ACES: Intermediate Python Programming In JupyterLab

Accelerating Workflows on a Composable Cyberinfrastructure

Zhenhua He 07/14/2024











ACES TechLab

Lab I. JupyterLab (30 mins)

We will load required modules and activate virtual environment and run JupyterLab on HPRC ACES portal.

Lab II. Data Exploration with Python(30 mins)

We will go through some examples with a popular Python library Pandas for data exploration.

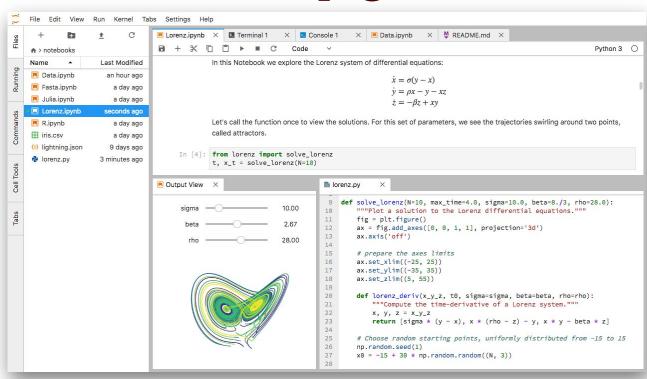
Lab III. Machine/Deep Learning (30 minutes)

We will learn how to use PyTorch to build and train a simple image classification model with deep neural network (DNN).



Lab I. JupyterLab







L1 - Resources

- Texas A&M High Performance Research Computing (HPRC)
- HPRC Microcredentials and Courses
- ACES Quick Start Guide
- ACES Portal (ACCESS)
- ACCESS Documentation
- HPRC YouTube Channel
- help@hprc.tamu.edu



NSF ACES

Accelerating Computing for Emerging Sciences

Our Mission:

- Offer an accelerator testbed for numerical simulations and AI/ML workloads
- Provide consulting, technical guidance, and training to researchers
- Collaborate on computational and data-enabled research.



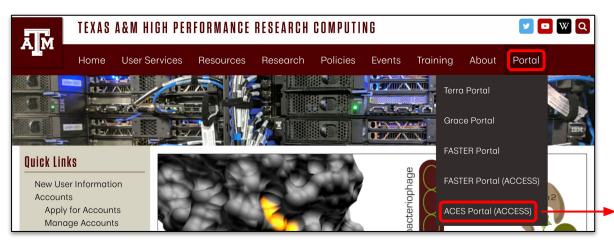


ACES Accelerators

Component	Quantity	Description
Graphcore IPU	32	16 Colossus GC200 IPUs, 16 Bow IPUs. Each IPU group hosted with a CPU server as a POD16 on a 100 GbE RoCE fabric
Intel PAC D5005 FPGA	2	Accelerator with Intel Stratix 10 GX FPGA and 32 GB DDR4
BittWare IA-840F FPGA	2	Accelerator with Agilex AGF027 FPGA and 64 GB of DDR4
NextSilicon Coprocessor	2	Reconfigurable accelerator with an optimizer continuously evaluating application behavior.
NEC Vector Engine	8	Vector computing card (8 cores and HBM2 memory)
Intel Optane SSD	48	18 TB of Intel Optane SSDs addressable as memory w/ MemVerge Memory Machine.
NVIDIA H100 + A30	30 + 4	NVIDIA GPUs for HPC, DL Training, AI Inference
Intel PVC	110	Software Development Platform for PVC

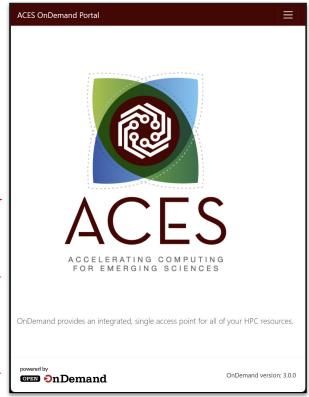


ACES Portal

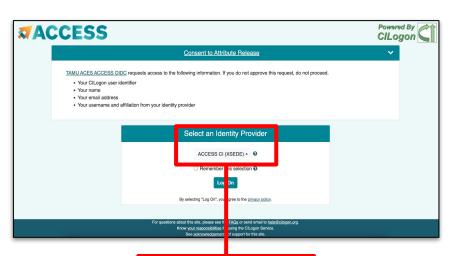


ACES Portal <u>portal-aces.hprc.tamu.edu</u> is the web-based user interface for the ACES cluster

