

#### A Compressed And Multidimensional Container For Big Medium Size Data

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# The Big Debate (I)

Evolution of proprietary file formats by Bio-Formats Source: Euro-Bioimaging Industry Board



### "Put simply, the format landscape has scaled beyond a manageable level."

-OME's position regarding file formats

https://blog.openmicroscopy.org/community/file-formats/2019/06/25/formats/

# The Big Debate (II)

- My position on adding new formats is that we absolutely need them for adapting not only to different use cases but also to the rapidly evolving storage technology.
- In particular, I claim that we need more in-memory compressed formats because I see memory more and more as another storage layer, and high-speed compression can help on using memory more efficiently.
- Having said that, we \*absolutely need\* to adopt open formats for reducing the maintenance burden and facilitate its adoption more quickly.

# What is Caterva?

- It is an open source C library and a format that allows to store large multidimensional, chunked, compressed datasets.
- Data can be stored either in-memory or on-disk, but the API to handle both versions is the same.
- Compression is handled transparently for the user by adopting the Blosc2 library.
- cat4py is a thin wrapper for Python.



### Caterva Brings Powerful Slicing Capabilities

- Caterva's main feature is to be able to extract all kind of slices out of high dimensional datasets, efficiently.
- Resulting slices can be either Caterva containers or regular plain buffers (for better interaction with e.g. NumPy).





#### **Compression and Partitions**



- Caterva normally splits the dataset into so-called partitions (aka chunks in Blosc idiom).
- Such partitions can be compressed individually.



#### Accessing Chunked Datasets

Those used to manipulate chunked multidimensional arrays know how critical choosing the partition size is.



https://github.com/Blosc/cat4py/blob/master/notebooks/compare\_getslice.ipynb

# Performance In-Memory



Caterva is meant to read data very fast!

# Performance On-Disk



Caterva still has room for optimization when reading from disk!

#### Brief Comparison Against Well Known Chunked Containers

	HDF5	Zarr	Caterva
Solid	Yes	No (1 file per chunk)	Yes
Mature	Yes	Yes	In process
In-memory version?	Yes (sequential?)	Yes (sparse)	Yes (sequential / sparse)
Hierarchical	Yes	Yes	No (use the filesystem)

# Blosc2

- Blosc2 is the next generation of the well-known Blosc (aka Blosc1).
- New features:
  - Enlargeable 64-bit containers: in-memory or on-disk
  - New compression codecs
  - New filters
  - Metalayers
  - User metadata



### A Few Words About Blosc1

- Blosc1 (or just Blosc) was the original meta-codec that was meant to be used to accelerate operation in PyTables (HDF5), but it became stand-alone pretty soon.
- Blosc1 is distributed with different codecs and filters so that it is very easy to be adopted by third-party libraries.
- It is already 10 years old, and since its inception it has seen a broad adoption in different libraries implementing data containers.



#### **Compression Speed**

Blosc(cname='zstd', clevel=5, shuffle=2) Blosc(cname='zstd', clevel=1, shuffle=2) Blosc(cname='zlib', clevel=5, shuffle=2) Blosc(cname='zlib', clevel=1, shuffle=2) Blosc(cname='lz4hc', clevel=5, shuffle=2) Blosc(cname='lz4hc', clevel=1, shuffle=2) Blosc(cname='lz4', clevel=9, shuffle=2) Blosc(cname='lz4', clevel=5, shuffle=2) Blosc(cname='lz4', clevel=1, shuffle=2) Blosc(cname='blosclz', clevel=9, shuffle=2) Blosc(cname='blosciz', clevel=5, shuffle=2) Blosc(cname='blosclz', clevel=1, shuffle=2) Blosc(cname='snappy', clevel=9, shuffle=2) Blosc(cname='zstd', clevel=5, shuffle=0) Blosc(cname='zstd', clevel=1, shuffle=0) Blosc(cname='zlib', clevel=5, shuffle=0) Blosc(cname='zlib', clevel=1, shuffle=0) Blosc(cname='lz4hc', clevel=5, shuffle=0) Blosc(cname='lz4hc', clevel=1, shuffle=0) Blosc(cname='lz4', clevel=9, shuffle=0) Blosc(cname='lz4', clevel=5, shuffle=0) Blosc(cname='lz4', clevel=1, shuffle=0) Blosc(cname='blosciz', clevel=9, shuffle=0) Blosc(cname='blosciz', clevel=5, shuffle=0) Blosc(cname='blosciz', clevel=1, shuffle=0) Blosc(cname='snappy', clevel=9, shuffle=0) Blosc(cname='lz4', clevel=0, shuffle=0) LZMA(format=1, check=-1, preset=1, filters=None) BZ2(level=1) Zlib(level=1)

