

# cabinetry

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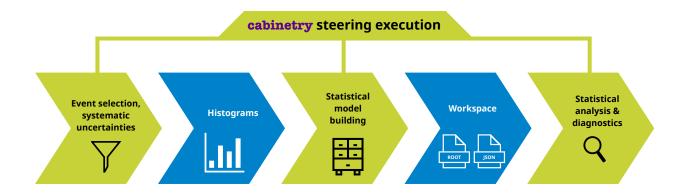
IRIS-HEP retreat <u>https://indico.cern.ch/event/896167</u> May 27, 2020





### What is cabinetry?

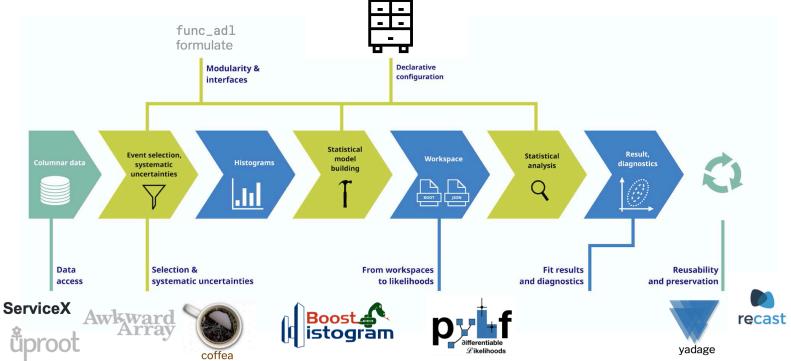
- analyzers use cabinetry to design template fit models via a declarative configuration file
  - analyzers specify selections for signal/control regions, (Monte Carlo) samples, systematic uncertainties
  - cabinetry steers the histogram creation (region  $\otimes$  sample  $\otimes$  systematic)
  - using the histograms, cabinetry produces a workspace (serialized fit model)
  - cabinetry steers statistical inference and provides diagnostics and visualization tools



- github: <u>https://github.com/alexander-held/cabinetry/</u>
- IRIS-HEP project page: <u>https://iris-hep.org/projects/cabinetry.html</u>

### cabinetry within IRIS-HEP

- a declarative configuration steers cabinetry and its interactions with other tools
- data is delivered to cabinetry via e.g. ServiceX after coffea processing, or straight from uproot as an awkward array
- pyhf is used for inference



## Hello world (1)

• simple fit model defined via configuration file in YAML format

▶ required histograms are implicitly specified as region ⊗ sample ⊗ systematic

- cabinetry creates histograms, visualize data, creates a workspace and runs a fit
- try out the example in the <u>repository</u>

#### •••

```
import cabinetry
```

cabinetry\_config = cabinetry.config.read("config\_example.yml")

# perform histogram post-processing
cabinetry.template\_postprocessor.run(cabinetry\_config, histo\_folder)

# visualize templates and data
cabinetry.visualize.data\_MC(cabinetry\_config, histo\_folder, "figures/", prefit=True
method="matplotlib")

```
# build a workspace
ws = cabinetry.workspace.build(cabinetry_config, histo_folder
```

# run a fit
cabinetry.fit.fit(ws

#### • • •

eneral: Measurement: "My fit" POI: "Signal strength"

#### Samples:

- Name: "Data"
   Tree: "pseudodata"
   Path: "ntuples/data.root
   Data: True
- Name: "Background" Tree: "background" Path: "ntuples/prediction.root Weight: "weight"
- Name: "Signal"
  Tree: "signal"
  Path: "ntuples/prediction.root"
  Weight: "weight"

#### Regions:

- Name: "Signal Region" Variable: "jet\_pt" Filter: "lep\_charge > 0" Binning: [0, 100, 200, 300, 400, 500]

