# Standard Operating Procedure (SOP) for User Onboarding and Job Execution on csis\_cluster1

#### Purpose: -

This SOP provides new users with a definitive guide to accessing the [Your Cluster Name] HPC cluster, understanding its resources, software, and policies, and successfully submitting computational jobs. This ensures consistent, secure, and efficient use of shared resources.

#### Scope: -

This procedure applies to all institute/department stakeholders who have been granted access to the aforementioned cluster.

# Prerequisites: -

- An approved user account with valid credentials.
- An SSH client (For Example: ssh on Linux/macOS, <u>PuTTY/MobaXterm</u> on Windows).
- Basic familiarity with the Linux CLI mode (Command Line Interface).

#### Definitions: -

#### Login Node/Master Node:

The entry point for editing files, compiling code, and submitting jobs.

#### Slurm:

The job scheduler (Simple Linux Utility for Resource Management) manages all compute resources.

#### Batch Job:

A computational task is submitted to a queue (partition) to run without user interaction.

#### **QoS (Quality of Service):**

A set of limits that override or supplement partition limits, often used to grant different priorities or capabilities (e.g., longer runtimes).

# **Compute Node:**

Worker nodes where jobs are executed. Access is allocated by the scheduler.

#### **Environment Modules:**

Software that allows users to dynamically load pre-installed applications and their dependencies into their shell environment.

#### Partition:

A queue of nodes with specific resource limits (max time, cores/GPU).

#### **SBATCH Script**:

A shell script containing directives for Slurm (`#SBATCH` lines) and commands to run your software.

#### Technical Specifications of Compute Node:

OEM & Model	DELL PowerEdge R7525
CPU:	2× AMD EPYC 7742 (128 cores total per node)
RAM:	256 GB DDR4 memory per node
OS:	Rocky Linux 8.10
Kernel:	4.18.0-553.el8_10.x86_64
NVIDIA Driver:	570.158.01
GPU:	2× NVIDIA A100 80GB GPUs per node
CUDA Cores:	6,912 per GPU
Tensor Cores:	432 per GPU
Total VRAM & CUDA Cores per Node:	160 GB VRAM; 13,824 CUDA cores
Workload Manager	Slurm 23.11.11 (https://slurm.schedmd.com/)

Processor Settings				
System BIOS Settings • Processor Settings				
PROCESSOR 1				
Family-Model-Stepping	17-31-0			
Brand	AMD EPYC 7502 32-Core Processo			
Level 2 Cache	32x512 KB			
Level 3 Cache	128 MB			
Number of Cores	32			
Microcode	0x8301052			
PROCESSOR 2				
Family-Model-Stepping	17-31-0			
Brand	AMD EPYC 7502 32-Core Processor			
Level 2 Cache	32x512 KB			

							+	+	
	Name Temp	Perf		Persist Pwr:Usa			Bus-Id Disp.A   Memory-Usage 		Uncorr. ECC Compute M. MIG M.
0 N/A	NVIDIA 30C	A100 P0	80GB PCI		/	0n 300W	00000000:21:00.0 Off   18MiB / 81920MiB	0%	0 Default Disabled
1 N/A	NVIDIA 29C	A100 P0	80GB PCI		/	0n 300W	00000000:81:00.0 off   18MiB / 81920MiB 	   0% 	0 Default Disabled
Proce GPU	sses: GI ID	CI ID	PID	Туре		Proces	ss name		GPU Memory Usage
Θ 1	N/A N/A	N/A N/A	3198 3198	G G			lib/xorg/Xorg lib/xorg/Xorg		4MiE 4MiE

#### Procedure: -

Step 1: Secure Connection (Login) to the Cluster

- I. Open your terminal (or SSH client).
- II. Connect using the following command: ssh <username>@172.24.16.132
- III. Enter your password and any 2FA token when prompted.

```
ThinkCentre-neo-50s-Gen-3:-$ ssh rs2@172.24.16.132

The authenticity of host '172.24.16.132 (172.24.16.132)' can't be established. ED25519 key fingerprint is SHA256:r4YlDhIWyytAU6+Grh8cS/TFeKpZ3UuN+5wFWpcjwkg. This key is not known by any other names

Are you sure you want to continue connecting (yes/no/[fingerprint])? yes Warning: Permanently added '172.24.16.132' (ED25519) to the list of known hosts. rs2@172.24.16.132's password:

Web console: https://csis.mn1.cluster_csis1:9090/ or https://172.24.16.132:9090/

Last login: Thu Aug 28 16:29:01 2025

[rs2@csis ~]$
```

