

cases	doc_1		doc_2				decision	id
	<div><div>authors<ul style="list-style-type: none">John WietingKevin GimpelGraham NeubigTaylor Berg-Kirkpatrick</div><div>titleParaphrastic Representations at Scale</div><div>publication_date2021-04-30 00:00:00</div><div>sourceSupportedSources.OPENALEX</div><div>journalarXiv (Cornell University)</div><div>volume</div><div>doiNone</div><div>urls<ul style="list-style-type: none">https://openalex.org/W3158209167</div><div>idid8140257015640013210</div><div>abstract</div><div>versions</div></div> <td><div>authors<ul style="list-style-type: none">John WietingKevin GimpelGraham NeubigTaylor Berg-Kirkpatrick</div><div>titleParaphrastic Representations at Scale</div><div>publication_date2021-04-30 00:00:00</div><div>sourceSupportedSources.INTERNET_ARCHIVE</div><div>journal</div><div>volume</div><div>doi</div><div>urls<ul style="list-style-type: none">https://web.archive.org/web/20210504041236/https://arxiv.org/pdf/2104.15114v1.pdf</div><div>idid1489107754186921318</div><td><div>abstract<p>We present a system that allows users to train their own state-of-the-art paraphrastic sentence representations in a variety of languages. 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