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## **Benchmark interactive analyses ongoing at INFN Napoli**

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**Spoke 2 - WP2 working meeting, 21st May 2024**

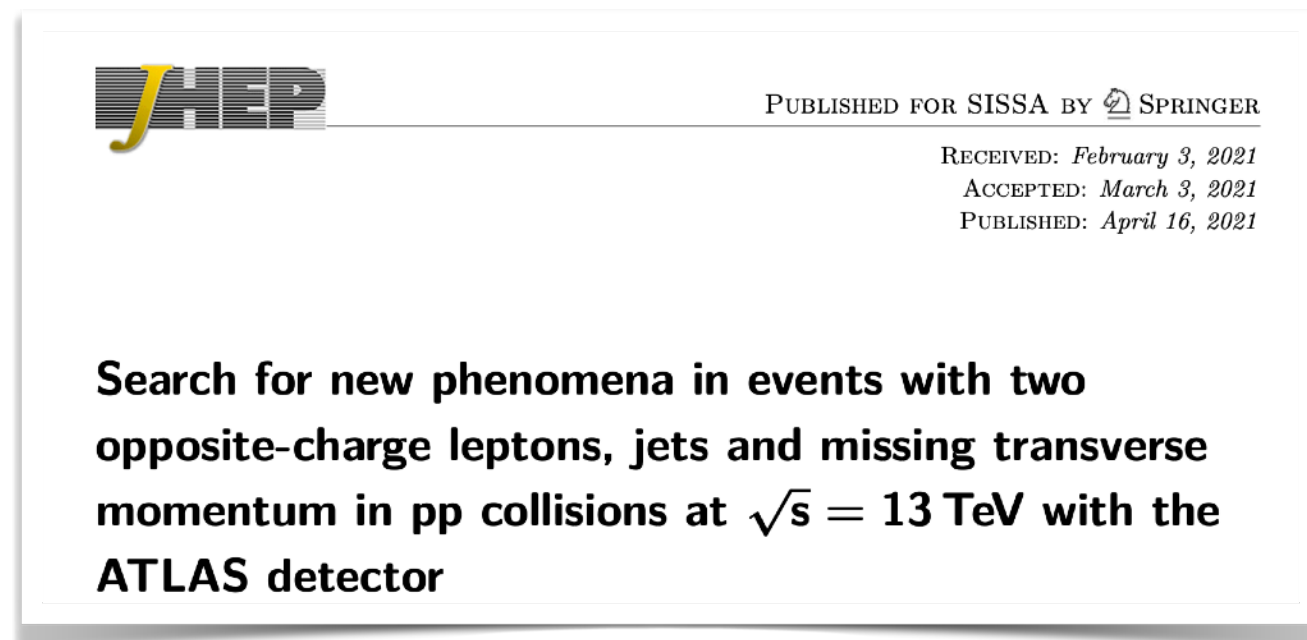
# Outline

- Motivations
- Use cases tested:
  - ATLAS: stop to 4-body SUSY analysis
    - ➔ collaboration with INFN Lecce
  - CMS: top quark+MET analysis
    - ➔ collaboration with INFN Perugia

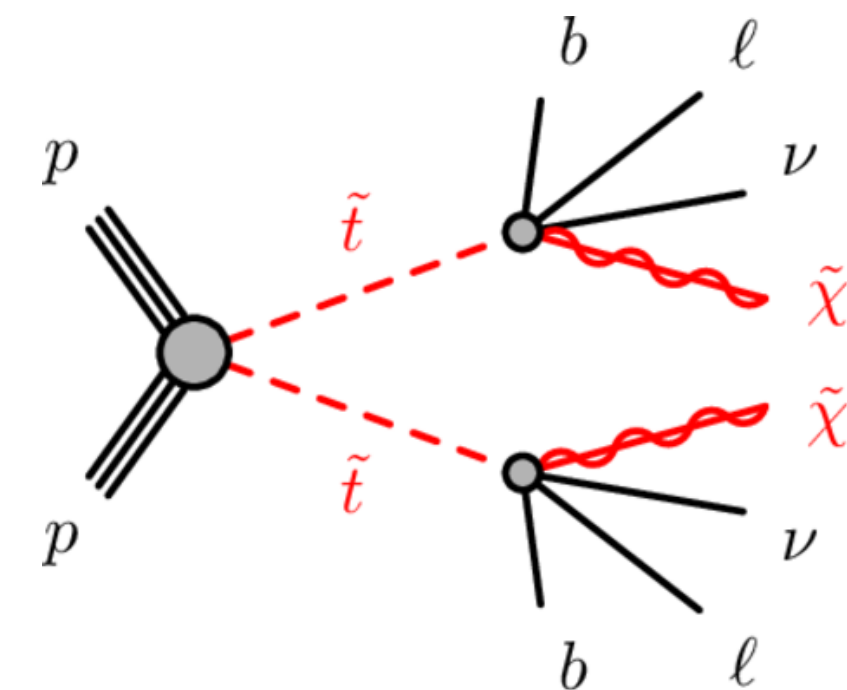
# Motivations

- Most of the LHC searches/measurements rely on locally developed scripts that process the datasets, with parallel tasks and on an asynchronous batch system
- **Challenges of HL-LHC and future** colliders are pushing to **re-think the HEP computing models**
  - 📌 Impact on several aspects, from software to the computing infrastructure
- From the software perspective, **interactive/quasi interactive analysis** is a promising paradigm
  - 📌 User-friendly environment
  - 📌 The implementation is simplified by adopting open-source industry standards: *Dask*, *Jupyter Notebooks* and *HTCondor*
  - 📌 Validating new frameworks (e.g. *ROOT RDataFrame* with multi-threading)
- **Preliminary feasibility studies** have been pursued exploiting **INFN Napoli** high rate platform
  - 📌 Distributed infrastructure which leverages *Dask*