Step 2: Understand the Environment

Initially, users will be on a master node as they successfully make their login. User is supposed to execute the job by preparing and using Slurm files, not directly on the node's terminal console. Each user has a **500 GB** quota limit for their home directory. It may be extended as the cluster is getting more disk availability.

To check the home directory usage details

```
du -sh $HOME
du -sh /nfs home/username
```

 To check the allotted home directory size details

```
quota -u $USER
quota -u /nfs_home/username
```

```
[rs1@csis ~]$ pwd
/nfs_home/rs1
[rs1@csis ~]$ du -sh $HOME
59G /nfs home/rs1
```

In the above fig, cited user has used around 75GB, **quota** (Soft Limit:- the **warning threshold**) is 499GB, and **limit** (Hard) 500GB

NB: - If the allotted user quota is near or exceeded with stored data, the user must archive or delete files before new jobs can run or data can be saved.

#### Step 3: Explore Available Software Modules and Resources

A key software stack is pre-installed and managed via Environment Modules. This simplifies loading complex software and its dependencies.

# Key Available Modules Include:

```
Froot@csis build1# module avail
              ------/nfs_home/software/modulefiles ------
                                   libraries/openblas/0.3.27
   apps/gamess/00
  apps/jupyter/lab
                                   libs/lapack/3.12.0
  apps/miniconda/default
                                   libs/scalapack/2.2.1
  envs/pytorch/1.0
                                   mpi/default
  envs/tensorflow/1.0
                                   mpi/mpich-x86_64
   lang/python/3.9
                                   mpi/mpich
   lang/python/3.13.5
                             (D)
                                   mpi/openmpi-x86 64
                                                           (D)
  libraries/boost/1.84.0
                                   mpi/openmpi
   libraries/fftw/3.3.10-double
                                   mpi/pmix-x86 64
  libraries/hdf5/1.14.3-serial
          -----/nfs_home/software/modulefiles/apps ------
0 jupyter/lab miniconda/default
   gamess/00
  pytorch/1.0 tensorflow/1.0
  python/3.9 python/3.13.5 (D)
                 ----- /nfs_home/software/modulefiles/mpi ------
  default mpich-x86 64
                           mpich openmpi-x86_64 (D)
                                                       openmpi pmix-x86 64
             ----- /nfs_home/software/modulefiles/libraries ------
  boost/1.84.0 fftw/3.3.10-double hdf5/1.14.3-serial
                                                        openblas/0.3.27
 Where:
  D: Default Module
If the avail list is too long consider trying:
"module --default avail" or "ml -d av" to just list the default modules.
"module overview" or "ml ov" to display the number of modules for each name.
Use "module spider" to find all possible modules and extensions.
Use "module keyword key1 key2 ... " to search for all possible modules matching any of
the "keys".
```

#### **Short Notes about available Modules**

#### 1) lang/ (Programming Languages)

- **python/3.9**: The Python 3.9 interpreter and core libraries. Useful for projects that require compatibility with this specific, older version.
- **python/3.13.5 (D)**: The default Python 3.13.5 interpreter. This is the latest version provided by the system and is the one that will load if you just type `module load lang/python`.

# 2) envs/(Pre-configured Environments)

- pytorch/1.0: A pre-installed environment for the PyTorch 1.0 deep learning framework.
   Loading this module gives you immediate access to this specific version of PyTorch and its dependencies without needing to install it yourself.
- **tensorflow/1.0**: A pre-installed environment for the TensorFlow 1.0 deep learning framework. Useful for running older code designed for the TF1 API, which is significantly different from modern TF2.

