

# Analysis Grand Challenge benchmarking tests on selected sites

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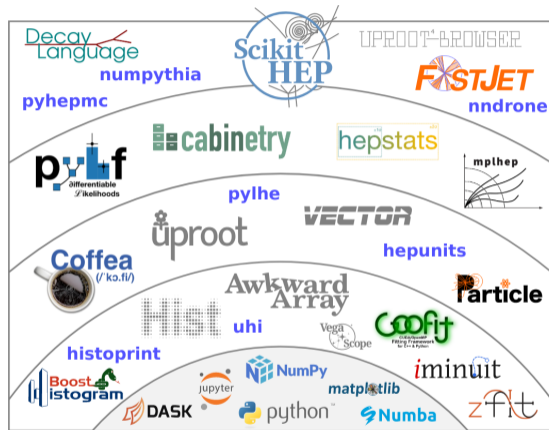
CHEP 2023



Bundesministerium  
für Bildung  
und Forschung

# Analysis Grand Challenge

- developed by the IRIS-HEP team
- effort to demonstrate feature-completeness and scalability of scikit-HEP tools
- main framework of the analysis: **coffea**, offers a high level interface for columnar analysis
- **github**, **readthedocs**



Scikit-HEP: Python ecosystem for HEP analyses

# Analysis Grand Challenge

$t\bar{t}$ -Analysis includes

- 1-lepton event selection
- top reconstruction
- cross-section measurement
- on-the-fly evaluation of systematic uncertainties

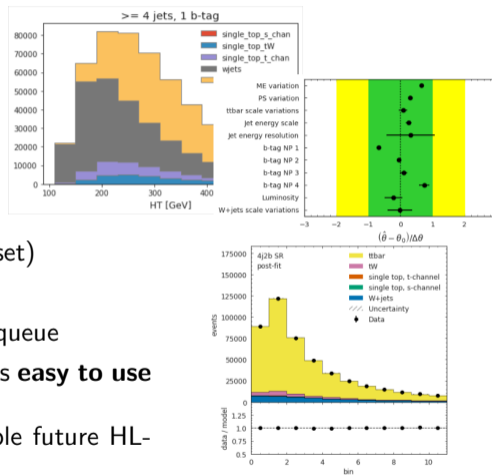
# Analysis Grand Challenge

ttbar-Analysis includes

- 1-lepton event selection
- top reconstruction
- cross-section measurement
- on-the-fly evaluation of systematic uncertainties
- total of 3.44 TB of CMS open data
- only  $\sim 138$  GB are actually read (4% of the total dataset)
- 948 mio events and 10 variables
- work is distributed across many workers with dask-jobqueue

... all this sits in a single Jupyter notebook  $\Rightarrow$  analysis code is **easy to use**  
should also be **scalable** and **fast**

The AGC analysis is meant as a showcase of how a possible future HL-LHC analysis could look like



Talks at CHEP directly related to the AGC:

- Elliott Kauffman, [Machine Learning for Columnar High Energy Physics Analysis](#), Monday 2pm
- Andrea Scabià, [I/O performance studies of analysis workloads on production and dedicated resources at CERN](#), Monday 3pm
- Oksana Shadura, [Coffea-Casa: Building composable analysis facilities for the HL-LHC](#), Tuesday 10am
- Alexander Held, [Physics analysis for the HL-LHC: concepts and pipelines in practice with the Analysis Grand Challenge](#), Tuesday 5pm
- Vincenzo Padulano, [First implementation and results of the Analysis Grand Challenge with a fully Pythonic RDataFrame](#), Tuesday 5:15pm

# Analysis Grand Challenge Benchmarks

Benchmarks performed on three different sites:

- LMU institute cluster at LMU Munich consisting of one very powerful node and desktop computers
- job-scheduler: SLURM
- reading of the data via xrootd from LRZ

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  - reading of the data via xrootd from LRZ
- LRZ** WLCG Tier-2 site in Munich
  - job-scheduler: SLURM
  - data is stored on regular Grid storage (HDD) as well as on a XCache server (SSD)

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data is stored on regular Grid storage (HDD) as well as on a XCache server (SSD)

