

## HDF5 & Blosc2

Unleashing the full potential of Blosc2 from HDF5

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### The Blosc Development Team



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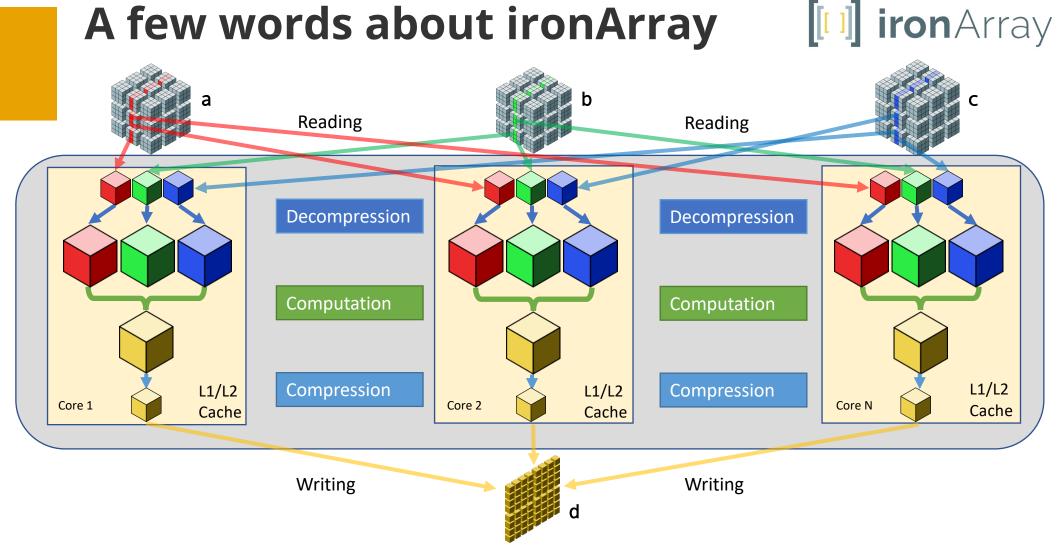
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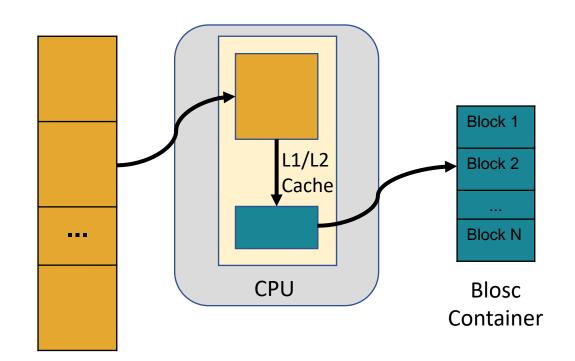


Intertwining compression and computation for improving performance

#### What is Blosc?



- Sending data from CPU to memory (and back) faster than memcpy().
- Split in blocks for better cache use: divide and conquer.
- It can use different filters (e.g. shuffle, bitsuffle) and codecs (e.g. LZ4, Zlib, Zstd, BloscLZ).



Binary dataset (Chunk)

#### **Origins of Blosc**



- 2009: it was very clear that compression was slowing down storage in PyTables/HDF5 a lot. Work began.
- **2010**: Blosc 1.0 was ready for production. Innovations:
  - Shuffle filter was optimized for SSE2 (\*much\* faster)
  - Multithreaded operation
- 2013: Blosc gained multi-codec (LZ4, Snappy and Zlib where included)
- 2015: hdf5-blosc plugin for HDF5 was released (hdf5plugin took over!)
- **2021**: Blosc2 appeared with **lots** of new features.

#### What is Blosc2?



#### Next generation of Blosc1.

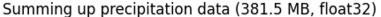
- New 63-bit frames that expand over the existing 31-bit chunks in Blosc1.
- Metalayers for adding info for applications.
- Area for adding metadata for users (variable length).

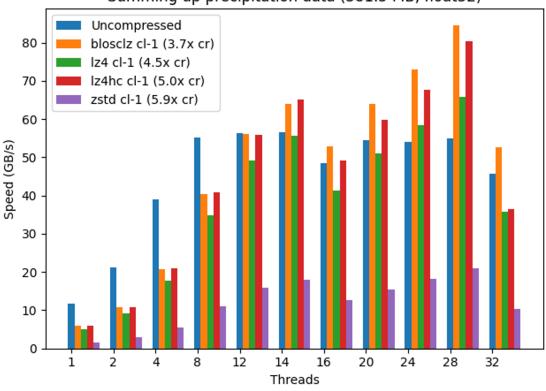
#### Blosc2 Frame

Filter Pipeline **Itemsize** Metalayers Chunk 0 Chunk 1 Chunk N **Chunk Index** UserMeta



#### **Example of Decompression Speed**



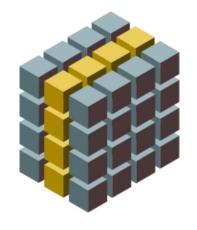


https://www.blosc.org/posts/breaking-memory-walls/



## **HDF5: Multidimensions and Chunking**

Data can be stored in hypercubes, making retrieval very convenient.