# ATLAS use-case



## SUperSYmmetry: Beyond Standard Model theory



Soft leptons coming from a virtual  $W^*$  boson decay

Compressed mass spectra:  
 $\Delta m < m_W + m_b$

- Three different analysis in the *Run 2 paper*, already published, according to mass splitting between *stop* ( $\tilde{t}_1$ ) and *neutralino* ( $\tilde{\chi}^0_1$ ), allowing different decay modes:

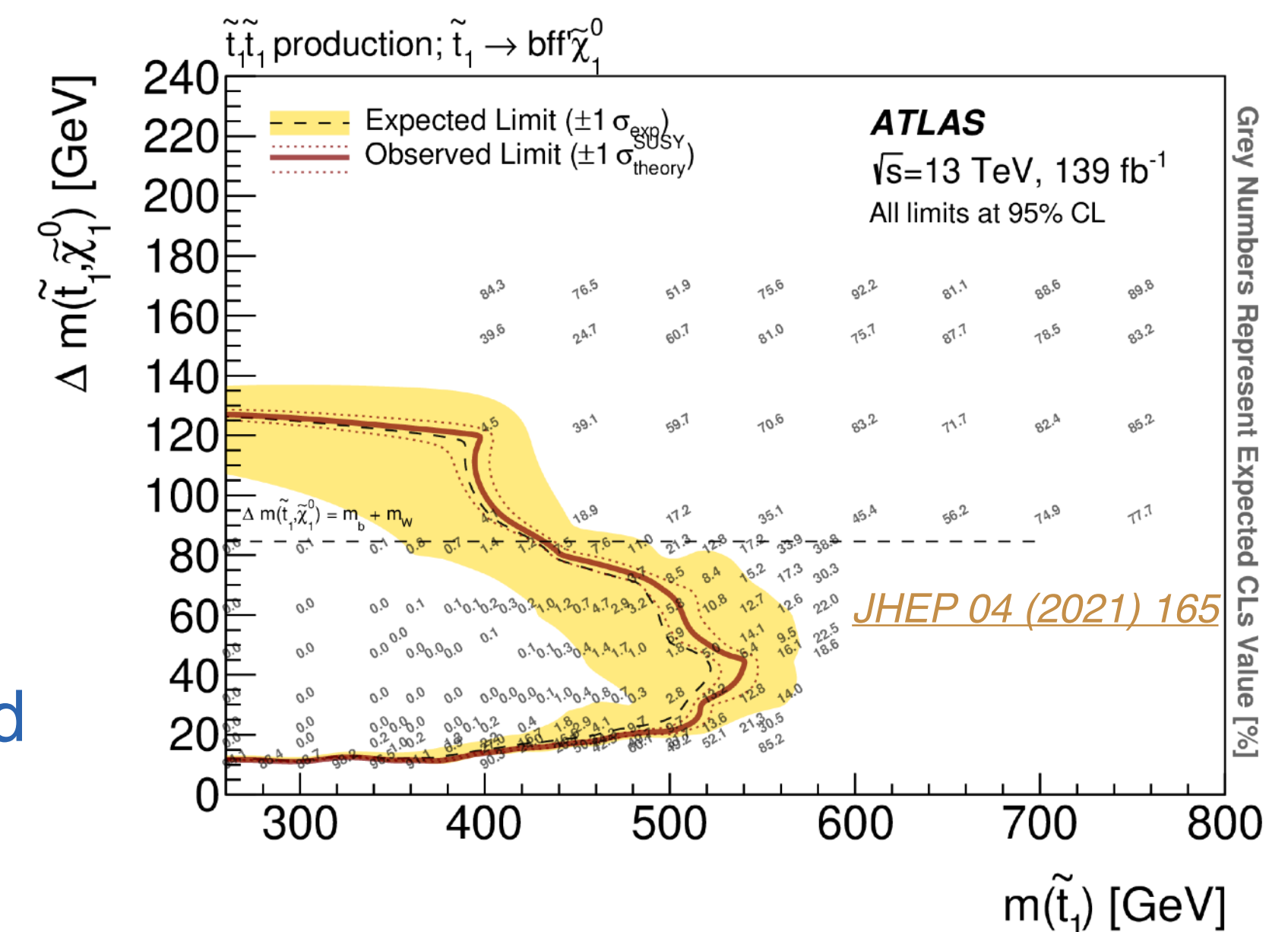
2 body  $\rightarrow \Delta m > m_t$

3 body  $\rightarrow m_W + m_b < \Delta m < m_t$

4 body, the one picked up  $\rightarrow \Delta m < m_W + m_b$

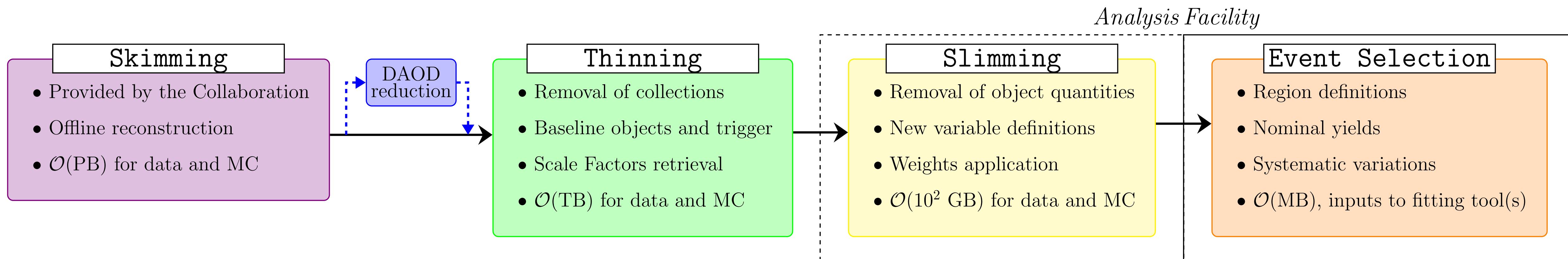
- Common final state signature: 2 OS leptons (electrons/muons), jets and missing transverse energy

