Things to do while you are waiting

- Course slides are available at: <u>https://hprc.tamu.edu/training/aces_intel.html</u>
- Get ready to SSH to the FASTER cluster
 - For ACCESS users:
 - Disconnect from your non-TAMU VPN
 - For TAMU users:
 - Log into TAMU VPN (if you're off campus)

Data Science for Python

using the FASTER and ACES clusters in preparation for Intel AI Analytics Toolkit

> by Richard Lawrence Date: 10/25/2022

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Fall 2022

Outline

- Getting Started with FASTER and ACES
- Jupyter Notebook Environment
- Data Structure with Pandas
- Machine Learning with Scikit Learn
- Machine Learning with XGBoost

Learning Resources

- ACCESS Documentation <u>https://access-ci.atlassian.net/wiki/spaces/ACCESSdocumenta</u> <u>tion/pages/95915115/FASTER+Texas+A+M</u>
- HPRC Wiki <u>https://hprc.tamu.edu/wiki/FASTER</u>
- HPRC on Youtube https://www.youtube.com/c/TexasAMHPRC



Getting Started with FASTER and ACES



FASTER Cluster

hprc.tamu.edu/wiki/FASTER:Intro

Node Type	Quantity
64-core login nodes	4 (3 for TAMU, 1 for ACCESS)
64-core compute nodes (256GB RAM each)	180 (11,520 cores)
Composable GPUs	200 T4 16GB 40 A100 40GB 10 A10 24GB 4 A30 24GB 8 A40 48GB
Interconnect	Mellanox HDR100 InfiniBand (MPI and storage) Liqid PCIe Gen4 (GPU composability)
Global Disk	5PB DDN Lustre appliances



FASTER (Fostering Accelerated Sciences Transformation Education and Research) is a 180-node Intel cluster from Dell featuring the Intel Ice Lake processor.





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ACES - Accelerating Computing for Emerging Sciences (Phase I)



Component	Quantity	Description
<u>Graphcore IPU</u>	16	16 Colossus GC200 IPUs and dual AMD Rome CPU server on a 100 GbE RoCE fabric
Intel FPGA PAC D5005	2	FPGA SOC with Intel Stratix 10 SX FPGAs, 64 bit quad-core Arm Cortex-A53 processors, and 32GB DDR4
Intel Optane SSDs	8	3 TB of Intel Optane SSDs addressable as memory using MemVerge Memory Machine.

ACES Phase I components are available through **FASTER**



Overview: Jupyter Lab on FASTER

- 1. Reach a login node
- 2. Make a copy of the exercise files
- 3. Reach a compute node
- 4. Open Jupyter Lab in browser

Accessing FASTER via SSH (TAMU users)

Two-Factor Authentication enabled using TAMU CAS.

- Off campus:
 - Set up and start VPN (Virtual Private Network): u.tamu.edu/VPnetwork
- SSH programs for Windows:
 - MobaXTerm (preferred, includes SSH and X11)
 - PuTTY SSH
 - Windows Subsystem for Linux

Accessing FASTER for TAMU users

- FASTER has two login nodes for TAMU users.
- SSH to either login node:

ssh -L <useridnum>:localhost:<useridnum>
netid@faster.hprc.tamu.edu



Accessing FASTER for ACCESS users

- ACCESS users must submit their ssh public key for installation in the FASTER jump host.
- FASTER has 1 login node for ACCESS users.
- SSH to login node via Jump Host:
 - \$ ssh
 - -L <useridnum>:localhost:<useridnum>
 - -J <fasterusername>@faster-jump.hprc.tamu.edu:8822
 - <fasterusername>@login.faster.hprc.tamu.edu

Files for the Exercises

- Navigate to your personal scratch directory \$ cd \$SCRATCH
- Files for this course are located at /scratch/training/intel-aiml-aces Make a copy in your personal scratch directory
 \$ cp -r /scratch/training/intel-aiml-aces \$SCRATCH
 Enter this directory (your local copy)
- Enter this directory (your local copy)
 \$ cd intel-aiml-aces
- Make a copy of the Intel AI examples (if attending afternoon)
 \$ git clone https://github.com/oneapi-src/oneAPI-samples.git

Reaching a Compute Node

- Execute slurm command to get a compute node
 \$ sbatch intel-jupyterlab-tunnel.slurm
- View the job output file
 \$ cat intel-jupyterlab.job.*
- Copy, paste, and execute the ssh command that appears near the top of the output file. Example:

\$ ssh -4 -L <port>:localhost:<port> <nodename>

Open Jupyter Lab in Browser

- Towards the end of the job output file (viewed like this)
 \$ cat intel-jupyterlab.job.*
- login instructions will appear. Example:
 To access the server, open this file in a browser: file:///home/<username>/.local/share/jupyter/runtime/jpserver-462321-open.html
 Or copy and paste one of these URLs: http://localhost:<port>/lab?token=67b0e1263053b6bc449c59999984bbfc30a97fa61fcd9e18
 - or http://127.0.0.1:<port>/lab?token=612b6b0iiic840c449c5a97fa61bbfc3fcd9e7b630530e18
- Copy and paste the link into your browser.

Jupyter Lab Environment



Intel Software

Intel Software integrated into HPRC Module Hierarchy

- module load intel/Toolkits
- (This command is in the slurm job file, already executed).

Provides access to a Conda environment where AI Toolkit and JupyterLab are installed.



Jupyter Lab File Navigator

Navigate to the "Hello_world.ipynb" file. Open by double-clicking.



Jupyter Exercises

Complete the exercises in the Hello_World.ipynb notebook.



Data Structure with Pandas



Pandas Series

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







Pandas DataFrame

- Primary Pandas data structure
- Like a dictionary of Series objects
- Tabular data structure
- Two-dimensional
- Size-mutable
- Heterogeneous





DataFrame Example

House sales data, King County

A	В	С	D	E	F	G	н
id	date	price	bedrooms	athrooms	sqft_living	sqft_lot	floors
7129300520	20141013T0	221900	3	1	1180	5650	1
6414100192	20141209T0	53800(3	2.25	2570	7242	2
5631500400	20150225T0	18000	2	1	770	10000	1
2487200875	20141209T0	604000	4	3	1960	5000	1
1954400510	20150218T0	51000	3	2	1680	8080	1
7237550310	20140512T0	1.23E+06	4	4.5	5420	101930	1
1321400060	20140627T0	257500	3	2.25	1715	6819	2
2008000270	20150115T0	29185(3	1.5	1060	9711	1
2414600126	20150415T0	229500	3	1	1780	7470	1

Pandas Exercises

Complete the exercises in the Pandas.ipynb notebook.



Machine Learning with Scikit Learn



Features of Scikit Learn



Classification	Regression	Clustering	Dimension Reduction	Model Selection	Preprocessing
Identifying category of an object Applications: Spam detection, image recognition. Algorithms: SVM, nearest neighbors, random forest, and more	Predicting a attribute for an object Applications: Drug response, Stock prices. Algorithms: SVR, nearest neighbors, random forest, and more	Grouping similar objects into sets Applications: Customer segmentation, Grouping experiment outcomes Algorithms: k-Means, spectral clustering, mean-shift, and more	Reducing the number of dimensions Applications: Visualization, Increased efficiency Algorithms: k-Means, feature selection, non-negative matrix factorization and	Selecting models with parameter search Applications: Improved accuracy via parameter tuning Algorithms: grid search, cross validation, metrics, and more	Preprocessing data to prepare for modeling Applications: Transforming input data such as text for use with machine learning algorithms. Algorithms: preprocessing feature
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Regression



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Scikit Learn Exercises

Complete the exercises in the Linear_regression.ipynb notebook.



Machine Learning with XGBoost



Decision-making

- Prediction function is step-wise
 (f, : a<x<b/li>
 - $f = \begin{cases} f_1 : a < x < b \\ f_2 : b < x < c \\ \dots \end{cases}$
- Objective Function is **dual**
 - \circ obj(f)=L(f)+ $\Omega(f)$
 - L is prediction error
 - \circ Ω is regularization



Images from https://xgboost.readthedocs.io/en/stable/tutorials/model.html

Decision Trees

- Complex Question?
 - Multiple Variables
 - Multiple Splits per Variable
 - Many Possible Tree Graphs
- "Learning" means growing the tree one Variable Split at a time



XGBoost Exercises

Complete the exercises in the Boosted_trees.ipynb notebook.



Shutdown JupyerLab

- In Browser
 - File \rightarrow Shutdown \rightarrow Yes
- Command line
 - \$ squeue -u <username>
 - \$ scancel <jobid>

Thank you

Contact: help@hprc.tamu.edu

