



# Universal Dependency Parsing from Scratch

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## System Highlights

Fully neural pipeline system for all stages required by the Shared Task, including: word/sentence segmentation, POS/UFeats tagging, lemmatization, and dependency parsing.

	LAS	MLAS	BLEX
Stanford	72.29 (7 <sup>th</sup> )	60.92 (2 <sup>nd</sup> )	64.04 (5 <sup>th</sup> )
Stanford+	74.16	62.08	65.28
Top System	75.84	61.25	66.09

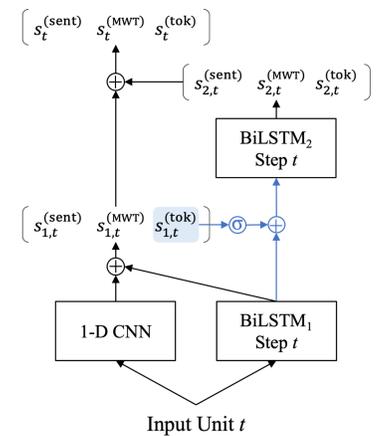
## Key Contributions

- Symbolic statistics  $\otimes$  Neural nets
- Biaffine POS/UFeats Classifier  $\rightarrow$  Consistency
- Lemmatizer + Edit Classifier  $\rightarrow$  Shortcut for long sequence
- Biaffine parser with relative location features
- Fully open-source implementation/models

## Tokenizer / Sentence Segmenter

Joint word and sentence segmentation as tagging.

sent					x
MWT	x				
tok	x	x	x	x	x
		ch	bin	im	Auto.



## Multiword-Token (MWT) Expansion

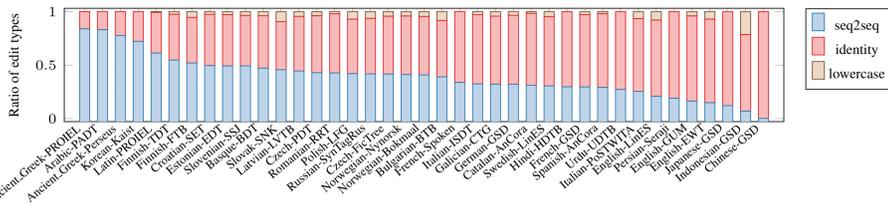
German example: *im*  $\rightarrow$  *in dem*

Combining a frequency lexicon (symbolic) with a seq2seq model as fallback (neural).

System	Tokens	Sentences	Words
Stanford+	<b>99.46</b>	<b>91.33</b>	<b>99.27</b>
- conv	99.45	91.03	98.67
- seq2seq	-	-	98.97

## Lemmatizer

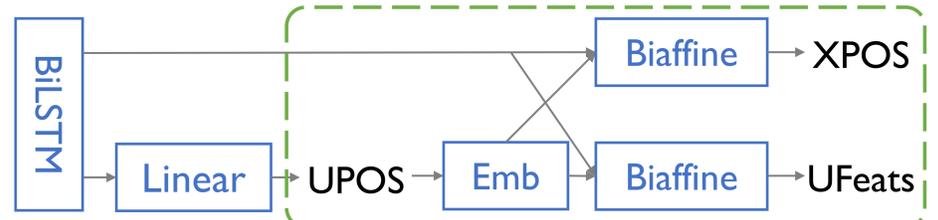
- 2 Frequency lexicons (word  $\times$  UPOS, word) + seq2seq
- Edit (e.g., lowercase, identity) classifier to skip decoder e.g., URL, email, proper nouns, sentence-leading capitalization, etc.



System	Big	Small	Low-Res	All
Stanford	<b>96.56</b>	<b>91.72</b>	<b>69.21</b>	<b>94.22</b>
- edit & seq2seq	89.97*	82.68*	63.50*	87.45*
- lexicons	95.37*	90.43*	66.02*	92.89*

## POS / UFeats Tagger

Biaffine classifier that conditions on UPOS embeddings to enforce consistency between predicted tags/features.



System	UPOS	XPOS	UFeats	AllTags	PMI
Stanford	<b>96.50</b>	<b>95.87</b>	<b>95.01</b>	<b>92.52</b>	<b>.0514</b>
- biaffine	96.47	95.71*	94.13*	91.32*	.0497*

## Dependency Parser

Deep Biaffine parser + terms to model relative positions of heads and dependents:

- Linearization: If there's an edge, are the words in the right order?
- Distance: If there's an edge, are the words close/far enough?

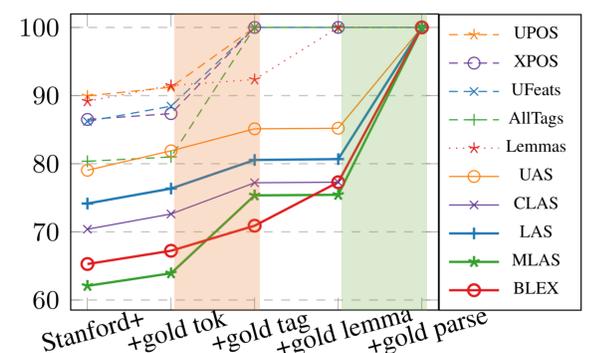
System	LAS	CLAS
Stanford	<b>87.60</b>	<b>84.68</b>
- linearization	87.55*	84.62*
- distance	87.43*	84.48*

## Full Pipeline

Highly accurate system on small and large treebanks alike.

Easy to make iterative improvements to the performance of the entire pipeline.

Treebanks	System	LAS	MLAS
Small	Stanford+	<b>83.90</b>	<b>72.75</b>
	Top System	69.53	49.24
Low-Res	Stanford+	<b>63.20</b>	<b>51.64</b>
	Top System	27.89	6.13
PUD	Stanford+	<b>82.25</b>	<b>74.20</b>
	Top System	74.20	58.75



## Code and Pretrained Models



Tensorflow tagger/parser code at <https://github.com/tdozat/Parser-v3>

PyTorch full pipeline at <https://github.com/stanfordnlp/stanfordnlp>

For more details see (also in QR code): <https://stanfordnlp.github.io/stanfordnlp>

Most of the potential for gains still remain with the challenging tasks of tagging and parsing.