



PPCES: Machine and Deep Learning

Hands-on Exercises



DAS KOMPETENZNETZWERK FÜR HOCHLEISTUNGSRECHNEN.

- **Hands-on material and slide decks can be found on the PPCES website**
 - <https://blog.rwth-aachen.de/itc-events/en/event/ppces-2024/>
- **Download and extract material on desired machine**
 - Separate folders for scikit-learn and PyTorch
 - scikit-learn available as regular Python (.py) and Jupyter Notebooks (.ipynb)

```
# change to the directory where you want to save your material
cd <working directory>

# download material and examples
wget https://hpc.rwth-aachen.de/ppces/ppces2024-ML-DL-labs.tar.gz

# unpack the tar file
tar -xzvf ppces2024-ML-DL-labs.tar.gz
```



Classic / Batch Mode

Login via SSH
SLURM Workload
Manager

Module System
Apptainer (Singularity)



Remote Desktop

Login via FastX
Graphical Interface

Rest is the same as in
Batch Mode



Interactive Mode

JupyterHub/Lab
Web Interface

Predefined Profiles

Accessing the CLAIR HPC Cluster

Accounts

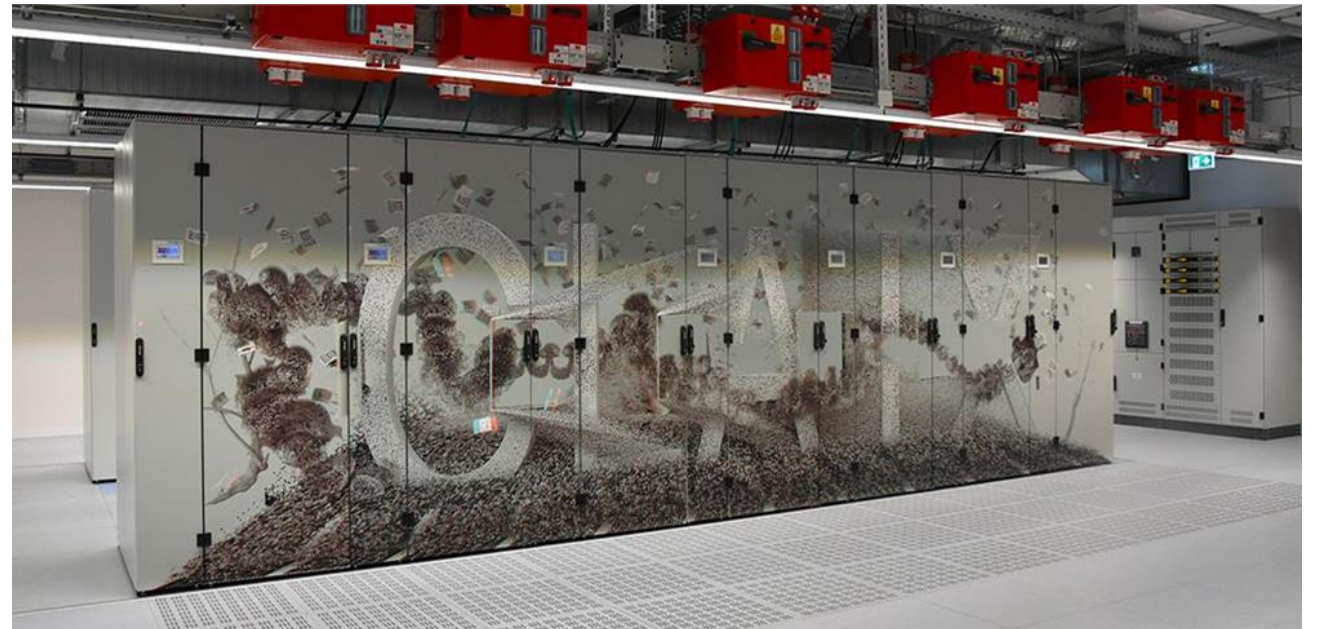
Members of RWTH (or affiliated persons)

- [HPC account required](#):
<https://regapp.itc.rwth-aachen.de/>
- [Two-factor-authentication](#) (2FA) required

External participants

- Temporary account will be provided

Dedicated hardware for the workshop



	OpenMP days	MPI days	ML day
Advanced reservation	PPCES24	PPCES24MPI	PPCES24GPU
Compute time project	lect0113	supp0006	lect0114

