GDR LIFT - ILFC Seminar

Interactions between formal and computational linguistics 1st June 2021

What do neural models tell us about the nature of language?

Juan Luis Gastaldi ETH Zürich (D-GESS, Turing Centre) juan.luis.gastaldi@gess.ethz.ch

Turing Centre





There is no "philosophy" of language. There is only linguistics.

Louis Hjelmslev

Principes de Grammaire Générale, 1928



Source: Bengio et al., 2003

WORD2VEC



Mikolov et al., 2013

WORD EMBEDDINGS



a cat catches a mouse

Source: Ferrone et al., 2017

WORD EMBEDDINGS



$$v_{house} - v_{city} + v_{countryside} \approx v_{farmhouse}$$

$$v_{king} - v_{man} + v_{woman} \approx v_{queen}$$
$$v_{king} - v_{queen} \approx v_{man} - v_{woman}$$



Word Embeddings



Word Embeddings



Type of relationship	Word	Pair 1	Wor	rd Pair 2	
Common capital city	Athens Greece		Oslo	Norway	
All capital cities	Astana	Kazakhstan	Harare	Zimbabwe	
Currency	Angola	kwanza	Iran	rial	
City-in-state	Chicago	Illinois	Stockton	California	
Man-Woman	brother sister		grandson	granddaughter	
Adjective to adverb	apparent	apparently	rapid	rapidly	
Opposite	possibly	impossibly	ethical	unethical	
Comparative	great	greater	tough	tougher	
Superlative	easy	easiest	lucky	luckiest	
Present Participle	think	thinking	read	reading	
Nationality adjective	Switzerland	Swiss	Cambodia	Cambodian	
Past tense	walking	walked	swimming	swam	
Plural nouns	mouse	mice	dollar	dollars	
Plural verbs	work	works	speak	speaks	

Source: Mikolov et al., 2013.

- 1. The automatic reconstruction of the underlying organization of language does not require more human intervention than the one implied in the most ordinary use of language as recorded in a practically raw linguistic corpus.
- 2. In that reconstruction, both semantic and syntactic contents of words are determined at once and as the result of the same procedure.
- 3. Word vector representations are not simply disposed in similarity neighbourhoods, but that the vector space itself is also structured following precise directions at the crossroads of which syntactic and semantic contents are established.

- Word Embeddings (word2vec, GloVe)
- Recurrent Neural Nets (LSTM)
- Encoding-Decoding (seq2seq)
- Transformers (attention, GPT-3)

THE SEMANTICS MEGA-THREAD



https://medium.com/huggingface/learning-meaning-in-natural-language-processing-the-semantics-mega-thread-9c0332dfe28e

THE SEMANTICS MEGA-THREAD



...



Emily M. Bender @emilymbender

En réponse à @jacobandreas et @_shrdlu_

I think that depends on what you mean by LM. If it's just: here's a bunch of sequences of words, learn to guess which word comes next this. is. no. access. to. meaning.

Traduire le Tweet

01:29 · 30/07/2018 · TweetDeck

Bender, E. M., & Koller, A. (2020). Climbing towards NLU: On Meaning, Form, and Understanding in the Age of Data. *Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics*. https://doi.org/10.18653/v1/2020.acl-main.463

- "You shall know a word by the company it keeps!" (Firth, 1957)
- "Words which are similar in meaning occur in similar contexts" (Rubenstein & Goodenough 1965)
- "Words with similar meanings will occur with similar neighbors if enough text material is available" (Schütze & Pedersen 1995)
- "A representation that captures much of how words are used in natural context will capture much of what we mean by meaning" (Landauer & Dumais 1997)
- "Words that occur in the same contexts tend to have similar meanings" (Pantel 2005)
- "The degree of semantic similarity between two linguistic expressions A and B is a function of the similarity of the linguistic contexts in which A and B can appear" (Lenci, 2010)

- Theory of (linguistic) meaning as "usage" (Wittgenstein)
 - "the meaning of a word is defined by the circumstances of its use" (Manning & Schütze, 1999)
- Two versions of the DH:
 - Weak: Correlation between context and word meaning (Spence & Owens, 1990)
 - Strong: Causality attributed to contextual distributions (Miller & Charles, 1991)
- Context: the domain or scope within which entities of the same nature can be
 presented together ("co-occur"), in such a way that they can be associated by a
 cognitive agent.

