

Comparing Effectiveness of Lossy and Lossless Reduction Techniques

BLOSC Compressor

80

100

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700

600

500

400

SA 300

200

100

Motivation/Goals

- Los Alamos National Labs is analyzing an additive manufacturing process
- They are taking pictures of this process to train a Machine Learning Algorithm
- The number of images needed far exceeds the data storage methods warranting data reduction methods
- The goal is to compare different reduction methods based on 3 different metrics
- PSNR (Peak Signal to Noise Ratio)
- Compression Rate
- Compression Ratio

Reduction Methods

- Hybrid Data Sampling
- Samples a percentage of the original data with complete resolution based on "importance"
- Forms a reconstructed version of the data from these saved segments
- SZ Lossy Compressor
 - Given error bounding method and error bound, will compress and decompress the data with slight error
- **BLOSC Lossless Compressor**
- Compresses and decompresses data with no error whatsoever

Important Formulas

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$PSNR = 10 \cdot \log_{10} \left(\right)$	$\left(\frac{MAX_I^2}{MSE}\right)$
Compression Ratio =	$\frac{\text{Uncompressed Size}}{\text{Compressed Size}}$
$Compression \ Rate =$	Uncompressed bits seconds to compress

Software Versions

- Software Versions
- Libpressio 0.79.0
- Python 3.9.2
- OpenMPI 4.0.3
- CUDA 11.5.0



10-

10

Rate

40 Timester 60

Key Takeaways

Importance Based Hybrid Data Sampling was fastest

Hybrid Data Sampling offered relatively low PSNR

SZ could reduce image by 5x with no error

BLOSC compressor could cut image size in half with

20

no error (faster than SZ)



Future Goals

- Finding an efficient way to combine both the Hybrid Data Sampling method with the lossy/lossless compressors
- Could be specific to LANL's dataset or more general
- Utilize the compressor's accuracy with the Hybrid Data Sampling's speed

Acknowledgements

Clemson University is acknowledged for generous allotment of compute time on the Palmetto cluster. This material is based upon work supported by the National Science Foundation under Grant No. SHF-1910197 and SHF-1943114.

Dataset From Los Alamos National Labs LA-UR-21-32202

40 Timestep Compression ratio from different reduction methods 20.0 17.5 9 15.0 4x Reduction Methods 12.5 Higher Hybrid Data Sampling Compression SZ Compressor 10.0 BLOSC Compressor 7 . 5.0 2.5 40 60 100 Timester

60

80

100

PSNR

20