

#### Towards Improving Reverse Time Migration Performance by High-speed Lossy Compression

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#### Outline

- Background
- Motivation and Goals
- Related Works
- Methodology
- Evaluation
- Summary





### Background

Seismic imaging is a procedure for estimating the seismic characteristics of the earth's sub-surface by measuring and computing the acoustic wavefields.

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- Putting vibroseis trucks or air guns and in the ground.
- Generating acoustic wave data by the sensors.
- Collecting acoustic wave data and generating earth's subsurface image.
- **Reverse time migration (RTM)** is a cutting-edge seismic imaging method.
  - A full two-way wave equation.
  - Simulating the forward propagation of seismic waves in the subsurface using the wave equation
  - Using the resulting wavefields from backward propagation to create the subsurface image.
  - Generating high resolution image for complex terrain.

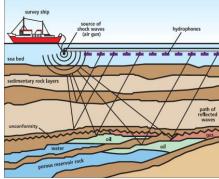
Science Argonne University of Laboratory OF IOWA

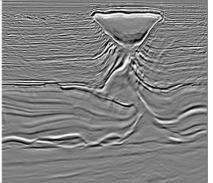
 RTM is widely-used in various domains, such as hydrocarbon exploration, geothermal exploration, and earthquake imaging.

#### Limitation of RTM: Very expensive.

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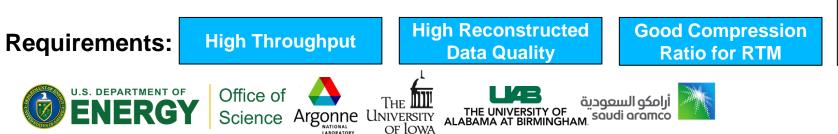


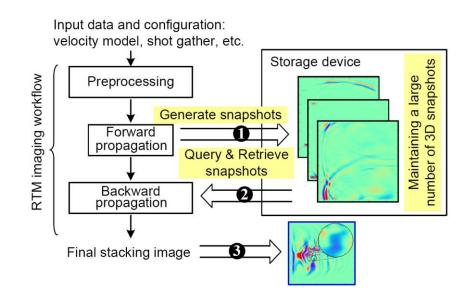
	WEM	Kirchhoff	RTM
Complex velocity focusing?	Yes	No	Yes
Can image steep (>70°) dips?	No	Yes	Yes
Accurate amplitude "out of the box"?	Yes	No	Yes



### Motivation and Goals

- Motivation: Why RTM is so expensive in practice?
  - RTM Workflow:
    - 1. Data preprocessing (input and configuration).
    - 2. Forward propagation -> generating snapshots
    - 3. Backward propagation -> retrieving snapshots
    - 4. Data postprocessing (generating stacking image)
  - To store the extremely large snapshot data:
    - Peripheral devices: significantly degrading performance.
    - Other nodes: I/O bandwidth, communication overhead.
    - Lossless compression: Limited compression ratio (2:1).
- Goal: Error-bounded Lossy Compression with RTM.
  - Drastically reducing the data size while maintain data quality, making it possible to store snapshot data to memory.
  - Improving RTM overall performance.





One snapshot data can be **2800TB** when:

- 10 x 10 km aperture
- 8 km maximum depth

An RTM overall execution have 6k snapshots:

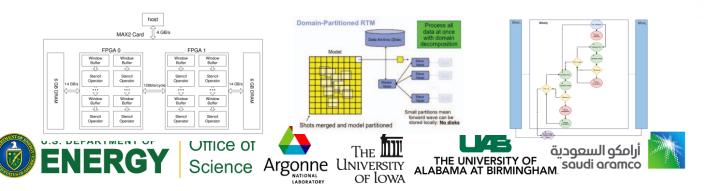
- Maximum frequency is 80 Hz
- Migration time is 6 seconds



