

Machine Learning at CSC

June 3, 2020 Mats Sjöberg – mats.sjoberg@csc.fi





Overview

What CSC service to use?

Puhti supercomputer

Data storage

GPU utilization

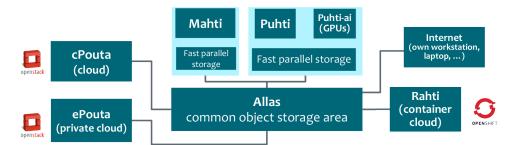
Multi-GPU and multi-node jobs

Singularity containers



- CSC's supercomputer Puhti
- Virtual server on Pouta

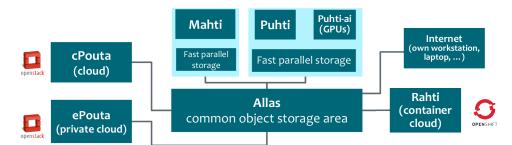
Container cloud Rahti





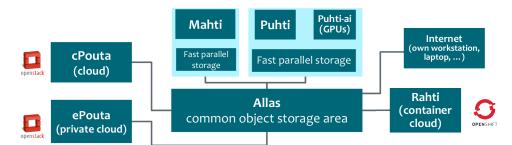
- CSC's supercomputer Puhti
 - Cluster with GPU-accelerated nodes
 - Multi-user environment
- Virtual server on Pouta

Container cloud Rahti



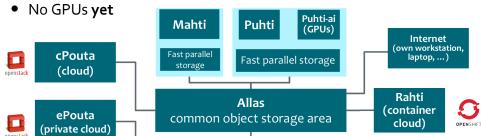


- CSC's supercomputer Puhti
 - Cluster with GPU-accelerated nodes
 - Multi-user environment
- Virtual server on Pouta
 - Your "own" server
 - Less powerful than Puhti
- Container cloud Rahti





- CSC's supercomputer Puhti
 - Cluster with GPU-accelerated nodes
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 - Your "own" server
 - Less powerful than Puhti
- Container cloud Rahti
 - Easy to run containers



Puhti supercomputer



Puhti supercomputer

- Puhti-AI, cluster with 80 nodes with 4 GPUs each → 320 GPUs in total
- Latest generation Nvidia V100 GPUs (Volta) with 32 GB of memory
- Fast network: 2 × 100 Gbps links to each node
- Each node has a fast 3.2 TB local NVME disk





Getting access to Puhti

```
https://docs.csc.fi/computing/overview/
```

To use Puhti you need to:

- Have a CSC account
- Be member of a CSC project, either by
 - · creating a new project, or
 - joining an existing project (ask the PI to add you!)
- Finally, the project needs to have Puhti access
- → MyCSC portal: https://my.csc.fi/



Accessing Puhti

- Using an ssh client such as OpenSSH or PuTTY
- Basic Linux skills are required!
- More info: https://docs.csc.fi/computing/connecting/

```
$ ssh <csc_username>@puhti.csc.fi
```

\$ ssh <csc_username>@puhti-login2.csc.fi



Supported frameworks

We currently support:

- Python Data collection of Python libraries for data analytics and machine learning
- TensorFlow deep learning library for Python
- PyTorch machine learning framework for Python
- MXNet deep learning library for Python
- RAPIDS suite of libraries for data analytics and machine learning on GPUs

https://docs.csc.fi/apps/#data-analytics-and-machine-learning



Example: TensorFlow

- First check the application page for instructions: https://docs.csc.fi/apps/tensorflow/
- Load the default version:
 module load tensorflow
- or specific version:module load tensorflow/2.0.0
- Note: some modules are Singularity-based!



What if some package is missing?

If you are using our module, but a trivial package is missing ...

- install it yourself, e.g.,pip install --user <packagename>
- ...or if it might be generally useful, send an email to servicedesk@csc.fi – we can install it for you!



What if some package is missing?

