QUERY-DRIVEN SPATIOTEMPORAL DATA ANALYTICS

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Publications Covered in This Talk

- Srisam Lakshminarasimhan, Neil Shah, S. Ethier, S. Klasky, R. Latham, R. Ross and N. F. Samatova, Compressing the Incompressible with ISABELA: In-situ Reduction of Spatio-Temporal Data, 17th International European Conference on Parallel and Distributed Computing [Euro-Par 2011], Nominated for Distinguished Paper Award.
- Sriram Lakshminarasimhan, John Jenkins, Isha Arkatkar, Z. Gong, H. Kolla, S-H Ku, S. Ethier, J. Chen, CS Chang, S. Klasky, R. Latham, R. Ross and N. F. Samatova, ISABELA-QA: Query-driven Data Analytics over ISABELA-compressed Scientific Data, [Supercomputing 2011], Under Review.
- List of all the other 10 publications in the NSF report of 5/25/11





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Motivating Factors

- Current large scale scientific simulation can easily produce several terabytes during a single run.
- I/O accesses have become a rate limiting factor to key scientific discoveries.
 - A paradigm shift is necessary for analysis in exascale.



Typical Simulation Data





- ~2TB per
 Checkpoint &
 Restart
- Every 1 hour
- Two copies

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- Keep the last copy
- Accessed once or never.



Analysis Data

- ~TB per run (now)
- Every 10th or 100th time step
- Repeated access.
- Analysis routines and I/O reads
- Matlab analysis routines



- ~ GBs
- Every 2nd time step
- Verify, and validate simulation.





Myth: "Scientific data is almost uncompressible" with lossless encoding



Still "Incompressible" with Lossy Encoding



Compression Techniques Performance Comparison

- GTS dataset, 170K vector, potential.
- Fixed overall correlation (>0.99*) & NRMSE (<0.01*) for lossy techniques

Metric	FPC	LZMA	ZIP	BZIP2	Wavelets*	B-splines*	ISABELA*
Lossless	Yes	Yes	Yes	Yes	No	No	No
CR _M	3.71	2.72	1.13	1.11	22.51	0	81.44
Compression (sec.)	0.58	7.01	1.03	3.96	0.62	0.78	0.93
Decompression (sec.)	0.56	1.38	0.49	1.18	0.58	0.82	1.05



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ISABELA's Magic Bullet



ISABELA: Communication-Free Workflow



Accuracy Robustness (CR=80%)

• ISABELA consistently gives almost a near-perfect correlation in Linear stages of the GTS simulation.



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Performance on Public Datasets (CR=80%)

	Correlation				NRMSE			
	WAVE	LETS	ISABELA		WAVELETS		ISABELA	
	Average	±	Average	±	Average	±	Average	±
msg_sppm	0.4	0.287	0.982	0.017	0.203	0.142	0.051	0.015
msg_bt	0.754	0.371	0.981	0.054	0.112	0.151	0.038	0.024
msg_lu	0.079	0.187	0.985	0.031	0.422	0.103	0.048	0.015
msg_sp	0.392	0.44	0.967	0.051	0.307	0.243	0.064	0.033
msg_sweep3d	0.952	0.07	0.998	0.006	0.075	0.036	0.004	0.003
num_brain	0.994	0.008	0.983	0.028	0.01	0.011	0.011	0.005
num_comet	0.988	0.018	0.994	0.025	0.02	0.02	0.01	0.006
num_control	0.614	0.219	0.993	0.017	0.083	0.037	0.009	0.002
num_plasma	0.605	0.062	0.994	0.004	0.277	0.038	0.033	0.004
obs_error	0.278	0.203	0.994	0.004	0.303	0.091	0.024	0.009
obs_info	0.717	0.136	0.993	0.006	0.181	0.078	0.026	0.016
obs_spitzer	0.992	0.001	0.742	0.004	0.005	0.001	0.03	0.032
obs_temp	0.611	0.114	0.994	0.011	0.096	0.025	0.009	0.003

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Effect of ISABELA Compression on Data Analysis with Combinations of Decompressed Variables



* Courtesy Hemanth Kolla, Jackie Chen @ SNL, C.S. Chang, S-H. Ku @ NYU

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MOTIVATION FOR QUERY-DRIVEN DATA ANALYTICS

- Exploratory data analytics is global-context and thus I/O bound:
 - Need *full context* of simulation data, not just a single time step.
 - Data-intensive analytics is constrained by I/O.
- Promise: Query-driven data reduction for downstream analysis
- Domain scientists' knowledge priors:

radial zones.

- Climate: Find Correlations between sea surface temperature anomalies and hurricane activity in North Atlantic region
- Fusion: Calculate the turbulence intensity flux at each radial zone and their time correlations
- Fusion: Calculate the time correlations of turbulence intensities between different



Challenges with B-Trees

 DBMS/ Indexing based techniques make it easy to locate "needles in a haystack"



Challenges with Fastbit

Record # / Bin	(- ∞, a)	[a, b)	[b, c)
1	0	1	0
2	0	1	0
3	0	1	0
4	0	1	0
5	1	0	0
6	1	0	0
7	0	0	1

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- Tailor made for read-only databases.
- Bitmap-based indexing.
- Encode each bitmap vector
 WAH compressed
 - Ability to perform fast logical operations directly on compressed space.
- Index size is **0.3 1.7x**
- **I/O** access dominate query cost



Challenges for Query-Based Analytics







LIMITATIONS OF EXISTING INDEXING APPROACHES

- Optimized for region retrieval
 - Output: points in space that meet the query (50<Temperature<70 & 250<Pressure<850)
 - BUT: Analysis requires actual values at those points
- I/O-bound for value retrieval
- Rely on heavy-weight index for fast query process
 - FastBit: 30%-200% of original data
 - B+ Trees: up to 300% of original data

Require access to the original data:

– Data + Index





ISABELA-QA: Goals



• Organize the data to allow decomposition of a query into parallel workloads.



ISABELA-QA: Architecture



Query Processing Cost with ISABELA-QA



ISABELA-QA: End-to-End Performance





ISABELA-QA: End-to-End Performance



ISABELA-QA vs. FastBit Features

Traditional Methods (FastBit)	ISABELA-QA
Optimized for region retrieval	Optimized for both region and especially value retrieval (2-5 orders of magnitude faster)
I/O bound	Trades I/O for CPU/GPU computation
Heavy-weight Index (30%-200% of original data)	Light-weight Index (3% of original data)
Requires original data	Operates over ISABELA-compressed data (15%-20% of original data)
Data + Index = ~270% of original data	Data + Index = 23% of original data
Serial	Embarrassingly Parallel

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Speedup with OpenMP

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Speedup of Computation for Value Query



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Speedup with GPU



Embarrassingly parallel. Bspline approximation ideally suited for GPU.