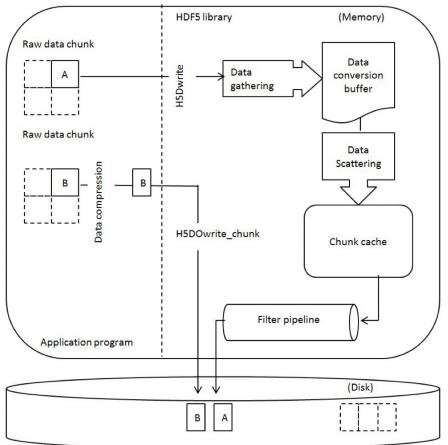
• But there is a price to pay for this flexibility: HDF5 is known to be **slow** when writing/retrieving (hyperslabs of) data.

#### Direct Chunk Write/Read Feature

- Allow the aplication to handle the chunk I/O and bypass the powerful (but slow!) chunk handling machinery in HDF5.
- The result is that data can be handled up to about 10x faster (with efficient pre and post processing in the app).



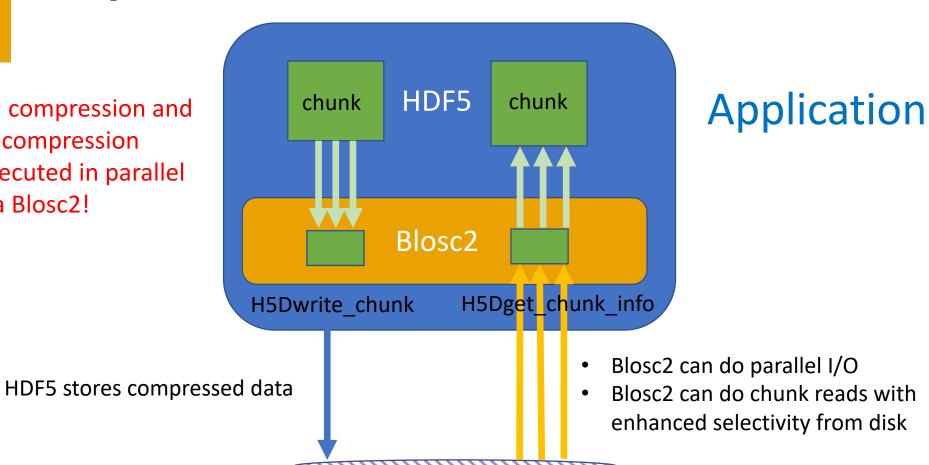




## Proposal 1: Use Blosc2 Inside Direct Chunk



All compression and decompression executed in parallel via Blosc2!



