

Leveraging public cloud resources for the processing of CMS open data

CHEP - October 19 - 25, 2024

Kati Lassila-Perini Helsinki Institute of Physics - Finland Tom Cordruwisch, Subash Jayawardhana Lapland University of Applied Sciences - Finland



| D pendata CERN Type something | Decade of <u>CMS op</u> - with a small d | <u>en data</u> edicated team | Help About - Search | | | |
|--|---|--|---|--|--|--|
| | 13 result(s) found | Sort by Most recent | | | | |
| Current parameters Clear all | CMS releases 13 TeV proton collision data from 2 f CMS completes Run-1 heavy ion open data collect | | v references.reference.dois:10.7483/OPENDATA.CMS ⁴ Q Literature Authors Jobs Seminars Conferences More | | | |
| Availability include on-demand datasets | CMS completes the release of its entire Run-1 pro First CMS open data from LHC Run 2 released CMS releases heavy-ion data from 2010 and 2011 | Date of paper | 85 results E cite all Citation Summary De Most Recent V Bridging Worlds: Achieving Language Interoperability between Julia and Python in Scientific #1 Computing Langu Othere Im Binardii Janzi 19, 2024 | | | |
| Type Dataset (42,583) Collision (342) | CERN Open Data Policy for the LHC Experiments CMS completes 2010-2011 proton-proton data rel CMS releases open data for Machine Learning | Number of authors | tanna Osaome, jim iviaski, jerty Ling (Apr 26, 2024) Contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>E</u> cite <u>c</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>E</u> cite <u>c</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>E</u> cite <u>c</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>E</u> cite <u>C</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>E</u> cite <u>C</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>E</u> cite <u>C</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>C</u> cite <u>C</u> dain <u>C</u> contribution to: ACAT2024 • e-Print: 2040.18170 [cs.PL] <u>D</u> pdf <u>D</u> cite <u>C</u> dain <u>C</u> contribution to: AcAt2040 contributio: AcAt2040 contribution to: AcAt2040 contri | | | |
| Derived (253) O Simulated (41,988) T Documentation (54) T About (4) In Activities (13) O | Observing the Higgs with over one petabyte of ne The Future of Particle Physics is "Open" Improving educational content with high-school te CMS releases new batch of research data from L | Single author 15 10 authors or less 74 Exclude RPP Exclude Review of Particle Physics 85 | e-Print: 2404.04138 [hep-ex] pdf cite daim Finetuning foundation models for joint analysis optimization in High Energy Physics Mathias Vigl (Tech. U., Munich (main)), Nicole Hartman (Tech. U., Munich (main)), Lukas Heinrich (Tech. U., Munich (main)) (Jan 24, 2024) Published in: Mach.Learn.Sci.Tech. 5 (2024) 2, 025075 • e-Print: 2401.13536 (hep-ex) | | | |
| Authors (3) Guide (27) Hala (2) | CMS releases first batch of high-level LHC open c | Document Type article 53 published ③ 40 conference paper 26 thesis 7 | Image: Port Port Port Port Port Port Port Port | | | |

3

≡





Open data quantity & time defined in the <u>CMS open data policy</u>

Public cloud and 2 – Kubernetes

Why and how?



Motivation: 1 - Why?





<u>MiniAOD</u>: richest Run-2 OD format (30-40 kB/evt) good for ~ all analyses requires CMSSW











Why cloud and Kubernetes?

| | | Do you have? | Want to get? | Why cloud / Kubernetes? |
|---|---------------------|--------------|--------------|---|
| 1 | Computing resources | × | \checkmark | Short-term, immediate resources Pay what you use Compatible with CMS OD environment |
| 2 | Tools | × | \checkmark | <u>Kubernetes</u> as a basic tool <u>Terraform</u> to deploy resources <u>Argo workflows</u> to manage jobs Open source and free |
| 3 | Skills | × | \checkmark | Applicable within and beyond research Attractive for early careers and young-minded Example setup provided - we did it for you! |
| 4 | Your custom CMS OD | × | \checkmark | Process what you need Download and store locally Analyze on your own resources |

How much time and money?

