

PyHEP2022 Workshop Report

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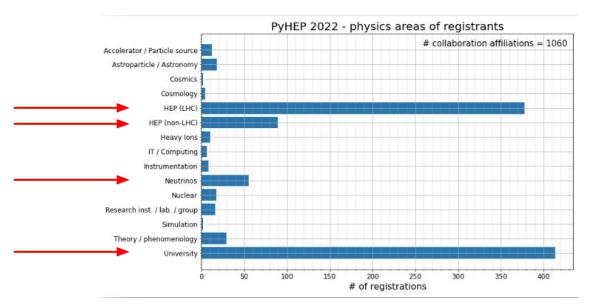
PyHEP Workshops



- Series of workshops started at CHEP2018
 - Successful event, so continued annually
 - Organised by the <u>PyHEP HSF Working Group</u> Eduardo Rodrigues was the founder and continues to play a critical role
- Motivation
 - Increasing interest in using tooling from Python data science
 - Recognition of Python as a first class language for analysis in HEP
 - Update the community on latest developments and discuss challenges
- First two workshops were mainly for package developers
- Since 2020, organised as an online event
 - Because, COVID...
 - But registrations jumped from 55 to 1000
 - Clearly tapped a significant vein of interest, particularly from students

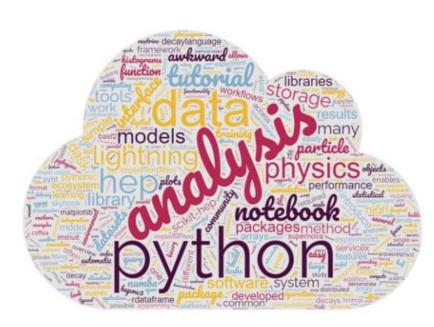
PyHEP2022

- Decided again for a virtual workshop
 - Low barrier for entry at the cost of lack of coffee-based interactions
 - 1122 people registered

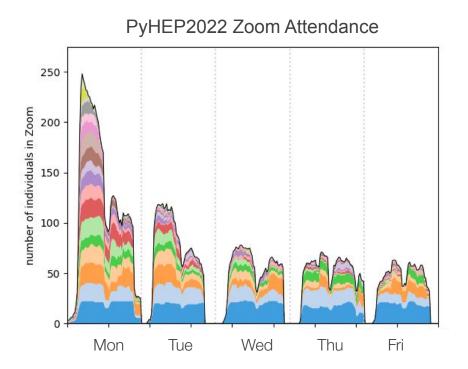


Format

- Open Abstract Call 42 submissions (a record!)
- Three presentation options
 - Tutorial 1 hour based on Jupyter notebook
 - Notebook Talk 30 minutes, demonstration and discussion, also from a notebook
 - Lightning talk 10 minutes, mixture of slides and notebooks used
- Bonus features
 - Slack for discussions
 - Sli.do for questions
 - RemotelyGreen for social mixing
 - Hackashop for leveling up from user to contributor



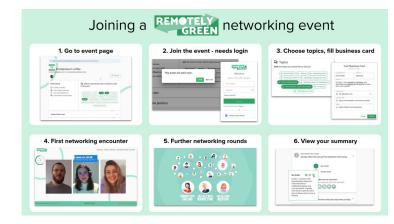
Registrations to Reality...



- Peak attendance almost 250
- Drops off day-by-day, to 50-70 people
 - Which is still a pretty active community
- In total, 421 people did attend at some point in the workshop
 - So a bit buffet style!
- Not clear why more than half the registrants didn't show up
 - We try to understand this through the survey
 - Seems that people who had previously attended were more likely to show up

Hackashop and Remote Mingling...

- We tried to overcome the virtual workshop disadvantages by using the <u>RemotelyGreen</u> social mixer app
 - Company founded by some of our CERN colleagues
 - Session was only very lightly attended (~22)
 - Positive feedback from those who did join, but didn't get wide traction too busy? Not interesting enough? YACT?
- Also wanted to engage users with developers, through an online hackashop event
 - Introductory talk by Aman Goel, followed by breakout rooms with projects
 - Also, around 20 people joined, but this was much more expected
 - At least some people now working on PRs with projects