Word Embeddings as Matrix Factorization



Source: topicmodels.west.uni-koblenz.de

Word Embeddings as Matrix Factorization



topicmodels.west.uni-koblenz.de

	 W	Х	У	Z	
	 \bigotimes	\bigcirc	Ø	\bigotimes	
а	 \bigcirc	1	1	\bigcirc	
b	 0	0	1	1	
С		\bigcirc	0		
	 \bigcirc	\bigcirc	0	\bigcirc	

Context	Term	Term	Context
(w,x)	(a,c)	(a,c)	(w,x,y,z)
(w,y)	(a,b,c)	(a,b,c)	(w,x, y,z)
(w,z)	(b, c)	(b,c)	(w,y, z)
(x,y)	(a ,b)	(a,b)	(x, y ,z)
(y,z)	(a, b ,c)	(a,b,c)	(w,x, y,z)

$$\begin{array}{ll} a = your & w = apartment \\ c = my & x = house \\ & y = chair \\ & z = stool \end{array}$$

your : house my : apartment

		W	Х	У	Z	
		\bigcirc	\bigcirc	\bigcirc	\bigcirc	
а		\bigcirc	1	1	\bigcirc	
b		\bigcirc	\bigcirc			
С		1	\bigcirc	\bigcirc	1	
	•••	\bigcirc	\bigcirc	\bigcirc	\bigcirc	

"The day she came to your house in Paris"

"She bought a house and a bungalow"

"Vous vous trompez"

• Saussure, Hjelmslev, Harris...



 $\sqrt{}$

~		Environments										
Seg- ments	#— ŗ	# — r	#]	$\left \begin{array}{c} \mathbf{e} \\ \mathbf{i} \end{array} \right - \mathbf{C}$	æ–C	a o – C u	s– e	s-æ	s— o u		t –	C^3-
ţ	\checkmark											
t		\checkmark		\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark			
K						\checkmark			\checkmark			
k		\checkmark	\checkmark		\checkmark			\checkmark				
К				\checkmark			\checkmark					
G						\checkmark						
g		\checkmark	\checkmark		\checkmark							
G				\checkmark								
r				\checkmark	\checkmark	\checkmark						\checkmark
ŗ											\checkmark	

1951)

- F. de Saussure
- R. Jakobson
- L. Hjelmslev
- L. Bloomfield
- Z. Harris

- Linguistic content (including essential aspects of linguistic meaning) is the effect of a virtual structure of classes and dependencies at multiple levels, underlying (and derivable from) the mass of things said or written in a given language.
- The task of linguistic analysis is not just that of identifying loose similarities between words out of distributional properties of a corpus, but rather this other one of explicitly drawing from that corpus the system of strict dependencies between implicit linguistic categories.

A priori it would seem to be a generally valid thesis that for every process there is a corresponding system, by which the process can be analyzed and described by means of a limited number of premises. It must be assumed that any process, can be analyzed into a limited number of elements recurring in various combinations. Then, on the basis of this analysis, it should be possible to order these elements into classes according to their possibilities of combination. And it should be further possible to set up a general and exhaustive calculus of the possible combinations.

Hjelmslev (1943)













Source: Wasserblat, M., Pereg, O., & Izsak, P. (2020)



	1	· . · ·	2	1. C. S.	3		4	
e la ste ● ● ●	the		bov		came	-	home	
	one	•	girl	Å	went		down	
	no		man		ran		1n)





the girl ran home

Procedure Step	Neural	Non-Neural
Segmentation	Sub-word Tokenization Character level DNNs	BPE
Classification	Word Embedding Distillation	Matrix models
Dependencies	Attention	Biorthogonal Typing Linear Logic

- Generic model
- Non-necessarily neural
- Producing explicit representations
- Supporting logical relations
- Capturing significant aspects of meaning

J. L. Gastaldi. "Why Can Computers Understand Natural Language?". In: *Philosophy & Technology* 34.1 (2021), pp. 149–214. DOI: 10.1007/s13347-020-00393-9 Link

J. L. Gastaldi and Luc Pellissier: **"The Calculus of Language: Explicit Representation of Emergent Linguistic Structure Through Type-Theoretical Paradigms"** In: *Interdisciplinary Science Reviews* (2021), DOI: 10.1080/03080188.2021.1890484

