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### An Intelligent, Adaptive, & Flexible Data Compression Framework

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Hariharan Devarajan, Anthony Kougkas, and Xian-He Sun

hdevarajan@hawk.iit.edu



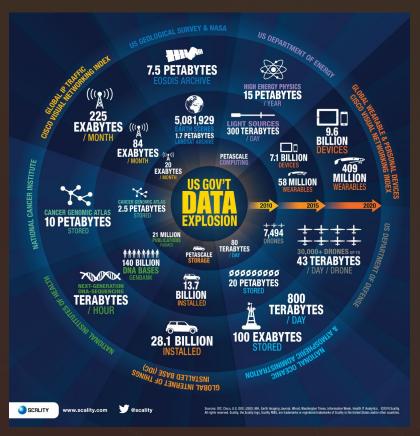
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### **Modern Big Data Applications**

- Explosion of data volume, variety, and velocity
  - Facebook is storing roughly 250 billion images.
  - 80% of all the world's data is unstructured.
  - Facebook users upload more than 900 million photos a day.
  - Square Kilometer Array (SKA) is estimated to reach 10 Pb/s.



#### Generates tremendous stress on storage sub-system





# Data Compression is popularly used to ease this stress





### **Data Compression**

- Categories of compression techniques
  - lossy and lossless algorithms.
- The lossless algorithms are standard in scientific and cloud applications.
- Popular examples of lossless algorithms
  - General Purpose: Bzip, Zlib, 7z, etc.
  - Specialized: Snappy, SPDP, LZO, etc.





### Challenges in Data Compression

- <u>Data-dependency</u>: Each Compression library is specialized for a certain input (i.e., data-type and data-format)
- Library-choice: Choice of library is complex as different situations might demand different compression needs.
- **<u>API diversity</u>**: Each library has its own definition of Interface.

These challenges highlight there is no "one compression for all".

Can we do something better?





## Outline

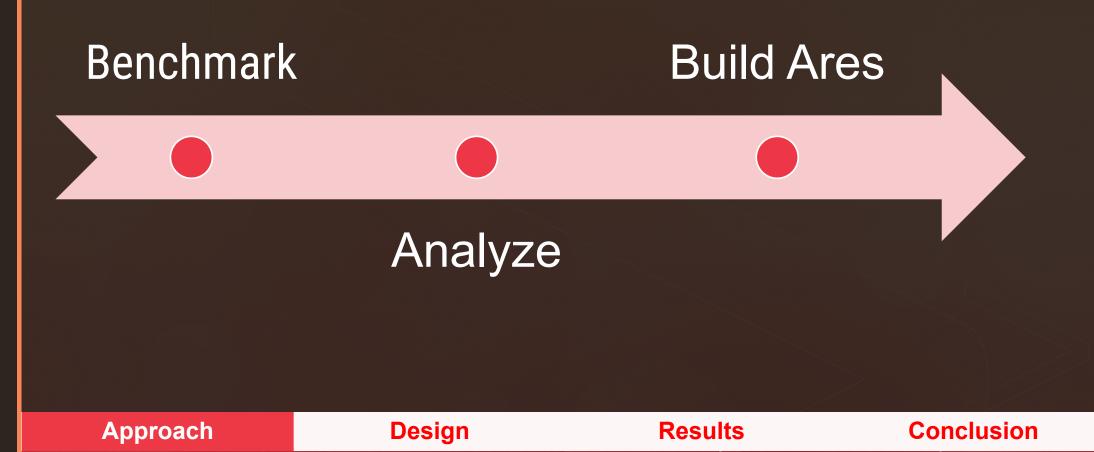
Ares

- □ <u>Approach</u>
- Design
- □ <u>Results</u>
- □ <u>Conclusion</u>





### **Approach: Overview**





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### **Approach: Benchmark**

- Library Corpus: bzip2, zlib, huffman, brotli, bsc, lzma, lz4, lzo, pithy, snappy, and quicklz.
- **Data-Types:** characters, integers along with their modifiers (short, long, signed, unsigned), sorted integers, floating point, and double floating points.
- **Data-Formats:** binary, HDF5, csv, json, xml, and Avro, Parquet.
- **Metrics:** Compression/Decompression Speed & Compression Ratio