**Vispa** analysis facility operated by RWTH Aachen; provides a web-based terminal, code editor and jupyter hub: <https://vispa.physik.rwth-aachen.de>

job-scheduler: HTCondor

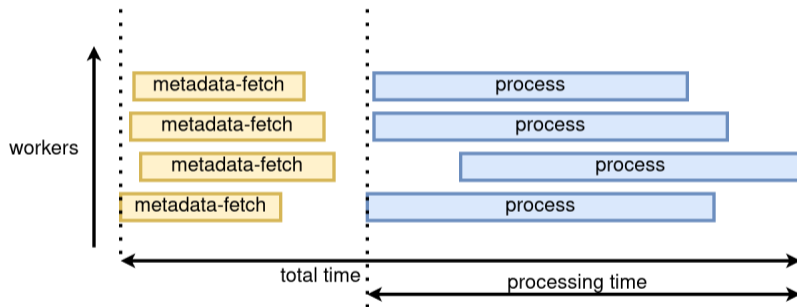
data is stored locally on SSDs and read directly; Vispa also has a very dedicated caching-system ([arXiv](#)) that I did not test with AGC yet



# Measurements

## Runtime

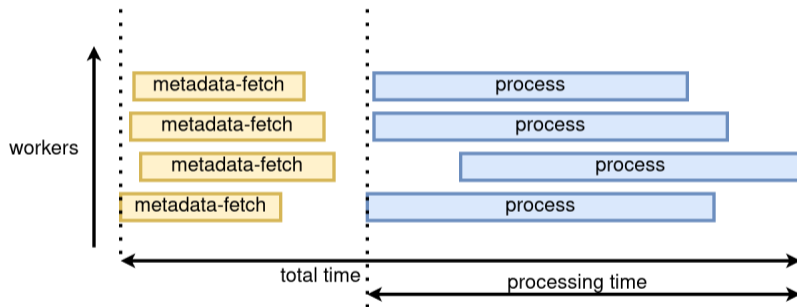
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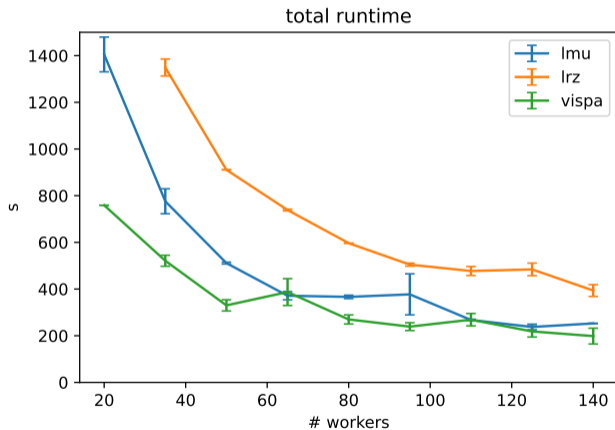


I can measure directly

- the total runtime
- the total processing time
- the sum of all process times across the workers (the sum of all blue rectangles here) via coffea's tooling

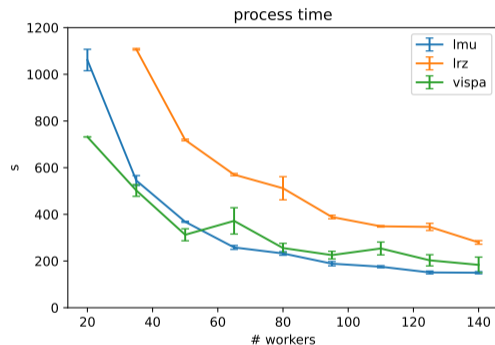
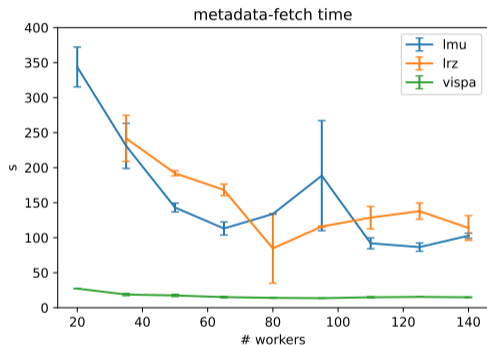
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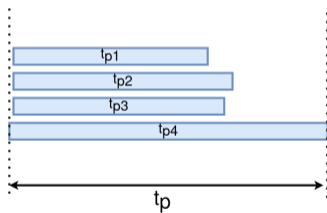
metadata-fetch time (= total - process time, also contains waiting and communication between disk workers) and process time