Storage

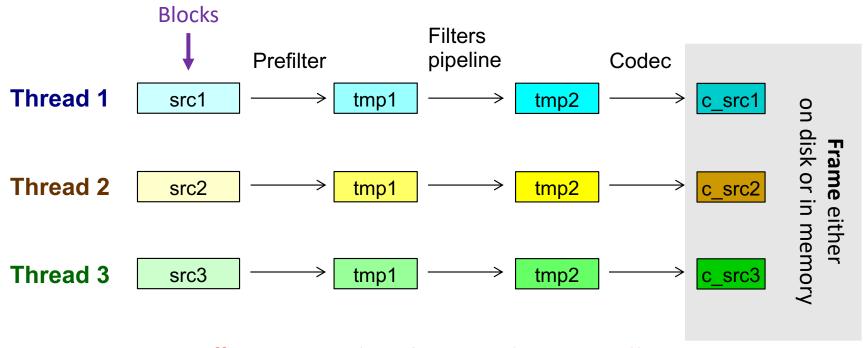


# **Blosc2 Advantages**



#### **Blosc2: Fine Tuned Cache Usage**

Compression: chunks are split in blocks for CPU cache sake

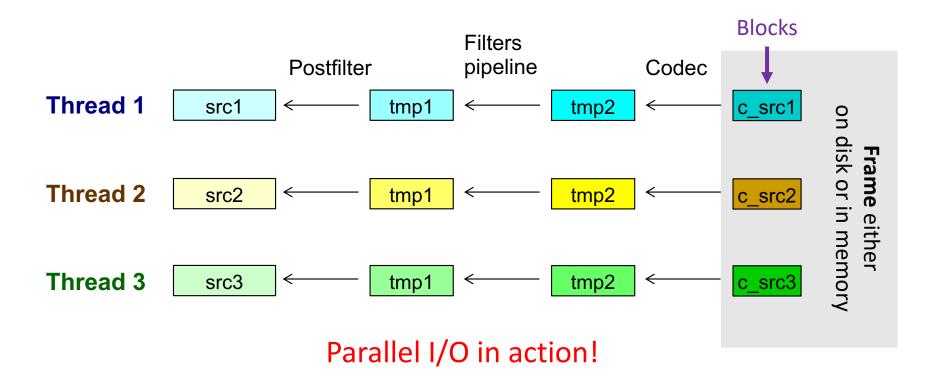


Buffers are reused **inside** CPU caches -> speed!



#### **Blosc2: Leveraging I/O Parallelism**

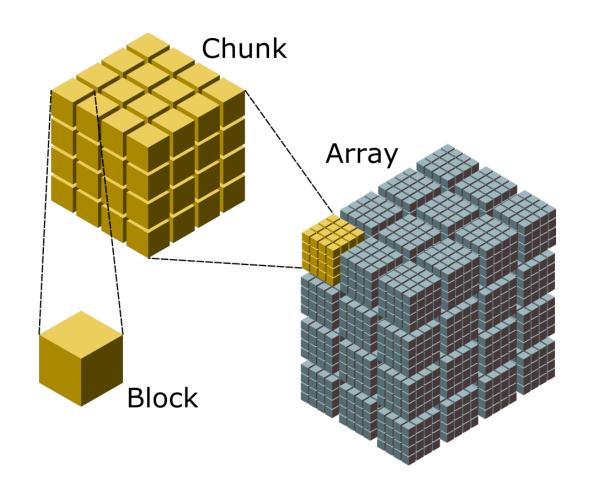
Decompression: blocks are read in parallel from storage





#### Caterva: Blosc2 Goes Multidimensional

- Metalayer representing multidimensionality
- Each Caterva array is split in chunks
- Each chunk is split in blocks
- All the partitions are multidimensional!

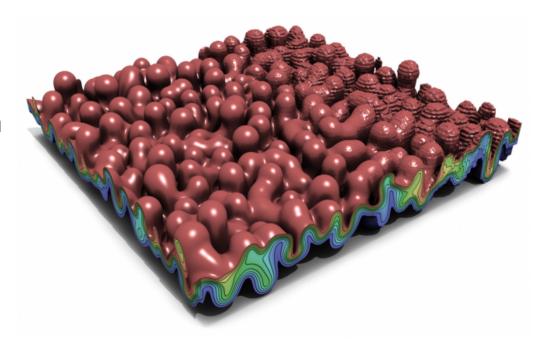




#### **Benefits of the Caterva Layer**

Get improved
 compression ratio
 because data is packed in a
 way that can show higher
 spatial locality.

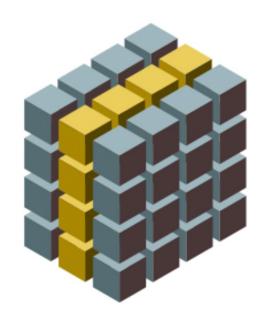
Also, get improved
 hyperslab query speed,
 i.e. some blocks can be
 masked out so as to not
 read them.

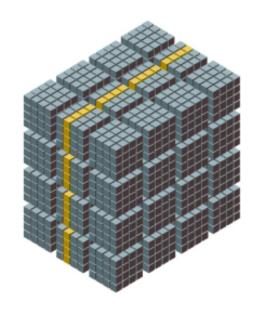


ZFP: a new registered plugin

## Masked & Paralel I/O in Multidim Datasets







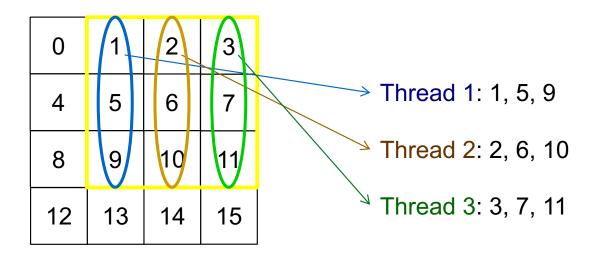
Much more selective and faster queries!