#### These total to over 1000 test cases

Results





### OF TECHNOLOGY Approach: Analyze

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- Workload Priority: defines different requirement that a workload prioritizes
- Score Formulation:

Binary Format								
CS	DS	CR	Workload	Char	Integer	Sorted int	Float	Double
1	0	0	Asynchronous communication	lz4	lz4	lz4	quicklz	lz4
0	1	0	Multicast in Network	lz4	lz4	pithy	pithy	brotli
0	0	1	Archival Store	bsc	Izma	bsc	Izma	bsc
0.5	0.5	0	Synchronous Communication	lz4	lz4	pithy	pithy	lz4
0	0.5	0.5	Dequeue Operation	lz4	lz4	lz4	quicklz	pithy
0.5	0	0	Queue Operation	lz4	lz4	lz4	pithy	lz4
0.3	0.3	0.3	Mixed workload	lz4	lz4	pithy	pithy	pithy

Approach

#### Design

#### **Results**





## **Approach: Build Ares**

- Goals
  - The framework should be able to learn and adjust itself to the input data compression characteristics.
  - The framework should be able to reconfigure itself, dynamically, to various compression needs of an application.
  - The framework should be able to unify all interfaces of the compression libraries it contains.

Results





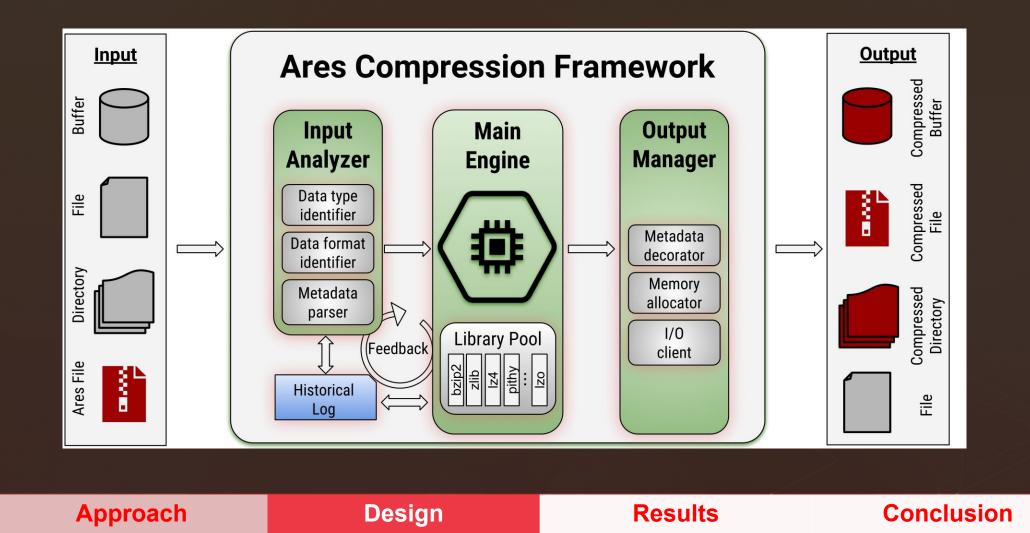


### **Ares Design**





### **Design: Overview**

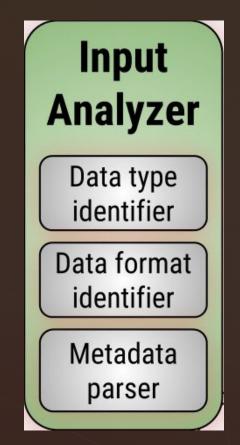




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Design

- Infers data type and format
- Uses a hybrid approach
  - static analysis and a dynamic feedback mechanism
- <u>Data-format:</u> mime-type, extensions, and metadata-rich.
- <u>Data-type</u>: decoding techniques, type-inference and metadata-rich.



