PyTables User's Guide

Francesc Alted

2002. October, 04th

Abstract

PyTables is a Python package which allows dealing with HDF5 tables. Such a table is defined as a collection of records whose values are stored in fixed-length fields. PyTables is intended to be easy-to-use, and tries to be a high-performance interface to HDF5. To achieve this, the newest improvements introduced in Python 2.2 (like generators or slots and metaclasses in new-brand classes) has been used. Pyrex creation extension tool has been choosen to access the HDF5 library.

1 Introduction

PyTables is a Python package which allows dealing with HDF5 tables. In this document, the term *table* means exactly the same than in HDF5 sense:

"A table is defined as a collection of records whose values are stored in fixed-length fields. All records have the same structure and all values in each field have the same data type."

Records in tables are also known, in the HDF5 naming scheme, as *compound* datatypes. So, you can define arbitrary records in Python, like in:

```
class Particle(IsRecord):
```

```
name = '16s' # 16-character String
lati = 'i' # integer
longi = 'i' # integer
pressure = 'f' # float (single-precision)
temperature = 'd' # double (double-precision)
```

fill it with your values, and save (large) collections of them in a file. Then, this data can be retrieved and post-processed quite easily with PyTables or another HDF5 application.

You probably noted that the terms "fixed-length" and strict "data types" present in the table definition seems to be strange concepts for an interpreted language like Python, but supporting them is fundamental when we want to save *lots* of data (mainly for scientific applications, but not only that), if we want to do that in a efficient (both in terms of CPU and I/O requeriments) way. PyTables allows that.

1.1 Features

PyTables has the next features:

- Support of HDF5 table entities: Allows working with large number of records which don't fit in memory.
- Supports a hierarchical data model: So, you can structure very clearly all your data. This is also very important when dealing with XML data.
- Allows field name, data type and range checkings: So you can be confident that if PyTables doesn't report an error, you can be confident that your data is probably ok.

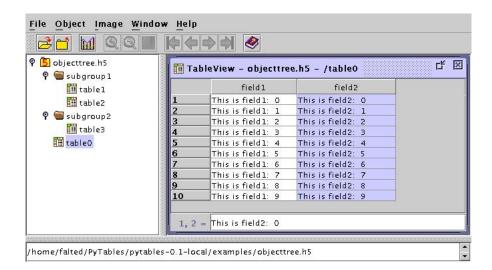


Figure 1: An HDF5 example with 2 subgroups and 3 tables.

- Support of files bigger than 2 GB: This is because HDF5 already can do that (if your platform suports the C long long integer, or, on Windows, __int64).
- *Big-Endian/Low-Endian safety:* PyTables has been coded (as HDF5 is) to care with little-endian/big-endian byte orderings. So, in principle, you can write a file in a big-endian machine and read it in other little-endian without problems ¹.

It's important to stress that PyTables doesn't support every functionality present on HDF5, so you would be able to create HDF5 files with PyTables, and read them with other HDF5 generic tools (like h4dump), but you can't hope PyTables can read every HDF5 file created with tools different than PyTables, as it supports only table objects (but I hope to extend the support to other objects too, like NumArray entities).

In the same sense, it should noted that PyTables do not pretends to be merely a wrapper of HDF5_HL library (don't confuse with HL-HDF5, the Swedish Meteorological and Hydrological Institute effort to provide another high Level interface to HDF5), but to provide a flexible tool to deal with HDF5 files. This is achieved by taking advantage of the powerful object orientation and introspection capabilities offered by Python.

1.2 The object tree

PyTables take advantage of the HDF5 hierarchical model to allow tables to be managed in a tree-like structure. It achieves that by creating an object tree imitating the HDF5 structure on disk. That way, the access to the HDF5 objects is made by walking throughout the PyTables object tree, and manipulating them. You should note that not all the data present on the HDF5 file is loaded in PyTables tree, but only the *metadata* (i.e. the data which describes the structure of the real data). The actual access to the real data is provided through the methods of those objects.

For example, imagine we have made a script (in fact, this script exists; its name is objecttree.py and you can find it in the examples/directory) which has created a simple HDF5 file, with the structure that appears in figure 1 (we have used the java program hdfview to obtain this image. If you re-open again this file (in read only mode, for example), the object tree with the HDF5 metadata will be constructed from this hierarchy.