### **Related Works**

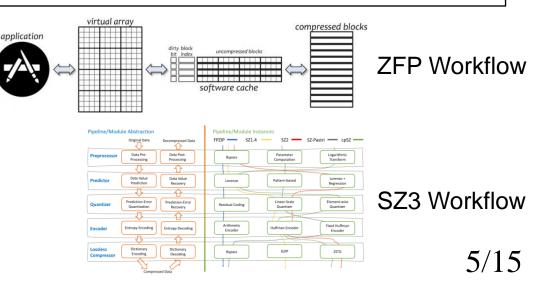
#### **Resolving Limited Memory in RTM**

- FPGA-based solution [FPGA'11].
  - Pro: Removing memory constraints.
  - Con: Computation perf. is compromised.
- **Domain-specific data partition** [IPDPS'12].
  - Pros: Paralleling RTM on Blue Gene.
  - Cons: Extra communication overhead.
- Distributed memory system [Cluster'19].
  - Pros: Utilizing multiple GPUs.
  - Cons: Data size remains the same.



#### **Error-bounded Lossy Compression**

- **ZFP** [TVCG'14]
  - Transform-based lossy compressor.
  - Orthogonal transformation + embedded encoding.
- **SZ** [IPDPS'16, '17]
  - Prediction-based lossy compressor.
  - Prediction + quantization + entropy encoding + lossless compression.





## Methodology: Outline

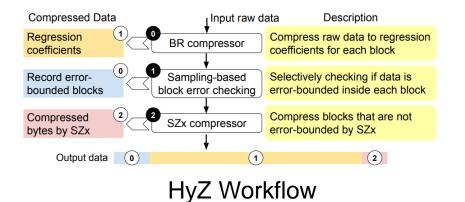
- **HyZ** is an error bounded lossy compressor that is specifically designed for RTM.
  - HyZ can satisfy the three key requirements for RTM execution.

High Reconstructed Data Quality

High Throughput

Good Compression Ratio for RTM (5+)

- We introduce HyZ compressor as follows:
  - 1. Limitations of SZx, which is an ultra-fast lossy compressor.
  - 2. BR: Block-wise regression-based lossy compression.
  - 3. HyZ: Combining BR and SZx for RTM execution.

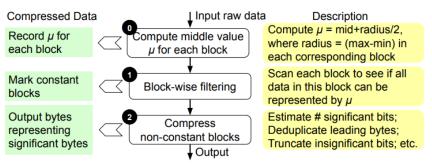




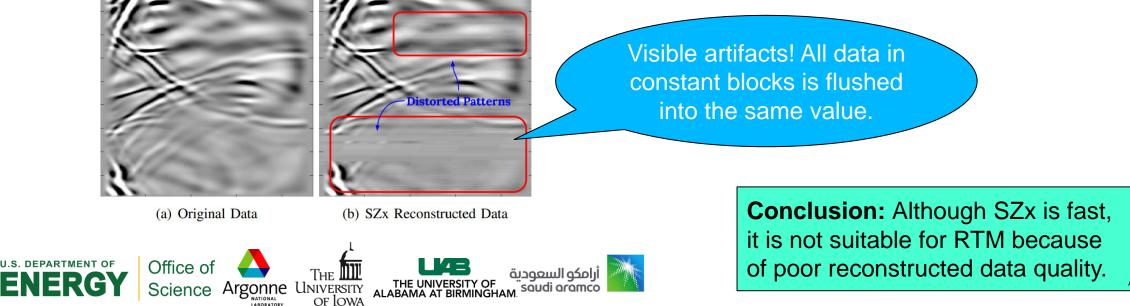
# **GRI2** Methodology: Limitations of SZx

- **SZx** [HPDC'18] is one ultra-fast compressor, which may be suitable for RTM execution.
  - SZx Workflow:
    - 1. Computing the middle value for each block.
    - 2. Checking if one block is constant or not.
    - 3. Compressing all non-constant blocks.

For constant blocks, only store its middle value.



