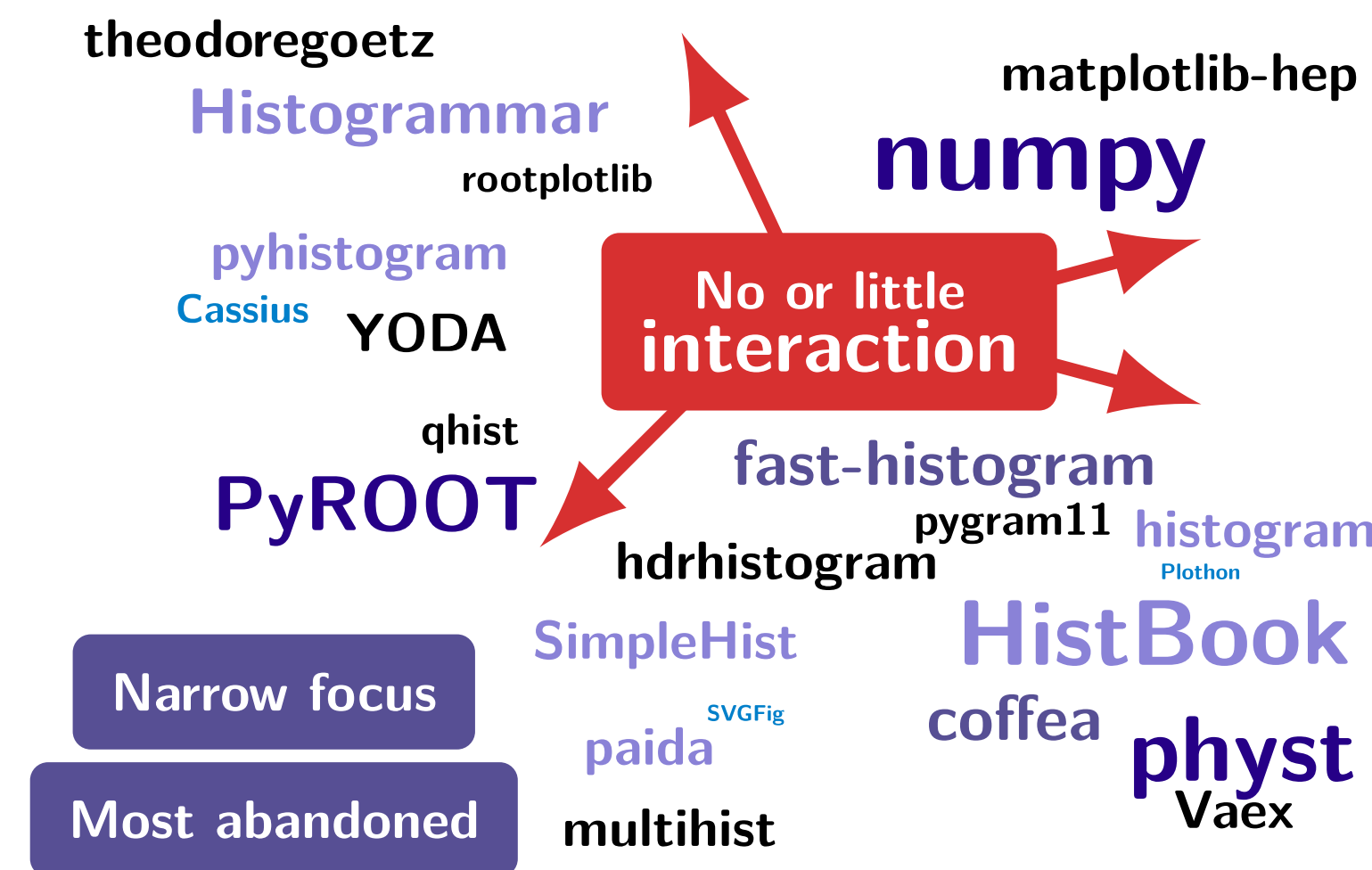




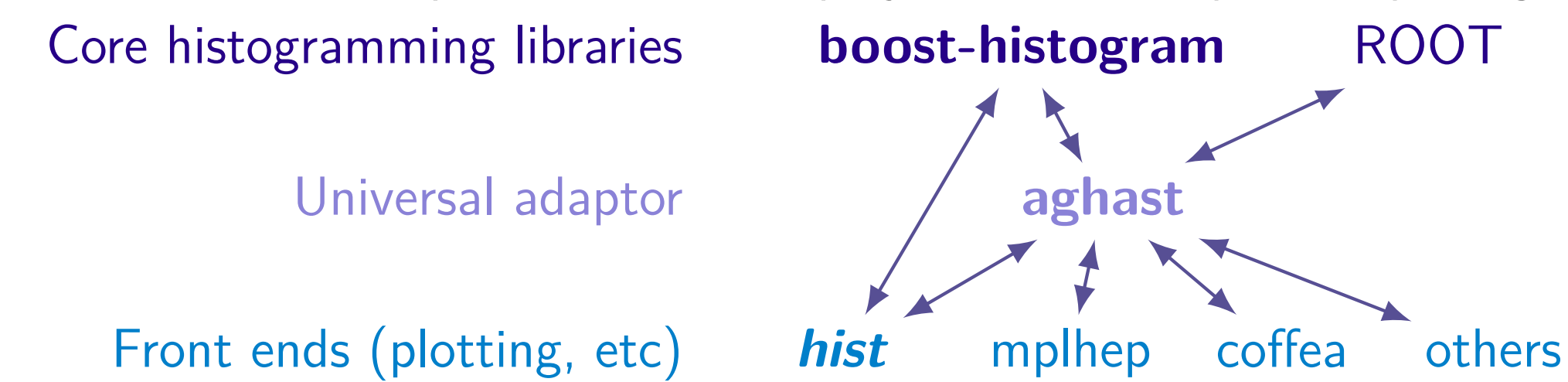
Many attempts at a solution

There are many histogramming libraries for Python, but all of the fall short in the key areas we care about: **Design, Flexibility, Performance, and Distribution.** Furthermore, they do not talk to each other.