Open OnDemand (OOD) is an advanced web-based graphical interface framework for HPC users



Authentication via CILogon



Log-in using your ACCESS CI credentials.



Select an Identity Provider

ACCESS CI (XSEDE) ~ ②

Select the Identity Provider appropriate for your account.



Get a Shell on ACES

Click on "Clusters" menu → _aces Shell Access

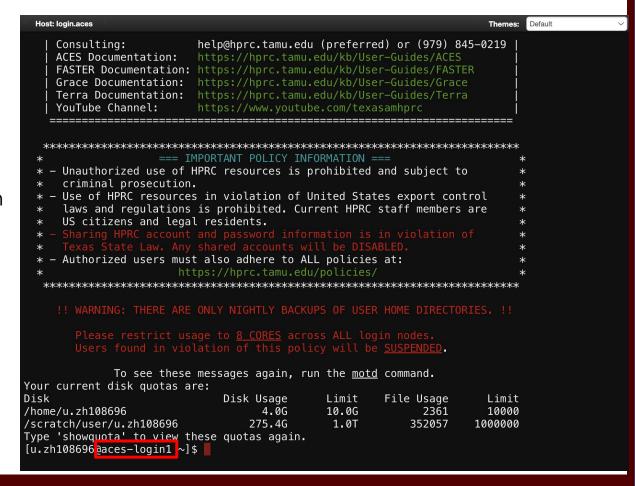




Success!

Welcome to the ACES login node.

Check which login node you are on.



Commands to copy the materials

Navigate to your personal scratch directory
 \$cd \$SCRATCH

• Files for this course are located at

```
/scratch/training/nh_wkshop
Make a copy in your personal scratch directory
$cp -r /scratch/training/nh wkshop $SCRATCH
```

Enter this directory (your local copy)

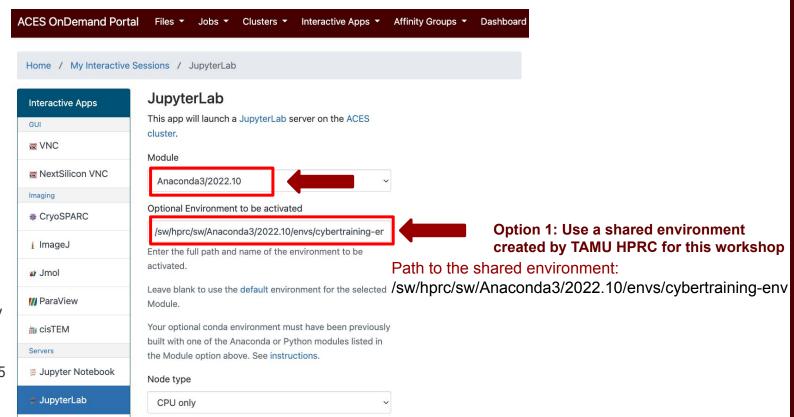
```
$cd nh_wkshop
```

Go to JupyterLab Page





JupyterLab Page



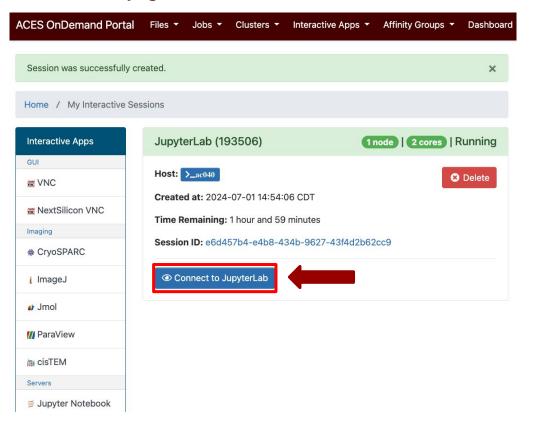
Other fields:

Node Type: CPU only Number of hours: 2 Number of cores: 2

Total memory (GB): 5

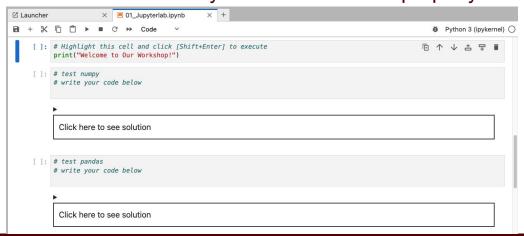


Connect to JupyterLab



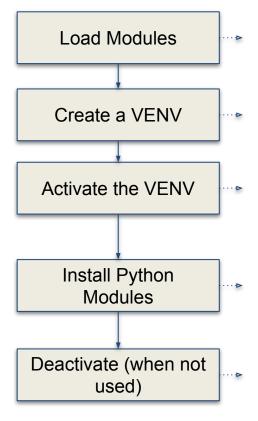
Review and Exercise

- Log into ACES through ACES Portal (ACCESS)
- Copy the training materials to your \$SCRATCH directory
- Launch JupyterLab app
- In the notebook named 01_Jupyterlab.ipynb, follow the instructions to import the required modules to make sure they have been loaded properly.





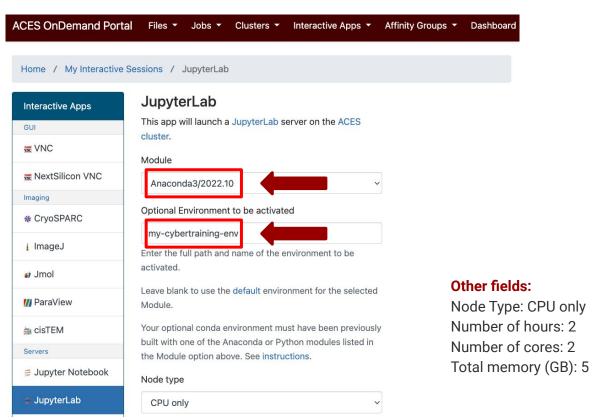
Option 2



```
# clean up and load Anaconda
cd $SCRATCH
module purge
module load Anaconda3/2022.10
# create a Python virtual environment
conda create -n my-cybertraining-env
# activate the virtual environment
source activate my-cybertraining-env
# install required package to be used in the portal
conda install -c conda-forge jupyterlab
conda install -c conda-forge numpy
conda install -c conda-forge pandas
conda install -c conda-forge xarray geopandas folium
conda install -c conda-forge netcdf4
(install other packages as well ...)
# deactivate the virtual environment
# source deactivate
```



JupyterLab Page





Lab II. Data Exploration













Data Structures

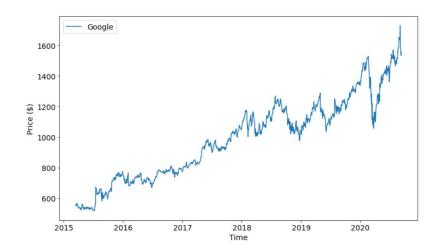
Pandas has two data structures that are descriptive and optimized for data with different dimensions.

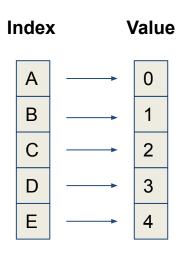
- **Series:** 1D labeled array
- DataFrame: General 2D labeled, size-mutable tabular structure with potentially heterogeneously-typed columns



Series in pandas

- One-dimensional labeled array
- Capable of holding any data type (integers, strings, floating point numbers, etc.)
- Example: time-series stock price data





DataFrame in pandas

- Primary Pandas data structure
- A dict-like container for Series objects
- Two-dimensional size-mutable

(source: https://ds.iris.edu/seismon/eventlist/index.phtml)