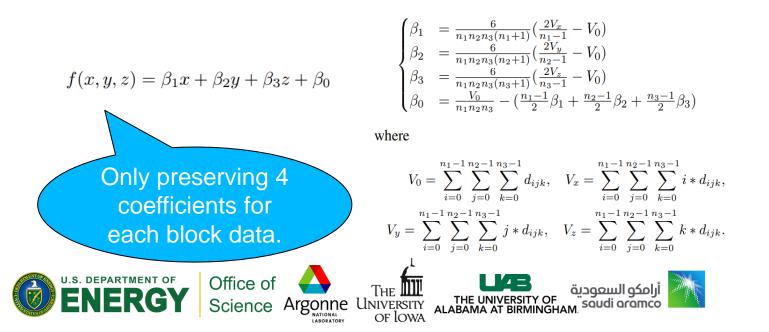
SZx Workflow

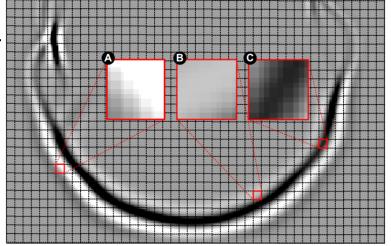




## Methodology: BR Compressor

- **BR:** A regression-based ultra-fast lossy compressor.
  - Key insight:
    - 1. The linearity of small data blocks in an RTM snapshot data.
    - 2. No dependency across different blocks, suitable for parallelism
  - BR workflow
    - 1. Dividing whole dataset into 3D blocks (e.g. 3x3x3).
    - 2. Predicting each block with a linear hyperplane.



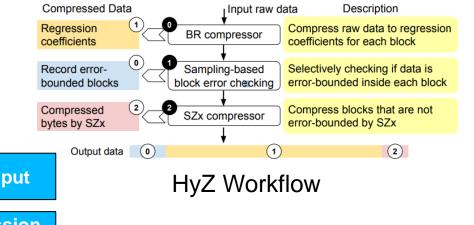


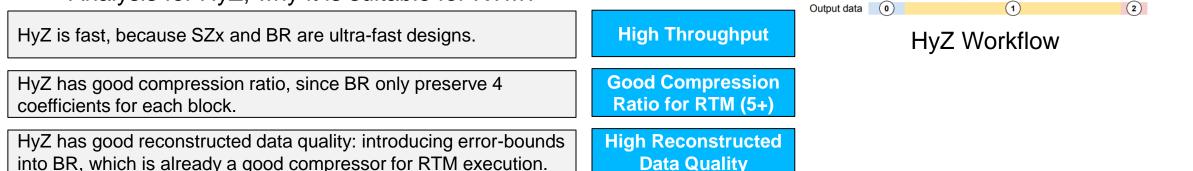
**Conclusion:** Fast enough, good data quality. But it is not error bounded, not enough for RTM post-hoc analysis.



## Methodology: HyZ Compressor

- **HyZ:** Combining two ultra-fast lossy compressor BR and SZx.
  - Idea: Designing an ultra-fast compressor with benefits from both BR and SZx.
  - HyZ workflow
    - 1. Compressing raw data by BR compressor.
    - 2. Checking if a BR-generated block is error-bounded or not.
    - 3. Compressing non-error-bounded blocks in BR.
  - Analysis for HyZ, why it is suitable for RTM?





Conclusion: Naturally suitable for RTM execution!







### **Evaluation: Setups**

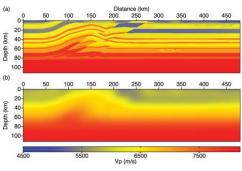
- Evaluation platform
  - Bebop supercomputer in Argonne National Laboratory. (two Intel Xeon E5-2695 V4 CPUs and 128GB RAM for each node)
- RTM code and dataset
  - Industrial parallel RTM code designed by Saudi Aramco.
  - Overthrust belt seismic data (dim: 449x449x235, 3600 snapshots in total).
- Baseline lossy compressor (36 threads for each compressor)
  - Transform-based: ZFP [TVCG'14]
  - Prediction-based: SZ [IPDPS'16, '17], SZ-Interp [ICDE'21], SZ-Interp-fast [TBD'23], SZx [HPDC'22]
- Evaluation metrics
  - Compression ratio.
  - Compressor speed: compression/decompression throughput (GB/s).
  - Data quality: PSNR, SSIM, visualization data quality.
  - RTM overall performance.











