

Hierarchical Data Format Version 5 (HDF5) Features, Tools, and Python Integration

Alvaro Aguilera alvaro.aguilera@tu-dresden.de http://tu-dresden.de/zih

Technische Universität Dresden

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WHAT IS HDF?

"The Hierarchical Data Format (HDF) is a set of file formats (HDF4, HDF5) designed to store and organize large amounts of numerical data."

- o stable, first versions in 1994
- ⊙ open source (BSD-equivalent license)
- widespread in the scientific community
- written in C, but supported in many programming languages
- lots of tools and know-how available

WIKIPEDIA
The Free Encyclopedia

Website: hdfgroup.org



Many scientific projects progress like this:

- Early development stage
 - small amounts of test data produced (e.g. some KB of particle information, time-series, etc.)
 - data stored as text in ASCII files (e. g. CSV)
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- ▷ **Solution:** binary format
 - ► faster I/O & small files
 - requires more programming and documentation effort



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 - · slow search across files
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▷ Solution:

consolidation of small files into bigger ones

compressed collections of groups

```
group101-100.tar.gz
group101-200.tar.gz
```



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 - ▷ implementation was time consuming
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 - implementation was time consuming
 - searching the data is still slow and tedious
 - usually poorly documented
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- **Improvement:** use an I/O-library like HDF5!



FILE ORGANIZATION

HDF5 files are containers for **objects**. An object can be either a **group** or a **dataset**.

A group is a collection of objects.

A dataset is a multidimensional array of data elements.

Attribute lists can be associated to any HDF5 object.

Groups define a hierarchical **path** similar to a directory structure:

```
/ root group
/foo object foo in group root
/foo/bar object bar in group foo in group root
```



EXPLORING AN HDF5 FILE WITH HDFView

Graphical tool to browse HDF4 & HDF5 files.

Several useful features like:

- File creation
- Add, edit and modify file data
- Integrated plotting capabilities
- Available for most computer platforms (Java)

To start the application open a terminal and use the command: hdfview



OTHER USEFUL TOOLS

- h5dump Dumps the contents of an HDF5 file to an ASCII file.
- h5ls Lists specified features in the file.
- h5diff Compares two files and report the differences.
- **h5repack** Copies a file to a new file with different options.
- h5repart Repartitions a file, creating a family of files.
- h5copy Copies objects from a file to a new file
- **h5mkgrp** Creates a new group in a file
- h5stat Reports statistics regarding a file and the objects in it.



HDF5 FOR PYTHON

- Available using the **h5py** library (http://h5py.org).
- Also distributed under BSD-like license.
- Used in this tutorial for simplicity.
- API follows to the traditional C interface.







CREATING A HDF5 FILE

file = h5py.File(name, mode='a', driver=None, libver=None,
userblock_size, **keywords)

Parameters:

- ⊙ **name:** file name or h5f.FileID
- ⊙ mode:
 - r Read-only, file must exist
 - r+ Read/write, file must exist
 - w Create file, truncate if exists
 - w- or x Create file, fail if exists
 - a Read/write if exists, create otherwise (default)
- o driver: desired driver
- libver: compatibility level
- userblock_size: Size in bytes of the user block (0 or power of $2 \ge 512$)
- **keywords:** options for the driver



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HDF5 FILE DRIVERS

file = h5py.File(name, mode='a', driver=None, libver=None,
userblock_size, **keywords)

Common drivers:

- None: Default and recommended. Usually unbuffered POSIX-I/O.
- ⊙ **core:** In-memory file
 - backing_store: False, contents are discarded when closing file. True, contents are saved to disk when closing file.
 - block_size: increment (in bytes) by which memory is extended (64KiB default)



FILE OBJECT

```
1 import h5py
2
3 f = h5py.File("test.h5", "w")
4
5 f.close()
```

Interesting object members:

- ⊙ **close()** file
- flush() buffers to disk
- o create_group(name)
- o create_dataset(name, options)



HDF5 Groups

```
group = File.create_group(name)
group = Group.create_group(name)
```

File object acts as the root group.

```
1 import h5py
2
3 f = h5py.File("example0.h5", "w")
4
5 group1 = f.create_group("group1")
6 subgroup1 = group1.create_group("subgroup1")
7 subgroup2 = f.create_group("/group2/subgroup2")
8 group2 = f["/group2"]
9
10 f.close()
```

HDF5 DATASETS

Homogeneous collections of data elements of immutable datatype.

```
dataset = Group.create_dataset(name, shape, dtype, data,
**kwds)
```

- o name: Name of the dataset
- shape: Dimensions of the array with data (tuple)
- dtype: Data type (optional, defaults to float)
- data: NumPy array with initial data (optional)
- o kwds: Additional parameters (optional)
 - chunks: chunking shape
 - ▷ compression: enabling dataset compression
 - maxshape: maximun dimension for resizing
 - ▶ **fillvalue:** default value when reading uninitialized data
 - ⊳ etc.