In figure 2 you can see an example of the object tree created by reading an HDF5 file (previously written with PyTables). It's important that you get familiar with this diagram to better understand how to work with PyTables. If you are going to be a PyTables user, take your time to study and understand it (bear in mind, however, that this diagram is not a standard UML class diagram; I've used a UML tool to draw it, that's all).

¹ Well, I didn't actually test that in real world, but if you do, please, tell me

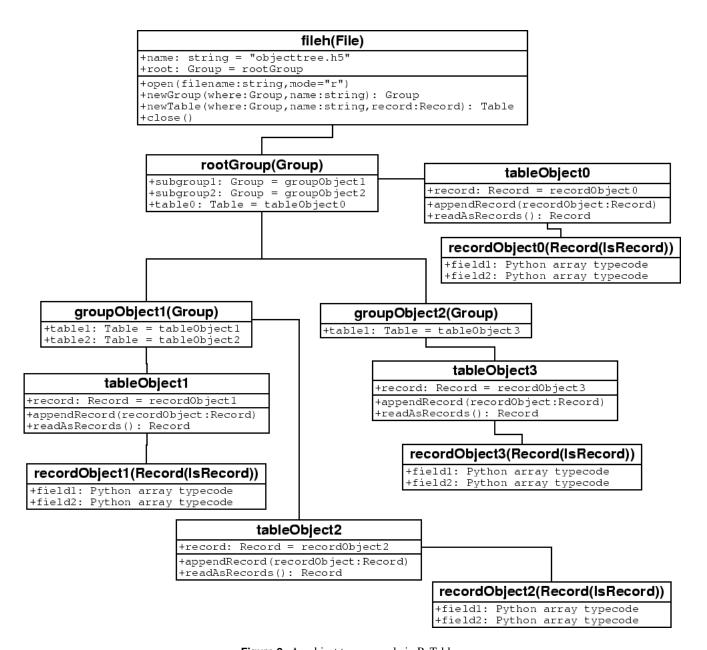


Figure 2: An object tree example in PyTables.

It's important to bear all this in mind while you are working with PyTables, because it will easy your work and will make you more proactive by avoiding programming mistakes.

2 Installation

This are instructions for Unix/Linux system. If you are using Windows, and get the library working, please, tell me about.

The Python Distutils are used to build and install tables, so it is fairly simple to get things ready to go.

1. First, make sure that you have hdf5 1.4.x and hdf5_hl libraries installed (I'm using hdf5 1.4.4 and hdf5_hl beta2 currently). If not, you can find them at http://hdf.ncsa.uiuc.edu/HDF5; compile/install them.

setup.py will detect these libraries and include files under either /usr or /usr/local; this will catch installations from RPMs and most hand installations under Unix. If setup.py can't find your libhdf5 and libhdf5_hl or if you have several versions installed and wants to select one of them, then you can give it a hint either in the environment (using the HDF5_DIR evironment variable) or on the command line by specifying the directory containing the include and lib directory. For example:

```
--hdf5=/stuff/hdf5-1.4.4
```

The libraries can installed everywhere on the filesystem, but remeber to always place them together. For example, if libhdf5.so is installed in /usr/lib, so does hdf5_hl. The same applies to the headers.

If your HDF5 libs were built as shared libraries, and if these shared libraries are not on the runtime load path, then you can specify the additional linker flags needed to find the shared library on the command line as well. For example:

```
--lflags="-Xlinker -rpath -Xlinker /stuff/hdf5-1.4.4/lib"
```

or perhaps just

```
--lflags="-R /stuff/hdf5-1.4.4/lib"
```

Check your compiler and linker documentation to be sure.

It is also possible to specify linking against different libraries with the --libs switch:

```
--libs="-lhdf5-1.4.6 -lhdf5_hl-beta2"
--libs="-lhdf5-1.4.6 -lhdf5_hl-beta2 -lnsl"
```

2. From the main pytables distribution directory run this command, (plus any extra flags needed as discussed above):

```
python setup.py build_ext --inplace
```

depending on the compiler flags used when compiling your Python executable, it may appear lots of warnings. Don't worry, almost all of them are caused by variables declared but never used. That's normal in Pyrex extensions.