#### Approach

#### Design

#### **Results**



ILLINOIS INSTITUTE OF TECHNOLOGY Design Main Engine



#### Approach

#### Design

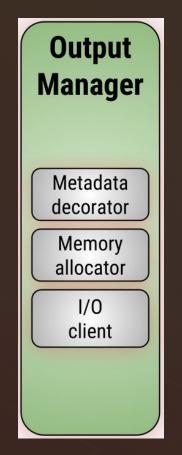
#### **Results**





Design

- Decorates the compressed data with headers, regarding the compression library used
  - 8 bytes header per data-type
- Checks the correctness of the format using parity checking
- Performs final I/O of the compressed/uncompressed data



#### Approach

#### Design

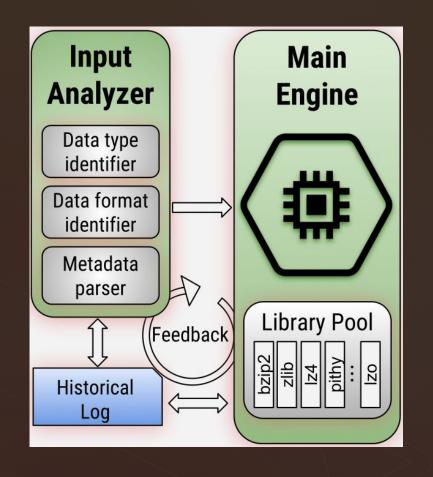
#### Results





### Design

- The engine updates the log with actual performance results
- Analyzer processes the log to identify the difference between expected and actual measurements.
- This makes the analyzer improve it predictions over time.



#### Design

#### **Results**





### **Evaluation**



ILLINOIS INSTITUTE V OF TECHNOLOGY Evaluation: Testbed

#### Machine Configuration (Per Node)

- dual Intel(R) Xeon(R)
  - CPU E5-2670 v3
  - 2.30GHz
  - 48 cores
- 128 GB RAM
- 10 Gbit Ethernet,
- 200 GB HDD

#### Deployment

- Scientific Setup:
  - 32 client nodes
  - 8 PFS nodes
- Cloud Setup:
  - 40-node Hadoop cluster
  - 1 Namenode

#### Approach

#### Design

#### **Results**





### **Evaluation: Goals**

- Overheads and Resource Utilization
  - Ares's analysis overheads + CPU + Memory utilization
- Compression/Decompression Intelligence
  - Data type and format aware data compression
- Compression/Decompression Adaptiveness
  - Workflow-specific data compression
- Compression/Decompression Flexibility
  - Ares for various real applications

Approach

#### Design

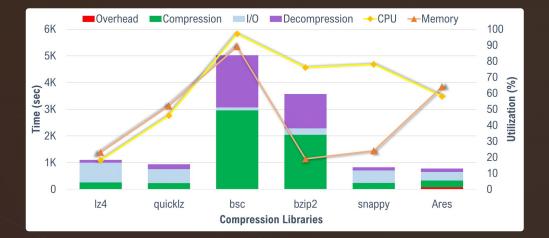
#### Results





### **Evaluation: Overheads**

- Description:
  - 64GB HDF5 input with four datasets: characters, integers, sorted integers, and doubles.
  - Workflow: read input data -> compress data -> write compressed data -> read compressed data -> decompress the data.
- Metrics overall time and utilization



Approach

#### Design

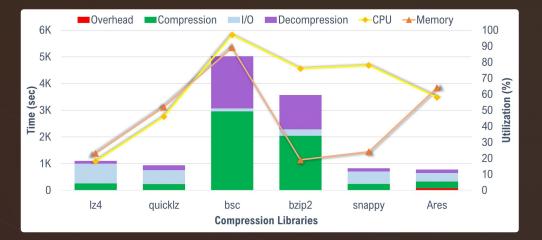
#### Results





### **Evaluation: Overheads**

- Observations
  - libraries demonstrates different overheads
  - Ares balances the tradeoff between CT/DT and CR by analyzing the input data with a 10% overhead.
  - Ares performs better as it uses a collection of libraries where they have strength.



Approach

#### Design

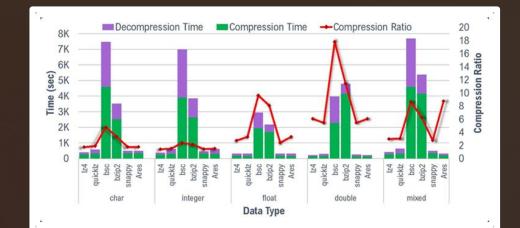
#### Results





### **Evaluation: Intelligence**

- Description:
  - Different Data-Types
  - 64 GB of buffer input
  - Configurations of this buffer:
    - Characters, integers, floats,
      doubles, and a mixed case
- We measure the CT, DT and CR.