Resources from a Google Cloud Research credit 303424260



Processing custom-NanoAOD:

- input: MiniAOD data
- full NanoAOD processing + Particle-Flow (PF) candidates
 - same as in the already-provided <u>PF-Nano datasets</u> on the portal
- image: <u>CMS OD image</u> with PFNano processing code precompiled
- using Argo workflow to run the processing
 - steps: get dataset metadata \rightarrow make a joblist \rightarrow process x N_{jobs} \rightarrow (test plot)





Workflow structure



- Vocabulary
 - **GKE** = Google Kubernetes Engine
 - **Cluster** consists of nodes (machines with local boot disks)
 - Jobs run on pods (containerized applications) on the nodes
 - **Persistent storage** is a disk available for all nodes (and pods)
- Two types of GKE clusters: standard and auto-pilot:

Using: GKE Standard clusters

- allows defining all cluster components (type and number of nodes)
- auto-scales (nodes deleted automatically) if so configured
- cost goes with time and depends on the cluster setup
- can create nodes with the container image from a secondary boot disk

Not tested: GKE Auto-pilot clusters

- creates the cluster based on the resource request
- did not allow for a NFS disk server (container in the privileged mode)
- auto-scales (nodes deleted automatically)
- creating nodes with the image from the secondary boot disk did not work out of the box



Disk types and cost

- NFS disk
 - predefined size
 - the cost / disk size, not / usage
 - requires an NFS server on the cluster
- Google Cloud Storage bucket
 - size not predefined
 - the cost / actual usage
- Storage:
 - negligible (for a cluster lifetime 1-2d)
- Download ("egress"):
 - costly

 \rightarrow Our choice: GCS bucket for the ease of use

| europe-west4 (Netherlands) | NFS | GCS bucket | | |
|-------------------------------|--|---|--|--|
| Storage /month | 0.44\$/10GB (22\$/500GB) | 0.20\$/10GB (10\$/500GB) | | |
| | SSD: 1.87\$/10GB | | | |
| Download cost | 0.12\$/GB 1 (60\$/500GB) 1 (60\$/500GB) | | | |
| Download | Local download speed | Idem | | |
| time | Slight dependence on distance | ependence on Independent of stance distance | | |



 \rightarrow 1 job / vCPU

Cluster and job configuration

- Node type
 - number of CPUs and amount of memory ("standard"/"highcpu"/"highmem")
- Match with the job resource needs
 - use a test workflow running a single job / node to see the resource needs
 - for a quick, provider-independent check:

| kubectl | top | node | | | (check node resources in use) |
|---------|-----|------|----|------|-------------------------------|
| kubectl | top | pods | -n | argo | (check single pod usage) |

\$ kubect1 top pods -n argo | grep runpfnano

pfnano-process-g2m2k-runpfnano-template-1439319107 998m 1394Mi pfnano-process-g2m2k-runpfnano-template-2136346710 999m 1460Mi

- resource requests define how argo distributes the jobs to the nodes
 - goal: close to full occupancy
 - lack of memory kills
 - lack of CPU slows down

resources: requests: cpu: "800m" memory: "1.8Gi" ephemeral-storage: "5Gi"

15



🔥 👉 Add data download 👈 🦺

🔥 👉 ~ 40\$ / 350GB 👈 🛝

An example job of 33M events, <u>MuonEG</u> (1.1 TB) 90-node e2-standard-4 regional cluster (4 vCPUs, 16GB memory / node) - auto-scale 80 CHF - 9 hours





Auto-scaling at work: cost = N nodes x time



Optimal configuration: a large cluster for a short time

80\$ / 1TB Approximate price of PFNano-type processing / 1 TB of input data



40\$

Example download (PFNano-type content - 35% of the original size)



- Did the jobs fail and why?
 - Some XRootD timeouts (fixed on server-side), rare cluster networking timeouts.
- Can we gain in speed by uploading input files to the cluster?
 - No, not for this workflow. \rightarrow See <u>back-up</u>
- How to handle the big CMSSW container image?
 - Use a secondary boot disk. \rightarrow See <u>back-up</u>
- Is there an overhead for Kubernetes and Argo services?
 - Nothing significant \rightarrow See <u>back-up</u>
- What about spot / preembtible nodes (cheap but deletable)?
 - Cheaper but unreliable, definitely worth a follow-up \rightarrow See <u>back-up</u>



- Message logger every event increases run time 1-2 %.
- Mounted disk on kubernetes pods is shared to the persistent storage at the end of the step
 - Internal networking in the cluster can be surprisingly slow.
- Multithreading in kubernetes clusters is not obvious
 - CMSSW jobs can be configured to run in parallel threads within a job.
 - But: in a cluster, 4 single-thread jobs go faster than a 4-thread job in a 4 vCPU node.

FAQ and other observations







ADAPT FOR YOUR USE and OFF YOU GO!

Run a test job with *your* processing code. Evaluate *your* output file size and adapt the disk size accordingly. Adapt the cluster size and type or use our suggested values.

Creating your custom CMS Open Data

KNOW YOUR PHYSICS and CMS OD

Have your research idea! Learn about CMS OD:

- <u>Workshops</u> / <u>docs</u> / <u>support</u> Need more than NanoAOD?

