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			authors	Guanbin Li     Yizhou Yu		
			title	Contrast-Oriented Deep Neural Networks for Salient Object Detection	]	
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		• Li, G.	source	SupportedSources.ARXIV	]	
	authors	• Yu, Y.	journal	None		
	title	Contrast-Oriented Deep Neural Networks for Salient Object Detection	volume		]	
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cases	source	SupportedSources.CROSSREF	urls	<ul> <li>http://arxiv.org/pdf/1803.11395v1</li> <li>http://arxiv.org/abs/1803.11395v1</li> </ul>	apping nse ts of S, e. We pilored prove	
	journal			• http://arxiv.org/pdf/1803.11395v1		
	volume			International Parameters (International Para		ES 170
	doi	10.1109/tnnls.2018.2817540	id	id-1575356177361262848		
	urls	<ul> <li>http://xplorestaging.ieee.org/ielx7/5962385/8540478/08337098.pdf? arnumber=8337098</li> <li>http://dx.doi.org/10.1109/tnnls.2018.2817540</li> </ul>	abstract	Deep convolutional neural networks have become a key element in the recent breakthrough of salient object detection. However, existing CNN-based methods are based on either patch-wise (region-wise) training and inference or fully convolutional networks. Methods in the former category are generally time-consuming due to severe storage and computational redundancies among overlapping patches. To overcome this deficiency, methods in the second category attempt to directly map a raw input image to a predicted dense saliency map in a single network forward pass. Though being very efficient, it is arduous for these methods to detect salient objects of		
	id	id-2526994083005545234		different scales or salient regions with weak semantic information. In this paper, we develop hybrid contrast-oriented deep neural networks to overcome the aforementioned limitations. Each of our deep networks is composed of two complementary components, including a fully convolutional stream for dense prediction and a segment-level spatial pooling stream for sparse saliency inference. We further propose an attentional module that learns weight maps for fusing the two saliency predictions from these two streams. A tailored alternate scheme is designed to train these deep networks by fine-tuning pre-trained baseline models. Finally, a customized fully connected CRF model incorporating a salient contour feature embedding can be optionally applied as a post-processing step to improve spatial coherence and contour positioning in the fused result from these two streams. Extensive experiments on six benchmark datasets demonstrate that our proposed model can significantly outperform the state of the art in terms of all popular evaluation metrics.		
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