Heterogeneous tabular data structure

C1 C₂ **C3** Index C4 **DEPTH** DATE and TIME (UTC) **♦ LAT ♦ LON** ♦ MAG 01-JUL-2024 17:10:59 13.97 52.28 10 4.6 0.1 True X 192 01-JUL-2024 15:22:40 -5.51 147.18 4.5 -32.63-69.34 4.3 10 01-JUL-2024 13:04:10 2.4 В 01-JUL-2024 11:54:40 -61.97 155.22 4.6 10 False 01-JUL-2024 11:37:41 -23.41-68.10 4.5 126 01-JUL-2024 10:20:59 37.86 73.24 4.6 10 1.9 True 42 01-JUL-2024 10:16:23 -5.56153.59 5.0 01-JUL-2024 05:53:52 10.79 125.37 4.8 8.3 01-JUL-2024 05:22:23 10.74 125.39 4.8 68 NA False W 01-JUL-2024 05:12:03 14.74 147.06 5.0 10 01-JUL-2024 02:38:03 -62.68165.56 5.1 10 Ε 9 6.8 False а

Columns



Pandas Learning Objectives

After this section, you will learn:

- DataFrame building
- DataFrame operations
 - Relabeling
 - Data grouping
- Data handling
 - Handle missing data
 - Handle duplicate data
 - Merege DataFrames



JupyterLab Exercises



Lab III. Machine/Deep Learning

Deep Learning

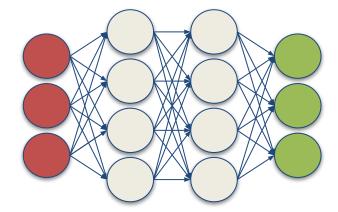
by Ian Goodfellow, Yoshua Bengio, and Aaron Courville http://www.deeplearningbook.org/

Animation of Neutron Networks

by Grant Sanderson https://www.3blue1brown.com/

Visualization of CNN

by Adam Harley
https://adamharley.com/nn-vis/cnn/3d.html



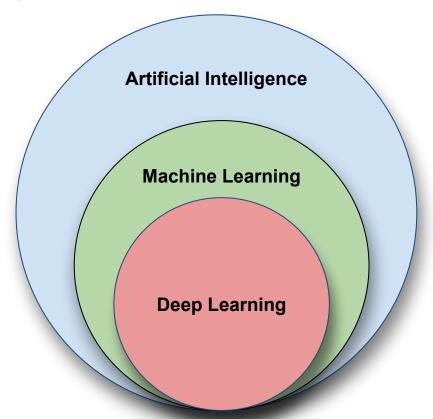






Relationship of AI, ML, and DL

- Artificial Intelligence (AI) is anything about man-made intelligence exhibited by machines.
- Machine Learning (ML) is an approach to achieve Al.
- Deep Learning (DL) is one technique to implement ML.



Types of ML Algorithms

Supervised Learning

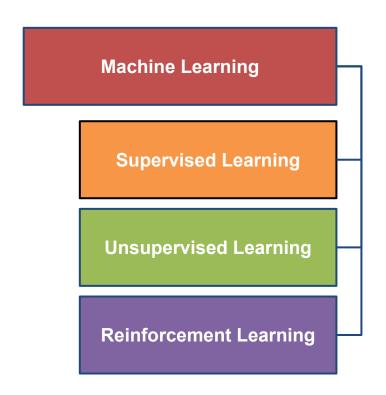
 trained with labeled data; including regression and classification problems

Unsupervised Learning

 trained with unlabeled data; clustering and association rule learning problems.

