A Deep Learning Approach for Intelligent Thinning of Satellite Data

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Background

- Multi/hyper spectral platforms provide rich datasets
- Data outpaces processing ability, operational cadence
- Can we compress spatial and spectral information and minimize artifacts?
- Deep learning → nonlinear dimensionality reduction

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Is Deep Learning a good fit for this problem?

- Lossy Compression
 - High compression ratios
 - Simpler techniques introduce artifacts
- Multi/hyperspectral Data
 - Nonlinear PCA
 - Spectral and spatial covariances
- Onboard processing
 - SWaP at the edge
 - 2-3 TOPS / W



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MyRadar's Mission & Use Case

- MyRadar
 - Democratizing environmental intelligence in a changing climate
 - GPU data processing + nowcasting + alerts
- HORIS Smallsat Platform
 - Hyperspectral imagers
 - Neuromorphic chipset
 - Demanding duty cycle

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• Faster detection -> alerting





Experimental Setup

- GOES-E Full Disk L2 MCMIP
 - ~5 TB / year
 - 2019 training
 - 2020 test
- Azure NC24rs_v2 VM
 - 4x GPU
 - Virtualized blob FS
 - Tensorflow 2.3
 - Hybrid Sequential/functional API
 - Data generator for streaming data

- 20-100 epochs per run
- Data standardization but no ZCA/PCA sphering
- Compare to Discrete Cosine Transform (JPEG)





Autoencoder

- Fully-convolutional = agnostic to input dimensionality
- Residual (local) skip connections
- Batch normalization
- Dropout





Results – 32/1 compression

Original



NN-32

DCT-32





Results – 32/1 compression



- 1% MAE for NN and DCT
- Gaussian AA filter



Results – 32/1 Compression

NN-32









Extreme Compression?

- Typical max DCT-II compression ratio is 144:1
- DCT MAE at 144:1 = 3%
- What is the compression ratio for NN with the same MAE?
- 2048/1



(The Hydraulic Press Channel)



Results – 3% MAE Max Compression

Original

DCTAA-144



NN-2048



How do we improve the texture?

- Adversarial Training
- FC PatchGAN
- GAN Loss \propto A * L1 loss + (1 discriminator loss), A>>1





EARLY Results – 32/1 Adversarial Training





EARLY Results – 32/1 Adversarial Training



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- Improving spectral power over L1 loss learner
- Data QC issue leading to spurious crosshatching
- Stay tuned!

Conclusions / Next Steps

- Autoencoder achieves larger compression ratios than DCT for comparable MAE
- Facilitates hyperspectral data processing with limited SWaP

- Model optimization
- Transfer learning to HORIS sensor suite
- On-orbit AI compression



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Extra – 3% MAE Max Compression



