

DIANA meeting, 1st October 2018

Jan Pipek, Showmax

? janpipek/physt

Background

2001-2007 masters in physics (HEP)

2007-2015 Ph.D. in medical physics (Geant4)

2015-2017 post-doc in medical physics (Geant4)

2017- data scientist @ Showmax, Prague

Motivation

```
import numpy as np
histogram = np.histogram(heights)
```

Tuple of arrays?

Motivation (cont'd)

- In 2016, no adequate histogramming in Python (?)
- Lots of particle/dose distributions (2D, 3D) to visualize
- Will to create a useful open source library on my own

=> Physt

Target use cases

- Data exploration
- Compact representation of distributions
- Visualization / presentation

General, non-field-specific audience.

Design goals

- simple & familiar API (~numpy, ~pandas)
- histogram as first-class object (ROOT-inspired)
- no complex dependencies
 - numpy necessary
 - matplotlib recommended
- extensibility (visualization, computing engines, IO)

Status

https://github.com/janpipek/physt

528 commits,

2 main branches

- version 0.3.43 (rich features)
- re-design 0.4 (goal: cleaner API)

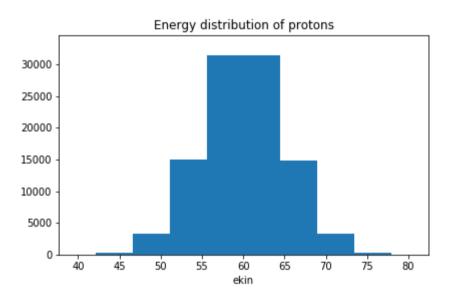
Example

```
import pandas as pd
from physt import h1

particles = pd.read_csv("protons.csv")
h = h1(particles["energy"], title="Energy distribution of protons")
```

```
Histogram1D(bins=(10,), total=100000, dtype=int64)
```

h.plot()



h.frequencies

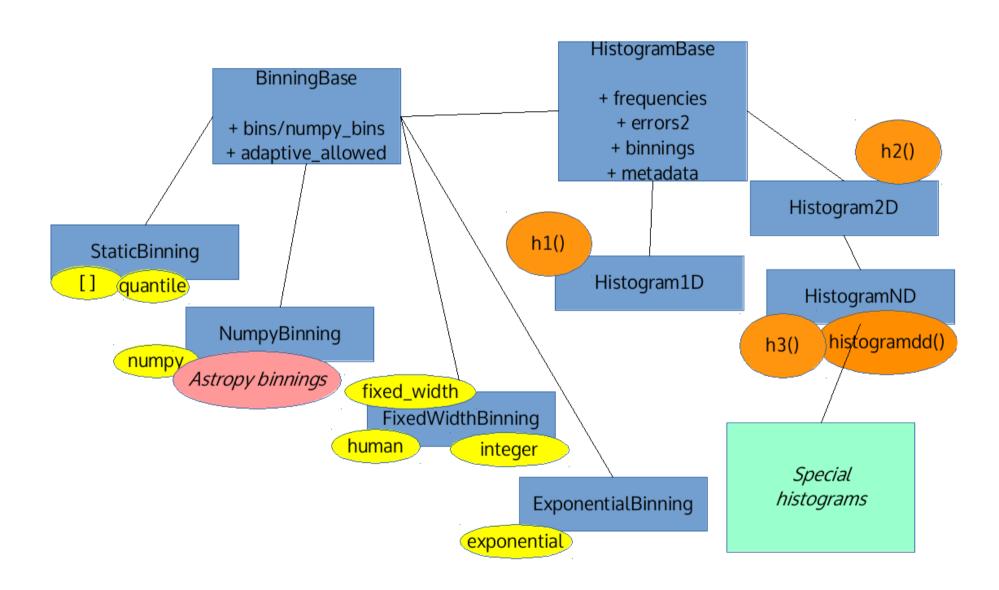
```
array([ 18, 346, 3383, 14978, 31434, 31348, 14827, 3318, 330, 18
```

