## Evaluation of a Sequence Tagging Tool for Biomedical Texts

Julien Tourille<sup>1,2</sup>, Matthieu Doutreligne<sup>1,3</sup>, Olivier Ferret<sup>4</sup> Nicolas Paris<sup>1,3,5</sup>. Aurélie Névéol<sup>1</sup>. Xavier Tannier<sup>5,6</sup>

<sup>1</sup>LIMSI, CNRS, Université Paris-Saclay – <sup>2</sup>Univ. Paris-Sud <sup>3</sup>WIND-DSI, AP-HP – <sup>4</sup>CEA, LIST – <sup>5</sup>Sorbonne Université – <sup>6</sup>INSERM, LIMICS

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#### Motivation

- Sequence tagging e.g. NER → initial step which allows to perform more complex analysis in the biomedical domain (e.g. relation extraction or text classification)
- Several SOTA neural sequence tagging models have been published
  - LSTM-CRF (Lample et al. 2016)
  - ► CNN-CRF (Ma and Hovy 2016)
  - LSTM-CRF + Char. Att. (Rei et al. 2016)
  - ► LSTM + LM Objective (Rei 2017)
- ► Lack of performance analysis of such models in the biomedical domain
- Lack of an off-the-shelf efficient implementation

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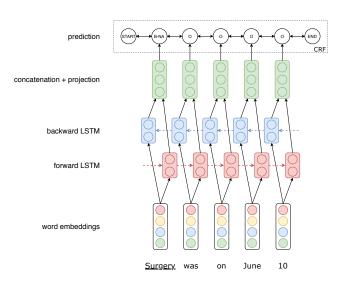
#### Contributions

- ➤ Yet Another Sequence Tagger (YASET): fast and accurate implementation of the LSTM-CRF model (Lample et al. 2016)
  - supports the use of handcrafted features
  - easy-to-use interface
- Evaluation on various biomedical corpora and the CoNLL 2003 corpus

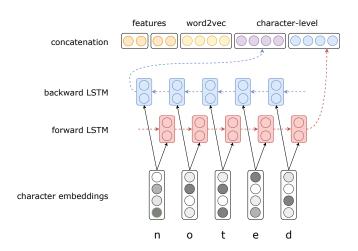
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## YASET - Main Component



## YASET - Input Embeddings



## YASET - Training

- ▶ Input data: CoNLL format
- **Out of vocabulary tokens**  $\rightarrow$  training of an *unknown* embedding by replacing singletons in the training corpus (according to probability p)
- Classical hyperparameters for neural models (early stopping, evaluation metric, hidden layer sizes, ...)
  - ightarrow Located in one configuration file

## Experiments

#### Two experiment sets

- 1. Performance evaluation on various corpora
  - ► MedPost (POS) (Smith et al. 2004)
  - ► NCBI disease (NER) (Doğan et al. 2014)
  - ► MERLoT (NER) (Campillos et al. 2018)
  - CoNLL (NER) (Sang and Meulder 2003)
- 2. Influence of training data size on performance (annotation is time-consuming and expensive): MERLoT corpus

## Experiments - Corpora

Lang.	Corpus	# sent.	# tok.	# a	nn.	# en.
EN	MedPost (POS) <sup>a</sup>	6,701	182,319	182,319	(100%)	51
EN	NCBI disease (NER) <sup>a</sup>	7,279	151,005	11,350	(7,5%)	4
FR	MERLoT medical (NER) <sup>b</sup>	5,137	123,942	56,680	(46%)	19
FR	MERLoT PHId (NER)b	25,599	177,158	31,723	(18%)	11
EN	CoNLL 2003 (NER) <sup>c</sup>	18,451	256,145	42,646	(17%)	4

<sup>&</sup>lt;sup>a</sup> MEDLINE abstracts

<sup>&</sup>lt;sup>b</sup> Medical reports

<sup>&</sup>lt;sup>c</sup> News articles

<sup>&</sup>lt;sup>d</sup> Protected Health Information

## Experiments - Hyperparameters

- Pre-trained word embeddings
  - ► CoNLL 2003: Google News
  - NCBI disease and MedPost: PubMed corpus
  - ► MERLoT: 138,000 clinical reports in French
- ▶ Other hyperparameters selected with hyperopt (Bergstra et al. 2013) on the MERLoT de-identification dataset
  - $\rightarrow$  Reused for other corpora
- $\blacktriangleright$  Robustness to the random seed  $\rightarrow$  30 runs (Reimers and Gurevych 2017)

## Experiments - Corpus Specific Results

Dataset	Model	F1
NCBI	Dang et al. (2018) Islamaj Dogan and Lu (2012) This paper	<b>84.41</b> 81.80 81.33
MERLoT med.*	Campillos et al. (2018) This paper	81.40 <b>82.87</b>
CoNLL 2003	Lample et al. (2016) Ma and Hovy (2016) Yang and Zhang (2018) Peters et al. (2018) This paper	90.94 91.21 91.35 <b>92.22</b> 87.31
MedPost*	Smith et al. (2004) This paper	97.43 <b>97.83</b>
MERLoT PHI	Grouin and Névéol (2014) This paper	94.00 <b>99.40</b>

<sup>\*</sup> Corpora where the experimental set-up differed between our experiments and that of prior work. For MEDPOST, we used 51 categories instead of 63; for MERLoT med. we removed nested entities.

## Experiments - Score Variability Across Runs

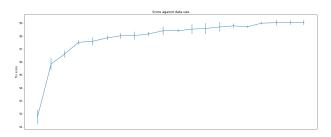
Task	Dataset	Mean F1	$\sigma$
NER	NCBI disease	81.33	0.83
NER	MERLoT medical	82.87	0.36
NER	CoNLL 2003	87.31	0.41
POS	MedPost	97.83	0.39
NER	MERLoT PHI	99.40	0.14

Performance of YASET on the 4 datasets.

## Experiments - Training Data Size

Dataset	5%	10%	25%	50%	100%
MERLoT PHI	91.79	95.86	97.60	98.41	99.06
	(0.55)	(0.47)	(0.30)	(0.33)	(0.19)

Performance (F1) against train set size (as percentage of the total training set indicated on the first row). Standard errors appear between parentheses.



These scores are computed over 6 training iterations per chunk. Vertical bars show the standard deviation  $\sigma$ .

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### Conclusion & Future Work

#### This paper

- Yet Another Sequence Tagger (YASET): accurate implementation of the LSTM-CRF model
- Hyperparameters optimized on one corpus can be transferred to others
- ▶ Performance improvement is logarithmic with the training data size

#### **Future directions**

- ► Nested entity model implementation (Campillos et al. 2018; Ju et al. 2018; Katiyar and Cardie 2018)
- ▶ Other model implementation (Ma and Hovy 2016; Rei and Yannakoudakis 2016)
- ► Elmo embeddings (Peters et al. 2018)

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julien.tourille@limsi.fr

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