- Baroni, Marco, Georgiana Dinu, and Germán Kruszewski (2014). "Don't count, predict! A systematic comparison of context-counting vs. context-predicting semantic vectors". In: Proceedings of the 52nd Annual Meeting of the Association for Computational Linguistics (Volume 1: Long Papers). Baltimore, Maryland: Association for Computational Linguistics, pp. 238–247.
- Bender, Emily M. and Alexander Koller (July 2020). "Climbing towards NLU: On Meaning, Form, and Understanding in the Age of Data". In: Proceedings of the 58th Annual Meeting of the Association for Computational Linguistics. Online: Association for Computational Linguistics, pp. 5185–5198.
- Bengio, Yoshua et al. (Mar. 2003). "A Neural Probabilistic Language Model". In: J. Mach. Learn. Res. 3, pp. 1137–1155. ISSN: 1532-4435.
- Gastaldi, Juan Luis (May 2020). "Why Can Computers Understand Natural Language?: The Structuralist Image of Language Behind Word Embeddings". In: Philosophy & Technology.
- Gastaldi, Juan Luis and Luc Pellissier (2021). "The Calculus of Language: Explicit Representation of Emergent Linguistic Structure Through Type-Theoretical Paradigms". In: Interdisciplinary Science Reviews.

- Hamilton, William L., Jure Leskovec, and Dan Jurafsky (2016). "Diachronic Word Embeddings Reveal Statistical Laws of Semantic Change". In: CoRR abs/1605.09096.
- Harris, Zellig (1960). Structural linguistics. Chicago: University of Chicago Press. ISBN: 0226317714 0226217714.
- Hewitt, John and Christopher D. Manning (June 2019). "A Structural Probe for Finding Syntax in Word Representations". In: Proceedings of the 2019 Conference of the North American Chapter of the Association for Computational Linguistics: Human Language Technologies, Volume 1 (Long and Short Papers). Minneapolis, Minnesota: Association for Computational Linguistics, pp. 4129–4138.
- Hjelmslev, Louis (1935). La catégorie des cas. Munchen: Wilhelm Fink Verlag.
- — (1953). Prolegomena to a Theory of Language. Baltimore: Wawerly Press.
- (1975). Résumé of a Theory of Language. Travaux du Cercle linguistique de Copenhague 16. Copenhagen: Nordisk Sprog-og Kulturforlag. ISBN: 0-299-07040-9.
- Lenci, Alessandro (2008). "Distributional semantics in linguistic and cognitive research". In: From context to meaning: distributional models of the lexicon in linguistics and cognitive science, Italian Journal of Linguistics 1.20. Ed. by Alessandro Lenci, pp. 1–31.

- Levy, Omer and Yoav Goldberg (2014a). "Linguistic Regularities in Sparse and Explicit Word Representations". In: Proceedings of the Eighteenth Conference on Computational Natural Language Learning, CoNLL 2014, Baltimore, Maryland, USA, June 26-27, 2014, pp. 171–180.
- — (2014b). "Neural Word Embedding As Implicit Matrix Factorization". In: Proceedings
 of the 27th International Conference on Neural Information Processing Systems Volume
 2. NIPS'14. Montreal, Canada: MIT Press, pp. 2177–2185.
- Manning, Christopher D. and Hinrich Schutze (1999). Foundations of Statistical Natural Language Processing. Cambridge, MA, USA: MIT Press. ISBN: 0-262-13360-1.
- Mikolov, Tomáš, Kai Chen, et al. (2013). "Efficient Estimation of Word Representations in Vector Space". In: CoRR abs/1301.3781.
- Mikolov, Tomáš, Quoc V. Le, and Ilya Sutskever (2013). "Exploiting Similarities among Languages for Machine Translation". In: CoRR abs/1309.4168.
- Mikolov, Tomáš, Ilya Sutskever, et al. (2013). "Distributed Representations of Words and Phrases and their Compositionality". In: CoRR abs/1310.4546.

- Mikolov, Tomáš, Wen-tau Yih, and Geoffrey Zweig (2013). "Linguistic Regularities in Continuous Space Word Representations". In: Proceedings of the 2013 Conference of the North American Chapter of the ACL: Human Language Technologies. Atlanta, Georgia: ACL, pp. 746–751.
- Sahlgren, Magnus (2008). "The Distributional Hypothesis". In: Special issue of the Italian Journal of Linguistics 1.20. Ed. by Alessandro Lenci, pp. 33–53.
- Saussure, Ferdinand de (1959). Course in General Linguistics. Translated by Wade Baskin. New York: McGraw-Hill.

GDR LIFT - ILFC Seminar

Interactions between formal and computational linguistics 1st June 2021

What do neural models tell us about the nature of language?

Juan Luis Gastaldi ETH Zürich (D-GESS, Turing Centre) juan.luis.gastaldi@gess.ethz.ch

Turing Centre