- Cut & Count based approach



Working with INFN Lecce

# 4-body search workflow



**Slimming**


ATLAS slimming code already in RDataFrame, but entirely written and compiled in C++  $\rightarrow$  NO dask distributed approach

**Event Selection**

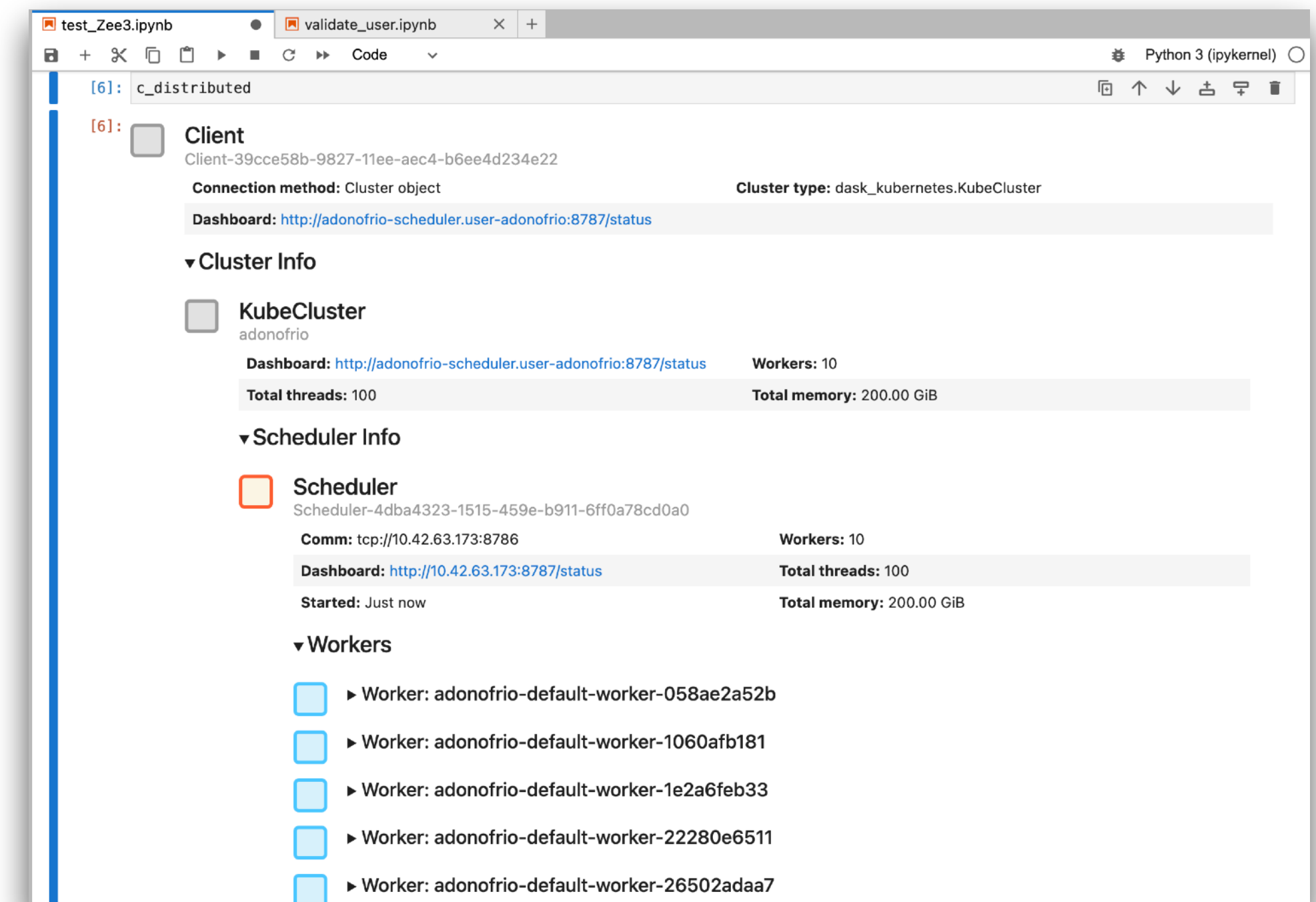
- Event selection for fitting tools
- RDataFrame + Dask applied to Wt bkg sample  $\sim 1.8 \text{ GB}$  copied to the INFN Na workspace
- Playing with syst. variations
- Code ready to play with other backgrounds



