	doc_1		doc_2		decision	id
cases			authors	Alessandro Lenci Magnus Sahlgren Patrick Jeuniaux Amaru Cuba Gyllensten Martina Miliani		
	authors	 Alessandro Lenci Magnus Sahlgren Patrick Jeuniaux Amaru Cuba Gyllensten Martina Miliani 	title	A comprehensive comparative evaluation and analysis of Distributional Semantic Models		
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	title	A comparative evaluation and analysis of three generations of Distributional Semantic Models	doi			
	publication_date 2021-05-20 00:00:00		urls	• https://www.semanticscholar.org/paper/44ba2fade1d9369bb1643094a70be7052aae4fba		
	source	SupportedSources.SEMANTIC_SCHOLAR	id	id-2126473246117565045	NOT DUPLICATES 402	402
	journal	Language Resources and Evaluation	abstract	Distributional semantics has deeply changed in the last decades. First, predict models stole the thunder from		
	volume	56		traditional count ones, and more recently both of them were replaced in many NLP applications by contextualized		
	doi	10.1007/s10579-021-09575-z		vectors produced by Transformer neural language models. Although an extensive body of research has been	e y	
	urls	https://www.semanticscholar.org/paper/98886ae28d00666a5ce5a216cd0137332bfe3eda		devoted to Distributional Semantic Model (DSM) evaluation, we still lack a thorough comparison with respect to tested models, semantic tasks, and benchmark datasets. Moreover, previous work has mostly focused on task-driven evaluation, instead of exploring the differences between the way models represent the lexical semantic		
	id	id6347277659586517241		space. In this paper, we perform a comprehensive evaluation of type distributional vectors, either produced by static DSMs or obtained by averaging the contextualized vectors generated by BERT. First of all, we investigate the performance of embeddings in several semantic tasks, carrying out an in-depth statistical analysis to identify the major factors influencing the behavior of DSMs. The results show that i.) the alleged superiority of predict		
	abstract	None				
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				based models is more apparent than real, and surely not ubiquitous and ii.) static DSMs surpass contextualized representations in most out-of-context semantic tasks and datasets. Furthermore, we borrow from cognitive neuroscience the methodology of Representational Similarity Analysis (RSA) to inspect the semantic spaces generated by distributional models. RSA reveals important differences related to the frequency and part-of-speech of lexical items.		
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