Our solution

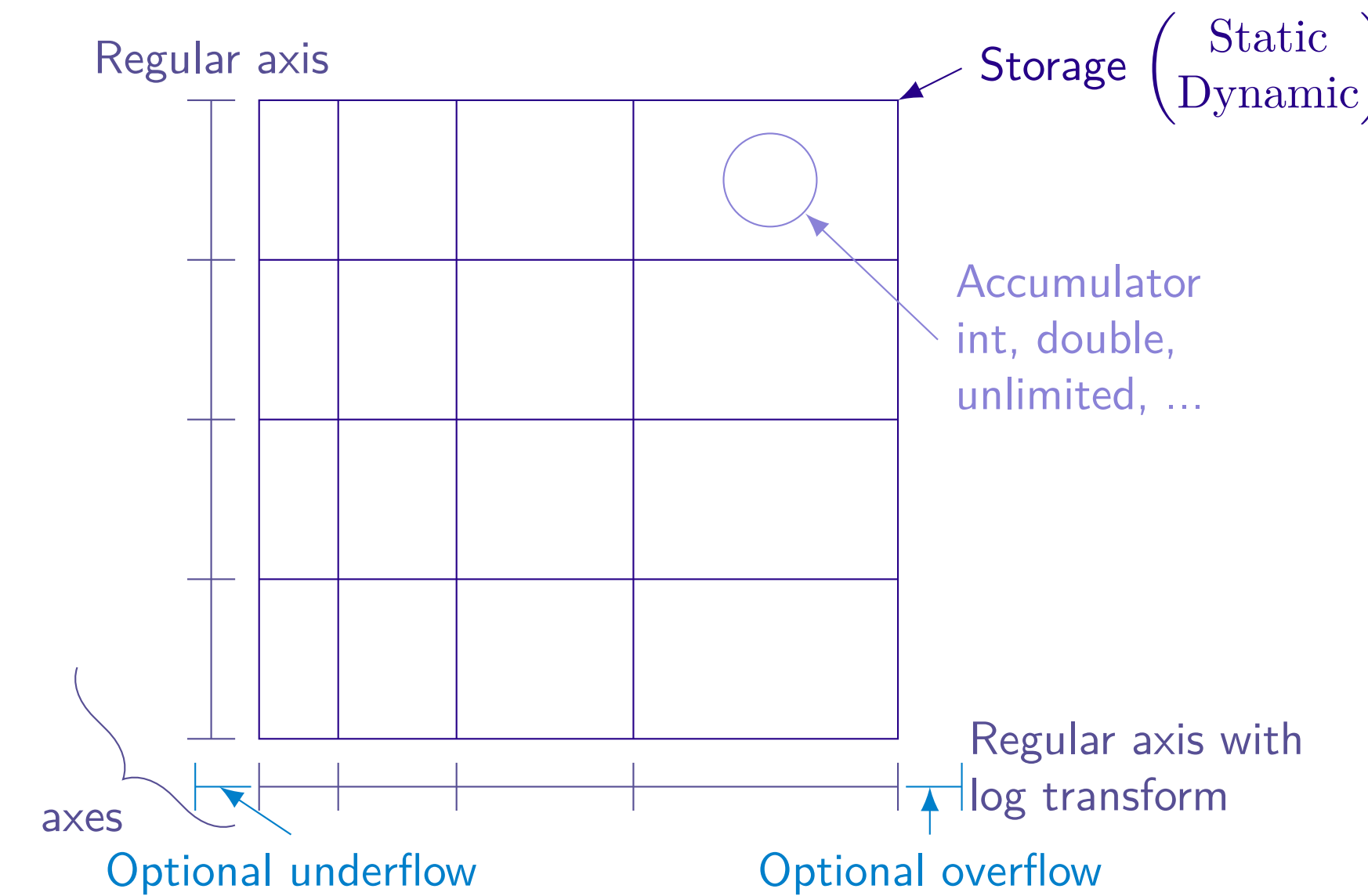
We have proposed the following family of libraries. Unlike previous attempts, we are building a modular solution and an adaptor. Bold indicates the library is related to this project, italics is a planned package.



A universal need

Histograms are a universal tool used across disciplines. However, for HEP, rather than just being a useful visualization tool, advanced histograms are often integral to the entire analysis. This is why we have some of the most highly developed histogram tools in C++ in ROOT, and why we need a high quality Python histogramming package.

Boost.Histogram for C++14 was developed by a HEP physicist and accepted as a general tool into the Boost C++ libraries, the most respected third-party library collection in the world. In close collaboration with the author, we have developed boost-histogram for Python.