<u>Caterva</u> (https://github.com/Blosc/caterva) and <u>ironArray</u> (https://ironarray.io)

#### **Block Masks and Parallel I/O**



Block maskout	F	Т	Т	Т	F	Т	Т	Т	F	Т	Т	Т	F	Т	Т	Т
Index	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15

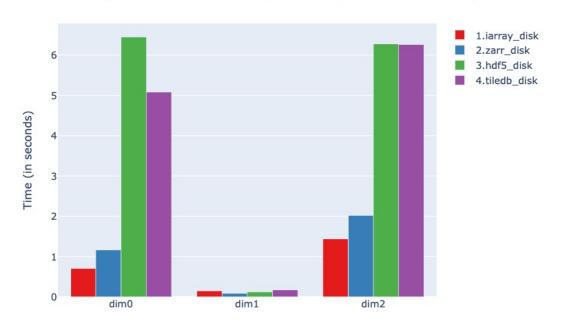


Specially effective when retrieving slices of multidim datasets.



#### Masked & Paralel I/O in Multidim Datasets

Slicing Performance on disk (with an optimized dimension)



Better performance in general (except for dimensions where retrieving a chunk is already optimal)

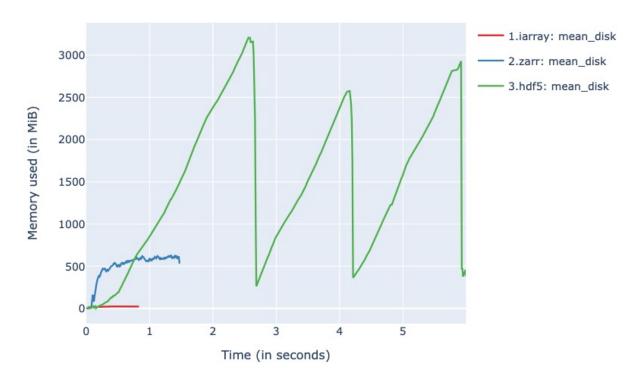
https://ironarray.io/docs/html/bench/03.Slicing Datasets and Creating Views.html



#### **Blosc2: Paralellism and Efficiency**

#### Mean of 3 arrays of 3 GB each (on disk)

- In the plot: 3 compressed arrays are decompressed, operated with, and the result is compressed again.
- ironArray is using Blosc2.
- When handled correctly, parallelism can buy not only speed, but also less memory resources!



https://ironarray.io/docs/html/bench/05.Reductions\_OnDisk.html

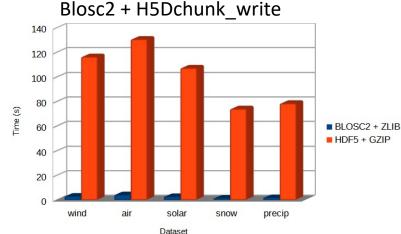


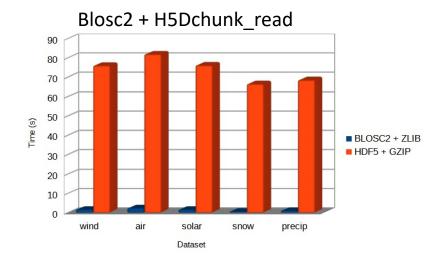
#### Blosc2 + HDF5 Direct Chunk Quick Benchmark

Quick test on data from ERA5 dataset, using different measurements (wind, temp, precip...). The datasets are about ~3 GB (uncompressed) each.

- Blosc2 + HDF5 speed-up is typically between 30x and 40x for writing.
- Blosc2 + HDF5 speed-up is typically between 40x and 60x for reading.

Note: this is using a laptop with 6 cores, but the measuments make sense, specially when using a fast storage (in this case OS FS cache).







## **Adaptability: Plugins in Local Registry**

#### Filters registry

User defined filter:

int urfilter2(
 blosc2\_filter \*filter) {
 ...
}

To register locally:

blosc2\_register\_filter(
 urfilter2)

BLOSC_SHUFFLE	1
BLOSC_BITSHUFFLE	2
BLOSC_DELTA	3
BLOSC_NDCELL	32
BLOSC_NDMEAN	33
urfilter1	160
urfilter2	161