A slice of overthrust data





### **Evaluation: Compression Ratio**

Compressor	Relative Error Bound 1E-3 4E-4		Absolute E 1E-5	Average	
SZx	7.58	6.22	9.57	7.72	7.77
SZ-Interp	213.53	116.60	177.92	113.20	155.31
SZ-Interp-fast	17.05	9.09	16.26	11.06	13.37
SZ	26.94	25.51	26.83	24.37	25.91
SZ(Serial)	70.36	50.68	81.47	54.94	64.36
ZFP	26.92	20.65	38.74	32.04	29.59
Compressor		Average			
Compressor	$3 \times 3 \times 3$	$4 \times 4 \times 4$	$5 \times 5 \times 5$	$6 \times 6 \times 6$	Average
BR	6.66	15.72	31.11	52.64	26.53

Compression ratio for other compressors.

HyZ	Relative Error Bound 1E-3 4E-4		Absolute 1E-5	Average	
$3 \times 3 \times 3$	5.01	4.61	5.17	4.99	
$4 \times 4 \times 4$	9.14	8.20	10.29	9.69	12.31
$5 \times 5 \times 5 \\ 6 \times 6 \times 6$	13.87 18.50	12.24 16.11	17.10 24.84	15.59 21.52	

Compression ratio for HyZ.

**Takeaway 1:** HyZ and BR both can meet compression ratio requirements (5+) in RTM execution across 3600 time steps, with 12.31 and 26.53 on average.





### **GRI** Evaluation: Compression Speeds

Compressor	Relative Error Bound 1E-3 4E-4		Absolute E 1E-5	Average			
SZx	12.48	11.08	16.00 8.78		12.09		
SZ-Interp	2.87	2.76	2.76	2.47	2.72		
SZ-Interp-fast	5.48 5.36		5.53 5.11		5.37		
SZ	3.63	3.57	3.61	3.57	3.60		
ZFP	2.82	2.56	4.33	4.04	3.44		
Block Size							
Compressor	$3 \times 3 \times 3$	$4\times 4\times 4$	$5 \times 5 \times 5$	$6 \times 6 \times 6$	Average		
BR	5.85	15.64	24.36	32.46	19.58		

Compressor	Relative En 1E-3	rror Bound 4E-4	Absolute E 1E-5	rror Bound 4E-6	Average
SZx	19.65	18.13	22.45	15.44	18.92
SZ-Interp	5.26	4.88	4.77	4.48	4.85
SZ-Interp-fast	9.00	9.10	9.61	8.03	8.94
SZ	5.14	4.75	5.01	4.76	4.92
ZFP	0.60	0.54	0.62	0.58	0.59
Commession		A			
Compressor	$3 \times 3 \times 3$	$4 \times 4 \times 4$	$5\times5\times5$	$6 \times 6 \times 6$	Average
BR	10.67	19.36	24.96	26.59	20.40

#### Compression speeds for other compressors

#### Decompression speeds for other compressors

	Compression Speed				Decompression Speed					
HyZ	Relative 1E-3	Error Bound 4E-4	Absolute 1E-5	Error Bound 4E-6	Average	Relative 1E-3	Error Bound 4E-4	Absolute 1E-5	Error Bound 4E-6	Average
$3 \times 3 \times 3$ $4 \times 4 \times 4$ $5 \times 5 \times 5$ $6 \times 6 \times 6$	4.54 8.45 13.12 17.64	4.28 8.37 11.93 16.41	4.72 8.83 13.42 17.94	4.33 8.34 12.32 16.40	10.69	7.45 12.71 14.64 15.85	7.33 12.16 14.43 15.47	7.65 12.94 15.01 16.01	7.26 11.83 13.47 15.03	12.45

Compression and decompression speeds for HyZ

**Takeaway 2:** HyZ and BR both have top-class compression and decompression speeds. HyZ can achieve 10.69 GB/s and 12.45 GB/s for compression and decompression.



