

Awkward plans

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Context



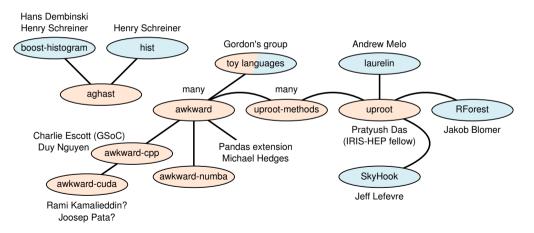
Awkward-array is the end of a long chain of ideas about columnar analysis:

- 2016 Femtocode was originally envisioned as a combined language, columnar processor, and distributed processing system: a full-service package for physics.
 - To be more realistic, I separated the language and columnar processing from the distributed processing, leaving the latter to others.
- 2017 Shredtypes \rightarrow Quiver (for Arrow) \rightarrow OAMap were iterations on expressing an abstract type system with an internal columnar representation.
 - Uproot 2 had a "minimal columnar data type": a JaggedArray class.
- 2018 Users found JaggedArrays useful despite the fact that it doesn't hide the columnar representation (this was a surprise to me).
 - Awkward-array built the abstract type system to columnar representation in the other direction: bottom-up, from physical arrays to abstract types.
- 2019 Awkward-array is the first in this series to be widely used; I'm getting a lot of feedback from Coffea and uproot 3 users.

Up to now: one specification, many implementations



Four of my "ongoing projects" are different implementations of awkward, all to adhere to the specification.adoc (which is now out of date).





Tutorial

😫 launch binder

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"Homogenization" because a common user complaint is that methods defined on one class (e.g. JaggedArray) don't work on another class (e.g. MaskedArray of JaggedArray).

While documenting (100 pages of examples), I ensured that a suite of "high-level" methods does something meaningful on every array class.

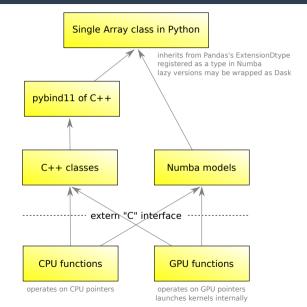




- Writing the same method on 14 classes × 4 implementations is a problem for developer FTEs now and maintainance later. (Remember, I haven't touched awkward-numba since February!)
- 2. Users need a better separation between high-level and low-level.
 - Structural classes, such as ChunkedArray of VirtualArrays of X to make array X lazy-loading, should be hidden.
 - Named methods, like x.cross(y) meaning "cross-join," compete for the same namespace with column names and domain-specific methods, like x.cross(y) meaning "3-D cross product."
- 3. Too many features are opt-in. Numba integration will fail and Pandas integration will have a subtle performance bug if the user doesn't **import awkward.numba** and **import awkward.pandas** first.

Solution to #1: implement at most twice





Put all CPU implementations behind a stateless **extern** "C" interface.

- C++ classes (e.g. JaggedArray) provide structure and ownership rules.
- Numba (e.g. JaggedArrayModel) extensions associate Python classes in @numba.jit functions with the same implementations.
- Structure classes mirrored between C++ and Python by pybind11.
- Structure classes hidden inside Array.

Maybe later write **extern** "C" functions for GPU as well.



```
Single Array class with a high-level type; narrow access to underlying structure.
>>> myarray
<Array [[1.1, 2.2, 3.3], [], [4.4, 5.5]] at 0x7fce7170bcf8>
>>> print (awkward.arraytype(array))  # datashape.readthedocs.io
3 * var * float64
>>> array.struct
JaggedArray([3, 99, 0], [5, 99, 3], [4.4, 5.5, 1.1, 2.2, 3.3])
The user only finds out that it is a Taraward path a large.
```

The user only finds out that it is a JaggedArray, and not a lazy JaggedArray, by digging into the structure classes provided by pybind11.

This also hides the starts=[3, 99, 0] and stops=[5, 99, 3].



It would be easier to implement and a cleaner API to put structure-manipulation operations in free-standing functions like

awkward.cross([x, y, z]) # size of output known at the beginning rather than

```
x.cross(y).cross(z)
```

have to make intermediate arrays

Apart from a single struct attribute and unnamed magic like add . array ufunc, etc., the Array class can be kept clear for columns and domain-specific methods:

mydata.btag lorentzVectors.boost(others) # physics methods vector3Ds.cross(others)

```
# field names as attributes
# reduced ambiguity
```

In short, everything after the dot would be physics.



Originally, I didn't want to automatically register Numba types and automatically make awkward arrays descend Pandas's ExtensionDtype because importing these libraries would add to the startup time, even if not used.

| | slow computer | medium computer | fast computer |
|------------------|---------------|-----------------|---------------|
| import pandas | 1.5 sec | 0.35 sec | 0.25 sec |
| import numba | 1.0 sec | 0.35 sec | 0.20 sec |
| pandas and numpy | 2.0 sec | 0.60 sec | 0.35 sec |

"Slow computer" is 1.1 GHz, "medium" is 2.6 GHz. Most users will be fine.



The original Numpy-based implementation and partial Numba-based one will have to be scrapped in favor of C++.

► A major change in API, even a good one, will require roll-out: awkward 0.x → awkward 1.0.

- ► After Charle Escott's GSoC ends, I'm the only developer.
- ▶ I estimate about 6 months for this transition.

Drawbacks of this plan



- The original Numpy-based implementation and partial Numba-based one will have to be scrapped in favor of C++. It served us well. The feedback we got from getting something in front of users quickly was invaluable.
- ► A major change in API, even a good one, will require roll-out: awkward 0.x → awkward 1.0.

Since we envisioned multiple awkward implementations, uproot already has hooks for switching between alternatives.

After Charle Escott's GSoC ends, I'm the only developer. It's my time to learn C++11. :)

I estimate about 6 months for this transition.
 We already know it has a userbase.



By putting the *primary* implementation in C++, awkward-array could be directly used by C++ projects.

- Easier to emit awkward arrays from ATHENA/CMSSW: e.g. for ServiceX.
- ► Could become a standard way to wrap C++ libraries for Pythonic analysis: e.g. JaggedArray of 4-vectors → FastJet → JaggedArray of jets.
- Actually write production code with awkward-arrays, to transparently replace CPUs with GPUs? (Giuseppe Cerati has been looking into this.)



July: two more tutorials: CoDaS-HEP and DPF (a total of 7 this year!) August: Charlie's GSoC project finishes: learn as much as we can about pybind11's strengths and weaknesses August: I finish the prototype SQL-for-events toy language September: Strange Loop and working on fundamentals of awkward 1.0 October: Still working on awkward 1.0 and maybe go to PyHEP November: CHEP and should be getting usable prototypes of awkward 1.0 to Coffea for testing

December: Should be transitioning PyPI packages awkward0 and awkward

— — — then it's the year 2020 — — —