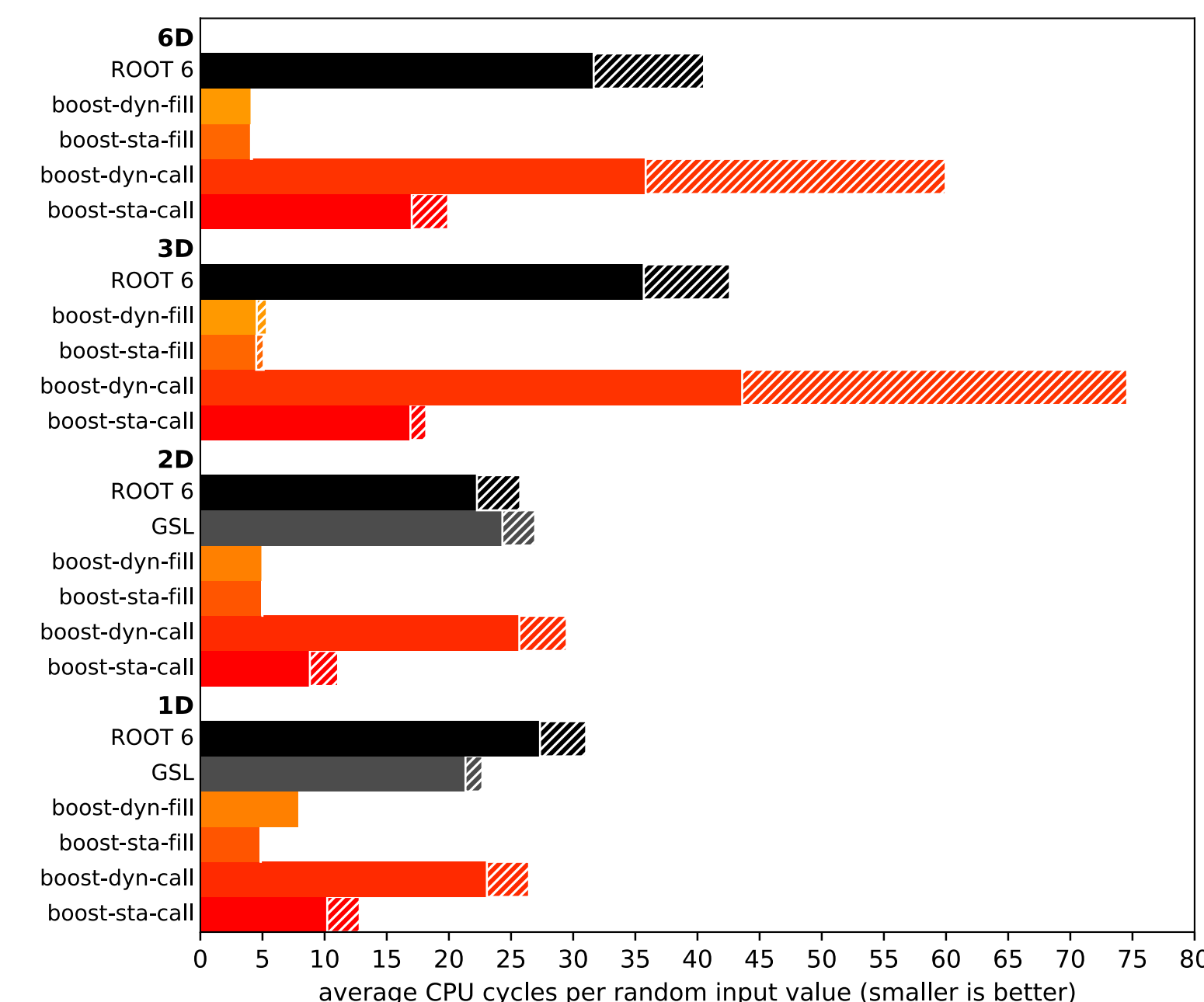
Storages

- Double:** Multipurpose, supports fractional weights.
- Int64:** Good for simple counts.
- AtomicInt64:** Threadsafe.
- Unlimited:** Starts as int8 and grows or converts to double as needed.
- WeightedSum:** Holds sum of weights squared (ROOT's optional weight tracking).
- Mean:** A "Profile" histogram, where means rather than sums are kept.
- WeightedMean:** Like Mean, but tracks the additional variance from varying weights.

Histograms are made up of components: a storage and 0 or more axes. All the ROOT histogram types (TH1D, TH2D, TH3D, THND, TProfile1D, etc.) can be represented using just a single histogram type, along with axes types and configuration not possible in ROOT. Everything is available in the formats Python users expect; data can be accessed without copy in any Python library.