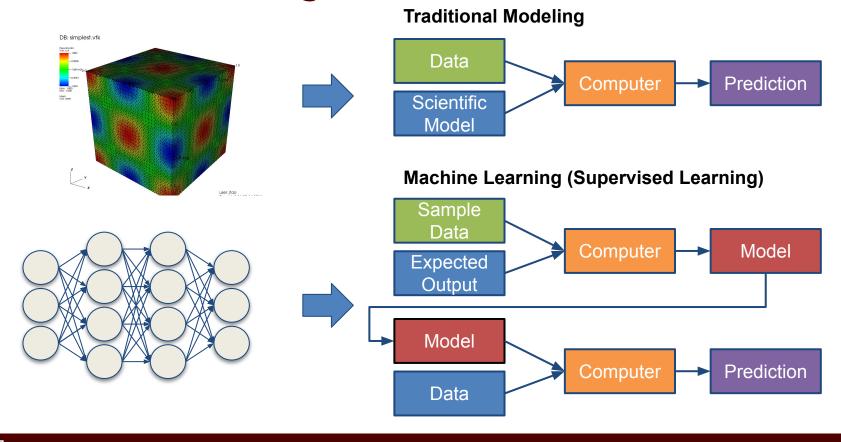
• Reinforcement Learning

 no training data; stochastic Markov decision process; robotics and business strategy planning.



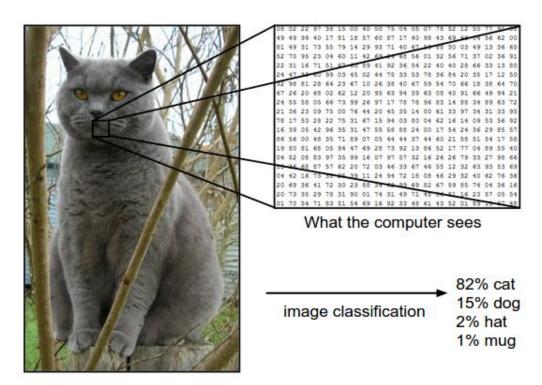


Machine Learning





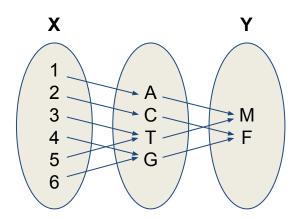
Inputs and Outputs



256 X 256
Matrix

DL model

4-Element Vector

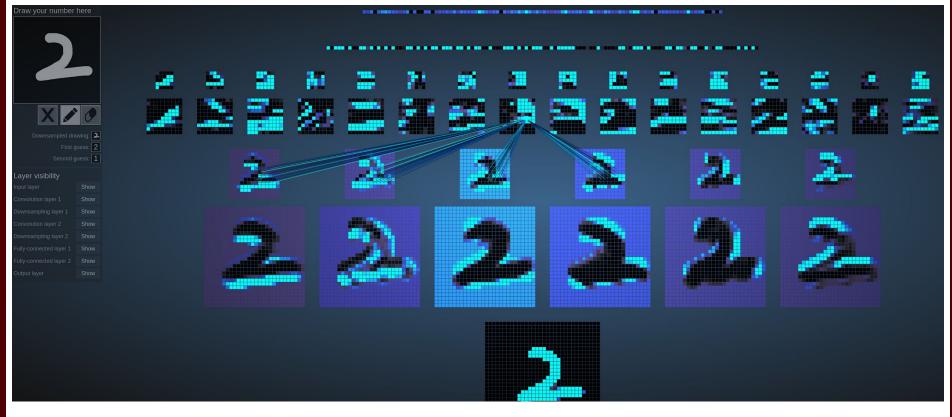


With deep learning, we are searching for a **surjective** (or **onto**) function **f** from a set **X** to a set **Y**.

Image from the Stanford CS231 Course



MNIST - CNN Visualization



(Image Credit: https://adamharley.com/nn_vis/cnn/3d.html)



CNN Explainer



(Image Credit: https://poloclub.github.io/cnn-explainer/)







High Performance Research Computing

https://hprc.tamu.edu

HPRC Helpdesk:

help@hprc.tamu.edu

Phone: 979-845-0219

Help us help you. Please include details in your request for support, such as, Cluster (ACES, FASTER, Grace, Terra, ViDaL), NetID (UserID), Job information (Job id(s), Location of your jobfile, input/output files, Application, Module(s) loaded, Error messages, etc), and Steps you have taken, so we can reproduce the problem.