#### Approach

#### Design

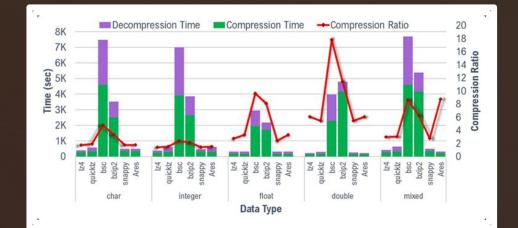
#### **Results**





## **Evaluation: Intelligence**

- Observation:
  - Different libraries excel in different data types.
  - trade-off between CT and CR
  - For mixed input each library takes a hit in performance
    - Ares optimizes by using best library for given data-type.



#### Design

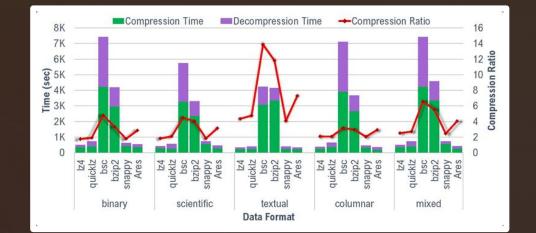
#### **Results**





## **Evaluation: Intelligence**

- Description:
  - Different Data-Format
  - 64 files (each 1 GB) in a directory
  - Composition of this folder:
    - POSIX , HDF5, pNetCDF, HTML,
      XML, JSON, Avro, and Parquet
- We measure the CT, DT and CR.



**Approach** 

#### Design

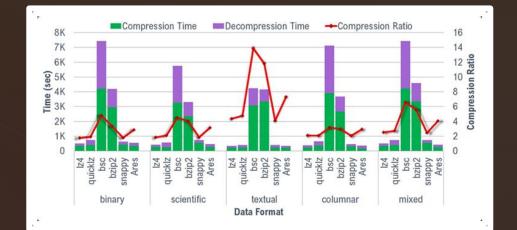
#### **Results**





## **Evaluation: Intelligence**

- Observation:
  - Different libraries excel in different data formats.
  - trade-off between CT and CR
  - For mixed directory each library takes a hit in performance
    - Ares optimizes by using best library for given data-format.



#### Approach

#### Design

#### Results





**Conclusion** 

### **Evaluation: Adaptiveness**

• Description:

Approach

- Different Workflow Priorities
- 64 GB of CSV file input
- Four columns of this file:
  - Index (sorted integer), location (char), population size (integer), income (double)

Design

• We measure the CT, DT and CR.



**Results** 





Conclusion

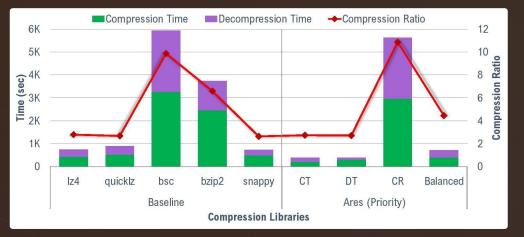
### **Evaluation: Adaptiveness**

• Observation:

Approach

- This multi type data has an effect on every compression library
- Different prioritization of Ares results in difference in performance metrics
- In Balanced mode, Ares is a Jack of all trades.

Design



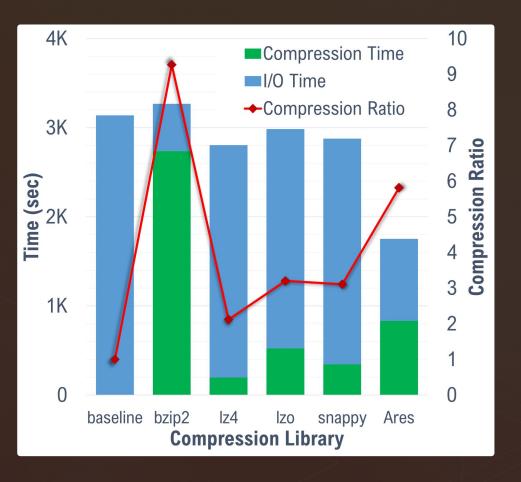
**Results** 





### **Evaluation: Scientific Application (VPIC)**

- VPIC simulation
  - Each process is producing 1 GB at each time step.
  - The overall data size is 1.5 TB
  - HDF5 file is organized with 7 datasets
    - two datasets of integers, two of floats and three of doubles.
- We show Compression Time, Compression Ratio and I/O Time.



