

Deep Learning for Natural Language Processing

NLP Tasks with Sequence Outputs



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AUTONOMOUS SYSTEMS
AND SOFTWARE PROGRAM

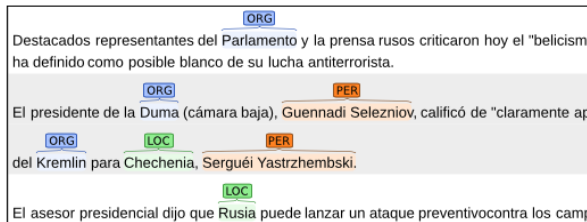
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this lecture block

- ▶ first instance of **structured outputs**: sequences
- ▶ now: quick overview with applications for some different sequence-related NLP tasks:
 - ▶ **sequence labeling** tasks
 - ▶ **segmentation** tasks
 - ▶ **bracketing** tasks
- ▶ later: the implementation of models for these tasks

Astronomers announce the detection of water in the atmosphere
NNS VBP DT NN IN NN IN DT NN



sequence labeling tasks

- ▶ in **sequence labeling** or **sequence tagging**, we are given a sequential input and our goal is to predict an output sequence of the same length
- ▶ in NLP, the input is most often a sequence of words

$$w_1 \quad w_2 \quad w_3 \quad \cdot \cdot \cdot \quad w_n$$

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$$\begin{array}{ccccccc} w_1 & w_2 & w_3 & \cdot & \cdot & \cdot & w_n \\ \downarrow & \downarrow & \downarrow & & & & \downarrow \\ l_1 & l_2 & l_3 & & & & l_n \end{array}$$

application example: part-of-speech tagging

- ▶ in part-of-speech tagging, we are given a sequence of words and want to assign **part-of-speech tags** (grammatical categories) to each word

She plays in many plays .

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






































- ▶ the most widely studied sequence labeling task
- ▶ used to be very important in applications

part-of-speech tagging: datasets

- ▶ there are **many** dataset for PoS tagging!
- ▶ PoS datasets for many languages, using a standardized set of tags: <https://universaldependencies.org/>

Current UD Languages

Information about language families (and genera for families with multiple branches) is mostly taken from [WALS Online](https://wals.info/) (IE = Indo-European).

▶		Afrikaans	1	49K		IE, Germanic
▶		Akkadian	1	1K		Afro-Asiatic, Semitic
▶		Albanian	1	<1K		IE, Albanian
▶		Amharic	1	10K		Afro-Asiatic, Semitic
▶		Ancient Greek	2	416K		IE, Greek
▶		Arabic	3	1,042K		Afro-Asiatic, Semitic
▶		Armenian	1	52K		IE, Armenian
▶		Assyrian	1	<1K		Afro-Asiatic, Semitic
▶		Bambara	1	13K		Mande
▶		Basque	1	121K		Basque
▶		Belarusian	1	13K		IE, Slavic
▶		Bhojpuri	2	4K		IE, Indic
▶		Breton	1	10K		IE, Celtic
▶		Bulgarian	1	156K		IE, Slavic
▶		Buryat	1	10K		Mongolic
▶		Cantonese	1	13K		Sino-Tibetan
▶		Catalan	1	531K		IE, Romance
▶		Chinese	5	285K		Sino-Tibetan
▶		Classical Chinese	1	74K		Sino-Tibetan
▶		Coptic	1	40K		Afro-Asiatic, Egyptian
▶		Croatian	1	199K		IE, Slavic

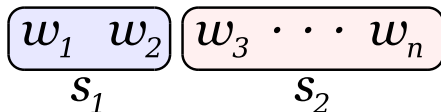
segmentation tasks

- ▶ in **segmentation** tasks, our goal is to split the text into pieces
- ▶ and optionally also assign labels to the pieces

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application example: word segmentation

- ▶ in scripts where whitespace is not used between words, **word segmentation** is nontrivial
- ▶ example by Wang and Xu (2017):

我 有 一 台 计 算 机 。

(I) (have) (a) (computer) (.)

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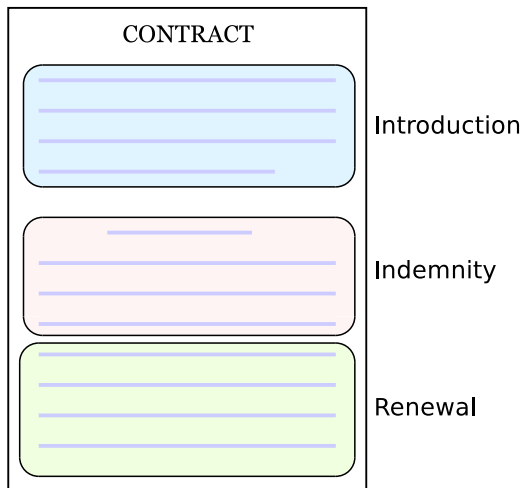
application example: contract clauses

- ▶ inspired by Jansson (2019)