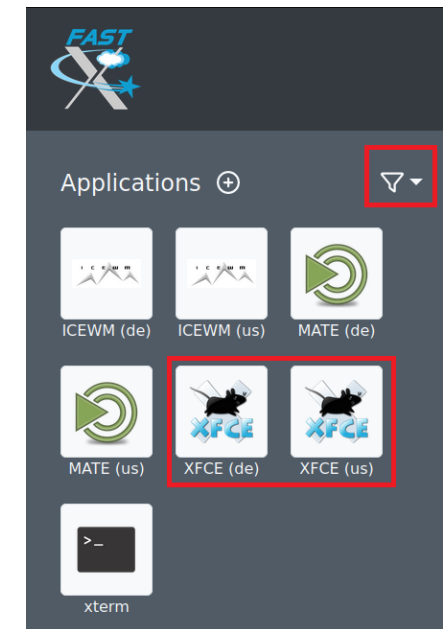
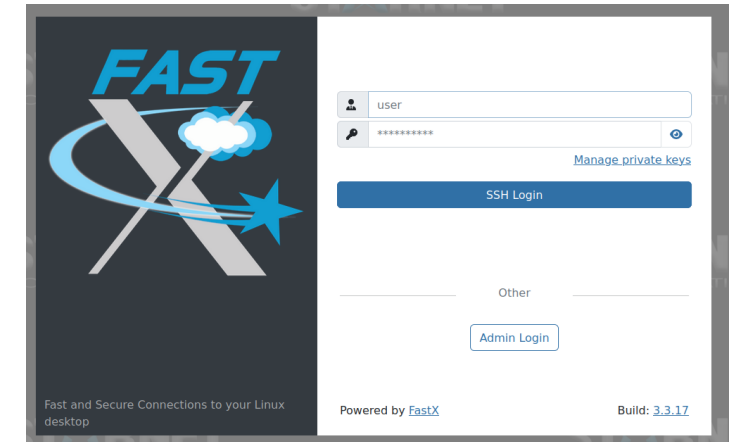
Example in batch script file:

```
#SBATCH -reservation=PPCES24
#SBATCH -account=lect0113
```

Remote Desktop Access

FastX3 (desktop client or browser)

- Browser client: <https://login23-x-2.hpc.itc.rwth-aachen.de:3300>
- Desktop client: login nodes (see next slide)
- Select the correct application
 - Recommended: XFCE
 - May need to select “all applications” in Filter at the top right corner of the Application box
- Start the session
 - Click on the session (may have to click twice)
- Open up a terminal
 - Bottom of the screen
 - Click the “terminal” icon



Cluster Access

Login nodes (full list [here](#))

- login23-x-2.hpc.itc.rwth-aachen.de (CLAIX-2023, X-Server)
- login23-2.hpc.itc.rwth-Aachen.de (CLALIX-2023)
- login23-3.hpc.itc.rwth-Aachen.de (CLAIX-2023)
- login18-2.hpc.itc.rwth-aachen.de (CLAIX-2018)
- login18-3.hpc.itc.rwth-aachen.de (CLAIX-2018)

Access via SSH

Access from within the RWTH network / eduroam via ssh:

- ssh -Y ab123456@login23-x-2.hpc.itc.rwth-aachen.de
- GitBash for Windows: <https://git-scm.com/download/win>

Cluster Access: External Participants Only!

Account

- Username: `hpclab[01-50]`
 - Distribution of accounts by speaker/helpers
- Password: see paper sheet

Access

- Login node: `login18-beta.itc.hpc.rwth-aachen.de`
 - Use desktop client or
 - Web interface: <https://login18-beta.hpc.itc.rwth-aachen.de:3300>
- Jump from there to a CLAIX-2023 frontend node (see node list on previous slide)

```
ssh -Y hpclab[01-50]@login23-2.hpc.itc.rwth-aachen.de
```

Note: Do not store any private/ sensitive data on the cluster

- **Large Jobs / high demand** → JupyterHub HPC
 - Special JupyterHub instance
 - Starts containers on HPC cluster nodes
 - <https://jupyterhub.hpc.itc.rwth-aachen.de:9651>
- **Alternatives**
 - Another RWTH JupyterHub
 - Cheaper standard servers under the hood
 - <https://jupyter.rwth-aachen.de/hub/login>
 - Profiles / kernles might differ
 - Install JupyterLab in your local virtual environment
 - Execution happens on your local machine
 - Visual Studio Code integration available