#### Systematics:

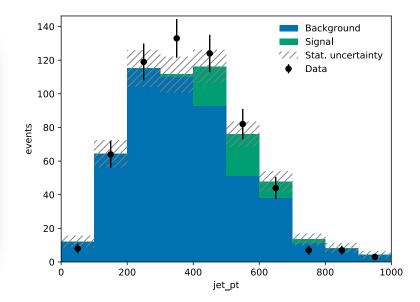
- Name: "Luminosity" OverallDown: -0.05 OverallUp: 0.05 Samples: ["Signal", "Background"] Type: OVERALL

#### NormFactors:

- Name: "Signal strength" Nominal: 1 Min: 0 Max: 5 Samples: "Signal"

## Hello world (2)

### visualization example (pre-fit)



### (partial) output

<pre>INF0 - cabinetry.fit INF0 - cabinetry.fit</pre>		0.989652 +/- 0.090191 1.010105 +/- 0.068933 1.083871 +/- 0.069312 1.016181 +/- 0.069627 1.015840 +/- 0.087416 0.950477 +/- 0.102159 0.732744 +/- 0.188015 0.914896 +/- 0.242512 0.824391 +/- 0.345761 0.278680 +/- 0.843117
INFO - cabinetry.fit	- Signal strength :	1.078683 +/- 0.381934

## Status and goals

- cabinetry started a month ago
  - feature set is being actively expanded

#### • goals:

- become a convenient tool for LHC-style binned template fit definition and steering
- Ieverage other existing IRIS-HEP tools for faster time-to-insight
  - factorize as much as possible, and support multiple backends where sensible
- develop and promote API to help factorize tasks into independent modules
- declarative approach
  - but allow analyzers to supply their own functions for key tasks that interact with the declarative configuration
- other existing tools:
  - range of frameworks with similar scope exits: <u>CMS combine</u>, many within ATLAS (HistFitter, TRExFitter, WSMaker, ...)
     want to provide a python / non-ROOT alternative, based on experience from existing tools

## Getting data to cabinetry

### • **cabinetry** makes requests to build histograms, which should be fulfilled by external backends

▶ need a common API

• simple uproot backend currently used as reference

### Challenges:

defining the data location in a file type agnostic way

- how to deal with substructure in the file, what if the file is not a file but a location in memory?

- Ianguage for defining observables, cuts, weights
  - using data["MET"] > 100 or MET > 100 or ...
  - implicit assumptions of how strings map to objects within the file
  - may not need to support complex operations this can be done in upstream tools

## Years 3, 4, 5

#### • year 3:

- obtain community feedback, promote community involvement in development
- converge on core design decisions
- expand feature set to cover most common use cases
- gather experience with fully featured LHC-style analysis

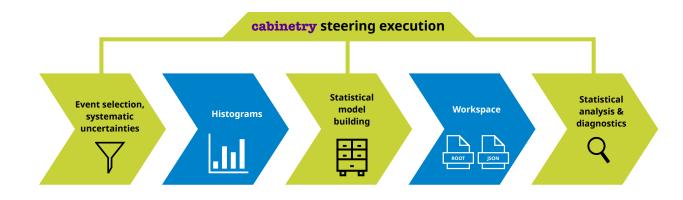
#### • years 4+5:

- move towards a stable product
- Integrate in IRIS-HEP analysis systems grand challenge

### Summary

#### • cabinetry is

- a new effort aiming to interface many existing IRIS-HEP tools
- a modular, python-based approach to building workspaces for statistical inference with template fits
- both a library (e.g. workspace creation from histograms) and a framework (steering other tools)
- welcoming contributions and thoughts!



## Backup

## Fully declarative approach?

• design of a declarative configuration that points to the right path for the data needed for any histogram?

• analyzers might store their data in many different structures

<pre>nominal/ signal.root region_1 region_2 [] systematic_variation/ signal.root region_1 []</pre>	<pre>region_1/    signal_nominal.root    signal_systematic_variation.root    background.root    region_2/     signal_nominal.root    signal_systematic_variation.root    background.root</pre>
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• difficult to support any possible structure via pre-defined options

• more powerful approach: analyzers define their own option and provide a function to parse them

• see <a href="https://github.com/alexander-held/cabinetry/issues/16">https://github.com/alexander-held/cabinetry/issues/16</a> for more