# 3) mpi/ (Message Passing Interface Libraries)

- **openmpi-x86\_64 (D)**: The default version of the OpenMPI library, compiled for x86\_64 architecture. This is the most common high-performance MPI implementation for distributed computing and parallel jobs.
- **openmpi**: Likely a symbolic link or alternative version of the OpenMPI library.
- **mpich-x86\_64**: The MPICH implementation, another high-performance and widely portable MPI library, compiled for x86 64.
- **mpich**: Likely a symbolic link or alternative version of the MPICH library.
- **pmix-x86\_64**: The Process Management Interface (PMIx) library, often used as a runtime for MPI implementations and other tools in resource-managed environments (like Slurm).
- **default**: A fallback module that likely points to the system's default MPI implementation (probably `openmpi-x86 64`).

# 4) libraries/ (Scientific & Numerical Libraries)

- openblas/0.3.27: The OpenBLAS library, an optimized implementation of the BLAS
   (Basic Linear Algebra Subprograms) API. Dramatically accelerates numerical and linear
   algebra operations in code like NumPy, R, and many scientific applications.
- **LAPACK**: Built upon BLAS, this library provides higher-level routines for solving systems of simultaneous linear equations, least-squares solutions of linear systems of equations, eigenvalue problems, and singular value problems, forming the core of many quantum chemistry algorithms.
- ScaLAPACK (Scalable LAPACK): A parallel extension of LAPACK for distributed-memory MIMD parallel computers. Building this library was critical for enabling the parallel execution of linear algebra operations across multiple compute nodes, which is necessary for studying large molecular systems with applications like GAMESS.

- **fftw/3.3.10-double**: The FFTW (Fastest Fourier Transform in the West) library, version 3.3.10, compiled for double-precision calculations. The standard library for high-performance Fourier transforms.
- hdf5/1.14.3-serial: The HDF5 library, version 1.14.3, compiled for serial (non-parallel) use. Used for managing complex data structures and storing large amounts of numerical data.
- **boost/1.84.0**: The Boost C++ source libraries, version 1.84.0. A large collection of peer-reviewed, portable C++ libraries that provide support for tasks like linear algebra, multithreading, and graph algorithms.

# 5) apps/ (Applications & Tools)

- miniconda/default: Provides the `conda` package manager and a minimal base Python environment. Essential for creating and managing your own isolated software environments for Python or R, allowing you to install specific package versions without requiring system administrator privileges.
- **jupyter/lab**: For launching JupyterLab sessions. This module is typically used with a job scheduler (like Slurm) to start an interactive session on a compute node, providing a web-based development environment for code, data, and documentation.
- **GAMESS** (General Atomic and Molecular Electronic Structure System)
  It is a quantum chemistry software package. This is a comprehensive computational chemistry program for ab initio molecular orbital calculations, including methods such as Hartree-Fock, DFT, MP2, CCSD(T), and many other advanced quantum mechanical methods. Essential for researchers in computational chemistry, materials science, and molecular modeling to perform electronic structure calculations, geometry optimizations, frequency calculations, and molecular property predictions.

Step 4: How to View and Load Available Software Modules:

- module avail
- module avail python
- module show lang/python/3.13.5
- module load miniconda/default
- module load jupyter/lab
- module list
- module unload jupyter/lab
- module purge

- # View all available modules
- # Search for specific modules (e.g., python)
- # Get information about a specific module
- # Load the default Miniconda module
- \*Mention the required module name explicitly for hassle-free setup.
- # Load JupyterLab for interactive work
- # See currently loaded modules
- # To Unload a specified module
- # Unload all modules

Step 5: How to View Cluster Specifications & Partitions:

sinfo

- # Overview of all partitions and node states
- sinfo -o "%20P %5D %14F %8z %10m %10c %10G"
- # Detailed view: Partition, Nodes, State, Memory, CPUs, GPUs

```
[rs1@csis ~]$ sinfo
PARTITION AVAIL TIMELIMIT
                            NODES
                                    STATE NODELIST
debug
             up
                     30:00
                                 3
                                     idle csis.cn[1-2],csis.mn1
                                     idle csis.cn[1-2],csis.mn1
                                 3
short*
                   4:00:00
             up
long
             up 1-00:00:00
                                 3
                                     idle csis.cn[1-2],csis.mn1
gpu-short
                                 3
                                     idle csis.cn[1-2],csis.mn1
                   4:00:00
gpu-long
                  12:00:00
                                 3
                                     idle csis.cn[1-2],csis.mn1
             up
                                 3
                                     idle csis.cn[1-2],csis.mn1
bigmem
                   8:00:00
             up
[rs1@csis ~]$
```

```
[rs1@csis ~]$ sinfo -o "%20P %5D %14F %8z %10m %10c %10G"
PARTITION NODES NODES(A/I/O/T) S:C:T MEMORY
                        NODES NODES(A/I/O/T) S:C:T
                                                                        CPUS
                                                                                     GRES
debug
                               0/3/0/3
                                                           240000
                                                                                     gpu:a100-8
                        3
                                                 2:32:2
                                                                        128
short*
                        3
                               0/3/0/3
                                                 2:32:2
                                                           240000
                                                                        128
                                                                                     gpu:a100-8
long
                               0/3/0/3
                                                2:32:2
                                                           240000
                                                                        128
                                                                                     gpu:a100-8
gpu-short
                        3
                               0/3/0/3
                                                 2:32:2
                                                           240000
                                                                        128
                                                                                     gpu:a100-8
                        3
                                                           240000
gpu-long
                                                 2:32:2
                                                                        128
                                                                                     gpu:a100-8
                               0/3/0/3
                        3
                               0/3/0/3
                                                 2:32:2
                                                           240000
                                                                        128
                                                                                     gpu:a100-8
bigmem
[rs1@csis ~]$ |
```

Step 6: How to View GPU Specifications

•	sinfo -o "%20N %10G" # Shows gpu:a100-80gb:2	# To see physical GPU
	5,	count per node
•	nvidia-smi	# To see GPU
		utilization, including
		memory
•	nvidia-smi -q   grep -E "(Product Name FB Memory Count)"	# To see detailed GPU
		properties

# Step 7:

Create a Job Script with Correct Partitions and QoS. The user may refer to the table below to know the most effective parameters with respect to the proposed job.

#### **QOS to Partition Mapping**

QOS Name	Max Time	Max CPUs/Job	Max Memory/Job	Max GPUs/Job
debug	30 min	16 CPUs	48GB	0 GPUs
short	4 hours	32 CPUs	96GB	0 GPUs
long	24 hours	16 CPUs	48GB	0 GPUs
gpu-short	4 hours	16 CPUs	64GB	2 GPUs
gpu-long	12 hours	8 CPUs	80GB	1 GPU
bigmem	8 hours	64 CPUs	192GB	0 GPUs

# **Setup Instructions for Users:**

1) Create a job script ('my\_job.sh') that requests appropriate resources and loads necessary modules.

nano my conda job.sbatch

2) Make it executable:

chmod +x my\_conda\_job.sbatch

3) Submit the job:

sbatch my\_conda\_job.sbatch

- 4) Monitor the job progress:
- squeue -u \$USER

# View status of your jobs

sinfo -o "%20P %10a %10c %10m %10G %60"

• tail -f slurm-<jobid>.out

• tail -f slurm-<jobid>.err

cat slurm-<jobid>.out

cat slurm-<jobid>.err

# Show resource utilization (CPU, memory, GPU loads)

#Check the output file of the job in progress

#Check the error file of the job in progress

#Check the output file of the completed job

#Check the error file of the completed job

5) To know the completed job, to cancel the ongoing job, and to know about pending jobs:

seff <jobid> # To get a detailed efficiency report after the job completes

• scancel <jobid> # Cancel a pending or running job

squeue -t pending -o "%.8i %.12P %.20j %.8u %.2t %.10M %.6D %25R"

# See all pending jobs and their reasons

squeue -t pending --start # See estimated start times for pending jobs

- 6) Quick job submission examples:
- sbatch --partition=debug --time=00:30:00 --mem=16G script.sh
- sbatch --partition=short --time=04:00:00 --gres=gpu:1 script.sh
- sbatch --partition=long --time=24:00:00 --mem=48G script.sh
- sbatch --partition=bigmem --mem=180G script.sh