Approach

Design

**Results** 

Conclusion

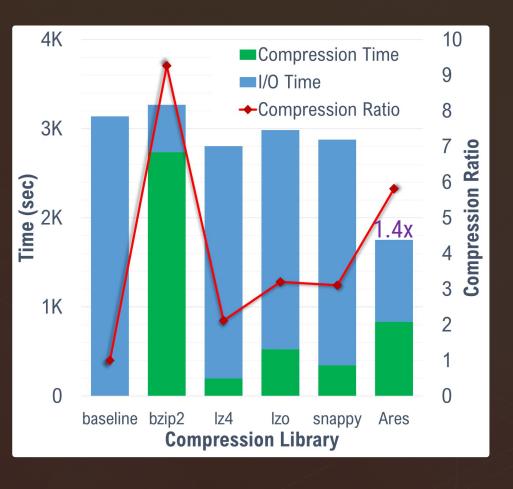
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### **Evaluation: Scientific Application (VPIC)**

- Observation
  - Compression reduces the I/O time.
  - Heavy compression is costly, and a balance must be found to be beneficial to the application
  - Opposite picture can be seen when using lz4, lzo, and snappy as compression filters.
  - Ares prioritizes both CT and CR.



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#### Design

#### **Results**

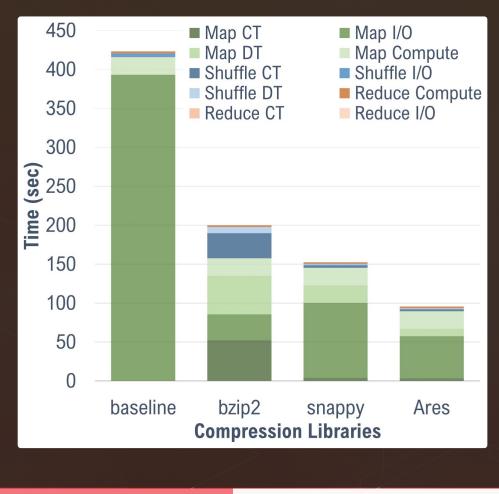


## **Evaluation: Hadoop Application (Word Count)**

- Map-Reduce implementation of the word-count kernel 32 mappers and 8 reducers
  - 1.5 TB of HTML files (Wikipedia articles)
  - Workflow:

Approach

- MAP: reads its input data and counts individual word occurrences and create intermediate files
  - a compressed input and a high DS.
- SHUFFLE: all intermediate files are sorted
  - quick compression to minimize I/O traffic.
- REDUCE: merge the final count across all intermediate files and write the final word count back to a file in HDFS



#### Design

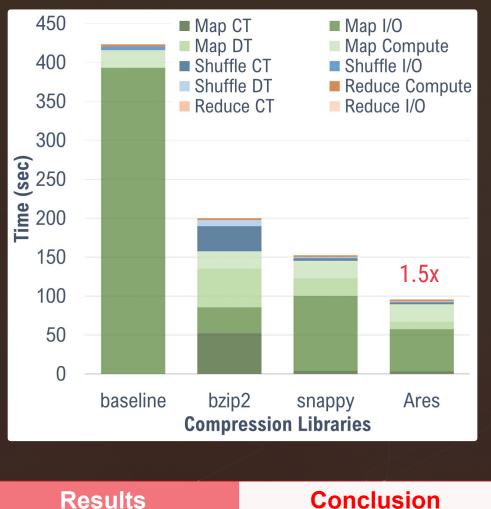
#### **Results**





### **Evaluation: Hadoop Application (Word Count)**

- **Observations:** 
  - compression on the input data reduces the I/O time in map phase.
  - tradeoff CT/DT and CR
  - Ares achieving the best overall performance
  - highlights the importance of striking a balance of compression speed and ratio.
    - Compression libraries do not offer dynamic adaptiveness based on the workload type.



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Approach





Conclusion

- We investigated how different data-types, data-format, and workload characteristics affect the choice of the "ideal" compression library.
- We have developed Ares, a dynamic, adaptive, and flexible compression framework, that can transparently meet various compression needs of big data applications.
- Under real world applications Ares performed 2-6x faster than competitive solutions with a 10% analysis cost.









#### Hariharan Devarajan, hdevarajan@hawk.iit.edu

**Q & A** 





