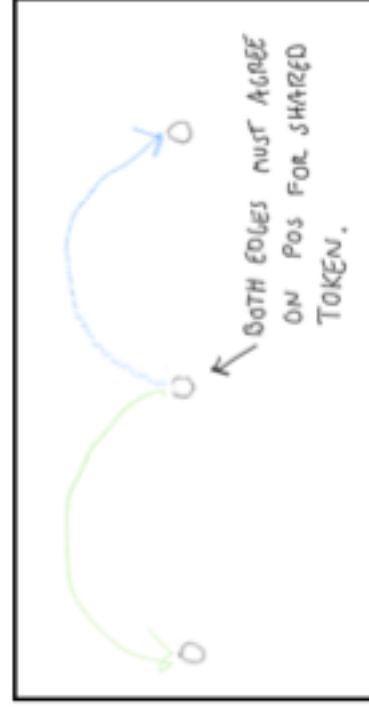


$\text{score}_T(\text{the, was}) = \text{score}_{DA}(\text{ordet, var}) \alpha(\text{var, was}) \alpha(\text{ordet, word})$

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

Projecting layers of annotation

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

Edges $e_{i,k,j,l} \in \{0, 1\}$

Vertices $v_{i,k} \in \{0, 1\}$



Flow $\phi_{i,k,j,l} \in \mathbb{R}^+$

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

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

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

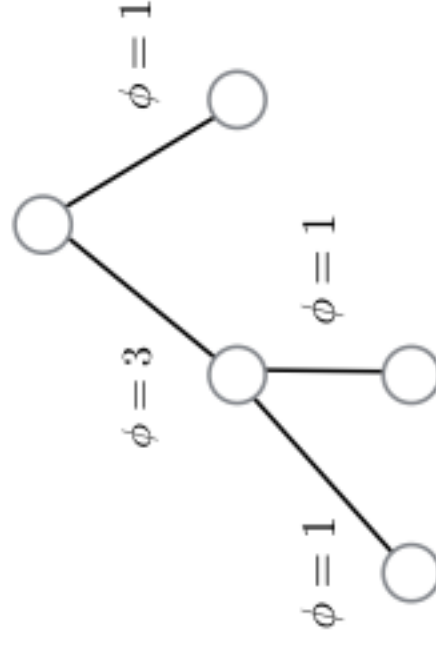
Active edges choose token POS

$$\begin{aligned} v_{i,k} &\geq e_{i,k,j,l} & \forall i \neq 0, j, k, l \\ v_{i,l} &\geq e_{i,k,j,l} & \forall i, j, k, l \end{aligned}$$

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

(Martins, 2012) 8

Results

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

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

Conclusion

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

Approach stays simple and heuristics-free.

These initial experiments show promising results.

Future work

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

Penalise inconsistent structures instead of disallowing.

Questions?