3. To run the test suite change into the test directory and run this command, (assuming your shell is bash or compatible):

```
export PYTHONPATH=..
python test_all.py
```

If you would like to see some verbose output from the tests simply add the word verbose to the command line. You can also run only the tests in a particular test module by themselves. For example:

```
python test_types.py
```

4. To install the entire PyTables Python package, change back to the root distribution directory and run this command as the root user:

python setup.py install

That's it!. Now, read on the next section to see some program examples.

3 Usage

3.1 A first example

Let's start by showing a simple example. For simplicity and direct comparison, I'll choose the same that is exposed in an HDF5_HL example (http://hdf.ncsa.uiuc.edu/HDF5/hdf5_hl/doc/RM_hdf5tb_ex.html).

So, we want to create a table whose records are particle properties. Each particle (record) has a name, a position (specified by latitude and longitude), pressure and temperature.

We start by define this record in PyTables by declaring a subclass of IsRecord. But first, the necessary imports:

As you see, we define the Particle class as a subclass of IsRecord which is a metaclass, but this is not important now. The name of each Particle attribute will be the name of the record field and its value will become its datatype. '16s' typecode means a 16-character string, 'i' an integer, 'd' a double, and so on. For a complete list of data types supported see table 1.

Now, we open an HDF5 file in write mode:

```
fileh = File(filename = "example1.h5", mode = "w")
    and get the object which is the root directory in HDF5 hierarchy:
group = fileh.getRootGroup()
    then, create a new table object

table = fileh.newTable(group, 'table', Particle(), "Title example")
    get the the Particle instance associated with the table
particle = fileh.getRecordObject(table)
    and fill the table with 10 particles
```

```
for i in xrange(10):
    # First, assign the values to the Particle record
    particle.name = '%16d' % i
    particle.lati = i
    particle.longi = i
    particle.pressure = float(i)
    particle.temperature = float(i)
    # This injects the Particle values
    fileh.appendRecord(table, particle)
    and finally, close the file:
fileh.close()
```

That's it!. We can see here the complete example for a better inspection, with a few additional comments:

```
from tables import File, IsRecord
class Particle(IsRecord):
   name
               = '16s' # 16-character String
                       # integer
    lati
               = 'i'
    longi
              = 'i'
                       # integer
    pressure = 'f'
                        # float (single-precision)
    temperature = 'd'
                       # double (double-precision)
# Open a file in "w"rite mode
fileh = File(name = "example1.h5", mode = "w")
# Get the HDF5 root group
root = fileh.getRootGroup()
# Create a new table
table = fileh.newTable(root, 'table', Particle(), "Title example")
#print "Table name ==>", table._v_name
# Get the record object associated with the table: all three ways are valid
#particle = table.record
particle = fileh.getRecordObject(table) # This is really an accessor
#particle = fileh.getRecordObject("/table")
# Fill the table with 10 particles
for i in xrange(10):
    # First, assign the values to the Particle record
   particle.name = 'Particle: %6d' % (i)
   particle.lati = i
   particle.longi = 10 - i
   particle.pressure = float(i*i)
    particle.temperature = float(i**2)
    # This injects the Record values. Both ways do that.
    #table.appendRecord(particle)
    fileh.appendRecord(table, particle)
# Finally, close the file
fileh.close()
```

In figure 3 you can see the table we have created in this example. You will find in the directory examples the working version of the code (source file example1.py).

3.2 A somewhat more complex exercise

Now, time for a more sophisticated example. Here, we will create a couple of directories (groups, in HDF5 jargon) hanging directly from the root directory called Particles and Events. Then, we will put 3 tables in

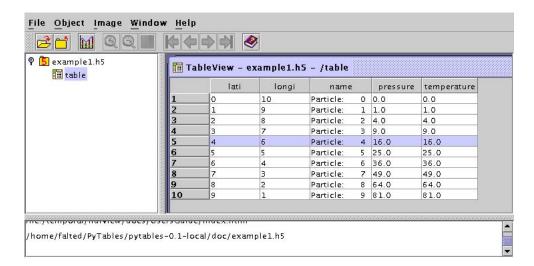


Figure 3: A simple table in HDF5.

each group; in Particles we will put instances of Particle records and in Events, instances of Event. After that, we will feed the tables with 257 (you will see soon why I choose such a "esoteric" number) entries each. Finally, we will read the recently created table /Events/TEvent3 and select some values from it using a comprehension list.

