	doc_1		doc_2		decision	id
cases	authors	<ul> <li>Nihal Jain</li> <li>Dejiao Zhang</li> <li>Wasi Uddin Ahmad</li> <li>Zijian Wang</li> <li>Feng Nan</li> <li>Xiaopeng Li</li> <li>Ming Tan</li> <li>Ramesh Nallapati</li> <li>Baishakhi Ray</li> <li>Parminder Bhatia</li> <li>Xiaofei Ma</li> <li>Bing Xiang</li> </ul>	authors	<ul> <li>Nihal Jain</li> <li>Dejiao Zhang</li> <li>Wasi Uddin Ahmad</li> <li>Zijian Wang</li> <li>Feng Nan</li> <li>Xiaopeng Li</li> <li>M. Tan</li> <li>Ramesh Nallapati</li> <li>Baishakhi Ray</li> <li>Parminder Bhatia</li> <li>Xiaofei Ma</li> </ul>		
	title	ContraGen: Effective Contrastive Learning For Causal Language Model		Bing Xiang		
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		• http://arxiv.org/pdf/2210.01185v1 • http://arxiv.org/pdf/2210.01185v1	urls	https://www.semanticscholar.org/paper/0b8772b7790c69f40897b5eb7f8fd57f24138f3d		
	id	id-4614918937080640725	id	id-5145387860803530201		
	abstract	Despite exciting progress in large-scale language generation, the expressiveness of its representations is severely limited by the \textit{anisotropy} issue where the hidden representations are distributed into a narrow cone in the vector space. To address this issue, we present ContraGen, a novel contrastive learning framework to improve the representation with better uniformity and discrimination. We assess ContraGen on a wide range of downstream tasks in natural and programming languages. We show that ContraGen can effectively enhance both uniformity and discrimination of the representations and lead to the desired improvement on various language understanding tasks where discriminative representations are crucial for attaining good performance. Specifically, we attain \$44\%\$ relative improvement on the Semantic Textual Similarity tasks and \$34\%\$ on Code-to-Code Search tasks. Furthermore, by improving the expressiveness of the representations, ContraGen also boosts the source code generation capability with \$9\%\$ relative improvement on execution accuracy on the HumanEval benchmark.	abstract	Despite exciting progress in large-scale language generation, the expressiveness of its representations is severely limited by the anisotropy issue where the hidden representations are distributed into a narrow cone in the vector space. To address this issue, we present C ONTRA G EN , a novel contrastive learning framework to improve the representation with better uniformity and discrimination. We assess C ONTRA G EN on a wide range of downstream tasks in natural and programming languages. We show that C ONTRA G EN can effectively enhance both uniformity and discrimination of the representations and lead to the desired improvement on various language understanding tasks where discriminative representations are crucial for attaining good performance. Speciï¬cally, we attain 44% relative improvement on the Semantic Textual Similarity tasks and 34% on Code-to-Code Search tasks. Furthermore, by improving the expressiveness of the representations, C ONTRA G EN also boosts the source code generation capability with 9% relative improvement on execution accuracy on the HumanEval benchmark.		
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