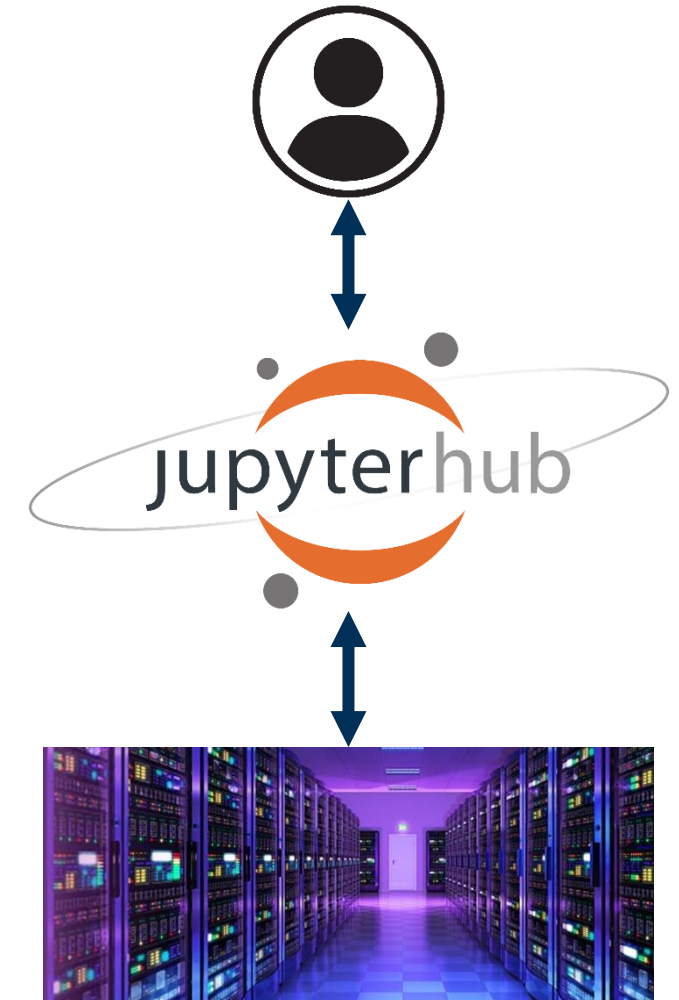
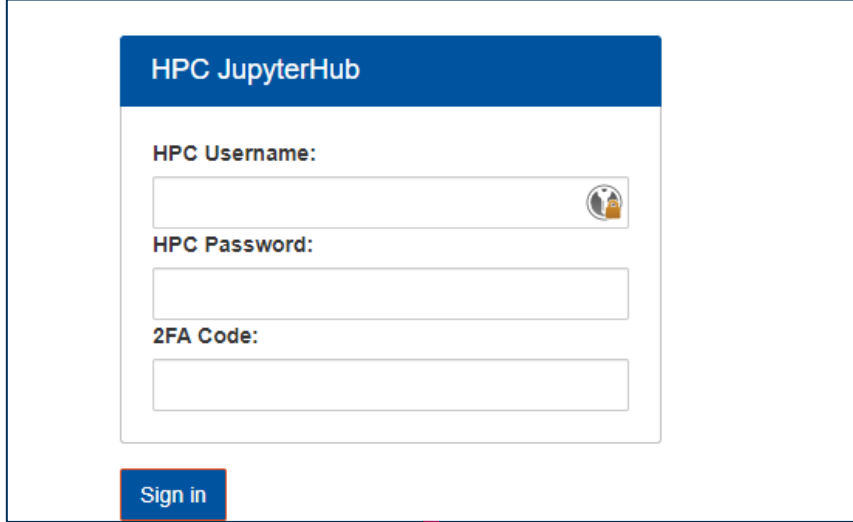



Image: scientific-computing.com

- **Step 1:** Login to JupyterHub HPC
 - Similar to SSH or FastX login
 - You need your HPC 2FA again here

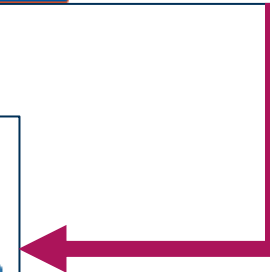


The screenshot shows a login form titled "HPC JupyterHub". It contains three input fields: "HPC Username:" with a lock icon on the right, "HPC Password:", and "2FA Code:". Below the fields is a blue "Sign in" button.