# Infrastructure and kubernetes

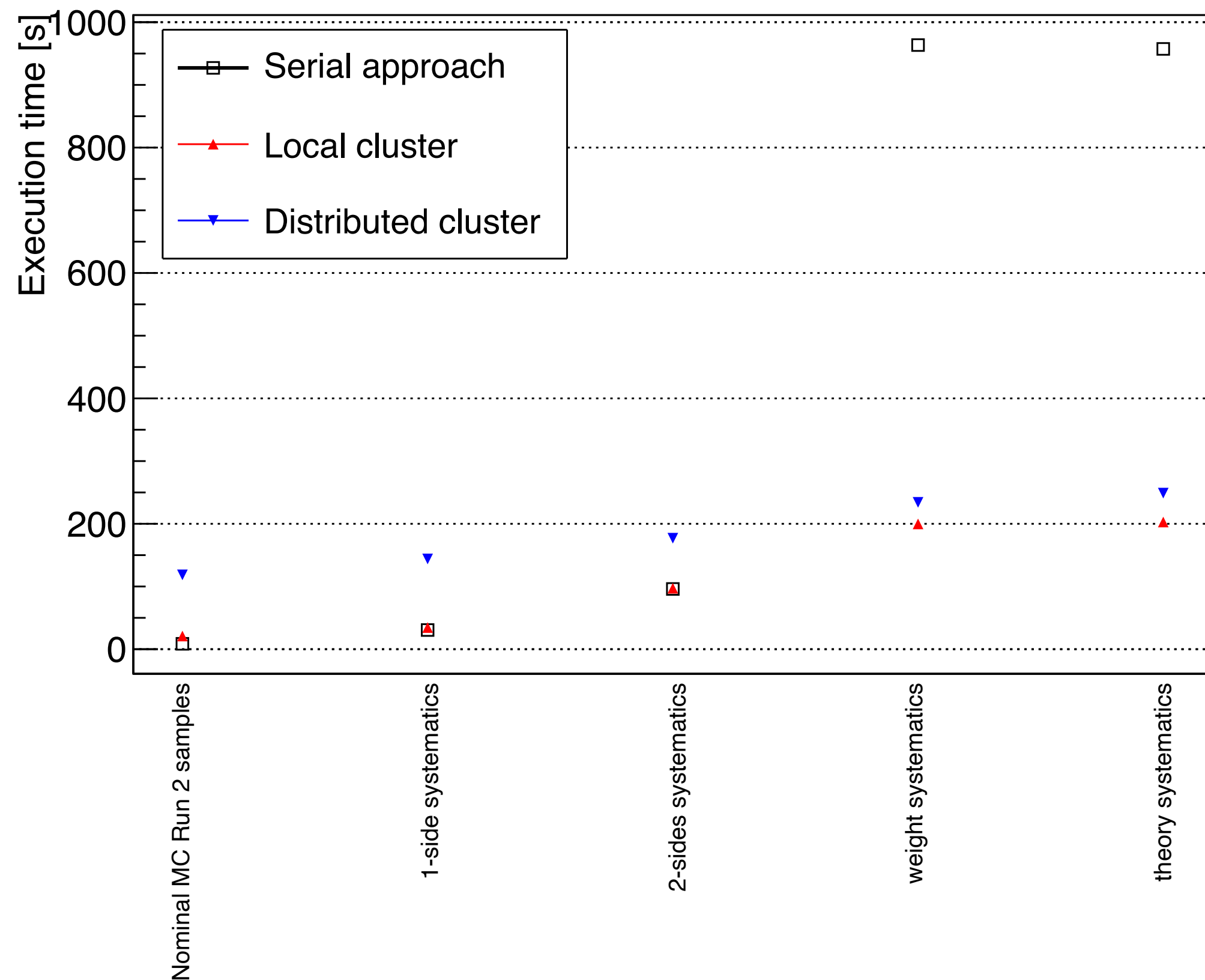
- The local (INFN Naples) deployment is based on the *Open-Stack IaaS* paradigm
- Starting from the already existing *I.Bi.S.CO* installation, several updates were performed
- The physical cluster is made up of 2 identical virtual machines, each equipped with 1CPU quadCore and 8GB RAM, currently expanded up to 12 cores and 64GB
- Rocky Linux 8.6 is the operating system
- 2 nodes are equipped with **Docker** (20.10) for containerisation and **Kubernetes** (1.26.3) for the orchestration
-  One node plays as controlplane, etcd & worker; the other node acts as a plain worker
- The cluster is equipped with **JupyterHub** & **JupyterLAB** where the user can play with **Python**, **ROOT** & **Dask** libraries

- Kubernetes infrastructure: 5+1 virtual machines
- 32 cores & 64 GiB in each compute
- If more users are connected, scheduling jobs priority
- Clusters are dynamic objects

