

Google R&D: PanDA and Dask setups

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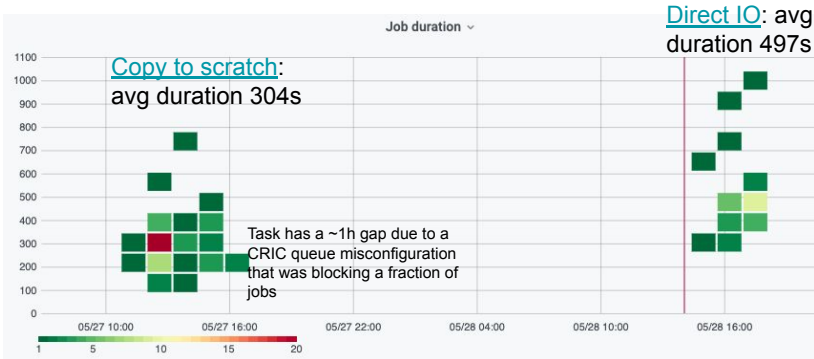


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PanDA + Rucio setup

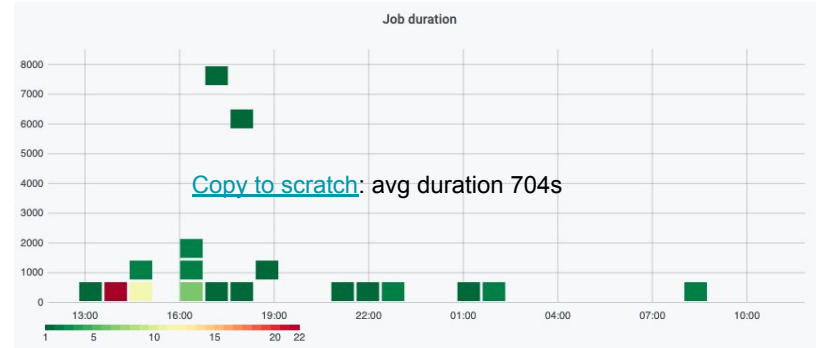
PanDA GKE analysis queue: GOOGLE100

- Cluster details
 - 0-10 **autoscaled, preemptible** nodes: n2-standard-8 (=8 cores, 32GB RAM)
 - Scaling down under discussion with Usman and Jason Nichols (GKE specialist)
 - Local **SSD** at each node
- Queue status “Brokeroff”: you need to specifically set the queue when submitting to PanDA
- Data needs to be pre-placed to RSE GOOGLE_EU
 - Jobs will stay in “**assigned**” if data not present
 - Requires **special permission/quota** in Rucio to interact with this storage element
- 11TB input task completed successfully (thanks to Nikolai Hartmann)
 - ~12GB input per job



GOOGLE100

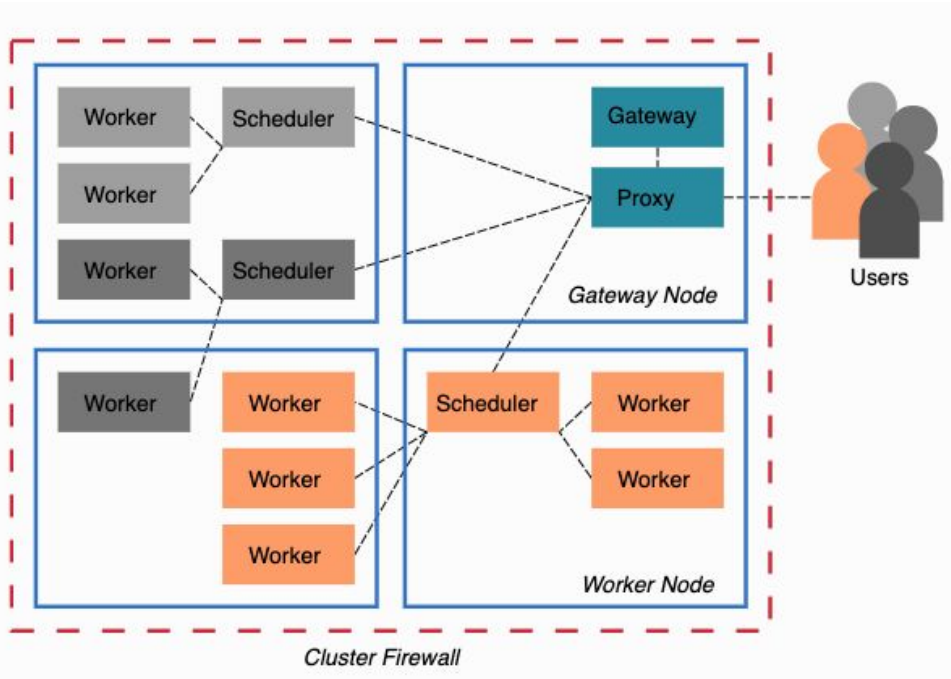
Colours are #occurrences, not failure rate!!!