100 bins, 10M events
8-core 2.4 GHz i9 macOS
Numpy: **146 ms**
PyROOT: **123 ms**
boost: **65 ms** or **58 ms**
boost MT: **10.3 ms**

100x100 bins, 10M events
8-core 2.4 GHz i9 macOS
Numpy: **1.18 s**
PyROOT: **157 ms**
boost: **93 ms** or **76 ms**
boost MT: **14.7 ms**



Comparison of fill performance against ROOT and GSL, in C++ using Boost.Histogram. The dyn-fill bars correspond to boost-histogram in Python.

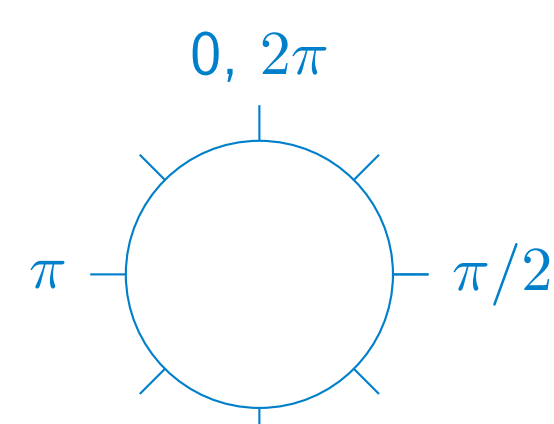
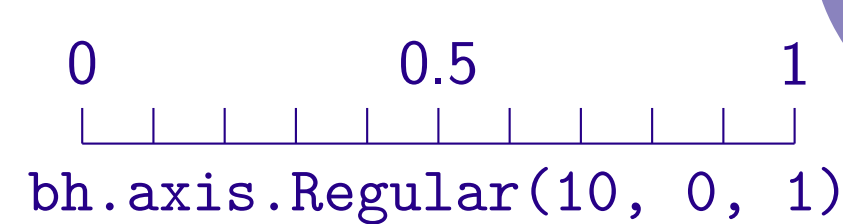
```

bh.axis.Regular
  under/overflow
  growth=True
  circular=True
  transform=Log()
  transform=Sqrt()
  transform=Pow(v)
  <any C transform>

bh.axis.Integer
  under/overflow
  growth=True

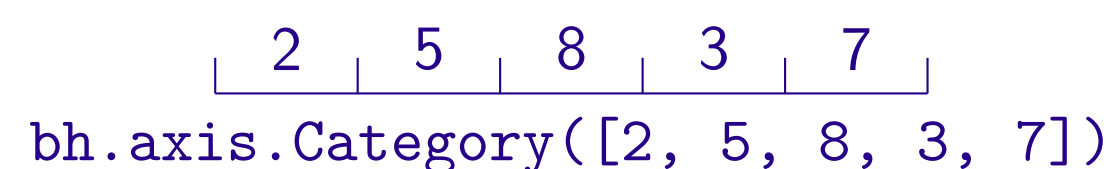
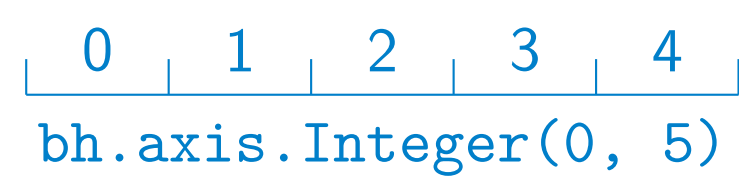
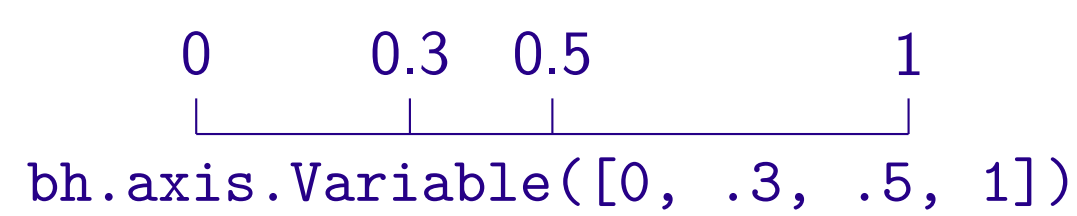
bh.axis.Variable
  under/overflow
  growth=True

bh.axis.IntCategory
bh.axis.StrCategory
  growth=True
  