Main JupyterLab Server
You can start your main JupyterLab server here:



A blue button labeled "Start Main Server" is positioned at the bottom right of the box.



– Step 2: Configure and start your JupyterLab server

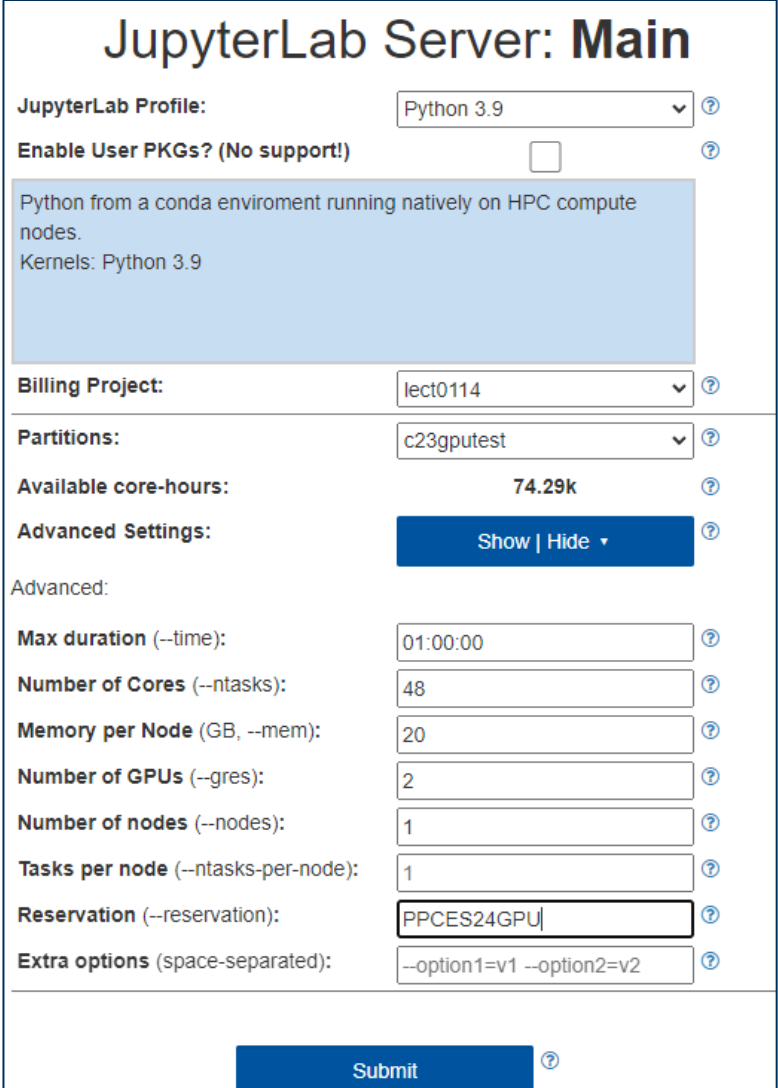
– Request resources on the HPC cluster

– General Settings

- JupyterLab Profile Python 3.9
- Billing Project lect0114
- Partition c23gputest

– Advanced Settings

- Number of cores max. 48
- Number of GPUs max. 2 (not used in scikit-learn)
- Max duration max. 1h
- Reservation PPCES24GPU



JupyterLab Server: Main

JupyterLab Profile: Python 3.9 ⓘ

Enable User PKGs? (No support!) ⓘ

Python from a conda environment running natively on HPC compute nodes.
Kernels: Python 3.9

Billing Project: lect0114 ⓘ

Partitions: c23gputest ⓘ

Available core-hours: 74.29k ⓘ

Advanced Settings: [Show | Hide ▾](#) ⓘ

Advanced:

