

An <u>on-disk</u> binary data container, query engine and computational kernel

Overview

What PyTables is?

Data structures in PyTables



Compressing data

Advanced capabilities in PyTables

What it is

A binary data container for on-disk, structured data

Can perform operations with data *on-disk*

Based on the standard de-facto HDF5 format

Free software (BSD license)

About HDF5 (Hierarchical Data File version 5)

- A versatile data model that can represent complex data objects as well as associated metadata
- A portable file format with no limit on the number or size of data objects in the collection
- Implements a high-level API with C, C++, Fortran 90, and Java interfaces
- Free software (BSD, MIT kind of license)

PyTables distinctive features

 Supports a good range of compressors: Zlib, bzip2, LZO and Blosc

 Powerful query capabilities for Table objects, including indexing

Can perform out-of-core operations very efficiently

What it is not

Not a relational database replacement

Not a distributed database

 Not extremely secure or safe (it's more about speed!)

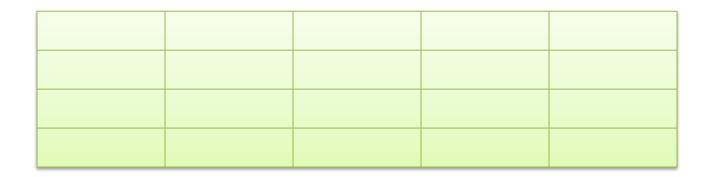
Not a mere HDF5 wrapper

DATA STRUCTURES

Data structures

- High level of flexibility for structuring your data:
 - Datatypes: scalars (numerical & strings), records, enumerated, time...
 - Tables support multidimensional cells and nested records
 - Mutidimensional arrays
 - Variable length arrays

The Array object



- Easy to create:
 - file.createArray(mygroup, 'array', numpy_arr)
- Shape cannot change
- Cannot be compressed

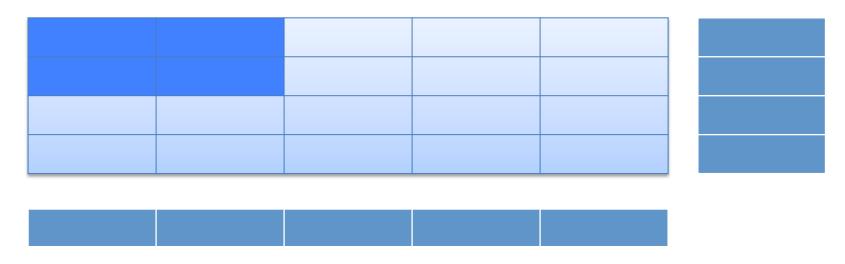
The CArray object



- Data is stored in chunks
- Each chunk can be compressed independently

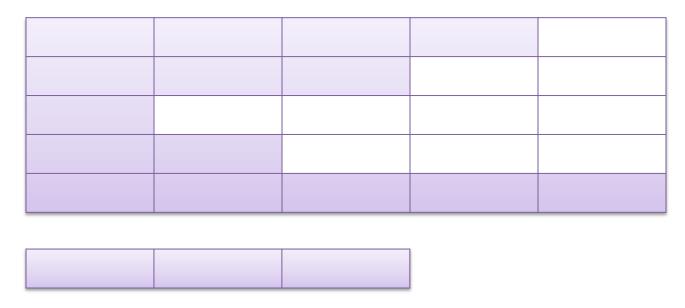
Shape cannot change

The EArray object



- Data is stored in chunks
- Can be compressed
- Shape can change (either enlarged or shrunk)
- Shape must be kept regular