```



```

bh.axis.Regular(8, 0, 2*np.pi, circular=True)
  
```



Design

Performance

Flexibility

Distribution

`pip install boost-histogram`

Wheels

Platform	32-bit	64-bit
Windows	2.7 3.6 3.7 3.8	2.7 3.6 3.7 3.8
macOS	2.7 3.6 3.7 3.8	2.7 3.6 3.7 3.8
Linux (1 & 2010)	2.7 3.5 3.6 3.7 3.8	2.7 3.5 3.6 3.7 3.8

Conda

`conda install boost-histogram --channel conda-forge`

Platform	32-bit	64-bit
Windows	3.6 3.7 3.8	2.7 3.6 3.7 3.8
macOS	2.7 3.6 3.7 3.8	2.7 3.6 3.7 3.8
Linux	64-bit / ARM / PowerPC	2.7 3.6 3.7 3.8

+ support for source builds, with only C++14 requirement

The Azure CI-based wheel build system designed for boost-histogram is now being used in several other Scikit-HEP projects.

Unified Histogram Indexing (UHI)

```

v = h[b] # Returns bin contents, indexed by bin number
v = h[loc(b)] # Returns the bin containing the value
v = h[loc(b) + 1] # Returns the bin above the one containing the value
v = h[underflow] # Underflow and overflow can be accessed with special tags

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[loc(v):] # Slices can be in data coordinates, too
h2 = h[:rebin(2)] # Modification operations (rebin)
h2 = h[a:b:rebin(2)] # Modifications can combine with slices
h2 = h[:,sum] # Projection operations
h2 = h[a:b:sum] # Adding endpoints to projection operations
h2 = h[0:len:sum] # removes under or overflow from the calculation
h2 = h[v, a:b] # A single value v is like v:v+1:sum
h2 = h[a:b, ...] # Ellipsis work just like normal numpy

h[b] = v # Returns bin contents, indexed by bin number
h[loc(b)] = v # Returns the bin containing the value
h[underflow] = v # Underflow and overflow can be accessed with special tags
h[...] = array(...) # Setting with an array or histogram sets the contents
  