Lets go,

```
from tables import File, IsRecord
class Particle(IsRecord):
                = '16s' # 16-character String
    name
                = 'i'
    lati
                         # integer
    longi
                = 'i'
                         # integer
                = 'f'
                         # float (single-precision)
    pressure
    temperature = 'd'
                         # double (double-precision)
class Event(IsRecord):
    name
               = '16s' # 16-character String
    TDCcount
               = 'B'
                         # unsigned char
    ADCcount
              = 'H'
                         # unsigned short
                = 'f'
                         # float (single-precision)
    xcoord
                = 'f'
                         # float (single-precision)
    ycoord
# Open a file in "w"rite mode
fileh = File(name = "example2.h5", mode = "w")
# Get the HDF5 root group
root = fileh.getRootGroup()
# Create the groups:
for groupname in ("Particles", "Events"):
    group = fileh.newGroup(root, groupname)
# Now, create and fill the tables in Particles group
gparticles = fileh.getNode("/Particles")
# You can achieve the same result with the next notation
# (it can be convenient and more intuitive in some contexts)
#gparticles = root.Particles
# Create 3 new tables
for tablename in ("TParticle1", "TParticle2", "TParticle3"):
    # Create a table
```

```
table = fileh.newTable("/Particles", tablename, Particle(),
                           "Particles: "+tablename)
    # Get the record object associated with the table:
    particle = fileh.getRecordObject(table)
    # Fill the table with 10 particles
    for i in xrange(257):
        # First, assign the values to the Particle record
        particle.name = 'Particle: %6d' % (i)
        particle.lati = i
        particle.longi = 10 - i
        particle.pressure = float(i*i)
        particle.temperature = float(i**2)
        # This injects the Record values
        fileh.appendRecord(table, particle)
    # Flush the table buffers
    fileh.flushTable(table)
# Now, go for Events:
for tablename in ("TEvent1", "TEvent2", "TEvent3"):
    # Create a table. Look carefully at how we reference the Events group!.
    table = fileh.newTable(root.Events, tablename, Event(),
                           "Events: "+tablename)
    # Get the record object associated with the table:
    event = table.record
    # Fill the table with 10 events
    for i in xrange(257):
        # First, assign the values to the Event record
        event.name = 'Event: %6d' % (i)
        #event.TDCcount = i
        event.ADCcount = i * 2
        event.xcoor = float(i**2)
        event.ycoord = float(i**4)
        # This injects the Record values
        fileh.appendRecord(table, event)
    # Flush the buffers
    fileh.flushTable(table)
# Read the records from table "/Events/TEvent3" and select some
e = [ p.TDCcount for p in fileh.readRecords("/Events/TEvent3")
      if p.ADCcount < 20 and 4<= p.TDCcount < 15 ]
print "Last record ==>", p
print "Selected values ==>", e
print "Total selected records ==> ", len(e)
# Finally, close the file (this also will flush all the remaining buffers!)
fileh.close()
```

Throughout the comments, you can see that PyTables let's you do things in, generally, more than one way. I don't know if that's good or not, but I'm afraid it is not. This is in part due to the fact that PyTables is in first stages of development, and probably as the API matures, there will be less choices.

If you have read the code carefully it looks pretty good, but it won't work. If you run this example, you will get the next error:

```
Traceback (most recent call last):
   File "example2.py", line 68, in ?
     event.xcoor = float(i**2)
AttributeError: 'Event' object has no attribute 'xcoor'
```

This error is saying us that we tried to assign a value to a non-existent field in an Event object. By looking carefully at the Event attributes, we see that we mis-spelled the xcoord field (we wrote xcoor instead). So we correct this in the source, and run it again.