Workshop Highlights

Training and Features

- With significant user interaction, tutorials and training in generic Python topics are a popular part of the workshop
- <u>Level up your Python</u> (Henry Schreiner), is an excellent <u>intermediate training</u>
 <u>resource</u> (all notebooks)
 - Core: Logging, debugging, profiling, generators, decorators, packaging
 - Numerical Python: numpy, panda, numba, PyBind11

Decorators

This is likely the simplest syntactic sugar you'll see today, but maybe one with some of the furthest reaching consequences. Let's say you have a bit of code that looks like this:

```
def f(): ...
f = g(f)
```

So g is a function that takes a function and (hopefully) returns a function, probably a very similar one since you are giving it the same name as the old "f". In Python 2.5, we gained the ability to write this instead:

```
@g
def f(): ...
```

That's it. The thing after the @ "decorates" (or transforms) the function you are defining and the output is saved with the name f.

Training and Features

HSF Training Experience

- Providing generic training to HEP community
- Working with The Carpentries, thanks to IRIS-HEP funding
- Building a training community

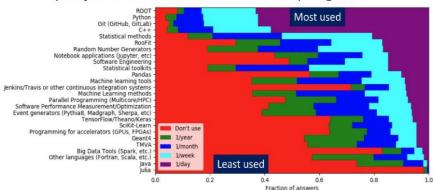
<u>Using C++ from Numba</u>

- How to use <u>cppyy</u> (as used in PyROOT) with numba
 - Mix-in C++ code and classes into numba jited functions

Forward look: What's new in Python 3.11

- Faster! (avg. 25% speed-up over Python 3.10)
- Better error messages and exception 'notes'
- Better static typing
- AsynclO.Taskgroups, Tomllib, WebAssembly (pyodide)

Frequency of use of some scientific software/computing tools



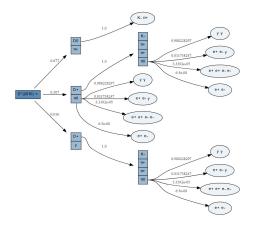
```
cppyy.cppdef("""
template<typename T>
T square(T t) { return t*t; }
""")

@numba.jit(nopython=True)
def tsa(a):
    total = type(a[0])(0)
    for i in range(len(a)):
        total += cppyy.gbl.square(a[i])
    return total
```

HEP Tooling in Python

- Continued improvements in Python tooling for HEP, making
 - things more convenient
 - Most of these tools joined to the <u>SciKit-HEP</u> project
- Particle, Decay Language, PhaseSpace
 - Particle Python interface to PDG data, including 'translation' between different PIDs
 - DecayLanguage Python interface to particle decays
 - Phase Space Simulate n-body decays, interfaces with DecayLanguage

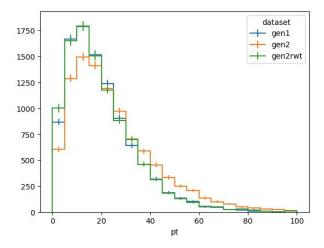




HEP Tools

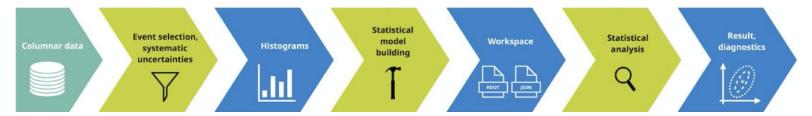
Correctionlib

- Well-structured JSON data format for correction factors that are typical in HEP
- C++ and Python
- Very good runtime performance
 - N.B. Also discussed that the Analysis Ecosystem II workshop
- Nicer Pythonic interfaces to important HEP packages
 - pyhepmc3
 - Hand crafted Pythonic interface (cf. autogenerated)
 - o Pythia8
 - Updated interface to Pythia, using PyBind11



Example of generator reweighting, implemented with CorrectionLib

Analysis Workflows



Data pipelines

- ServiceX+FuncADL
- Skyhook
- Spark
- coffea+WorkQueue
- Analysis Grand Challenge

Examples from experiments

- Belle II
- LHCb
- CMS analysis with RDF+Dask

Tools and interoperability

- Scikit-HEP tools
- Vector
- XRootD+Dask
- uproot+Dask
- Awkward RDataFrame

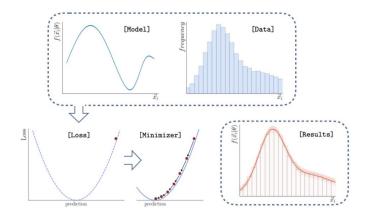
Analysis Workflows - selected highlights