The VLArray object



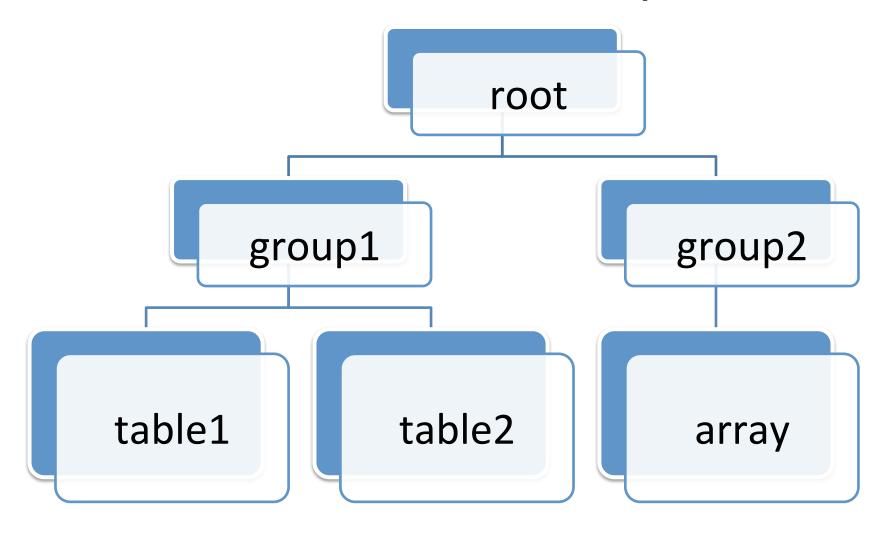
- Data is stored in variable length rows
- Can be enlarged or shrunk
- Data cannot be compressed

The Table object

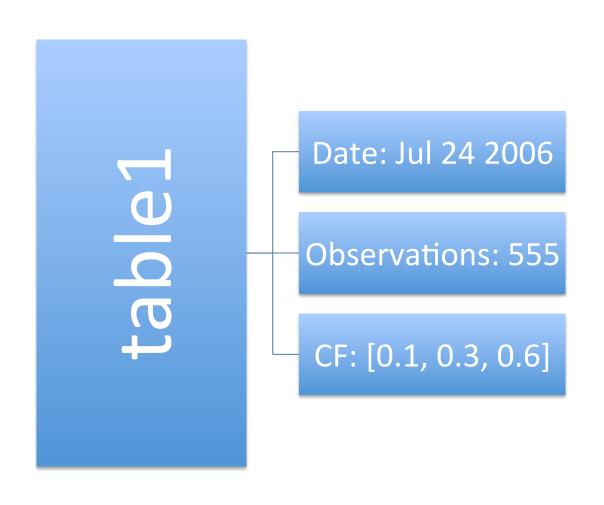
Col1 (int32)	Col2 (string 10)	Col3 (bool)	Col4 (complex64)	Col5 (float32)

- Data is stored in chunks
- Can be compressed
- Can be enlarged or shrunk
- Fields cannot be of variable length

Dataset hierarchy



Attributes: Metadata about data



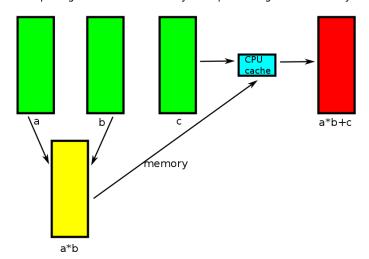
OUT-OF-CORE COMPUTATIONS

Operating with disk-based arrays

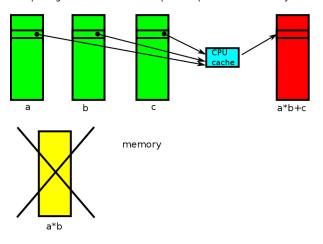
- tables.Expr is an optimized evaluator for expressions of disk-based arrays.
- It is a combination of the Numexpr advanced computing capabilities with the high I/O performance of PyTables.
- Similarly to Numexpr, disk-temporaries are avoided, and multi-threaded operation is preserved.

Avoiding temporaries with Numexpr

Computing "a*b+c" with NumPy. Temporaries goes to memory.

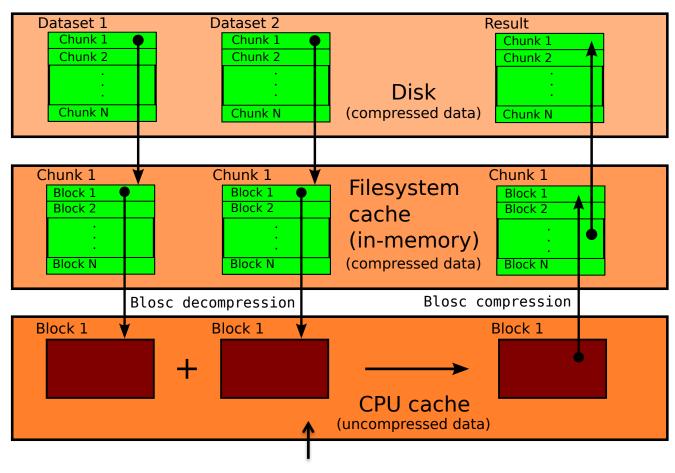


Computing "a*b+c" with Numexpr. Temporaries in memory are avoided.