Grid

Dask+JupyterHub

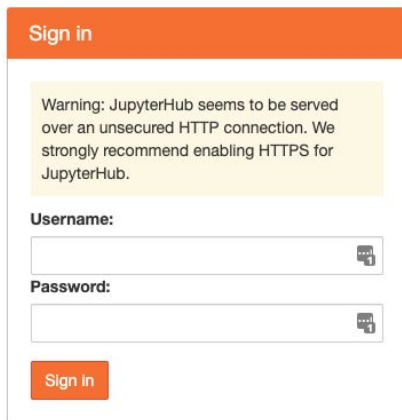
Dask + Jupyter setup



- Sets up common Dask cluster and JupyterHub for all users
- Users have access to JupyterHub and Dask, but not to GCP/GKE
- Disadvantage: less flexibility for individual customization. Needs central maintenance of a set of images that work for everybody
- Current installation on modest cluster: 3 e2-standard-8 nodes with 100GB disk

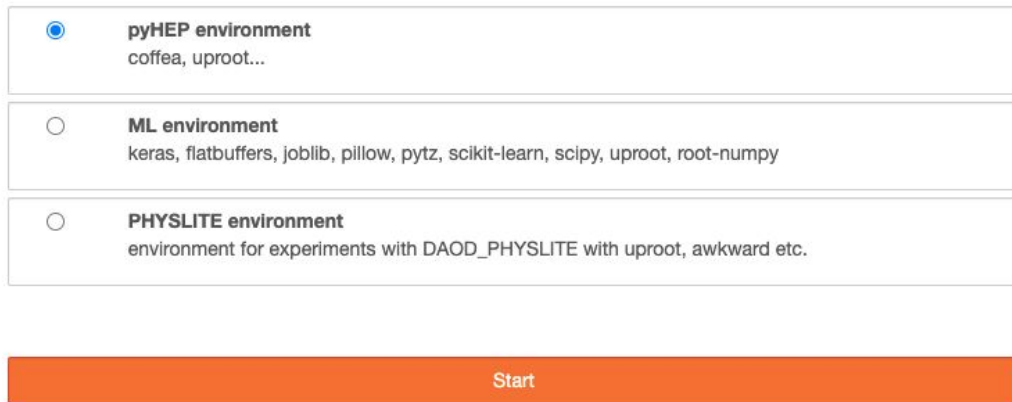
JupyterHub

<http://jupyter.gcp4hep.org/>



- Local accounts. I need to add new users
- Integration with other identity providers possible. Could be done only if there is real usage

Server Options



| |
|--|
| <input checked="" type="radio"/> pyHEP environment coffea, uproot... |
| <input type="radio"/> ML environment keras, flatbuffers, joblib, pillow, pytz, scikit-learn, scipy, uproot, root-numpy |
| <input type="radio"/> PHYSLITE environment environment for experiments with DAOD_PHYSLITE with uproot, awkward etc. |

