Histogramming

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Overview

- Part 1: Overview of histograms
 - Components of a Histogram
 - Histograms in Python
 - ▶ Boost.Histogram in C++14
 - Introducing: boost-histogram for Python
 - Outlook, with hist and aghast
- Part 2: Hands-on with boost-histogram





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What is a histogram?

- A histogram is a set of accumulators over data in ranges
 - Usually continuous in Physics, could also be categories
 - Accumulators often are a sum of values can contain other components
- Input values are digitized by axes (AKA binnings)
 - Categories
 - Real values
 - Variable sized bins (usually give edges)
 - Regular binning (#bins, start, stop)
 - May have special features (overflow, circular, etc.)





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Histogram components

A 'histogram is a collection of 1+ axes and an accumulator.

Performance

- Variable axis list of edges is most general but requires a sorted search.
- Regular axis: regular spacing







```
Histograms in (classic) PyROOT
```

• 1D Regular

```
h = ROOT.TH1D("", "", 10, 0, 1)
h.fillN(arr)
```

- 1D Variable
- h = ROOT.TH1D("", "", (1,2,3,4,5,6))
 h.fillN(arr)
 - 2D Regular

```
h = ROOT.TH2D("", "", 10, 0, 1, 20, 0, 2)
h.fillN(arr)
```





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Histogram in Numpy

• 1D Regular

bins, edges = np.histogram(arr, bins=10, range=(0,1))

• 1D Variable

bins, edges = np.histogram(arr, bins=(1,2,3,4,5,6))

```
• 2D regular
```

b, e1, e2 = np.histogram2d(x, y, bins=(10,20), range=((0,1),(0,2)))





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Numpy Pros and Cons

Pros

- Comes with Numpy
- Good for interactive operations (auto binning)
- Reasonably fast
- Density option, weight support too

Cons

- Manipulation of plain arrays
- One time fill
- 2D+ not optimized for regular binning
- 1D, 2D, and ND syntax variations
- MPL had to mimic: plt.hist





PyROOT Pros and Cons

Pros

- Full histogram object
- Iterative fill option
- Weights option
- Can track sum of weights too

Cons

- ROOT requirement (Conda-forge helps)
- Can be slow in Python (and C++)
- Poor interactive exploration
- Odd syntax, odd memory model
- Max 3D





Histogram Libraries

- Narrow focus: speed, plotting, or language
- Many are abandoned
- Often issues with design, backends, distribution
- No/little interaction

theodoregoetz matr	olotlib-hep			
rootplotlib NUN pyhistogram	npy			
YODA	Histogrammar			
PyROOT fast-histogram				
SimpleHist	HistBook			

paida multihist





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physt Vaex

Physt

- Histograms as objects
- Pure Python Dropped Python 2 this year :)
- Very slow fills (slower than numpy)

hist = histogram(heights)
hist.plot(show_values=True)

- Powerful plotting
- Easy conversion to Pandas and many more (ROOT through uproot)
- Special histograms, like polar histograms



Figure 1: Physt example default plot





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Fast-Histogram

- Exactly like numpy, but faster
 - C kernel
 - Takes advantage of regular binning
 - Can be 20-25x faster for 2D histograms
 - Missing some features / combinations









Histogramming

HistBook (archived)



Histogramming

The first Scikit-HEP library for histograms

- Designed for shared axis histogram collections
- Plotting with Vega-Light

Now deprecated and in archive mode, functionality may return in Hist (see next slides).

```
>>> array = np.random.normal(0, 1, 1000000)
>>> histogram = Hist(bin("data", 10, -5, 5))
>>> histogram.fill(data=array)
>>> histogram.step("data").to(canvas)
```





SciKit-HEP Histogramming plan

- boost-histogram: Fast filling and manipulation (core library)
- hist: Simple analysis frontend
- aghast: Conversions between histogram libraries
- UHI: Unified Histogram Indexing: A way for histograms to be indexed cross-library (boost-histogram and hist to begin with)







Histogramming

Boost.Histogram C++14

- Multidimensional templated header-only histogram library: **O**/boostorg/histogram
- Designed by Hans Dembinski, inspired by ROOT and GSL







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Boost.Histogram example

```
#include <boost/histogram.hpp>
#include <boost/histogram/ostream.hpp>
#include <random>
int main() {
    namespace bh = boost::histogram;
    auto hist = bh::make histogram(bh::axis::regular<>{20, -3, 3});
    std::default_random_engine eng;
    std::normal_distribution<double> dist(0, 1);
    for(int n = 0; n < 10'000; ++n)
        hist(dist(eng));
    std::cout << hist << std::endl;</pre>
    return 0;
}
```





Boost.Histogram example (output)

histogram(regular(20, -3, 3, options=underflow | overflow))

+			
[-inf,	-3)	9	
[-3,	-2.7)	19	=
[-2.7,	-2.4)	36	==
[-2.4,	-2.1)	110	=====
[-2.1,	-1.8)	191	======
[-1.8,	-1.5)	275	=========
[-1.5,	-1.2)	518	
[-1.2,	-0.9)	644	======================================
[-0.9,	-0.6)	914	
[-0.6,	-0.3)	1107	
[-0.3,	0)	1183	
[0,	0.3)	1185	
[0.3,	0.6)	1120	
[0.6,	0.9)	874	
[0.9,	1.2)	663	======
[1.2,	1.5)	491	
[1.5,	1.8)	322	======================================
[1.8,	2.1)	172	
[2.1,	2.4)	79	====
[2.4,	2.7)	38	==
[2.7,	3)	28	=
[3,	inf)	22	=
			+





Histogramming

boost-histogram: Python bindings

Design

- A histogram should be an object
- Manipulation and plotting should be easy

Performance

- Fast filling
- Compiled composable manipulations

Flexibility

- Axes options: sparse, growing, labels
- Storage: integers, weights, errors...