# Measurements

## Runtime

measure for the amount of overhead  
relative to the runtime

$$\frac{t_p}{\sum t_{p_i}/n}$$



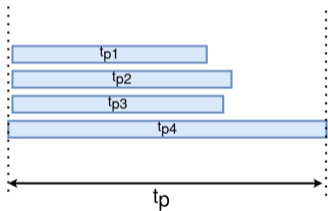
$\sum t_{p_i}/n =: \bar{t}$  is the average process  
time per worker – *pure computing*  
time without overhead

# Measurements

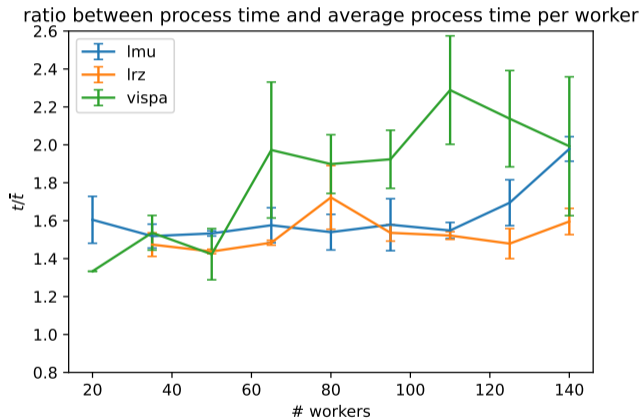
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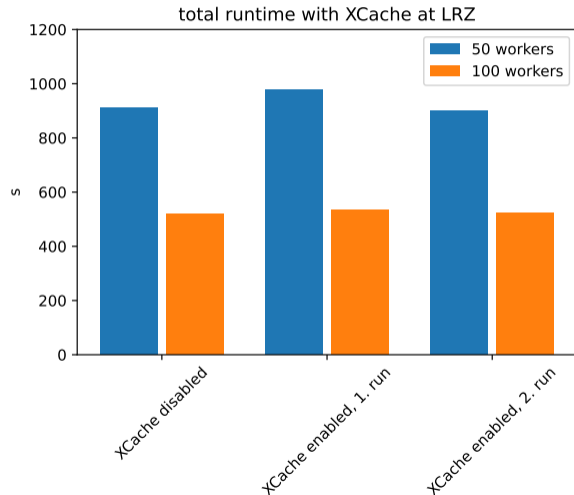
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overhead

# Measurements

## XCache



runtimes at LRZ with and without XCache enabled: makes no significant difference  
⇒ with this setup, the analysis is hardly I/O limited

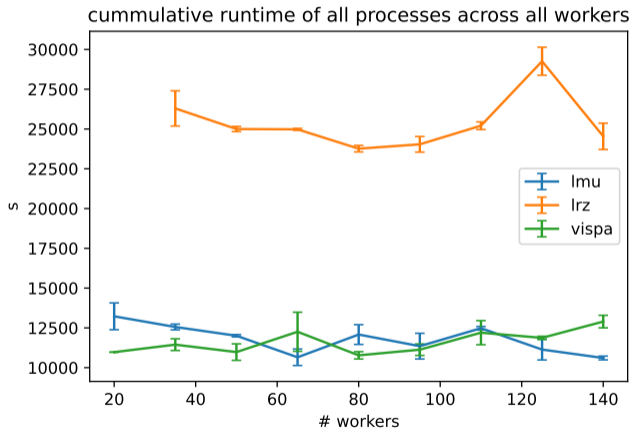
# Questions?