h.bins

```
array([[ 38.83518235, ..., 81.791677 ]])
```

h.binning

```
NumpyBinning(array([ 38.83518235, ..., 81.791677 ]))
```

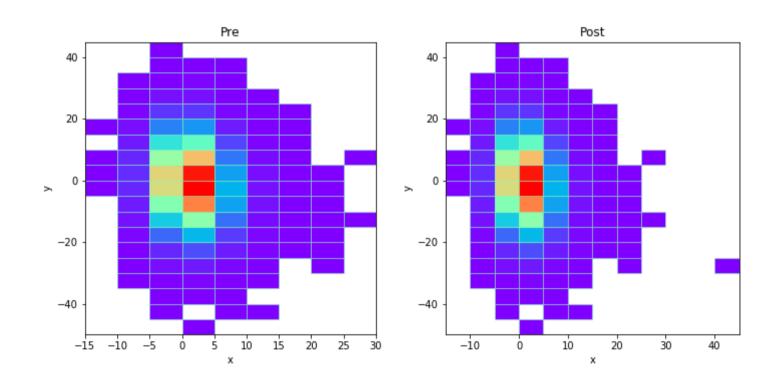


Binning schemas

- numpy (+ optimized bin counts)
- fixed-width (adaptive)
 - human (special case)
 - integer (special case)
- exponential
- quantile

Adaptive binning

```
hx = h2(particles["x"], particles["y"], "fixed_width", 5, adaptive=True)
hx.plot(figsize=(5, 5), show_zero=False, show_colorbar=False, cmap="raink"
hx << (43.4, -27.5)
hx.plot(figsize=(5, 5), show_zero=False, show_colorbar=False, cmap="raink")</pre>
```



Other features

- arithmetics (+ * /)
- statistics (mean, bin variance...)
- projections, slicing
- coordinate transformations (cylindrical, spherical)

Computation engines

- Currently, **numpy** is doing most of the work.
- Experimental usage of dask for "big" data.
- tensorflow?
- HDembinski/histogram?

Interoperability

- pandas, xarray, numpy
- ROOT? Geant4 histograms CSV
- file I/0: JSON, protobuf, HDF5

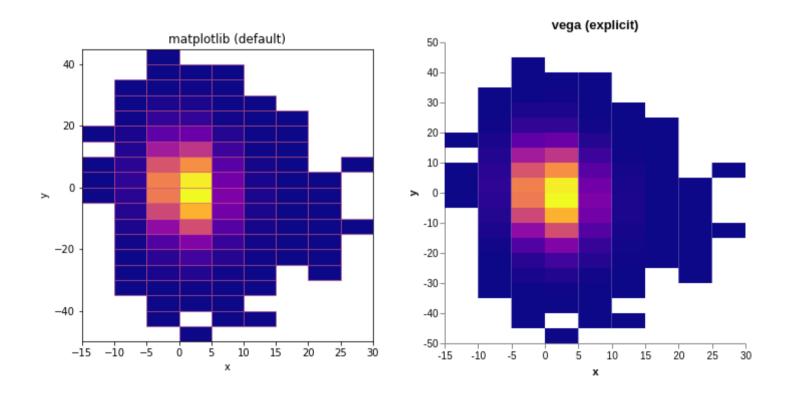
Plotting backends

- matplotlib (standard)
- vega (for notebooks)
- plotly (way to go?)
- ascii (wish I had it)