And again, we find another problem:

```
Traceback (most recent call last):
    File "example2.py", line 69, in ?
        table.appendRecord(event)
File "/usr/lib/python2.2/site-packages/tables/Table.py", line 210, in appendRecord
        self._v_packedtuples.append(recordObject._f_pack2())
File "/usr/lib/python2.2/site-packages/tables/IsRecord.py", line 121, in _f_pack2
        self._f_raiseValueError()
File "/usr/lib/python2.2/site-packages/tables/IsRecord.py", line 130, in _f_raiseValueError
        raise ValueError, \
ValueError: Error packing record object:
[('ADCcount', 'H', 256), ('TDCcount', 'B', 256), ('name', '16s', 'Event: 256'), ('xcoord
        Error was: ubyte format requires 0<=number<=255</pre>
```

This other error is saying that one of the records is having trouble to be converted to the datatypes stated in the Event class definition. By looking carefully to the record object which is responsible of the problem, we see that we are trying to assign a value of 256 to the 'TDCcount' field which has a 'B' (C unsigned char) typecode and the allowed range for it is 0<=TDCcount<=255. This is a very powerful capability to automatically check for ranges: the message error is explicit enough to figure out what is happening. Here you can solve the problem by promoting the TDCcount to 'H' which is a unsigned 16-bit integer, or avoid the mistake you probably made in assigning a value greater than 255 to a 'B' typecode.

If we change the line:

```
event.TDCcount = i
  by the next one:
event.TDCcount = i % (1<<8)</pre>
```

you will see that our problem has disappeared, and the HDF5 file has been created. As before, you will find in the directory examples the working version of the code (source file example2.py).

Finally, admire the structure we have created in figure 4.

Feel free to visit the rest of examples in directory examples, and try to understand them. I've tried to make the cases as ortogonal as possible to give you an idea of the PyTables capabilities and its way of dealing with HDF5 objects.

4 Library Reference

This package implements an important class to deal with HDF5 files, called File, one class to help defining records, with field, type and range checks, which is called IsRecord. There exists other important classess called Group and Table which do their work silently behind the scenes. The user has to be aware of its existence, but generally speaking, they won't need to call their methods explicitly.

4.1 The File class.

The File class hosts the most part of PyTables user interface. It is in charge of create, open, flush and close the HDF5 files. In addition it provides accessors to functionality present in Group and Table classes.

This class defines the next methods²:

² On the following, the term Leaf will refer to a Table instance. Right now the only supported Leaf object is Table, but that will change in the short future.

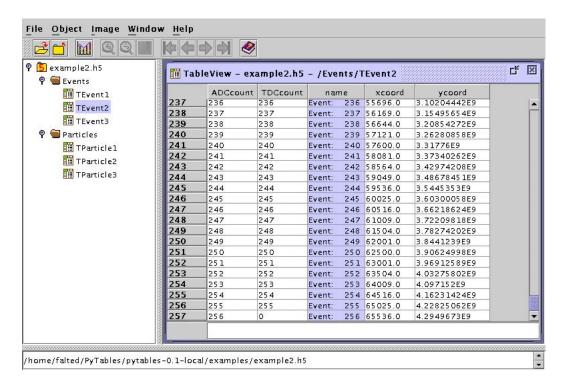


Figure 4: Tables structured in a hierarchical order.

File(filename, mode="r"): This is the constructor and opens an HDF5 file. The supported access modes are: "r" means read-only; no data can be modified. "w" means write; a new file is created, an existing file with the same name is deleted. "a" means append (in analogy with serial files); an existing file is opened for reading and writing, and if the file does not exist it is created. "r+" is similar to "a", but the file must already exist.

getRootGroup(): Returns a Group instance which will act as the root group in the hierarchical tree. If file is opened in "r", "r+" or "a" mode, and the file already exists, this method dynamically builds a python object tree emulating the structure present on file. It *must* be always called after the File object is constructed.