Max duration (--time): 01:00:00 ⓘ

Number of Cores (--ntasks): 48 ⓘ

Memory per Node (GB, --mem): 20 ⓘ

Number of GPUs (--gres): 2 ⓘ

Number of nodes (--nodes): 1 ⓘ

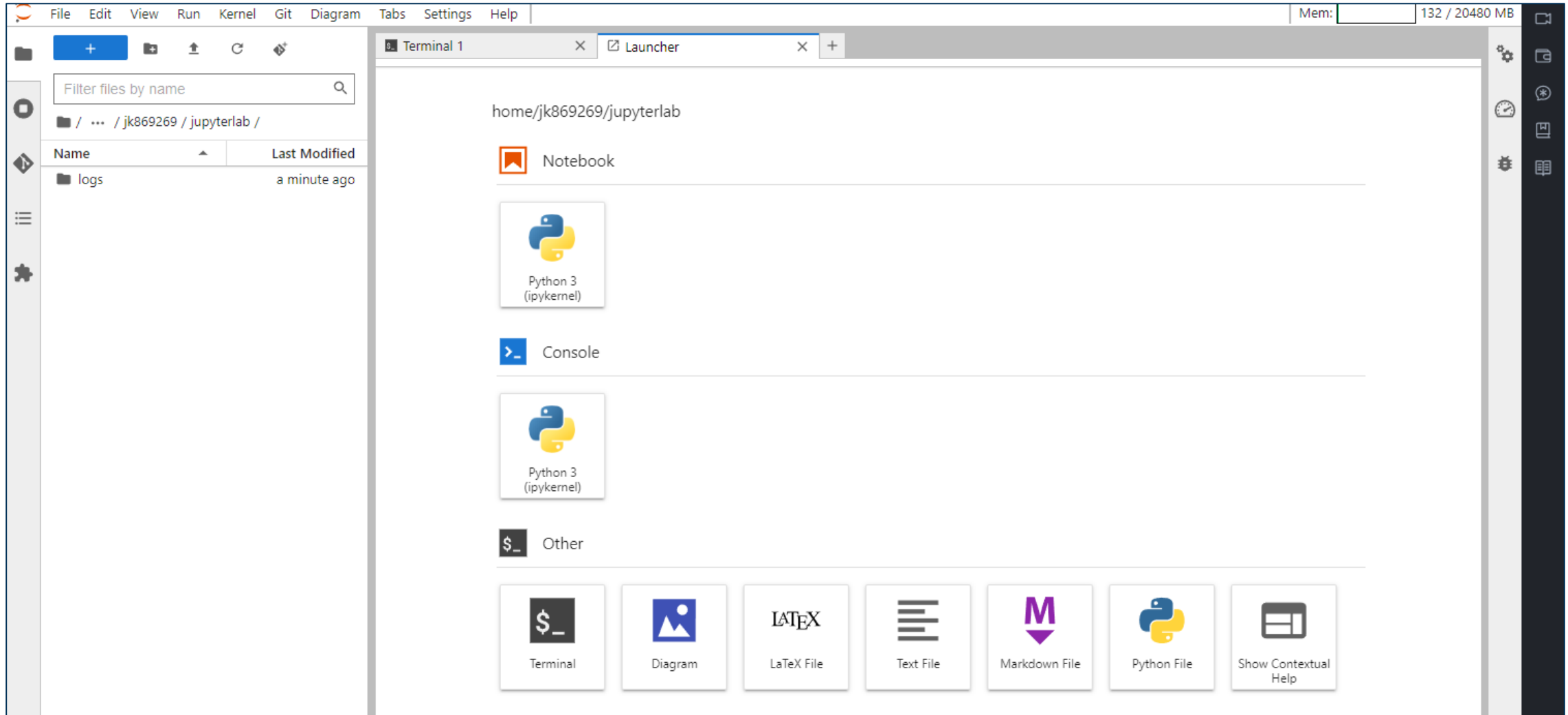
Tasks per node (--ntasks-per-node): 1 ⓘ

Reservation (--reservation): PPCES24GPU ⓘ

Extra options (space-separated): --option1=v1 --option2=v2 ⓘ

[Submit](#) ⓘ

Hands-on: scikit-learn – Getting Started with JupyterHub



The screenshot displays the JupyterLab web interface. On the left is a file browser with a search bar and a table of files. The main area is a launcher with several options: Notebook, Console, and Other. At the bottom, there is a row of icons for Terminal, Diagram, LaTeX File, Text File, Markdown File, Python File, and Show Contextual Help. The top menu bar includes File, Edit, View, Run, Kernel, Git, Diagram, Tabs, Settings, and Help. The top right corner shows memory usage: Mem: 132 / 20480 MB.