- Many presentations regarding different tools in the Scikit-HEP family
 - Low-level array views
 - Vector manipulations
 - Remote I/O: Dask understands XRootD → enables end-to-end pipeline with Dask collections
- Data pipelines depend on experiment/site
 - Thus need for common analysis interfaces
- ServiceX embeds data delivery as part of the analysis notebook
 - yaml configuration file to specify dataset files/trees
 - Data **caching** is available, path to local cache directory must be supplied
- Skyhook's vision: zero-copy in-memory views on distributed datasets
 - Avoids serialization costs in principle
 - Relies on a solid organization of data lake and the associated object store (Ceph)

Fitting and Histograms

Statistical analysis

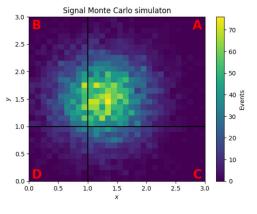
- <u>iminuit</u>
- <u>jacobi</u>
- <u>zfit</u>
- abcd_pyhf
- <u>pyhf<->combine</u>



Depiction of fitting workflow

Histograms

- Scikit-HEP histogram tools
- <u>Uhepp</u>



Signals with regions ABCD

Fitting and Histograms

- jacobi
 - Numerical differentiation of any Python callable,
 - Numerically robust, although not as fast as other solutions
- iminuit
 - Minimizer wrapper around Minuit2 (no ROOT dependency)
 - Used also in other packages below
- zfit:
 - General purpose fitting tool
 - Particularly flexible, can be used with other packages (iminuit, hepstats).
 - Compatible with Universal Histogram Interface
- abcd_pyhf:
 - o **pyhf wrapper**, specialized for background estimation
- pyhf and combine analyses can be converted both ways
 - CMS data card to json translation layer
 - Extra "shapes files" to store binned data (ROOT file)

Fitting and histograms

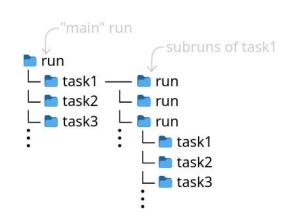
- Uhepp: a format for sharing histograms
 - Data + graphics settings in the same format
 - o Bin edges, values, metadata stored in semi-structured file
 - Also provides a **web service** to upload/download histograms in the application
- boost.histogram + Hist
 - A lot of overlap between the two (Hist is based on boost.histogram)
 - Again a lot of interoperability with plotting libraries
 - (matplotlib, inline html in the notebook, terminal-friendly representation)

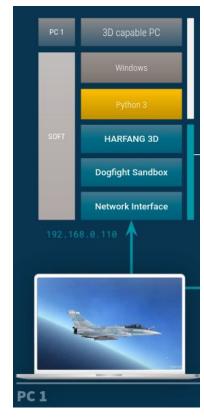
```
v root:
 badge:
 v bins:
   ▶ edges: [] 41 items
   ▶ rebin: [] 11 items
 ► metadata:
 ▶ ratio: [] 1 item
 ▶ ratio axis:
 ▶ stacks: [] 2 items
   type: "histogram"
 ▶ variable:
   version: "0.4"
   y axis: {}
 ▼ yields:
   ▶ bkg:
  v data:
    base:
                42 items
               42 items
    stat: []
   ▶ sig:
```

Bonus Features

- The SuperNova Early Warning
 System & Software for Studying
 Supernova Neutrinos
- 3D and VR Industrial Use Cases in Python
- The Bureaucrat
 - Task + file manager aimed at HEP workflows
- EOS A software for Flavor Physics
 Phenomenology
 - Flavor physics theory predictions in a Jupyter notebook
- Scalable, Sparse IO with larcy
 - Sparse IO layer, data format, Python/C++, parallel I/O with MPI. Aimed at ML pipelines







Summary

- PyHEP2022 brought 100s of package developers and users together again
 - Videos now all posted to <u>YouTube</u>
 - Survey results indicate a positive view of the event overall
- Participation and wide range of topics shows this piece of our ecosystem remains very vibrant
 - Continuing importance of Python in the future for the field
- User experience is always emphasized
 - Lots of areas to engage in
 - Extreme interoperability between contents of presentations, frequently overlapping