# Backup

# Measurements

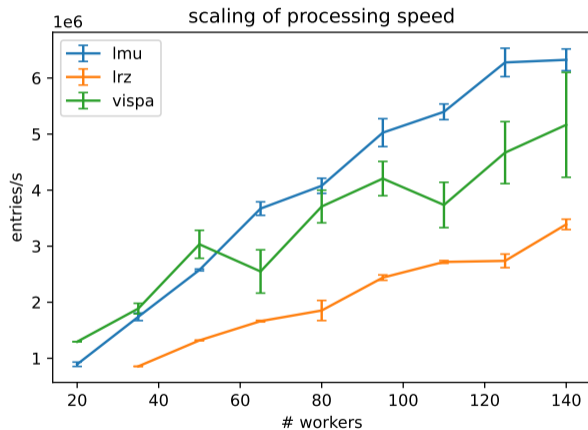
## Runtime



summed process time across all workers

# Measurements

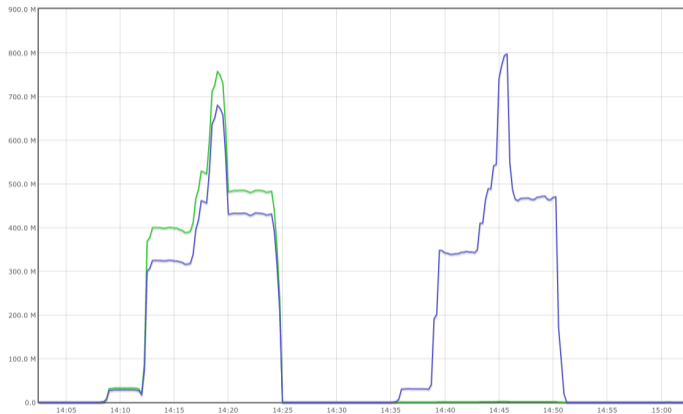
## Runtime



Scaling of process time

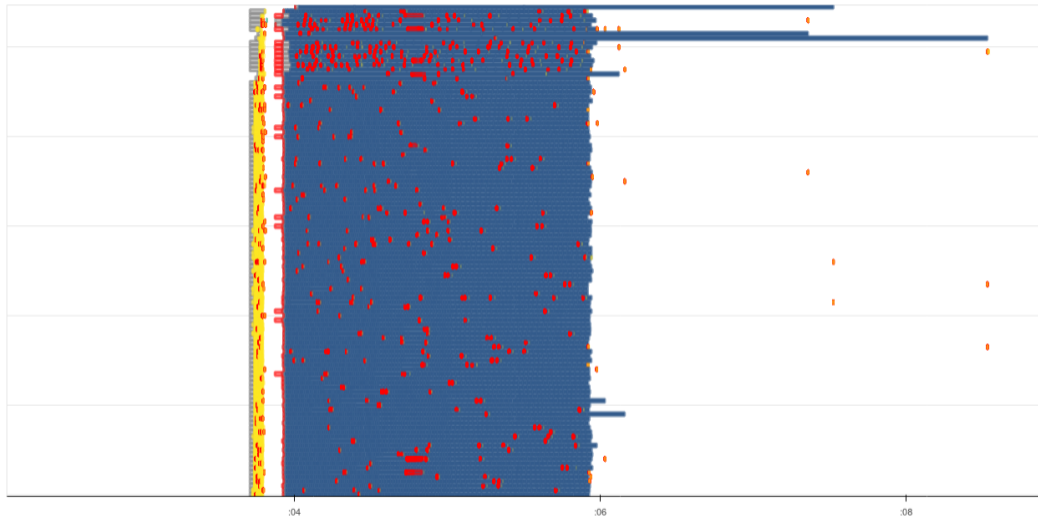
# Measurements

## XCache

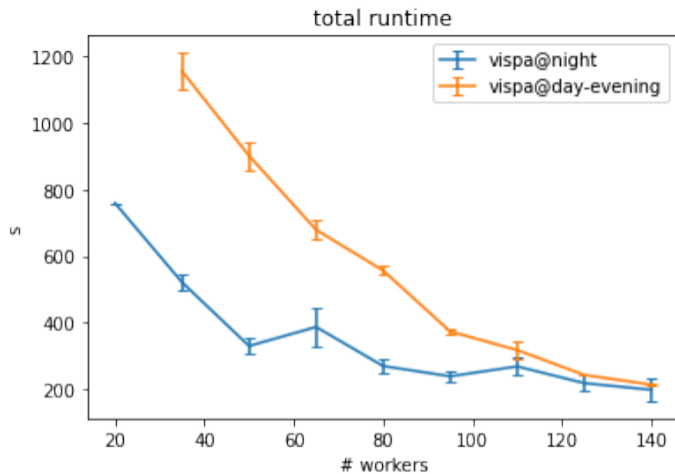


XCache load during two consecutive runs with 50 workers

Task Stream



example Dask dashboard with long tail



total runtimes during the day and night @Vispa