Start

- Available images
 - pyHEP environment: dependencies suggested in [this tutorial](#)
 - ML environment: dependencies requested by Fang-Ying
 - PHYSLITE environment: image provided by Nikolai
- Images hosted in GCP Container Registry
- Image management is **very time consuming**

Jupyter notebook specs

- Each user notebook runs on an **independent** pod with image selected at startup
 - “**Burstable**” QoS with 1GB RAM base request
 - How much you can burst depends on overall cluster usage and occupancy of the node
- User home directory: 10GB
 - Independent persistent disk
 - Value is configurable
 - Disk can also be manually extended
 - Anything outside the home directory gets **cleaned up** once notebook stops
 - A potential conda user environment installed on home directory would survive

```
(notebook) jovyan@jupyter-fbarreir:~$ df -h
Filesystem      Size  Used Avail Use% Mounted on
overlay          95G   16G   79G  17% /
tmpfs            64M    0    64M   0% /dev
tmpfs            16G    0    16G   0% /sys/fs/cgroup
/dev/sdc         20G   46M   20G   1% /home/jovyan
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/atlas.cern.ch
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/atlas-condb.cern.ch
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/atlas-nightlies.cern.ch
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/grid.cern.ch
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/sft.cern.ch
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/sft-nightlies.cern.ch
cvmfs2           5.9G   2.7G   3.2G  46% /cvmfs/unpacked.cern.ch
/dev/sda1       95G   16G   79G  17% /etc/hosts
shm             64M    0    64M   0% /dev/shm
tmpfs           16G    0    16G   0% /proc/acpi
tmpfs           16G    0    16G   0% /proc/scsi
tmpfs           16G    0    16G   0% /sys/firmware
```


Spinning up a Dask cluster: LOCAL

LOCAL: your cluster lives in your jupyter pod

The screenshot shows a JupyterLab interface with a Dask dashboard on the left and a code editor on the right. The dashboard includes a sidebar with navigation icons and a main area with various performance metrics like PROGRESS, CPU, MEMORY BY KEY, BANDWIDTH TYPES, NPROCESSING, AGGREGATE TIME PER ACTION, COMPUTE TIME PER KEY, NBYTES, TASK STREAM, GPU UTILIZATION, WORKERS, PROFILE SERVER, PROFILE, GRAPH, BANDWIDTH WORKERS, GPU MEMORY, and CLUSTER MAP. Below these metrics is a 'CLUSTERS' section with a '+ NEW' button and a card for 'LocalCluster 1' showing scheduler address, dashboard URL, number of cores (8), memory (33.68 GB), and number of workers (4). The code editor shows Python code to create a Dask Client. A blue box with the text 'Run your dask code' is positioned below the code, with an arrow pointing to the 'drag' label.

```
[1]: from dask.distributed import Client

client = Client("tcp://127.0.0.1:41677")
client
```

[1]: Client Cluster
Scheduler: tcp://127.0.0.1:41677 Workers: 4
Dashboard: http://127.0.0.1:8787/status Cores: 8
Memory: 33.68 GB

[]: Run your dask code

drag

Not sure how these values were chosen, probably by retrieving node size

Spinning up a Dask cluster: distributed

Distributed: each worker gets an independent pod, so you can scale to multiple nodes

File Edit View Run Kernel Tabs Settings Help

http://jupyter.gcp4hep.org/services/d

Untitled3.ipynb Python [conda env:notebook]

```
[1]: from dask_gateway import GatewayCluster
cluster = GatewayCluster(worker_cores=1, worker_memory=2, image="eu.gcr.io/gke-dev-311213/dask-gateway-coffea:20210518")
cluster.scale(1)
client = cluster.get_client()
```

1-4 cores 1-8 GB Limits defined by admin

/srv/conda/envs/notebook/lib/python3.8/site-packages/distributed/client.py:1129: VersionMismatchWarning: Mismatched versions found

| Package | client | scheduler | workers |
|---------|--------|-----------|---------|
| lz4 | 3.1.1 | 3.1.3 | None |