File Edit View Run Kernel Git Diagram Tabs Settings Help Mem: 132 / 20480 MB

Filter files by name

/ ... / jk869269 / jupyterlab /

Name	Last Modified
logs	a minute ago

home/jk869269/jupyterlab

Notebook

Python 3 (ipykernel)

Console

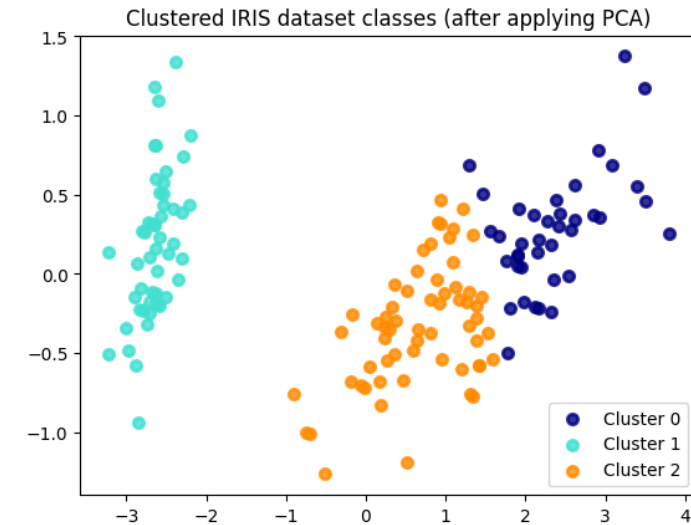
Python 3 (ipykernel)

Other

Terminal Diagram LaTeX File Text File Markdown File Python File Show Contextual Help

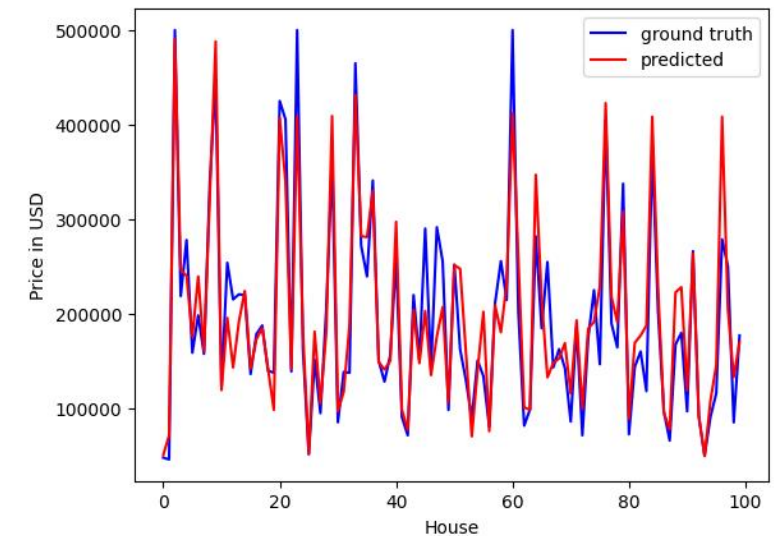
– Exercise 1: Clustering with Iris Dataset

- Load dataset
- Train clustering model (here: Kmeans clustering)
- Apply dimensionality reduction (PCA)
- Visualize and compare against true labels



– Exercise 2: Regression with California Housing Dataset

- Load dataset
- Apply preprocessing techniques (Standardization)
- Train regression model (here: RandomForest)
- Model performance evaluation and visualization



– Exercise: Train ResNet model with CIFAR-10 dataset

– Model: ResNet

- Popular model for image classification
- Winner of ILSVRC 2015 (ImageNet Large Scale Visual Recognition Challenge)
- Tackles vanishing gradient problem

– Dataset: CIFAR-10

- Several images of 10 different classes
- Airplane, automobile, bird, cat, dog, frog, ...