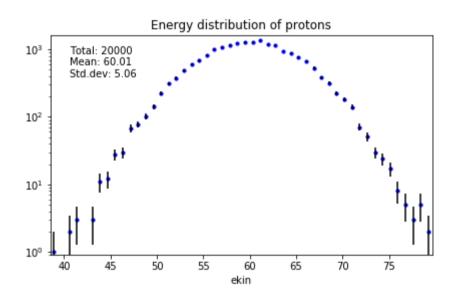
```
hx = h2(particles["x"], particles["y"], "fixed_width", 5)

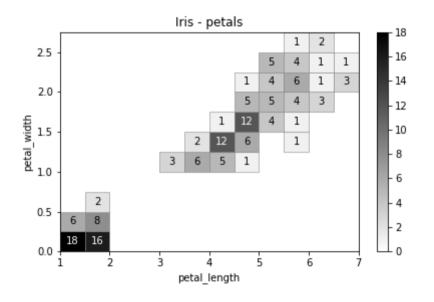
# Matplotlib
hx.plot(show_zero=False, cmap="plasma", title="matplotlib (default)")

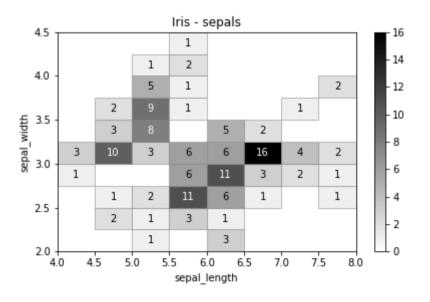
# Vega
hx.plot(backend="vega", show_zero=False, cmap="plasma", title="vega (expl
```

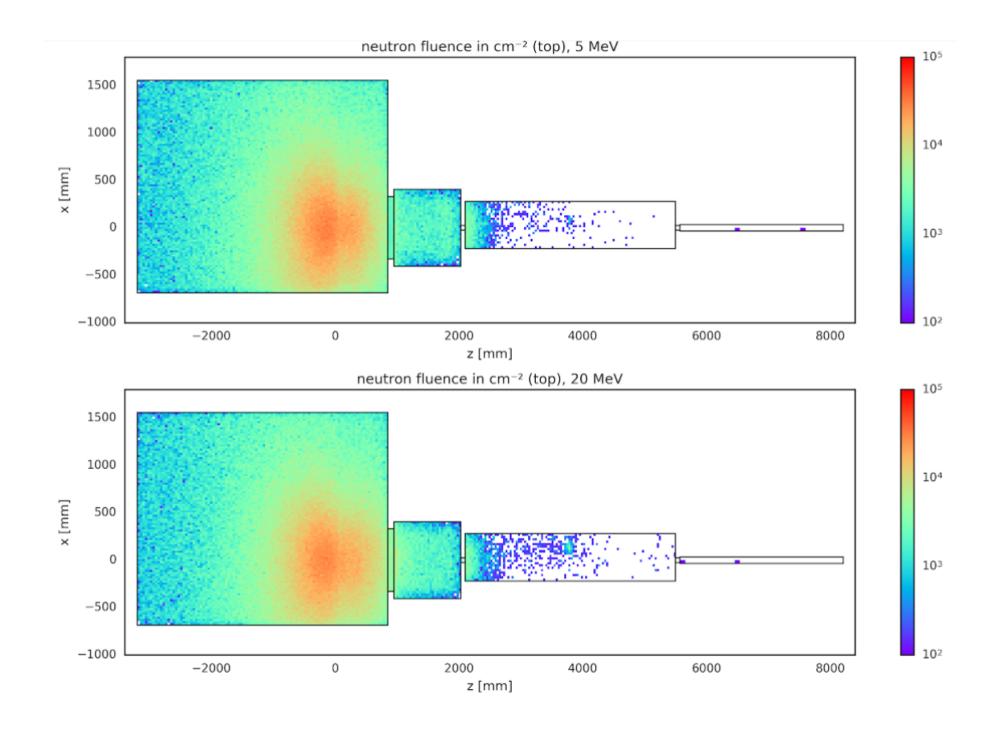


```
particles = pd.read_csv("protons.csv")
h = h1(particles["ekin"][::5], 50, title="Energy distribution of protons'
h.plot.scatter(errors=True, yscale="log", s=10, show_stats=True)
```









```
data = ... # Some random x, y, z points

h = special.spherical_histogram(data)
h = h.projection("theta", "phi")

h.plot.globe_map(density=True, figsize=(7, 7), cmap="rainbow")
```