Warning... more on this later

```
warnings.warn(version_module.VersionMismatchWarning(msg[0] ["warning"]))
```

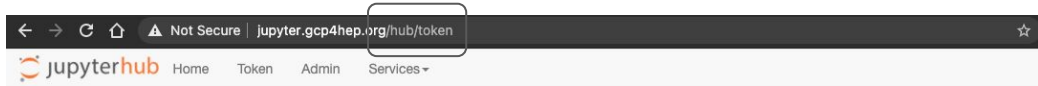
```
[2]: client
```

```
[2]: Client Scheduler: gateway://traefik-dhub-dask-gateway.default:80/default.6295b69fd9d5454fb2cab7cb80829d1b
Dashboard: /services/dask-gateway/clusters/default.6295b69fd9d5454fb2cab7cb80829d1b/status
Including the host!
```

Cluster Workers: 1
Cores: 1
Memory: 2.15 GB

CLUSTERS ↻ + NEW

Spinning a dask cluster from an arbitrary python shell



```
export JUPYTERHUB_API_TOKEN=<YOUR TOKEN>
[user@machine gke-dask]# python3
Python 3.6.8 (default, Nov 16 2020, 16:55:22)
[GCC 4.8.5 20150623 (Red Hat 4.8.5-44)] on linux
Type "help", "copyright", "credits" or "license" for more information.
>>> from dask_gateway import Gateway
>>> gateway = Gateway("http://dask.gcp4hep.org/services/dask-gateway", auth='jupyterhub')
>>> cluster = gateway.new_cluster(image='xxx/yyy:zzz')
>>> client = cluster.get_client()
>>> # RUN YOUR COMPUTATION
```

Software compatibility

- The client (Jupyter or your python shell) and the Dask workers need to have compatible SW (dask, tensorflow...) installed
- I tried to generate compatible images for both Jupyter images building on the default pangeo/daskgateway images

| Jupyter image | Name | Worker image | Description |
|--|----------------------|---|---|
| pangeo/base-notebook:2020.11.06 | Not selectable | daskgateway/dask-gateway:0.9.0 | Basic Dask installation. We have overwritten this option with our images |
| eu.gcr.io/gke-dev-311213/jupyter-coffea:20210518 | pyHEP environment | eu.gcr.io/gke-dev-311213/dask-gateway-coffea:20210518 | This image is based on the dependencies used in this PyHEP tutorial . It includes coffea, python-graphviz, mimesis on top of the default pangeo image. |
| eu.gcr.io/gke-dev-311213/jupyter-ml:20210518 | ML environment | eu.gcr.io/gke-dev-311213/dask-gateway-ml:20210518 | Image based on Fang-Ying's request which includes root, keras, flatbuffers, joblib, pillow, pytz, scikit-learn, scipy, energyflow, root-numpy, sklearn, awkward, uproot on top of the default pangeo image. |
| eu.gcr.io/gke-dev-311213/jupyter-physlite:20210526 | PHYSLITE environment | eu.gcr.io/gke-dev-311213/dask-gateway-physlite:20210526 | Image by Nikolai including PHYSLITE SW (numpy h5py numba uproot awkward pyarrow coffea aiohttp) . |

<https://github.com/gcp4hep/analysis-cluster/wiki/Daskhub-images>

Conclusions

- Infrastructure is ready and required features implemented
 - Desirable Dask features (https, oAuth) can be implemented depending on evolution of activity
 - Data management setup in Dask to be explored further
- First tests in progress
 - PanDA:
 - Nikolai: 11TB input task done, next is 100TB task
 - Lukas also has a potential idea
 - Dask:
 - Nikolai: Small validation done
 - Fang-Ying: evaluating whether the notebooks fit her needs
 - Working on rucio client dependency issues
- Paul has setup a separate single-user cluster
 - This model can be more appropriate for a potential PanDA integration
 - Still requires more experience and agreeing on architecture