– Single GPU (Task 1), Distributed (Task 2)

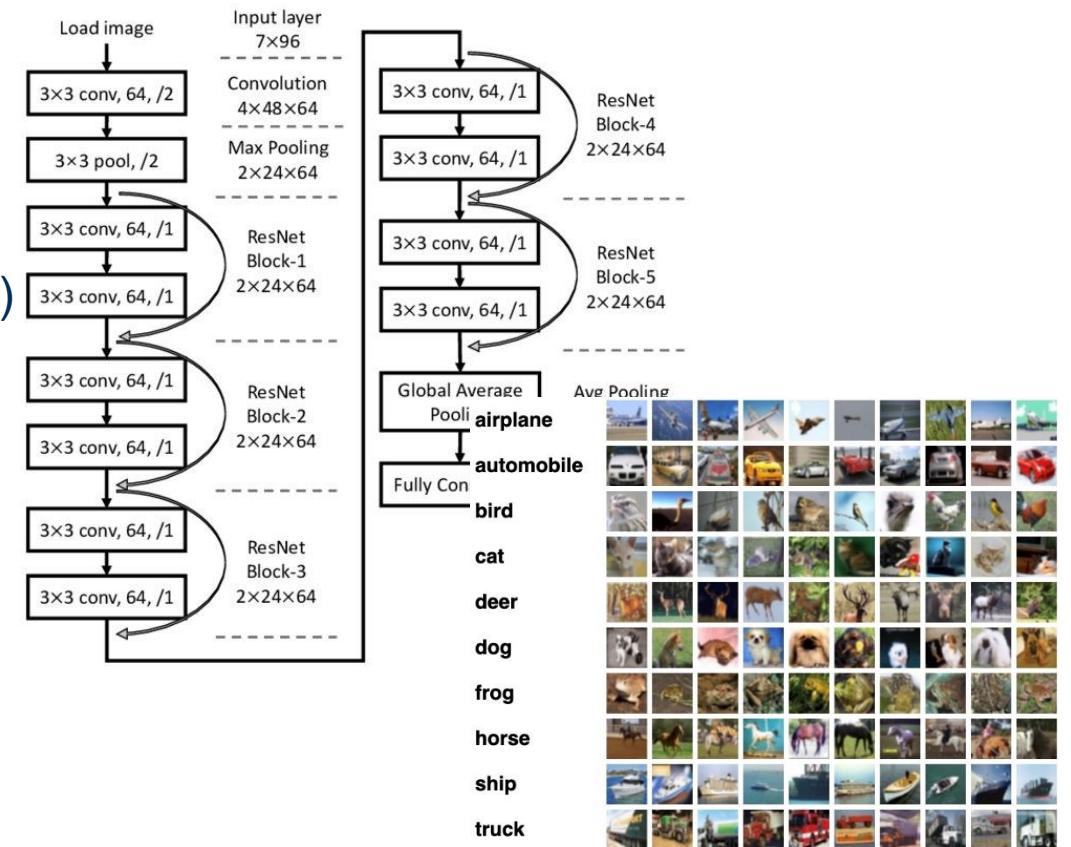


Diagram: Choi, Hyungeun & Ryu, Seunghyoung & Kim, Hongseok. (2018). Short-Term Load Forecasting based on ResNet and LSTM

- **Note:** We will use a container for this exercise!

	File	Description
Provided by us	<code>set_vars.sh</code>	Shell script that sets environment variables for DDP and container
	<code>submit_job.sh</code>	Submits a batch job to SLURM, which loads the required container module and executes the Python code
Your job	<code>train_model.py</code>	Python code that is responsible for training and testing the model

