

Motivation and Objectives

- actions in our planet.

Landsat 5 RGB Bands and Landcover Training Labels









Deep Learning for Land Cover Classification Diego F. Chamorro¹, Patrick C. Gray² . Department of Electrical and Computer Engineering, Duke University, Durham, North Carolina 2. Nicholas School of the Environment, Duke University, Durham, North Carolina **RCNN Graph Architecture** Current land cover classification algorithms have low accuracy, are rarely run, and lack temporal context. This is a problem because measuring land cover change in our Α Α environment is crucial to understanding the effects of climate change, and human "Strange Blocks" Deep learning algorithms have shown promise in their ability to classify land cover. Our Densely aim is to apply these tools and improve on current land cover classification algorithms. connected LSTM The best performing model is a graph model composed of two portions: A stack of layers Convolutional LSTM layers, and a stack of "Strange Blocks". Both include temporal context, which is necessary due to the intraclass variance that occurs during the year. Model Statu Water 0.99 0.00 Developed -0.00 Forest 0.00 Cultivated -0.02 Barren 0.00 Wetland Wate Resolution (meters) K Keras 30 30 **Tensor**Flow 30 30 Comet Microsoft Azure 30 120* (30) 30 Marine Robotics & Remote Sensing





US						
0.00	0.00	0.00	0.01	0.00	- 0.	8
0.97	0.00	0.00	0.03	0.00		
0.00	0.96	0.02	0.01	0.01	-0.	6
0.02	0.00	0.89	0.02	0.06	- 0.	.4
0.08	0.01	0.03	0.86	0.01		
0.00	0.02	0.02	0.01	0.95	-0.	2
eloped	FOREST	inated .	Barren W	etland		
Predicted label					└───0.	0

Diego Chamorro