newTable(where, name, tableTitle = "", compress = 1, expectedrows = 10000): Returns a new Table instance with name *name* in *where* location. *where* parameter can be a path string, or another group instance. Other optional parameters are: *tableTitle* which set a TITLE attribute on the HDF5 table entity. *compress* is a boolean option and specifies if data compression will be enabled or not. *expectedrows* is an user estimate about the number of records that will be on table. This parameter is used to set important internal parameters, as buffer size or HDF5 chunk size. If not provided, the default value is appropriate to tables until 100 KB in size. If you plan to save bigger tables by providing a guess to PyTables will optimize the HDF5 B-Tree creation and management process time and memory used.

newGroup(where, name): Returns a new Group instance with name *name* in *where* location. *where* parameter can be a path (for example "/Particles/TParticle1" string, or another Group instance.

getNode(where): Returns the object node (Group or Leave) in *where* location. *where* can be a path string, Group instance or a Table instance.

getGroup(where): Returns the object group in *where* location. *where* can be a path string or a Group instance. If *where* doesn't point to a Group, a ValueError error is raised.

getTable(where): Returns the object table in *where* location. *where* can be a path string or a Table instance. If *where* doesn't point to a Table, a ValueError error is raised.

listNodes(where): Returns all the object nodes (groups or tables) hanging from *where*. *where* can be a path string or group instance.

listGroups(where): Returns all the groups hanging from where. where can be a path string or group instance.

listLeaves(where): Returns all the Leaves objects hanging from *where*. *where* can be a path string or group instance.

walkGroups(where): Recursively obtains groups (not leaves) hanging from where.

getRecordObject(table): Returns the record object associated with the *table*. *table* can be a path string or table instance.

appendRecord(table, record): Append the *record* object to the *table* output buffer. *table* can be a path string or table instance.

readRecords(table): Generator thats return a Record instance from a *table* object each time it is called. *table* can be a path string or table instance.

flushTable(table): Flush the table object to disk. *table* can be a path string or table instance.

flush(): Flush the buffers for all the objects on the HDF5 file tree.

close(): Flush all the objects in HDF5 file and close the file.

4.2 The IsRecord class.

This class is in fact a so-called *metaclass* object. There is nothing special on it, except that their subclasses attributes are transformed during its construction phase, and new methods for the are defined based on the values of the attributes. In that way, we can *force* the resulting instance to only accept assignments on the declared attributes (in fact, it has a few more, but they are hidden with prefixes like "___", "_v_" or "_f_", so please, don't use attributes names starting with these prefixes). If you try to do an assignment to a non-declared attribute, PyTables will raise an error.

To use such a particular class, you have to declare a descendent class from *IsRecord*, with many attributes as fields you want in your record. To declare their types, you simply assign to these attributes their *typecode*. That's all, from now on, you can instantiate objects from you new class and use them as a very flexible record object with safe features like automatic name field, data type and range checks (see the section 3.2 for an example on how it works).

The supported data types are the same which are supported by the array module in Python, with some additions, which will be briefly discussed shortly. The typecodes for the supported data types are listed on table 1.

The additions to the array module typecodes are the 'q', 'Q' and 's'. The 'q' and 'Q' conversion codes are available in native mode only if the platform C compiler supports C long long, or, on Windows, __int64. They are always available in standard modes. The 's' typecode can be preceded by an integer to indicate the maximum length of the string, so '16s' represents a 16-byte string.

Also note that when the 'I' and 'L' codetypes are used in records, Python uses internally Long integers to represent them, which can (or can not, depending on what you are trying to do) be a source of inefficiency in your code.

Type Code	Description	С Туре	Size (in bytes)	Python Counterpart
,c,	8-bit character	char	1	String of lenght 1
'b'	8-bit integer	signed char	1	Integer
'B'	8-bit unsigned integer	unsigned char	1	Integer
'h'	16-bit integer	short	2	Integer
'H'	16-bit unsigned integer	unsigned short	2	Integer
'i'	integer	int	4 or 8	Integer
'I'	unsigned integer	unsigned int	4 or 8	Long
'1'	long integer	long	4 or 8	Integer
'L'	unsigned long integer	unsigned long	4 or 8	Long
'q'	long long integer	long long	8	Long
'Q'	unsigned long long integer	unsigned long long	8	Long
'f'	single-precision float	float	4	Float
'd'	double-precision float	double	8	Float
's'	arbitrary lenght string	char[]	*	String

 Table 1: Data types supported by PyTables