None

Compression speed (multi-threaded Blosc)



http://alimanfoo.github.io/2016/09/21/genotype-compression-benchmark.html

### **Decompression Speed**

Blosc(cname='zstd', clevel=5, shuffle=2) Blosc(cname='zstd', clevel=1, shuffle=2) Blosc(cname='zlib', clevel=5, shuffle=2) Blosc(cname='zlib', clevel=1, shuffle=2) Blosc(cname='lz4hc', clevel=5, shuffle=2) Blosc(cname='lz4hc', clevel=1, shuffle=2) Blosc(cname='lz4', clevel=9, shuffle=2) Blosc(cname='lz4', clevel=5, shuffle=2) Blosc(cname='lz4', clevel=1, shuffle=2) Blosc(cname='blosciz', clevel=9, shuffle=2) Blosc(cname='blosclz', clevel=5, shuffle=2) Blosc(cname='blosclz', clevel=1, shuffle=2) Blosc(cname='snappy', clevel=9, shuffle=2) Blosc(cname='zstd', clevel=5, shuffle=0) Blosc(cname='zstd', clevel=1, shuffle=0) Blosc(cname='zlib', clevel=5, shuffle=0) Blosc(cname='zlib', clevel=1, shuffle=0) Blosc(cname='lz4hc', clevel=5, shuffle=0) Blosc(cname='lz4hc', clevel=1, shuffle=0) Blosc(cname='lz4', clevel=9, shuffle=0) Blosc(cname='lz4', clevel=5, shuffle=0) Blosc(cname='lz4', clevel=1, shuffle=0) Blosc(cname='blosclz', clevel=9, shuffle=0) Blosc(cname='blosclz', clevel=5, shuffle=0) Blosc(cname='blosclz', clevel=1, shuffle=0) Blosc(cname='snappy', clevel=9, shuffle=0) Blosc(cname='lz4', clevel=0, shuffle=0) LZMA(format=1, check=-1, preset=1, filters=None) BZ2(level=1) Zlib(level=1)

None

Decompression speed (multi-threaded Blosc) - 61.1X 53.1X • 50.2X 37.6X --- 41.8X - 35.0X 32.8X 31.0X 29.1X 30.6X - 30.7X 28.3X ---- 10X ••••• 47.4X • --- 38.7X 42.2X 28.3X --- -- 34.2X 20.5X ---------18.9X 18.5X ----- ---- 18.1X •••••• 13.4X 19.7X ..... ...... 8.2X 10X •••• 10X •• 48.7X 70.9X 29.0X 10X 0 2000 4000 6000 8000 10000 12000 Speed (M/s)

http://alimanfoo.github.io/2016/09/21/genotype-compression-benchmark.html



In-memory

# MetaLayers in Blosc2

- Metalayers are small metadata for informing about the kind of data that is stored on a Blosc2 container.
- Example: Caterva defines a metalayer for multidimensional data, but another library can add another metalayer for specifying the type.
   [https://github.com/Blosc/cat4py/blob/master/notebooks/ array-metalayer.ipynb]
- They are handy for defining layers with different specs: multi-dimensions, types, geo-spatial...



Multiple layers to target different data aspects

# Caterva MetaLayer

Caterva specifies a metalayer on top of a Blosc2 container for storing multidimensional information:

```
typedef struct {
    int8_t ndim;
    //! < The number of dimensions
    uint64_t dims[CATERVA_MAXDIM];
    //! < The size of each dimension
    int32_t pdims[CATERVA_MAXDIM];
    //! < The size of each partition dimension
} caterva_dims_t;</pre>
```

Such metalayer can be modified so that the shapes can be updated (e.g. an array can grow or shrink).

# Frame Format and MetaLayers Specs

- The format for a Blosc2 frame is completely specified at:
  - <u>https://github.com/Blosc/c-blosc2/blob/master/</u> <u>README\_FRAME\_FORMAT.rst</u>
- The format for a Caterva metalayer:
  - <u>https://github.com/Blosc/Caterva/blob/master/</u> <u>README\_CATERVA\_METALAYER.rst</u>

#### **Everything specified in the <u>msgpack</u> format.**

# IronArray

- High-performance matrix and vector computations.
- Optimized for compressed in-memory multi-dimensional data.
- Language specific API's leveraging a powerful C kernel.



### Example of IronArray API for Python

```
import iarray as ia
shape = [10 * 1000 * 1000]
pshape = [100 * 1000]
cparams = dict(clib=ia.LZ4, clevel=1, nthreads=2)
xa = ia.linspace(ia.dtshape(shape=shape, pshape=pshape_), 0., 10., **cparams)
# Evaluate a polynomial
ya = ((xa - 1.35) * (xa - 4.45) * (xa - 8.5)).eval()
```

- There are wrappers for Java too.
- An R wrapper could also be interesting in the future.

# IronArray Availability

- A beta version is planned for November.
- It will follow a production-ready version later in the year.
- Commercially available, with a possible Community Edition.

#### Stay tuned!

# Overview

- Caterva is a C library and a metalayer for handling multidimensional data on top of Blosc2 containers.
- cat4py leverages Caterva and Blosc2 for building multidimensional containers that can be very large.
- You can use metalayers for adapting Blosc2/Caterva containers to your own needs.
  - Beware: Do not reinvent the wheel and build on top of existing metalayers!

# Help Needed!

- Caterva and specially cat4py are quite new and still need a fair amount of love in terms of documentation, API consistency and testing in general before they are apt for general consumption.
- Please join us in our sprint on Friday morning if you are interested in helping us to reach production quality as soon as possible.

You are welcome!

# Acknowledgements

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- Last but not least, NumFOCUS for providing funding for developing Blosc2 and Caterva.









# Thank You!

#### **Questions?**



