## Analysis Grand Challenge at REANA

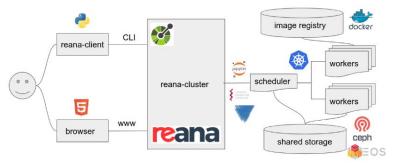
Andrii Povsten

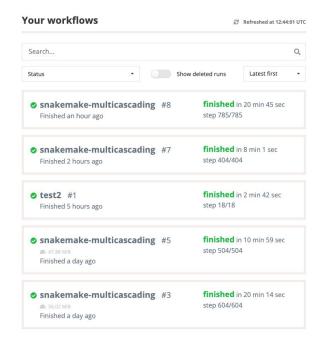
*Mentors:* Alex Held, Matthew Feickert (University of Wisconsin- Madison), Oksana Shadura (University Nebraska-Lincoln), Tibor Simko (CERN)

FAIR and Open Science in High Energy Physics, OAC-2226378, OAC-2226379 and OAC-2226380

## **REANA - Reusable analysis**

- Allow complex multi-stage physics analysis to be executed with a single command, using REANA service
- Enable to submit parameterized computational workflows to run on remote compute clouds or using other backends
- REANA uses container technologies to provide exact runtime environment necessary for various analysis steps
- Supports several different container technologies (Docker, Singularity), compute clouds (Kubernetes/OpenShift,), shared storage systems (Ceph, EOS) and structured workflow specifications (CWL, Yadage, Snakemake)





## REANA instance at CERN - https://reana.cern.ch

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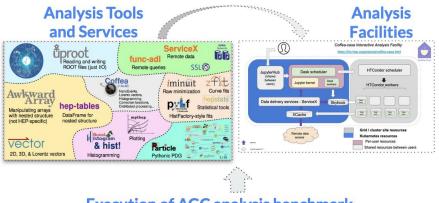
- Using custom user tokens
- CERN gitlab native integration - allows to integrate REANA into your GitLab pipelines.
- Excellent documentation https://docs.reana.io/
- User support: <u>https://forum.reana.io/</u>

# For setting REANA client: pip install reana-client

<pre>(reana) andrewpovsten@pucomphep03 cms-open-data-ttbar % reana-client ping</pre>
REANA server: https://reana.cern.ch
REANA server version: 0.9.2
REANA client version: 0.9.1
Authenticated as: Andrii Povsten <andrii.povsten@cern.ch></andrii.povsten@cern.ch>
Status: Connected
(reana) andrewpovsten@pucomphep03 cms-open-data-ttbar %

## Analysis Grand Challenge (AGC)

- Main AGC analysis task: ttbar cross-section measurement in single lepton channel
  - Includes single top reconstruction
- **AGC goal** to allow coping with HL-LHC data sizes by rethinking data pipeline (e.g. increasingly complex analysis benchmark)
  - As well to provide flexible, easy-to-use, low latency analysis facilities

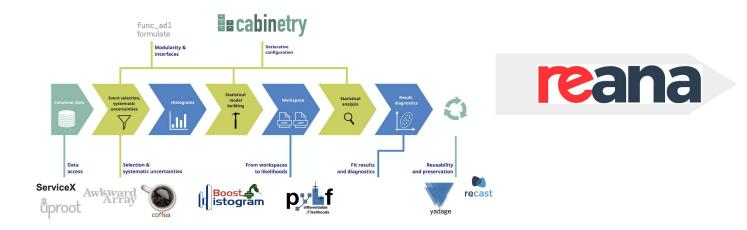


**Execution of AGC analysis benchmark** 

## Analysis Grand Challenge

- Columnar data extraction from large dataset
- Processing of that data (event filtering, construction of observables, evaluation of systematic uncertainties) into histograms
- Statistical model construction and statistical inference
- Relevant visualisation for this steps

### + Adding analysis preservation step to AGC pipeline



## Next step: REANA specification file

The REANA reproducible analysis platform requires to have reana.yaml file present in your analysis source code (REANA specification file).

Its purpose is to answer the Four Questions:

- 1. What is your input data? (e.g. dataset samples)
- 2. What is your analysis code? (e.g. python notebook, compiled executable, script)
- 3. What is your computing environment? (e.g. docker image)

4. Which computational steps do you take to arrive at results? (e.g. data processing or statistical model construction and statistical inference)



## Next step: Workflow management engine choice

- Our choice was to use **Snakemake workflow management system** (integrated in REANA)
  - Help to keep a record of used scripts and their input dependencies
  - Run multiple steps in sequence, parallelising where possible
  - Automatically detect if something changes and then reprocess data if needed
- Snakemake key feature is a "rule" description, which enables the parallelisation within REANA, running each rule in a separate virtual node.
- <u>Snakemake allows you to create a set of rules, each one defining a</u> <u>"step" of your analysis.</u>
- The rules need to be written in a file called Snakefile:
  - The input: Data files, scripts, executables or any other files.
  - The expected output: It's not required to list all possible outputs. Just those that you want to monitor or that are used by a subsequent step as inputs.
  - Shell: A command to run to process the input and create the output.