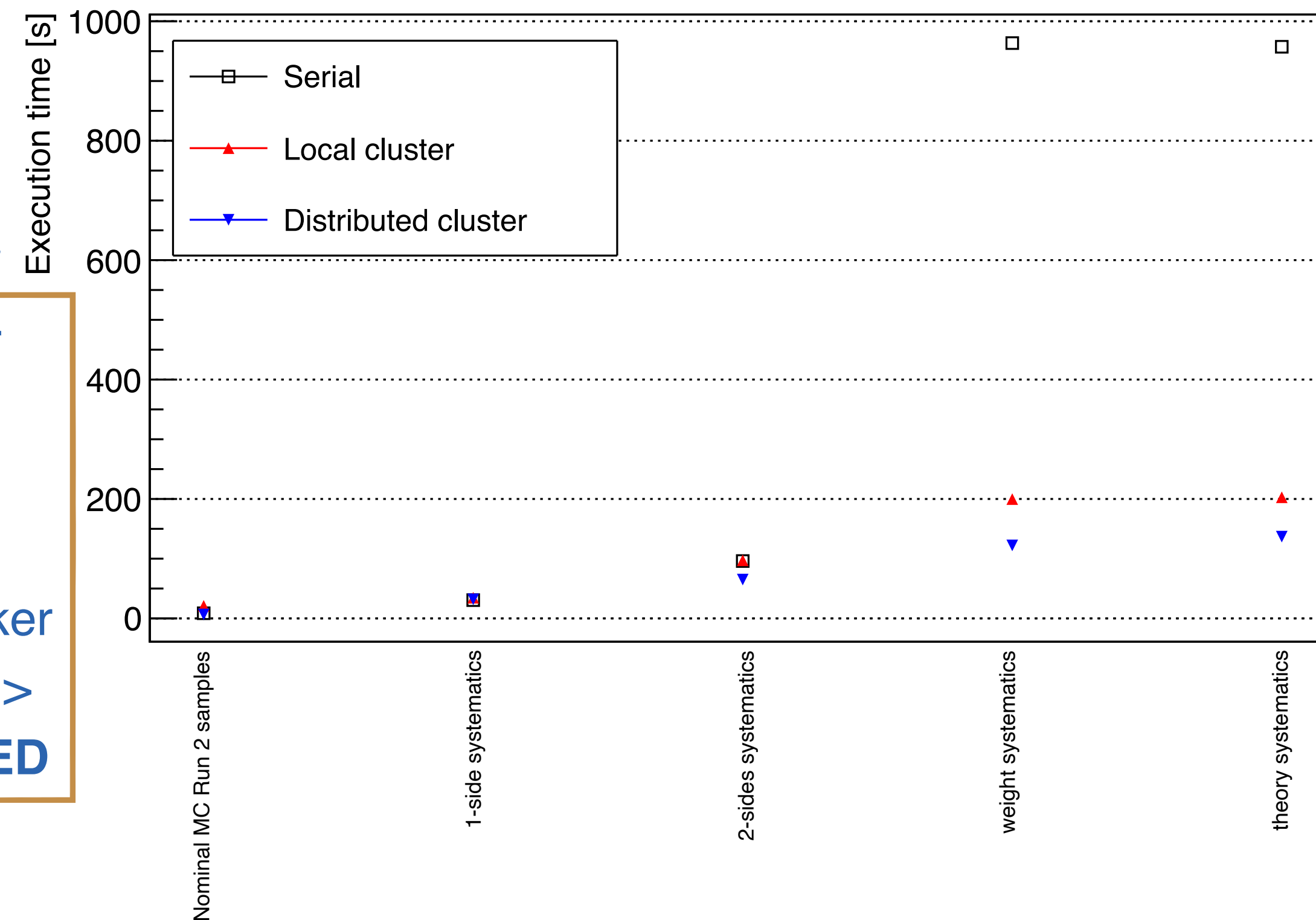


# Preliminary results

Defined Metric	
Overall execution time	Time elapsed from the start of the execution (execution triggered) to the end of execution



Limiting factor for the distributed approach: time spent to copy the inputs to the worker nodes (~120 s) → **TO BE IMPROVED**



- Exploiting the distributed approach, the execution time improves *wrt* the standard/serial approach if we iterate over a significant number of systematic variations (each step in the x-axis includes previous contributions)

# Scheduler and Working Nodes Reports

Distributed approach

## Dask Performance Report

Select different tabs on the top for additional information

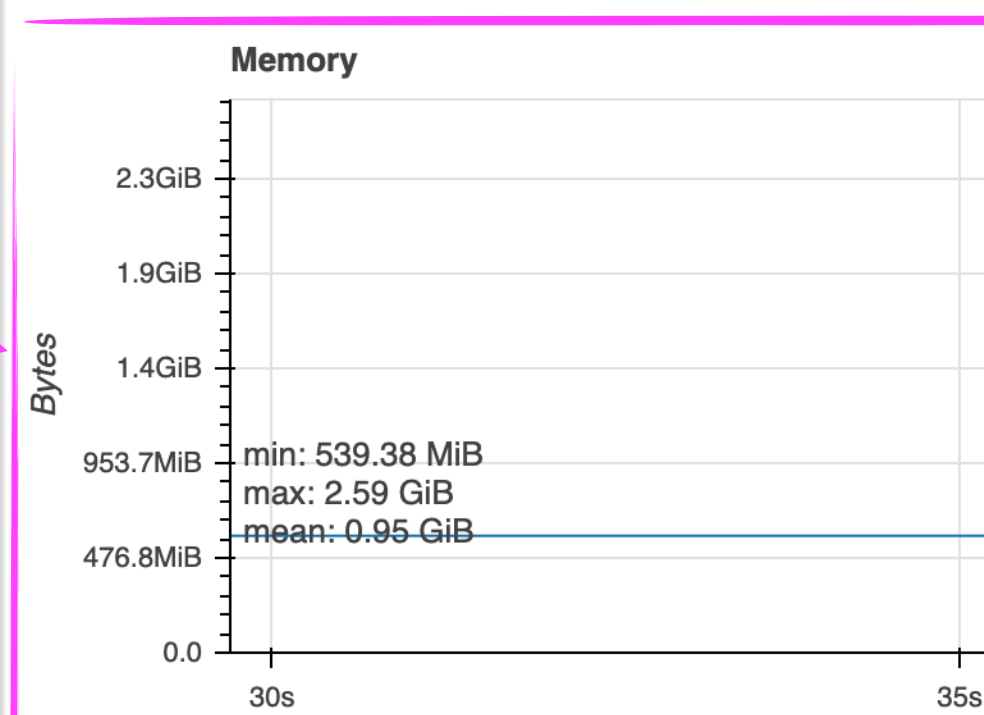
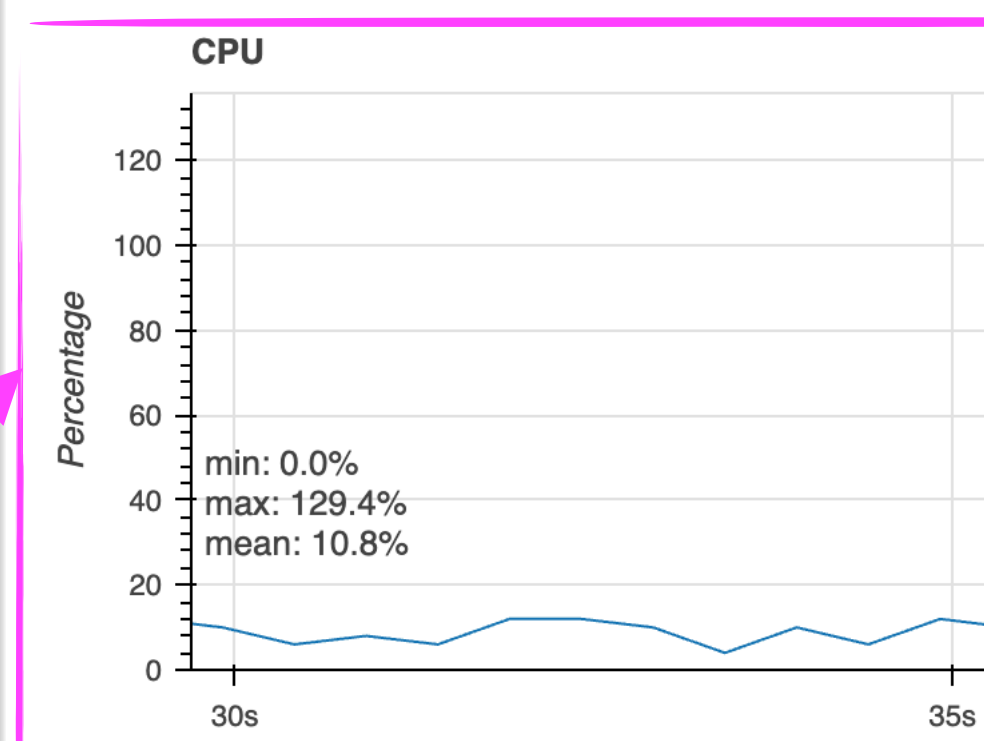
Duration: 252.87 s