Tables.Expr follows the same approach, but with disk instead of memory

Performing out-of-core computations with PyTables



Virtual machine: numexpr

ADVANCED QUERY CAPABILITIES

Different query modes

Regular query:

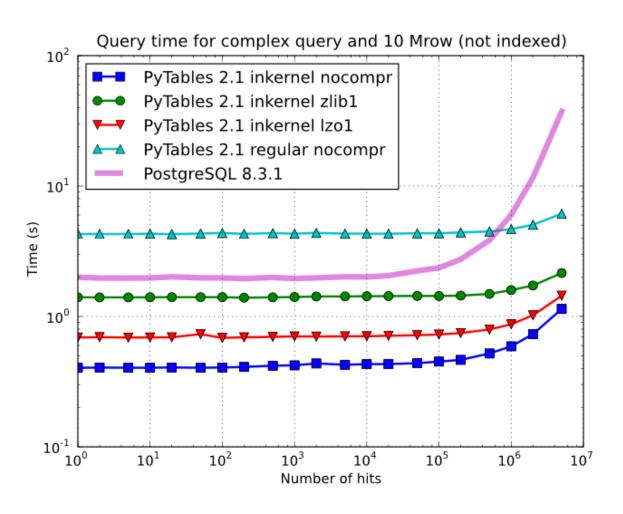
In-kernel query:

```
• [ r['c1'] for r in table.where('(c2>2.1)&(c3==True)') ]
```

Indexed query:

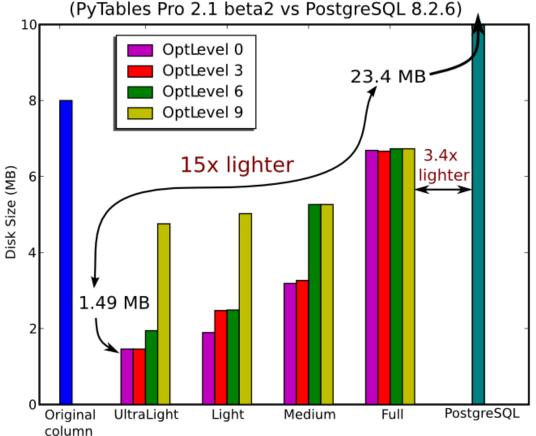
- table.cols.c2.createIndex()
- table.cols.c3.createIndex()
- [r['c1'] for r in table.where('(c2>2.1)&(c3==True)')]

Regular and in-kernel queries

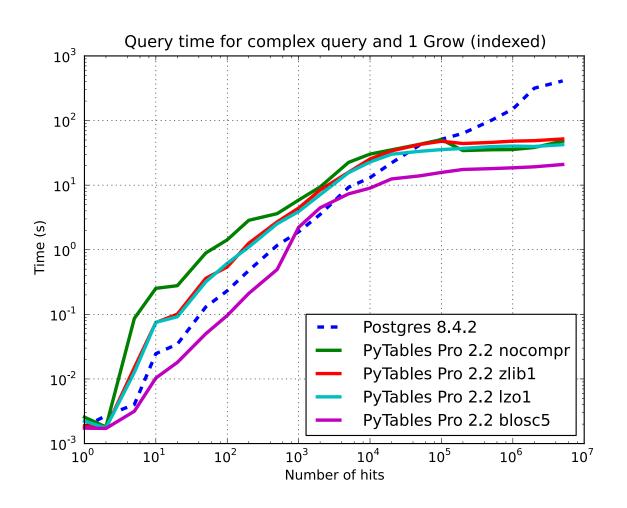


Customizable indexes

Sizes for index of a 1 Grow column with different optimizations (PyTables Pro 2.1 beta2 vs PostgreSQL 8.2.6)



Indexed query performance



Concepts to take home

PyTables is optimized to deal with data on disk

 Most of the operations use the iterator/ generator machinery in Python: the goal is not to bloat memory with data

 Queries, indexes and out-of-core operations are good examples of the above