[illegible]

application example: contract clauses

- ▶ inspired by [Jansson \(2019\)](#)



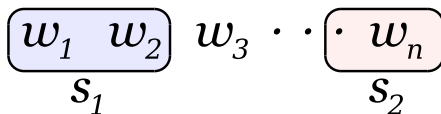
bracketing or markup tasks

- ▶ in **bracketing** or **markup** tasks, the goal is to select and categorize some pieces of text (**spans**)

$$w_1 \quad w_2 \quad w_3 \quad \cdot \cdot \cdot \quad w_n$$

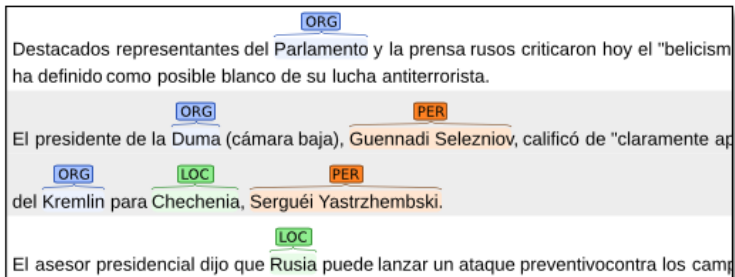
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named entity recognition

- ▶ the canonical example of a bracketing task is probably **named entity recognition**



other examples of bracketing tasks

- ▶ aspect-based sentiment analysis

Their spring rolls

Target

are

absolutely delicious

Positive

semantic role labeling and “slot-filling”

- ▶ given a sentence and a target word, find the “slot-fillers”
- ▶ general tasks such as **semantic role labeling** and domain-specific **slot-filling information extraction** systems

AB Volvo

Seller

sold

Event

Volvo Cars

Goods

to

Ford

Buyer

in

1999

Time

reducing segmentation and bracketing to sequence labeling

- ▶ in practice, systems for segmentation and bracketing are often implemented as sequence labelers

The cases of **metastatic cancer** of the **gall bladder**
Pathology Organ

reducing segmentation and bracketing to sequence labeling

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The	cases	of	metastatic cancer	of	the	gall	bladder
			Pathology			Organ	
0	0	0	B-PAT	I-PAT	0	0	B-ORG I-ORG

reducing segmentation and bracketing to sequence labeling

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The	cases	of	metastatic cancer	of	the	gall	bladder
			Pathology			Organ	
O	O	O	B-PAT	I-PAT	O	O	B-ORG I-ORG

- ▶ **BIO coding** (a.k.a. IOB2) (Tjong Kim Sang and Veenstra, 1999) uses **B**eginning, **I**nside and **O**utside tags

Chinese word segmentation again

- ▶ example by Wang and Xu (2017):

我	有	一	台	计	算	机	。
(I)	(have)	(a)		(computer)			(.)

Chinese word segmentation again

- ▶ example by Wang and Xu (2017):

S	S	B	E	B	M	E	S
我	有	一	台	计	算	机	。
(I)	(have)	(a)		(computer)			(.)

evaluating sequence labelers

- ▶ sequence labeling tasks: usually **accuracy** at the token level

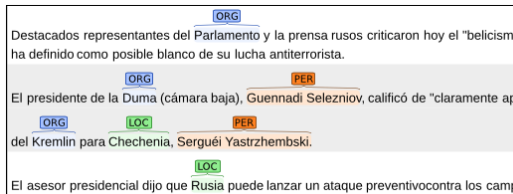
$$\text{accuracy} = \frac{\# \text{ correct tags}}{\# \text{ tokens}}$$

Astronomers	announce	the	detection	of	water	in	the	atmosphere
NNS	VBP	DT	NN	IN	NN	IN	DT	NN

evaluating bracketing and segmentation systems

- segmentation and bracketing: usually **precision** and **recall**

$$P = \frac{\# \text{ correct spans}}{\# \text{ proposed spans}} \quad R = \frac{\# \text{ correct spans}}{\# \text{ reference spans}}$$

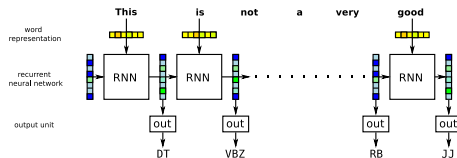


Overall: P = 0.8755, R = 0.8837, F1 = 0.8796

LOC: P = 0.9161, R = 0.9336, F1 = 0.9248

PER: P = 0.9216, R = 0.9387, F1 = 0.9301

plan: sequence labeling and friends



- ▶ **this block:**
 - ▶ simple neural models for sequence labeling
 - ▶ improving representations
- ▶ **next block:** more sophisticated learning approaches

references

- E. Jansson. 2019. [Domain adapted language models](#). Master's Thesis, Chalmers University of Technology.
- E. Tjong Kim Sang and J. Veenstra. 1999. [Representing text chunks](#). In *EACL*.
- C. Wang and B. Xu. 2017. [Convolutional neural network with word embeddings for Chinese word segmentation](#). In *IJCNLP*. pages 163–172.