### Tasks Information

- number of tasks: 621
- compute time: 118.06 s
- deserialize time: 2.39 s

### Scheduler Information

- Address: tcp://127.0.0.1:43821
- Workers: 2
- Threads: 2
- Memory: 4.39 GiB
- Dask Version: 2022.11.0
- Dask.Distributed Version: 2022.11.0



## Connecting to working nodes

- Out of 9 worker nodes, we get up to 91% CPU occupancy on one node and up to 5% on the other nodes

```
lalmalinux@kuber-node-2 ~|$ top
```

```
top - 10:47:40 up 17 days, 22:40, 1 user, load average: 0.47, 0.28, 0.20
Tasks: 504 total, 1 running, 503 sleeping, 0 stopped, 0 zombie
%Cpu(s): 3.1 us, 0.4 sy, 0.0 ni, 96.2 id, 0.2 wa, 0.1 hi, 0.0 si, 0.0 st
MiB Mem : 64298.3 total, 52125.8 free, 6999.9 used, 5863.5 buff/cache
MiB Swap: 0.0 total, 0.0 free, 0.0 used. 57298.4 avail Mem
```

PID	USER	PR	NI	VIRT	RES	SHR	S	%CPU	%MEM	TIME+	COMMAND
3467064	almalin+	20	0	1077988	627812	378248	S	91.4	1.0	0:05.15	python
3436893	almalin+	20	0	6841396	553904	170460	S	3.0	0.8	0:15.16	python
2163103	almalin+	20	0	544308	118628	29288	S	2.3	0.2	94:53.01	python3.10
1244708	almalin+	20	0	6485488	390772	27588	S	2.0	0.6	337:41.39	python
1574	root	20	0	4011788	148624	72036	S	1.7	0.2	755:31.40	kubelet
1342269	almalin+	20	0	544320	116988	29364	S	1.7	0.2	250:13.90	python3.10
1245200	almalin+	20	0	1165092	378416	19604	S	1.3	0.6	279:15.23	python
3462166	almalin+	20	0	2453528	1.9g	29908	S	1.3	3.1	0:13.91	python3.10
3467062	almalin+	20	0	438548	87068	27772	S	1.3	0.1	0:02.15	python
1245202	almalin+	20	0	1189084	437148	19584	S	1.0	0.7	260:06.77	python
3461916	almalin+	20	0	4722168	61600	19928	S	1.0	0.1	0:06.10	dask-worker
1173	root	20	0	3428412	75068	32896	S	0.7	0.1	63:20.86	containerd
2163062	almalin+	20	0	4797108	65568	19992	S	0.7	0.1	53:46.23	dask-worker
3467329	almalin+	20	0	16408	7092	5576	R	0.7	0.0	0:00.17	top
3947642	almalin+	20	0	811908	148688	18232	S	0.7	0.2	5:05.35	jupyterhub-sing
1294	root	20	0	4272748	138224	58384	S	0.3	0.2	286:48.93	dockerd