□ Blosc official registered filters
□ User local filters

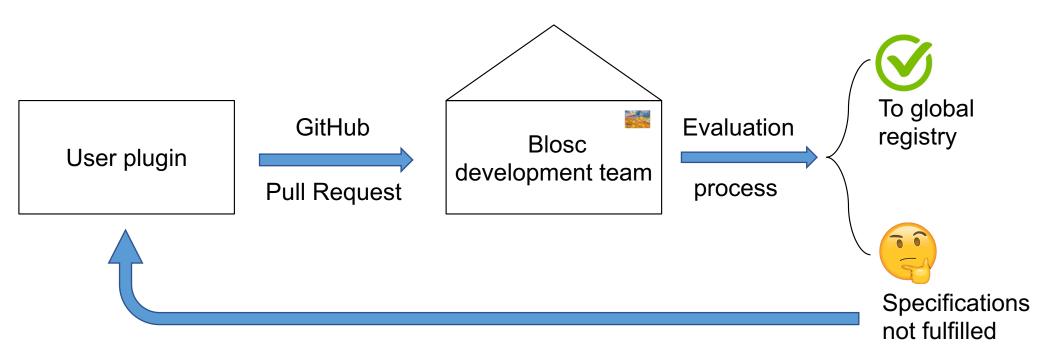
Can be used now:

cparams.filters[4] = 161;

(Similar functionality to the plugin interface in HDF5)

### **Registering Global Plugins in Blosc2**





Specs: https://github.com/Blosc/c-blosc2/blob/main/plugins/README.md



# Proposal 2: Help in Determing Optimal Compression Pipelines

We are offering a service for adapting to the user data, and determining:

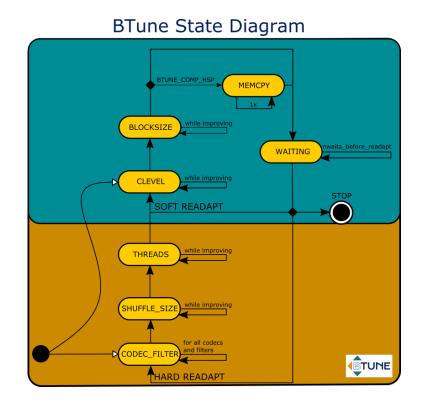
- Set of most useful codecs to be used
- Set of most useful filters to be used

We produce **specific versions** of **BTune**, a machine learning tool for selecting the best pipeline candidate on a **chunk by chunk** basis, that adapts to the needs of the user.



#### Fine Tuning Performance with BTune

- BTune can fine tune the different parameters of the underlying Blosc2 storage to perform as best as possible.
- Active during the compression pipeline.
   Automatically learns the best parameters on the go.



#### **Demo time**



https://btune.blosc.org

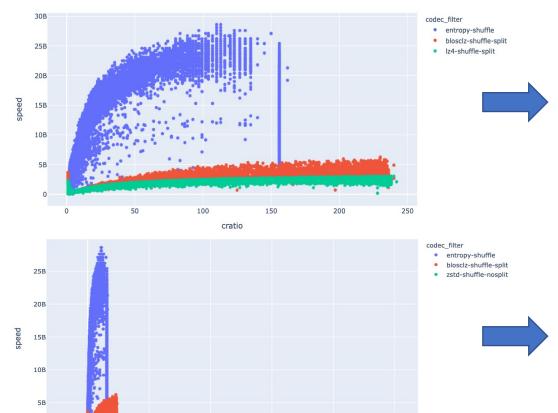


#### **Work in Improving BTune**



- Currently BTune needs some warm-up tests (hard and soft readapts) so as to come with a sensible guess.
- We are planning to shorten this warm-up period by using deep learning techiques.
- The idea is to come with some predictor for the entropy for every chunk and train a neural network. This will be used for reaching the sensible guess faster.





1500

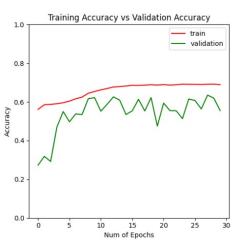
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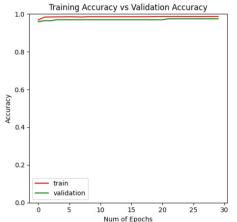
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Codecs with similar features:
Not good predictions



Codecs with different features:
Much better predictions!



## Conclusion

#### **Blosc2 Helps Saving Resources**



Blosc2 orchestrates a **rich set of codecs and filters** for:

- CPU parallelization via multithreading
- Reuse and sharing internal buffers for optimal memory consumption
- Parallel I/O
- More selective hyperslabs

The result is a highly efficient tool for compressing and accessing your data your way

#### **Summary of Proposals**



1. Use Blosc2 in combination with HDF5 direct chunking mechanism for efficient compression and parallel I/O.

2. Help in determing optimal compression pipelines by adapting to user data and using machine learning techniques.

The Blosc team would be glad to be involved in efforts towards these goals

# Thanks to donors & contracts & contractors!

















Jeff Hammerbacher

Without them, we could not have possibly put Blosc2 into production status: Blosc2 2.0.0 came out in June 2021; now at 2.1.1.



## **Enjoy data!**



https://blosc.org/