- **About: `submit_job.sh`**

- Option to run with 1 GPU or 2 GPUs (distributed)
- Option to enable explicit monitoring (`nvidia-smi`)

```
# execute with a single GPU
sbatch submit_job.sh

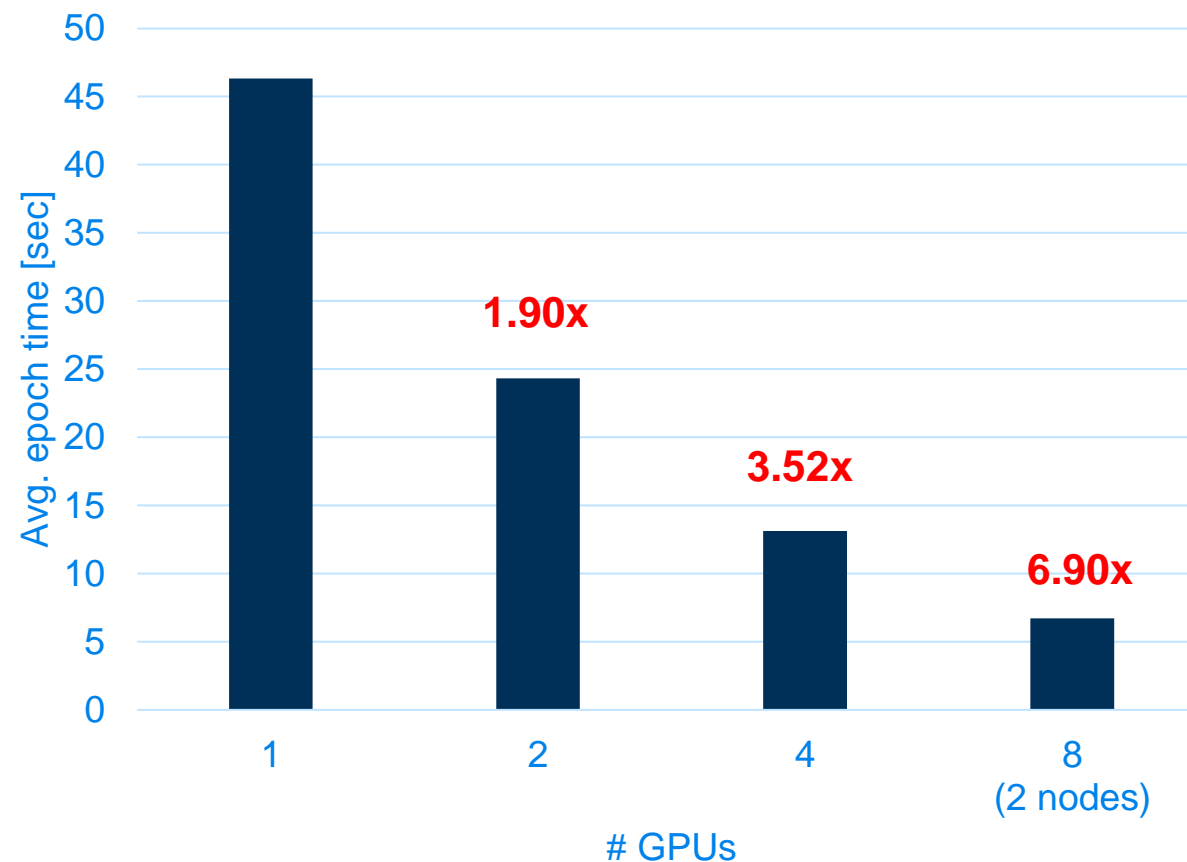
# execute with nvidia-smi monitoring
sbatch --export=ENABLE_MONITORING=1 submit_job.sh

# execute with 2 GPUs (distributed)
sbatch --export=RUN_DISTRIBUTED=1 submit_job.sh

# check for status
squeue --me -a
```

– Output and scaling behavior

```
...  
Epoch 2/2 Step 0 / 391  
Epoch 2/2 Step 20 / 391  
Epoch 2/2 Step 40 / 391  
Epoch 2/2 Step 60 / 391  
Epoch 2/2 Step 80 / 391  
Epoch 2/2 Step 100 / 391  
Epoch 2/2 Step 120 / 391  
Epoch 2/2 Step 140 / 391  
Epoch 2/2 Step 160 / 391  
Epoch 2/2 Step 180 / 391  
Epoch 2/2 Step 200 / 391  
Epoch 2/2 Step 220 / 391  
Epoch 2/2 Step 240 / 391  
Epoch 2/2 Step 260 / 391  
Epoch 2/2 Step 280 / 391  
Epoch 2/2 Step 300 / 391  
Epoch 2/2 Step 320 / 391  
Epoch 2/2 Step 340 / 391  
Epoch 2/2 Step 360 / 391  
Epoch 2/2 Step 380 / 391  
Epoch 2/2 Elapsed: 47.857 sec Acc: 0.642  
Epoch 2/2 Test Acc: 0.648
```



- **Problem:** PyTorch throws errors when executing distributed variant
 - “Port already in use”
 - **Reasons:** You will potentially be working on a shared machine (with only 2 of 4 GPUs)
- **Solution:** Select a different port

- **Problem:** Be careful with using `$ (pwd)` or `$PWD` inside a container
 - This might return `/rwthfs/...` which is not mounted by default
 - What's interesting:
 - `/work/<user-id>/` is mounted by default
 - `/rwthfs/rz/cluster/work/<user-id>/` is not although pointing to the same path
 - **Cause:** Can occur when working with multiplexers like `tmux` on the host system
- **Solution:**
 - Map the directories that you want to use in the container
 - Also see hands-on examples