

cases	doc_1		doc_2				decision	id
			authors	<ul style="list-style-type: none"><li>Junichi Tsujii</li><li>Yuki Arase</li></ul>			DUPLICATES	321
			title	Transfer Fine-Tuning: A BERT Case Study				
			publication_date	2019-09-03 00:00:00				
	authors	<ul style="list-style-type: none"><li>Yuki Arase</li><li>Jun'ichi Tsujii</li></ul>	source	SupportedSources.PAPERS_WITH_CODE				
	title	Transfer Fine-Tuning: A BERT Case Study	journal					
	publication_date	2019-09-01 00:00:00	volume					
	source	SupportedSources.OPENALEX	doi					
	journal	arXiv (Cornell University)	urls	<ul style="list-style-type: none"><li>https://arxiv.org/pdf/1909.00931v1.pdf</li><li>https://github.com/yukiar/TransferFT</li><li>https://aclanthology.org/D19-1542.pdf</li></ul>				
	volume		id	id-8731650874298029426				
	doi	10.18653/v1/d19-1542	abstract	A semantic equivalence assessment is defined as a task that assesses semantic equivalence in a sentence pair by binary judgment (i.e., paraphrase identification) or grading (i.e., semantic textual similarity measurement). It constitutes a set of tasks crucial for research on natural language understanding. Recently, BERT realized a breakthrough in sentence representation learning (Devlin et al., 2019), which is broadly transferable to various NLP tasks. While BERT's performance improves by increasing its model size, the required computational power is an obstacle preventing practical applications from adopting the technology. Herein, we propose to inject phrasal paraphrase relations into BERT in order to generate suitable representations for semantic equivalence assessment instead of increasing the model size. Experiments on standard natural language understanding tasks confirm that our method effectively improves a smaller BERT model while maintaining the model size. The generated model exhibits superior performance compared to a larger BERT model on semantic equivalence assessment tasks. Furthermore, it achieves larger performance gains on tasks with limited training datasets for fine-tuning, which is a property desirable for transfer learning.				
	urls	<ul style="list-style-type: none"><li>https://openalex.org/W2988707515</li><li>https://doi.org/10.18653/v1/d19-1542</li><li>https://www.aclweb.org/anthology/D19-1542.pdf</li></ul>	versions					
	id	id257162581231611739						
	abstract							
	versions							