- No? You could have skipped this talk.

- Yes? Adapt the example to your needs.

GET STARTED WITH OUR INSTRUCTIONS and EXAMPLES

Create a Google Cloud project. / Install: terraform, gcloud (or use Google Cloud shell), argo CLI, kubectl. Deploy resources using example Terraform scripts. Run our example job with Argo workflows.



Conclusion and outlook

CMS Open Data can be used without complications, NanoAOD format is a streamlined and condensed storage format that can be analyzed directly by open data users.

For analyses requiring detailed event content, we have demonstrated that using public cloud resources for custom NanoAOD processing is feasible both for time and cost.

We have shown how to optimize disposable cloud resources for a typical processing task.

Containerized CMS Open Data workflows can easily be run in a modern kubernetes environment.



Questions?

And thanks to <u>SlidesCarnival</u> for this free presentation template







Knowledge:

-instructions

- -actionable examples
- -understanding of experimental data



🕚 – Input data streaming vs upload

- Data is streamed with xrootd protocol from eospublic at CERN
 - No significant difference between locations close to of far from CERN
 - Processing time dominates over data access time.
- Any faster if input data uploaded?
 - Uploaded files to the container local disk before processing
 - Fairly fast upload with xrdcp (upload faster close to CERN)
 - But:
 - No speed-up for the processing time (even slightly slower from local files)
 - Explored differences with file:<filepath> (normal local file access) and root://<xrootdserver>/<filepath> (local xrootd server in the container): no significant difference
 - xrootd server version on eospublic more recent that in the container
- No significant gain.

Processing time: streaming vs local

Compare:

same 4 jobs several times (Run 1-4)

- streaming
- local (read with file;)
- local (read with root:) No gain observed.





- CMSSW container image is big
 - Initial pull can take 30 mins.
 - Once pulled on the node, it is available to all pods.
 - Run a start job to pull the image to each node.
- Is there a better solution?
 - Uploading image to Google artifact registry and accessing from there is not significantly faster.
 - Use a new GCP feature: a <u>secondary boot disk</u> with container images preloaded:
 - Build a disk (<u>tools exist</u>), enable image streaming and define the disk in the node pool configuration so that it uses this secondary disk.
 - Immediate start of the jobs #!



– Kubernetes / argo overhead

| NAMESPACE | NAME | CPU(cores) | MEMORY(bytes |
|------------------|--|------------|--------------|
| argo | argo-server-5f7b589d6f-6jqcm | 1m | 18Mi |
| argo | workflow-controller-864c88655d-gwjcr | 4m | 26Mi |
| gke-managed-cim | kube-state-metrics-0 | 1m | 72Mi |
| gmp-system [] | collector-2j5dw | 5m | 139Mi |
| gmp-system | gmp-operator-57874fcf58-8h8w4 | 2m | 62Mi |
| kube-system | event-exporter-gke-78fb679b7b-fdcr8 | 1m | 83Mi |
| kube-system | fluentbit-gke-28sjh | 12m | 109Mi |
| [] | | | |
| kube-system | gke-metrics-agent-x7rft | 7m | 113Mi |
| kube-system | konnectivity-agent-5f967456fc-2mc7x | 1m | 40Mi |
| [] | | | |
| kube-system | konnectivity-agent-autoscaler-897d4f648-t97gb | 1m | 39Mi |
| kube-system | kube-dns-5fc99b87cb-dgzqj | Зт | 166Mi |
| kube-system | kube-dns-5fc99b87cb-nb25q | 2m | 138Mi |
| kube-system | kube-dns-autoscaler-6f896b6968-crn2z | 1m | 39Mi |
| kube-system | kube-proxy-gke-cluster-4-cluster-4-9248283b-00xm | 1m | 15Mi |
| [] | | | |
| kube-system | 17-default-backend-6697bb6dfd-swb9h | 1m | 10Mi |
| kube-system | metrics-server-v1.30.3-7c8f6576cd-bdc85 | 4m | 78Mi |
| kube-system | pdcsi-node-2fv5q | 5m | 47Mi |

Compare: same jobs several times VM: no argo, no k8s vs GKE cluster No significant overhead.





Preemtible / spot nodes

- Considerably cheaper
 - → ¼ ½ price
- Nodes can be deleted any time
- A trial with a 90-node e2-standard-4 cluster:
 - 13 / 90 nodes terminated
 - \rightarrow 52 / 353 jobs failed
- Requires rerunning of the failed jobs.
 - The price advantage is worth some scripting for automated reruns.