If you need a specific setup, and our modules are not right for you ...

use a virtualenv:

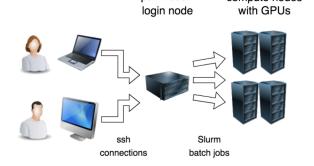
```
$ python3 -m venv myenv
$ source myenv/bin/activate
$ pip install ...
```

- Use conda: https://docs.csc.fi/support/tutorials/conda/
- use singularity containers:
 https://docs.csc.fi/computing/containers/run-existing/
- or if generally useful, send an email to servicedesk@csc.fi

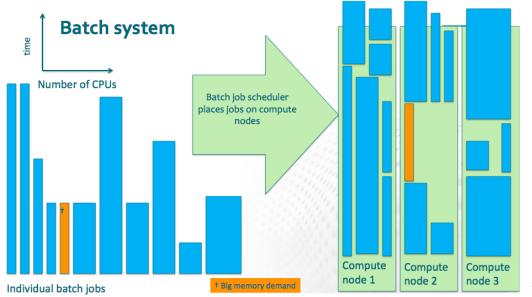


Don't run heavy computing jobs in the login nodes!

- Puhti uses the *Slurm* batch job system
- Jobs do not run instantly but are put in a queue
- Resources (runtime, memory, number of cores) need to be specified
 puhti.csc.fi compute nodes









Create a job script, for example run.sh:

```
#!/bin/bash
#SBATCH --account=cet>
#SBATCH --partition=qpu
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=10
#SBATCH --mem=64G
#SBATCH --time=1:00:00
#SBATCH -- gres = qpu:v100:1
module load tensorflow/2.0.0
srun python3 myprog.py <options>
```

https://docs.csc.fi/computing/running/creating-job-scripts/



Submit the job:

sbatch run.sh

Check the queue:

squeue -l -u \$USER

Cancel a job:

scancel <jobid>

https://docs.csc.fi/computing/running/submitting-jobs/

Data storage



Data storage on Puhti

- Disk space and number of files are limited on Puhti!
 - \rightarrow We want to ensure that the shared (Lustre) filesystem works efficiently for everyone!
- Useful command: csc-workspaces

	Owner	Path	Capacity	Number of files	Cleaning
home	Personal	/users/ <user-name></user-name>	10 GiB	100 000 files	No
projappl	Project	/projappl/ <project></project>	50 GiB	100 000 files	No
scratch	Project	/scratch/ <project></project>	1 TiB	1 000 000 files	Yes - 90 days

Data quotas can be increased via MyCSC!

https://docs.csc.fi/computing/disk/



Using Allas

- store big datasets in Allas, CSC's object storage
- download them to project scratch prior to computation
- you can also upload trained models (or keep in projappl)

```
$ module load allas
$ allas-conf
$ cd /scratch/<your-project>
$ swift download <bucket-name> your-dataset.tar
```



Large number of files

- Many datasets contain a large number of small files
- Shared filesystem (Lustre) performs poorly in this scenario
 → noticable slowdowns for all Puhti users!

Consider alternatives:

- packaging your dataset into larger files
- use NVME fast local storage on GPU nodes



Using more efficient data formats

Instead of many small files, use one or a few bigger files.

Examples:

- TensorFlow's TFRecord format
- HDF₅
- LMDB
- ZIP, for example via Python's zipfile library



Fast local NVME drive

- All GPU nodes have a local NVME drive
- Just add nvme: <number-of-GB> to sbatch --gres flag

```
#!/bin/bash
#SBATCH --account=cet>
#SBATCH --partition=qpu
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=10
#SBATCH --mem=64G
#SBATCH --time=1:00:00
#SBATCH --qres=qpu:v100:1,nvme:100
tar xf /scratch/<your-project>/your-dataset.tar -C $LOCAL SCRATCH
srun python3 myprog.py --data_dir=$LOCAL_SCRATCH <options>
```

GPU utilization

Multi-GPU and multi-node jobs

Singularity containers