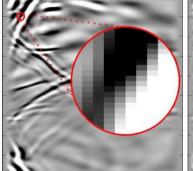
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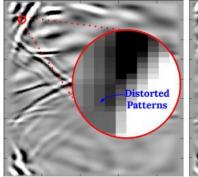


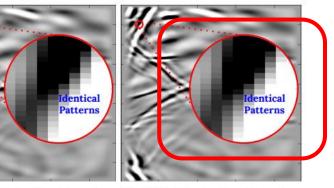




#### **Evaluation: Data Quality**



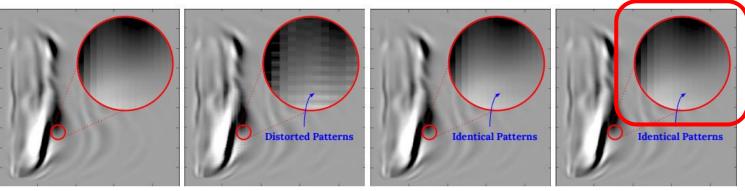




(a) Original Data

(b) BR, Block Size= $5 \times 5 \times 5$ , (c) HyZ, Block Size= $5 \times 5 \times 6$  HyZ, Block Size= $6 \times 6 \times 6$ (PSNR: 50.86dB, SSIM: 0.9546) 5, REL=1E-3 (PSNR: 87.91dB, 6, REL=1E-3 (PSNR: 68.78dB, SSIM: 0.9980) SSIM: 0.9861)

#### Single snapshot (snapshot=3000)



(a) Original Data

(b) BR, Block Size= $4 \times 4 \times 4$ , (c) HyZ, Block Size= $4 \times 4 \times 4$ , (d) HyZ, Block Size= $5 \times 5 \times 5$ , (PSNR: 74.50dB, SSIM: 0.9713) (PSNR: 124.13dB, SSIM: 0.9935) (PSNR: 117.35dB, SSIM: 0.9902)

#### Final stacking image



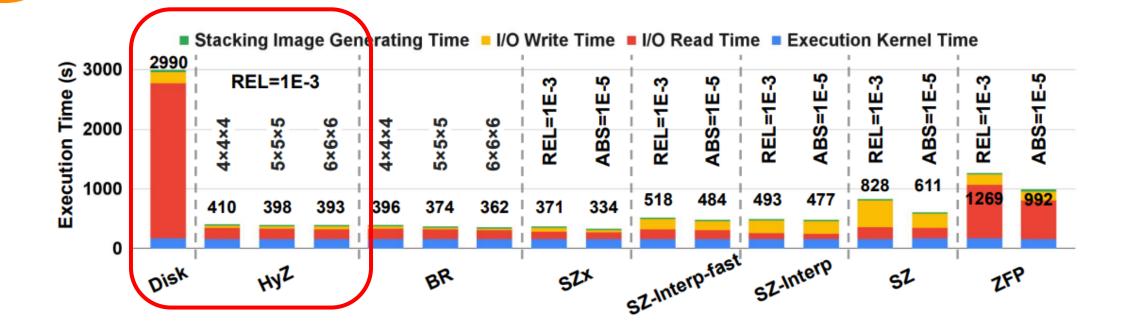






**Takeaway 3:** HyZ is superior to BR in preserving the data quality in both single snapshot and final stack imaging, also with higher PSNR and SSIM.

# GREVALUATION: RTM Overall Performance



**Takeaway 4:** HyZ and BR both can lead to a top-tier performances in RTM execution, increasing the performance on average by 6.46x and 6.93x, compared with RTM without lossy compression.





### Summary

- Conclusion
  - We use lossy compression to resolve the memory capacity issue in RTM execution.
  - We design HyZ, a domain-specific error-bounded compressor, for RTM execution.
  - Evaluation demonstrates HyZ is the best lossy compressor for RTM, with high speeds, high compression, and high reconstructed data quality.
- Future works
  - Accelerating RTM in GPU computing environments.
  - Optimizing HyZ performance.



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