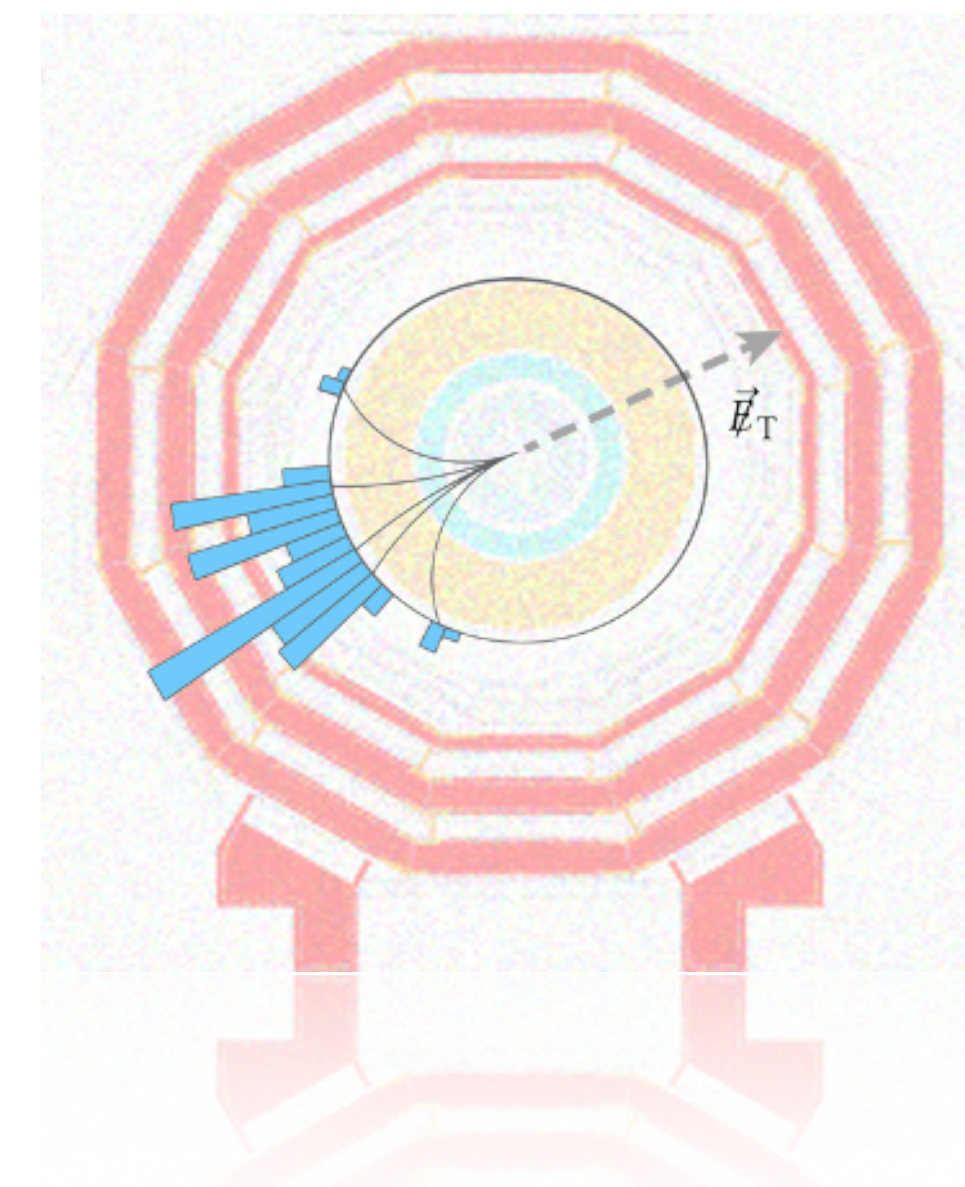
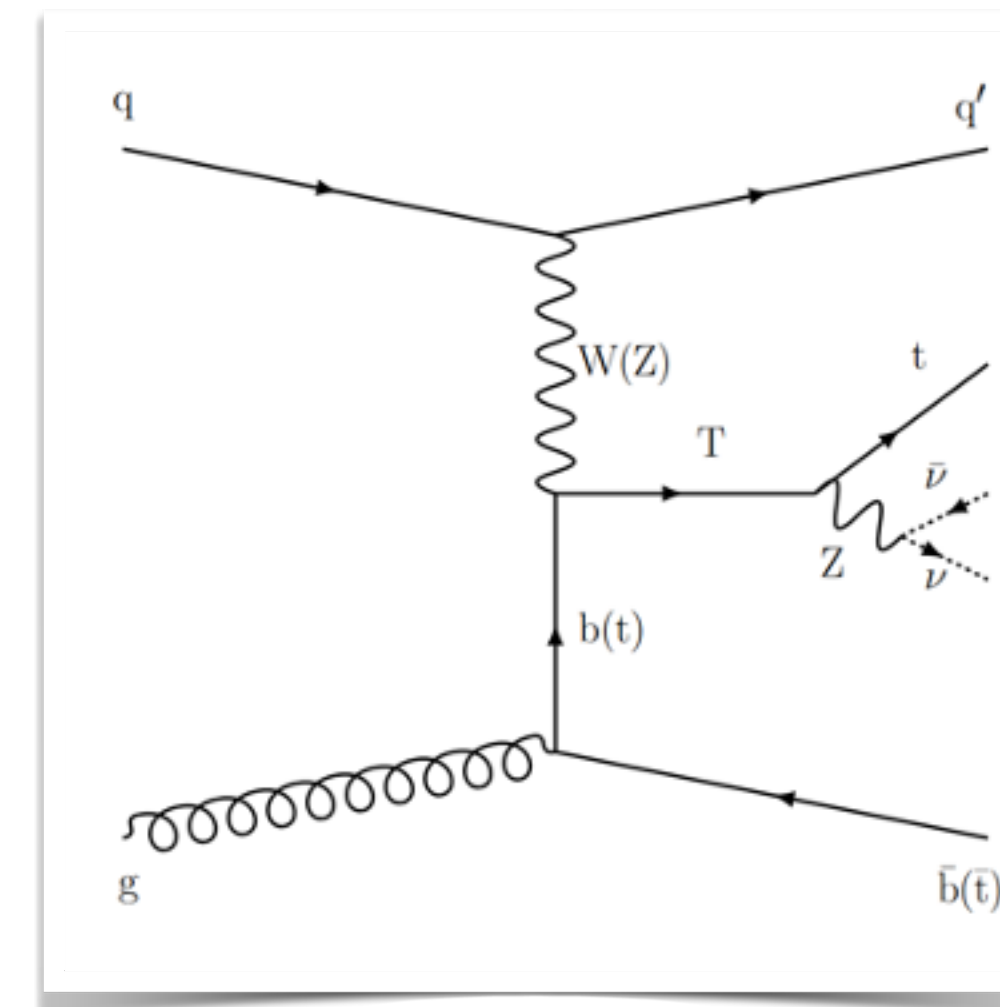
</home/jovyan/work/pnrr-atlas-analysis/dask-report.html>



# CMS use-case

- Early Run 3 analysis (2022-2023 data taking)
- Beyond Standard Model searches
- Vector-Like Quark T in  $T \rightarrow tZ$  channel
- Final state: hadronic Top quark and Z ( $\nu\nu$ )
- Development of the already published full Run 2 analysis [\*JHEP05\(2022\)093\*](#), with the idea to extend the results interpretation to more models predicting the same final state
- Dark Matter production in association with a Top quark
- Technicolor models [\*The Radiative Flavor Template at the LHC\*](#)

24/05 WP2.5 presentation [link](#)



# State of the art

Workflow in 2 steps:

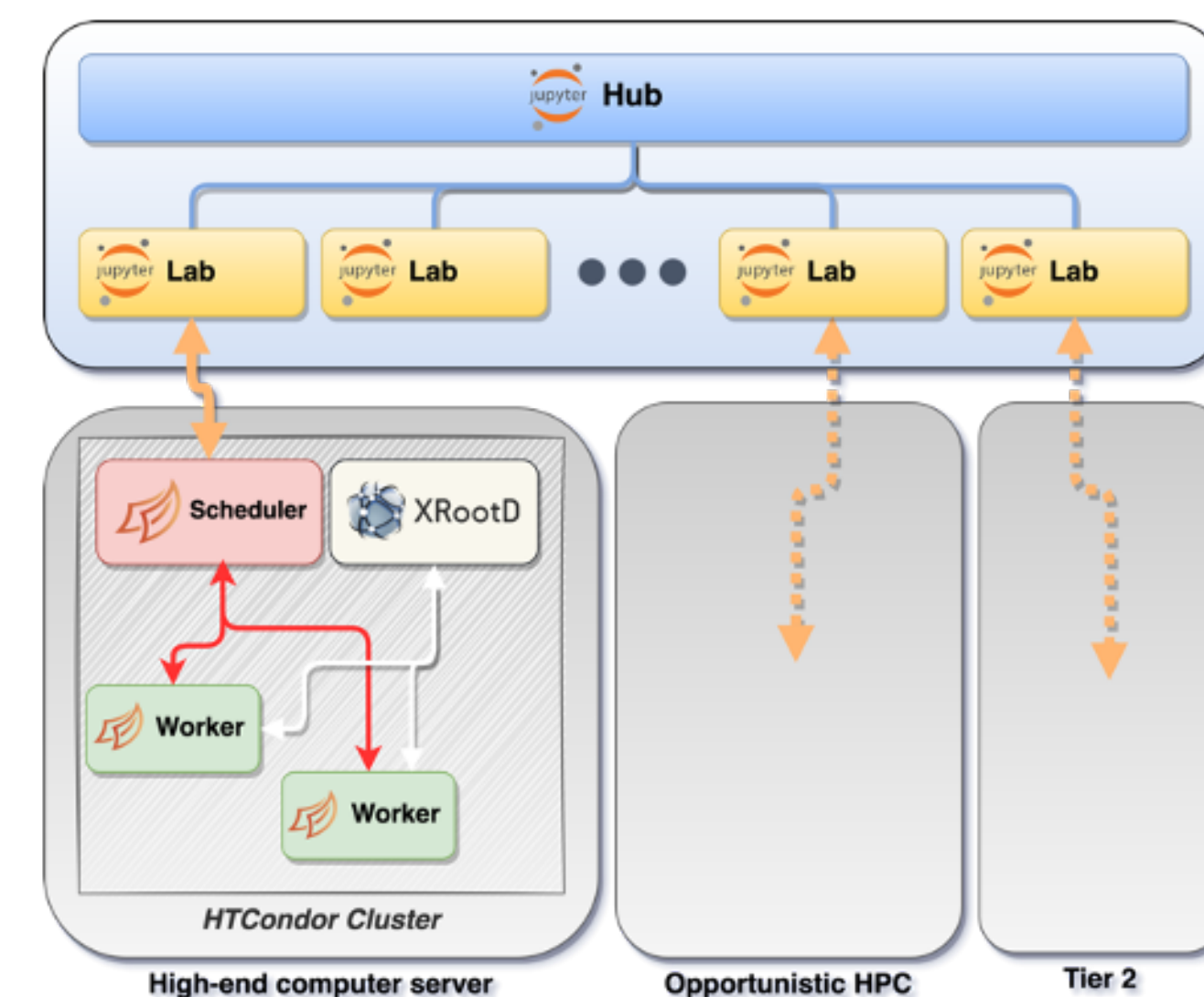
- Data preprocessing, evaluation through ML model.

Using CMS NanoAOD tools (pyROOT-based) and CRAB.

- Skimming and selection using Interactive Analysis

- Input: ntuple from the 1st step
- Selection + variables calculation through RDataFrame
- Distribution of the process using Dask
- Output: TH1D easy to manage, also possible to store snapshot using remote storages

Working on Perugia's analysis facility



- Analysis still far from the end, more processes will have to be added that will slow it down

- Currently the results are very promising

Time reduced from ~1d to ~3h and there is still room for development

# Conclusions & Next Steps

- Interactive analyses feasibility studies on the INFN Naples infrastructure succeeded
- Three use cases tested (ATLAS, CMS, FCCee), in different scenarios: different experiments and analyses
- Towards an INFN national cloud infrastructure with a datalake model to facilitate future analyses (hopefully starting from LHC Run 3)

## ➔ **Thoughts about ATLAS use case:**

- I/O implementation to be improved, feasible approach: including grid certificate and voms in the jupyter image
- We currently exploit a cut & count analysis as use case: this is relatively easy from a computational point of view and to fully benefit from the distributed approach we could think about workflows including for example ML techniques

## ➔ **Long term goals:**

- 🔧 The current analysis is a successful case of migration to a new more efficient paradigm
- 🔧 End-goal is to eventually test also the production steps, i.e. exploring the feasibility of a passage: WLCG ➔ Interactive Analysis
- 🔧 Also in view of HL-LHC and potential future applications

The background is a deep blue gradient. On the left side, there is a complex pattern of light trails and particles. These trails are composed of many thin, curved lines that appear to be moving or vibrating, creating a sense of depth and motion. Small, bright blue dots are scattered along these trails, some appearing as larger, more prominent spots. The overall effect is reminiscent of a digital or data visualization, possibly representing a network or a flow of information.

**Thank you!**

# Conferences and workshops contributions

- ECFA 2023 : talk about FCCee Zee benchmark use case —> **done**
- ICHEP2024: contribution accepted as poster (ATLAS SUSY analysis)
- CHEP2024 : contribution submitted via ATLAS computing speakers committee (ATLAS SUSY analysis)
- SIF2024 : contribution submitted (ATLAS SUSY analysis)