```

Analysis using axes

What traditionally would be multiple histograms can be described as axes in a single histogram!

```

value_ax = bh.axis.Regular(100, -5, 5)
bool_ax = bh.axis.Integer(0, 2,
                          underflow=False,
                          overflow=False)

run_number_ax = bh.axis.IntCategory([], growth=True)

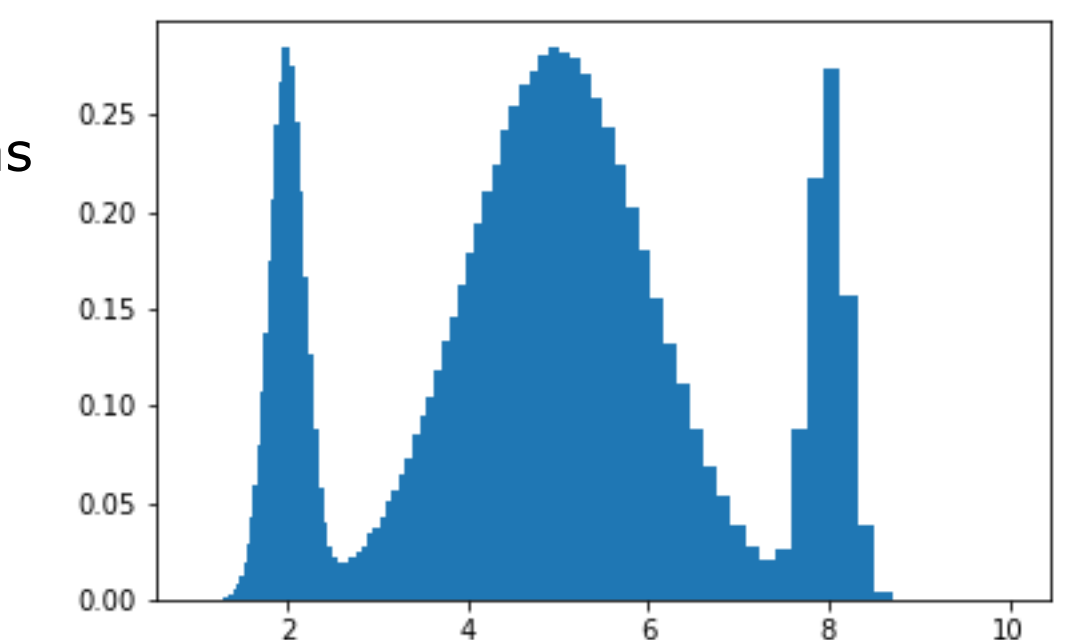
hist = bh.Histogram(value_ax, bool_ax, run_number_ax)

hist.fill(values, bools, run_numbers)

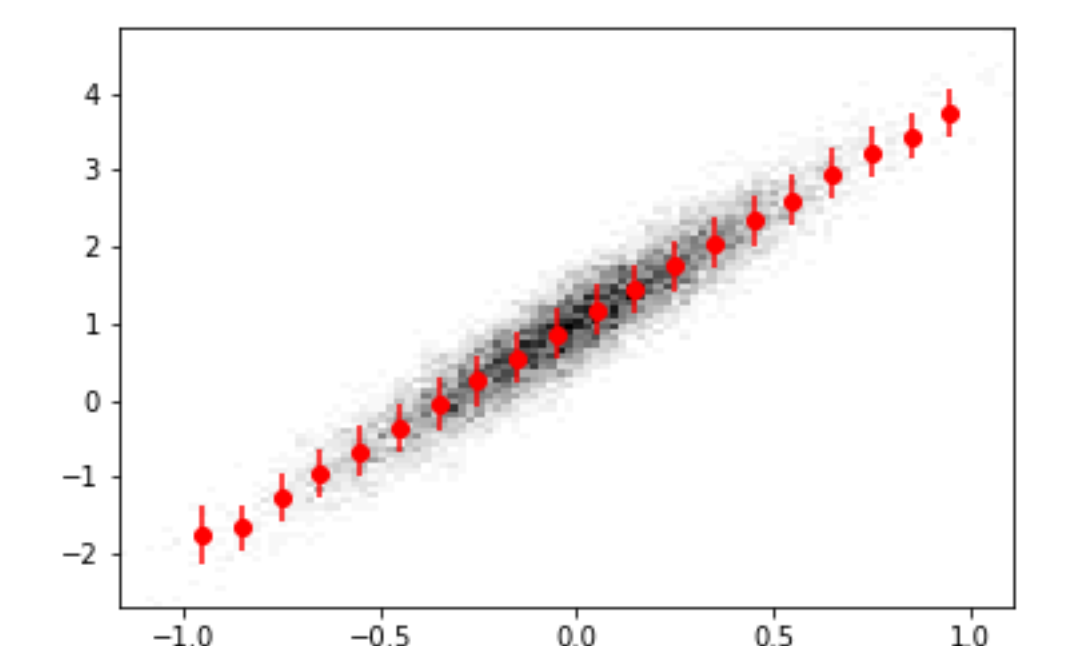
hist_true = hist[:, True, ::bh.sum] # Classic 1D hist
  
```

The future plans

Boost-histogram is ready for broad use; final polishing work is being done to enable smooth behavior when mixing types, etc. Boost-histogram has a well defined scope; it does not plot histograms or convert them; it has no dependencies. Aghost handles conversions, and Hist will assist in plotting and other common analysis tasks.



Log transform axis example



Profile histogram example