ACES: Intermediate Python Programming In JupyterLab

Accelerating Workflows on a Composable Cyberinfrastructure

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High Performance Research Computing DIVISION OF RESEARCH







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) on GPU.



Lab I. JupyterLab



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📃 Julia.ipynb	a day ago	$\dot{y} = \rho x$ $\dot{z} = -l$	y x2			
• 🖪 Lorenz.ipynb	seconds ago	~ *				
🖪 R.ipynb	a day ago	Let's call the function once to view the solutions. For this set of	parameters, we see the trajectories swirling around two points,			
🖽 iris.csv	a day ago	called attractors.				
{:} lightning.json	9 days ago					
🍦 lorenz.py	3 minutes ago	<pre>in [4]: from Lorenz import solve_Lorenz t, x_t = solve_Lorenz(N=10)</pre>				
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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



ACES Portal



Accessing via ACCESS







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.

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Host: login.aces				Themes: Default					
Consulting: ACES Documentation: FASTER Documentation: Grace Documentation: Terra Documentation: YouTube Channel:	<pre>help@hprc.tamu.edu https://hprc.tamu.e https://hprc.tamu.e https://hprc.tamu.e https://hprc.tamu.e https://www.youtube</pre>	(preferre edu/kb/Use edu/kb/Use edu/kb/Use edu/kb/Use e.com/texa	ed) or (979) er-Guides/ACE er-Guides/FAS er-Guides/Gra er-Guides/Ter asamhprc	845-0219 STER ice rra					
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Commands to copy the materials

• Navigate to your personal scratch directory

\$cd \$SCRATCH

• Files for this course are located at

/scratch/training/CyberTraining

Make a copy in your personal scratch directory

\$ cp -r /scratch/training/CyberTraining \$SCRATCH

• Enter this directory (your local copy)

\$ cd CyberTraining

Go to JupyterLab Page

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ACCELERATING COMPUTING FOR EMERGING SCIENCES

JupyterLab Page

ACES OnDemand Portal Files • Interactive Apps -Affinity Groups -Dashboard Jobs 🔻 Clusters -Home / My Interactive Sessions / JupyterLab JupyterLab Interactive Apps This app will launch a JupyterLab server on the ACES GUI cluster. WNC WNC Module RextSilicon VNC Anaconda3/2022.10 Imaging Optional Environment to be activated * CryoSPARC **Option 1: Use a shared environment** /sw/hprc/sw/Anaconda3/2022.10/envs/cybertraining-en ImageJ created by TAMU HPRC for this workshop Enter the full path and name of the environment to be Path to the shared environment: activated. # Jmol Leave blank to use the default environment for the selected /sw/hprc/sw/Anaconda3/2022.10/envs/cybertraining-env ParaView Module. Your optional conda environment must have been previously a cisTEM built with one of the Anaconda or Python modules listed in Servers the Module option above. See instructions. Jupyter Notebook Node type JupyterLab CPU only V

Other fields:

Node Type: CPU only Number of hours: 3 Number of cores: 2 Total memory (GB): 5



Connect to JupyterLab

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an cisTEM								
Servers								



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

ACES OnDemand Portal Files 🔻 Jobs 🔻 Clusters -Interactive Apps 🔻 Affinity Groups 🔻 Dashboard

Home / My Interactive Sessions / JupyterLab

Interactive Apps	Supytereds	
<u><u></u></u>	This app will launch a JupyterLab server on the ACES	
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Lab II. Data Exploration









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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
- Heterogeneous tabular data structure

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13.97

-5.51

-32.63

-61.97

-23.41

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-69.34

155.22

-68.10

73.24

153.59

125.37

125.39

147.06

165.56

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 آ

Lab III. Machine/Deep Learning

Deep Learning by Ian Goodfellow, Yoshua Bengio, and Aaron Courville <u>http://www.deeplearningbook.org/</u>

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 AI.
- **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

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Inputs and Outputs



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

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MNIST - CNN Visualization



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

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CNN Explainer



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

JupyterLab Exercises





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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.