Example of Snakemake rule:
<pre>rule ttbar_nominal_sample:</pre>
input:
"samples.ipynb"
output:
"histograms_ttbarnominal.root"
params:
<pre>key_to_extract = 'ttbar',</pre>
variation = 'nominal'
container:
"povstenandrii/papermill:20231102a"
shell:
"/bin/bash -1 && source fix-env.sh &&

papermill {input} sample1\_out.ipynb -p
key\_to\_extract {params.key\_to\_extract} -p
variation {params.variation} -k python3"

## **REANA** specification file

### Version without parallelisation

#### workflow:

type: serial

#### specification:

#### steps:

- name: demoanalyzer
- environment:
- 'hub.opensciencegrid.org/iris-hep/analysis-systems-b
  ase:latest'
  - commands:
  - /bin/bash -1 && source fix-env.sh && python
- ttbar\_analysis\_pipeline.py

outputs:

files:

- merged\_histograms.root

Extremely long execution time since AGC was written in mind without support for other workflow languages

### Version with parallelisation

#### inputs:

#### files:

- ttbar\_analysis\_pipeline.py
- ttbar\_analysis\_pipeline.ipynb
- nanoaod\_inputs.json
- nanoaod\_branch\_rations.json
- corrections.json
- Snakefile
- merged.ipynb
- directories:
  - histograms
  - models
- utils
- files
- workflow:
- type: snakemake
- file: Snakefile
- outputs:
- files:
  - merged\_histograms.root

### AGC notebooks modification:

- Rerun the same notebook n-times but with different parameters => instead of processing all files, samples we process one sample with one file
- Firstly we parallelized each sample from fileset:

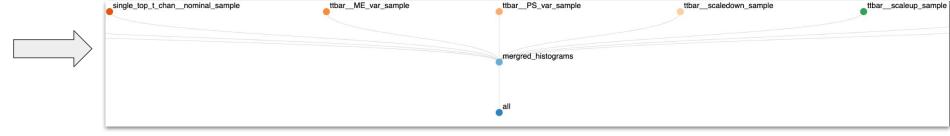
```
original_dict = fileset
selected_file = original_dict[sample_name]['files']
new_dict = {sample_name: {'files': [filename], 'metadata': original_dict[sample_name]['metadata']}}
```

- Second, parallelize each file for each sample:

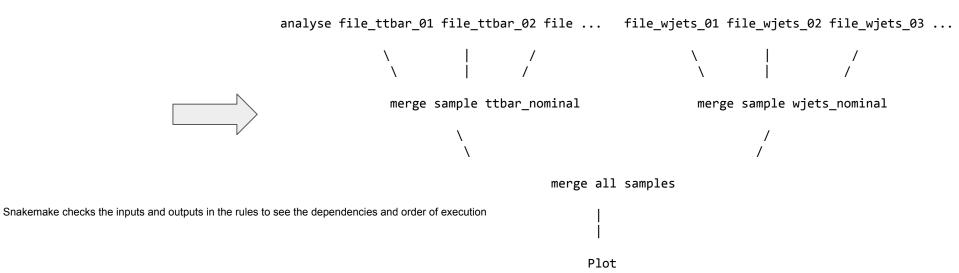
```
all_histograms, metrics = run(
    fileset={sample_name: new_dict[sample_name]},
    treename=treename,
    processor_instance=TtbarAnalysis(USE_INFERENCE, USE_TRITON)
)
```

The main idea is to see the whole picture of your analysis what steps suppose to be after another and modify it on the early stages to have a separate pieces which could be count as 1 job.

## Analysis Grand Challenge pipeline: Adapting to Snakemake



Each rule REANA sends to the Kubernetes cluster as separate node



### Next step: Execute REANA workflow

### To run the workflow use the: reana-client run -f reana.yaml -w your-workflow

(reana) andrewpovsten@pucomphep03 cms-open-data-ttbar % reana-client run -f reana-snakemake.yaml -w snakemake-multicascading ==> Creating a workflow... lob stats:

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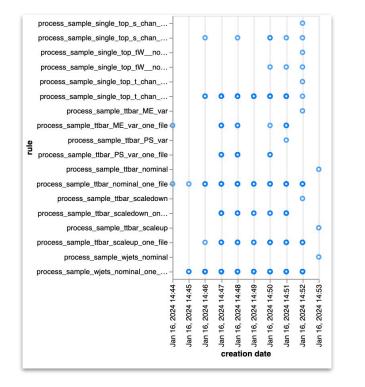
process\_sample\_ttbar\_scaleup
process\_sample\_ttbar\_scaleup\_one\_file

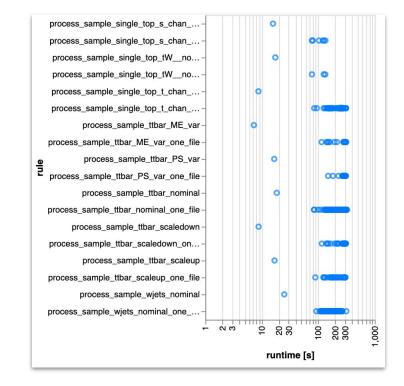
process\_sample\_wjets\_nominal

32

### Current results:

The processing, and merging all files in each sample takes around 21 minutes with REANA. But this results can be improved further.





### Conclusion

- Successfully implement the AGC at REANA
- Get parameterized AGC notebook and execute it with papermill tool

### Future Tasks

- Identify the possible ways of improving AGC adoption for REANA workflow and RECAST
- Demonstrate HEPData submission of AGC artefacts for reusability
- Further optimisation of AGC processing time in REANA
- Testing on larger clusters
- Testing AGC ServiceX and machine learning pipelines in REANA