DATA TYPES

Prefixed with h5py.h5t.

- Floating point: IEEE_F32LE, IEEE_F32BE, IEEE_F64LE, IEEE_F64BE
- Integer: STD_I8LE, STD_I16LE, STD_I32LE, STD_I64LE.
 STD_U8LE, ... (also with BE)
- Strings: C.S1 (Null-terminated fixed-length string),
 FORTRAN.S1 (Zero-padded fixed-length string), VARIABLE (Variable-length string)
- ⊙ etc.



CREATING A DATASET

```
1 import h5py
2 #
3 # Create a new file using defaut properties.
4 #
5 file = h5py.File('example1.h5', 'w')
6 #
7 # Create a dataset under the Root group.
8 #
9 dataset = file.create_dataset("dset", (4, 6), h5py.h5t.STD_I32BE)
10 print("Dataset dataspace is", dataset.shape)
11 print("Dataset datatype is", dataset.dtype)
12 print("Dataset name is", dataset.name)
13 print("Dataset is a member of the group", dataset.parent)
14 print("Dataset was created in the file", dataset.file)
15 #
16 # Close the file before exiting
17 #
18 file.close()
```

Writting to a dataset

```
1 import h5py
 2 import numpy as np
 3
 4 # Open an existing file using default properties.
 5 file = h5py.File('example1.h5', 'r+')
 7 # Open "dset" dataset under the root group.
 8 dataset = file['/dset']
9 # Initialize data object with 0.
10 data = np.zeros((4,6))
11
12 # Assign new values
13 for i in range (4):
      for j in range(6):
14
           data[i][j] = i*6+j+1
17 # Write data
18 print ("Writing data...")
19 dataset [...] = data
20
21 # Read data back and print it.
22 print("Reading data back...")
23 data_read = dataset[...]
24 print ("Printing data...")
25 print(data_read)
26
27 # Close the file before exiting
28 file.close()
```



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CHUNKING

By default HDF5 creates **contiguous** datasets.

Chunking divides the contiguous data into **chunks** that are irregularly stored on disk and indexed using a B-tree.

Example in which the data will be read and written in blocks with shape (100,100):

```
dset = f.create_dataset("chunked", (1000, 1000),
chunks=(100, 100))
```

Automatic chunking can be enabled using chunks=True. The recommended size for the chunk are between 10 KiB and 1 MiB.



CREATING COMPRESSED FILE

```
1 import h5py
 2 import numpy as np
 4 # Create hdf file
 5 file = h5py.File('example2.h5','w')
 7 # Create /DS1 dataset; in order to use compression, dataset has to be chunked.
 8 dataset = file.create_dataset('DS1', (32,64), 'i', chunks=(4,8), compression='gzip',
         compression_opts=9)
10 # Initialize data
11 data = np.zeros((32,64))
12 for i in range (32):
      for j in range (64):
14
           data[i][j]= i*j-j
15
16 # Write data.
17 print("Writing data...")
18 dataset [...] = data
19
20 file.close()
21
22 # Read data back; display compression properties and dataset max value.
23 file = h5py.File('example2.h5','r')
24 dataset = file['DS1']
25 print ("Compression method is", dataset.compression)
26 print ("Compression parameter is", dataset.compression_opts)
27 data = dataset[...]
28 file.close()
```



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CREATE ATTRIBUTES

HDF5 is a self-describing format.

This is achieved through attributes (metadata) attached to the groups and datasets.

```
1 import h5py
 2 import numpy as np
 4 # Open an existing file using defaut properties.
 6 file = h5py.File('example1.h5', 'r+')
 8 # Open "dset" dataset.
10 dataset = file['/dset']
12 # Create string attribute.
14 attr_string = "Meter per second"
15 dataset.attrs["Units"] = attr_string
17 # Close the file before exiting
19 file.close()
```



PRACTICAL SESSION

- Create an HDF5 file with a 100x100 integer dataset inside the group "experiment1"
- Initialize each element to the sum of its coordinates.
- Measure the execution time for the default and core driver (with and without storing on close)
- Turn on compression and compare time and resulting size using different chunking shapes.



SOURCES AND REFERENCES



[1] The HDF Group

HDF5 Tutorial

https://www.hdfgroup.org/HDF5/Tutor



[2] National Energy Research Scientific Computing Center (NERSC)

Introduction to Scientific I/O

http://www.nersc.gov/users/training/online-tutorials/introduction-to-scientific-i-o



[3] Andrew Collette et al.

HDF5 for Python

http://h5py.org



THANK YOU FOR YOUR ATTENTION



More information: http://www.sppexa.de
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