

cases	doc_1		doc_2		decision	id
	authors	<ul style="list-style-type: none">Alan B. Cannaday IICurt H. DavisGrant J. ScottBlake RuprechtDerek T. Anderson	authors	<ul style="list-style-type: none">Alan B. Cannaday IIC. DavisG. ScottBlake RuprechtDerek T. Anderson	DUPLICATES	141
	title	Broad Area Search and Detection of Surface-to-Air Missile Sites Using Spatial Fusion of Component Object Detections from Deep Neural Networks	title	Broad Area Search and Detection of Surface-to-Air Missile Sites Using Spatial Fusion of Component Object Detections From Deep Neural Networks		
	publication_date	2020-03-23 22:10:19+00:00	publication_date	2020-03-23 00:00:00		
	source	SupportedSources.ARXIV	source	SupportedSources.SEMANTIC_SCHOLAR		
	journal	None	journal	IEEE Journal of Selected Topics in Applied Earth Observations and Remote Sensing		
	volume		volume	13		
	doi		doi	10.1109/JSTARS.2020.3015662		
	urls	<ul style="list-style-type: none">http://arxiv.org/pdf/2003.10566v3http://arxiv.org/abs/2003.10566v3http://arxiv.org/pdf/2003.10566v3	urls	<ul style="list-style-type: none">https://www.semanticscholar.org/paper/8ee62eb0a7e55fb2ed3d2143cd567669795090fd		
	id	id-803600172549298944	id	id8567324706401912176		
	abstract	Here we demonstrate how Deep Neural Network (DNN) detections of multiple constitutive or component objects that are part of a larger, more complex, and encompassing feature can be spatially fused to improve the search, detection, and retrieval (ranking) of the larger complex feature. First, scores computed from a spatial clustering algorithm are normalized to a reference space so that they are independent of image resolution and DNN input chip size. Then, multi-scale DNN detections from various component objects are fused to improve the detection and retrieval of DNN detections of a larger complex feature. We demonstrate the utility of this approach for broad area search and detection of Surface-to-Air Missile (SAM) sites that have a very low occurrence rate (only 16 sites) over a ~90,000 km^2 study area in SE China. The results demonstrate that spatial fusion of multi-scale component-object DNN detections can reduce the detection error rate of SAM Sites by $>85\%$ while still maintaining a 100% recall. The novel spatial fusion approach demonstrated here can be easily extended to a wide variety of other challenging object search and detection problems in large-scale remote sensing image datasets.	abstract	Here, we demonstrate how deep neural network (DNN) detections of multiple constitutive or component objects that are part of a larger, more complex, and encompassing feature can be spatially fused to improve the search, detection, and retrieval (ranking) of the larger complex feature. First, scores computed from a spatial clustering algorithm are normalized to a reference space so that they are independent of image resolution and DNN input chip size. Then, multiscale DNN detections from various component objects are fused to improve the detection and retrieval of DNN detections of a larger complex feature. We demonstrate the utility of this approach for broad area search and detection of surface-to-air missile (SAM) sites that have a very low occurrence rate (only 16 sites) over a $\sim 90\text{â€‰}000\text{ km}^2$ study area in SE China. The results demonstrate that spatial fusion of multiscale component-object DNN detections can reduce the detection error rate of <i>SAM Sites</i> by $>85\%$ while still maintaining a 100% recall. The novel spatial fusion approach demonstrated here can be easily extended to a wide variety of other challenging object search and detection problems in large-scale remote sensing image datasets.		
	versions		versions			