Distribution

- Easy to use anywhere, pip or conda
- Should have wheels, be easy to build, etc.





Intro to the Python bindings

- Boost.Histogram developed with Python in mind
- Original bindings based on Boost::Python
 - Hard to build and distribute
 - Somewhat limited
- New bindings: ()/scikit-hep/boost-histogram
 - O-dependency build (C++14 only)
 - State-of-the-art PyBind11







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Design

• 500+ unit tests run on Azure on Linux, macOS, and Windows

Resembles the original Boost.Histogram where possible, with changes where needed for Python performance and idioms.

C++14

```
#include <boost/histogram.hpp>
namespace bh = boost::histogram;
```

```
auto hist = bh::make_histogram(
    bh::axis::regular<>{2, 0, 1, "x"},
    bh::axis::regular<>{4, 0, 1, "y"});
```

```
hist(.2, .3); // Fill will also be
hist(.4, .5); // availble in 1.7.2
hist(.3, .2);
```

Python

```
import boost.histogram as bh
```

```
hist = bh.histogram(
    bh.axis.regular(2, 0, 1, metadata="x"),
    bh.axis.regular(4, 0, 1, metadata="y"))
```

```
hist.fill(
[.2, .4, .3],
[.3, .5, .2])
```





Design: Manipulations

Combine two histograms hist1 + hist2

Scale a histogram hist * 2.0

Sum a histogram contents
hist.sum()

Access an axis ax = hist.axis(0) ax.edges # The edges array ax.centers # Centers of bins ax.widths # Width of each bin Fill 2D histogram with values or arrays
hist.fill(x, y)

Convert contents to Numpy array
hist.view()

Convert to Numpy style histogram tuple
hist.to_numpy()

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Pickle supported (multiprocessing)
pickle.dumps(hist, -1)

Copy/deepcopy supported hist2 = copy.deepcopy(hist)





Unified Histogram Indexing (UHI)

The language here (bh.loc, etc) is defined in such a way that any library can provide them - "Unified".

Access

v = h[b] # Returns bin contents, indexed by bin number v = h[bh.loc(b)] # Returns the bin containing the value v = h[bh.underflow] # Underflow and overflow can be accessed with special tags

Setting

h[b] = v h[bh.loc(b)] = v h[bh.underflow] = v





Unified Histogram Indexing (UHI) (2)

- h == h[:] # Slice over everything h2 = h[a:b] # Slice of histogram (includes flow bins) h2 = h[:b] # Leaving out endpoints is okay h2 = h[bh.loc(v):] # Slices can be in data coordinates, too h2 = h[::bh.project] # Sum an axis (name may change) h2 = h[::bh.rebin(2)] # Modification operations (rebin) h2 = h[a:b:bh.rebin(2)] # Modifications can combine with slices h2 = h[a:b, ...] # Ellipsis work just like normal numpy
 - Docs are here
 - Description may move to a new repository





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Performance

- Factor of 2 faster than 1D regular binning in Numpy 1.17
 - Currently no specialization, just a 1D regular fill
 - Could be optimized further
- Factor of 6-10 faster than 2D regular binning Numpy





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Distribution

- We *must* provide excellent distribution.
 - If anyone writes pip install boost-histogram and it fails, we have failed.
- Docker ManyLinux1 GCC 9.2: **()**/scikit-hep/manylinuxgcc
- Used in O/scikit-hep/iMinuit, see O/scikit-hep/azure-wheel-helpers

Wheels

- manylinux1 32 and 64 bit, Py 2.7 & 3.5-3.7
- manylinux2010 64 bit, Py 2.7 & 3.5-3.8
- macOS 10.9+ 64 bit, Py 2.7 & 3.6-3.8
- Windows 32 and 64 bit, Py 2.7 & 3.6–3.7

Source

- SDist
- Build directly from GitHub

Conda

• conda-forge package planned

python -m pip install boost-histogram

OR git+https://github.com/scikit-hep/boost-histogram.git@develop





hist is the 'wrapper' piece that does plotting and interacts with the rest of the ecosystem.

Plans

- Easy plotting adaptors (mpl-hep)
- Serialization formats via Aghast (ROOT, HDF5)
- Auto-multithreading
- Statistical functions (Like TEfficiency)
- Multihistograms (HistBook)
- Interaction with fitters (ZFit, GooFit, etc)
- Bayesian Blocks algorithm from SciKit-HEP
- Command line histograms for stream of numbers

Call for contributions

- What do you need?
- What do you want?
- What would you like?

Join in the development! This should combine the best features of other packages.





Aghast



Aghast is a histogramming library that does not fill histograms and does not plot them.

- A memory format for histograms, like Apache Arrow
- Converts to and from other libraries
- Uses flatbuffers to hold histograms
- Indexing ideas inspired the UHI

Binnings

IntegerBinning • RegularBinning • HexagonalBinning • EdgesBinning • IrregularBinning • CategoryBinning • SparseRegularBinning • FractionBinning • PredicateBinning • VariationBinning





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Now, we will go hands on with the first beta of boost-histogram!

Support

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