- 7) The user may leverage the below given Job Scripts to use the available modules:
- 1. Basic CPU Job with Miniconda (short partition)

```
#!/bin/bash
#SBATCH --job-name=conda job
#SBATCH --partition=short
#SBATCH --time=04:00:00
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=8
#SBATCH --mem=32G
#SBATCH --output=conda_%j.out
#SBATCH --error=conda_%j.err
# Initialize conda from NFS shared installation
source /nfs home/software/miniconda/etc/profile.d/conda.sh
# Activate your environment
conda activate my env
# Verify environment
echo "Conda environment: $CONDA_DEFAULT_ENV"
echo "Python path: $(which python)"
# Your code here
python my_script.py"=== Job Completed ==="
```

# 2. GPU Job with Miniconda (gpu-short partition)

```
#!/bin/bash
#SBATCH --job-name=conda gpu job
#SBATCH --partition=gpu-short
#SBATCH --gres=gpu:a100-80gb:1
#SBATCH --time=04:00:00
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=8
#SBATCH --mem=32G
#SBATCH --output=gpu conda %j.out
#SBATCH --error=gpu_conda_%j.err
echo "=== GPU Job Started ==="
echo "Job ID: $SLURM JOB ID"
echo "GPU allocated: $CUDA VISIBLE DEVICES"
# Load required modules
module load miniconda3/4.12.0
module load cuda/11.7
# Initialize conda
source /nfs_home/software/miniconda/etc/profile.d/conda.sh
# Activate environment with GPU support
conda activate pytorch gpu
# Verify GPU access
echo "Python: $(which python)"
nvidia-smi
python -c "import torch; print(f'CUDA available: {torch.cuda.is available()}'); print(f'GPU count:
{torch.cuda.device count()}')"
# GPU-accelerated script
python train model.py --batch-size 32 --epochs 10 --gpu-id 0
echo "=== GPU Job Completed ==="
```

# 3. Debug Job with Miniconda (debug partition)

```
#!/bin/bash
#SBATCH --job-name=conda debug
#SBATCH --partition=debug
#SBATCH --time=00:30:00
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=4
#SBATCH --mem=16G
#SBATCH --output=debug_%j.out
#SBATCH --error=debug_%j.err
echo "=== Debug Job ==="
echo "Quick test of conda environment"
# Source conda (adjust path as needed)
source /nfs_home/software/miniconda/etc/profile.d/conda.sh
# List available environments
echo "Available conda environments:"
conda env list
# Test basic environment
conda activate base
python -c "
import sys
print(f'Python {sys.version}')
import numpy as np
print(f'NumPy {np.__version___}')
print('Debug test completed successfully')
echo "=== Debug Completed ==="
```

4. Long-running CPU Job with Miniconda (long partition)

```
#!/bin/bash
#SBATCH --job-name=conda_long_job
#SBATCH --partition=long
#SBATCH --time=1-00:00:00
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=12
#SBATCH --mem=40G
#SBATCH --output=long_%j.out
#SBATCH --error=long_%j.err
echo "=== Long-running Job ==="
echo "Duration: 24 hours"
# Setup conda
source /nfs_home/software/miniconda/etc/profile.d/conda.sh
conda activate research env
# Long-running analysis
python extended_analysis.py \
 --threads $SLURM_CPUS_PER_TASK \
 --memory $(($SLURM_MEM_PER_NODE / 1024))G
echo "=== Long Job Completed ==="
```

# 5. Memory-Intensive Job with Miniconda (bigmem partition)

```
#!/bin/bash
#SBATCH --job-name=conda bigmem
#SBATCH --partition=bigmem
#SBATCH --time=08:00:00
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=32
#SBATCH --mem=180G
#SBATCH --output=bigmem_%j.out
#SBATCH --error=bigmem_%j.err
echo "=== Big Memory Job ==="
echo "Allocated memory: $SLURM MEM PER NODE MB"
# Setup conda for memory-intensive work
source /nfs home/software/miniconda/etc/profile.d/conda.sh
conda activate bigdata_env
# Memory-intensive processing
python process large data.py \
  --workers $SLURM_CPUS_PER_TASK \
  --memory $(($SLURM_MEM_PER_NODE / 1024))G
echo "=== Big Memory Job Completed ==="
```

#### 6. Multi-GPU Job with Miniconda (gpu-short partition)

```
#!/bin/bash
#SBATCH -- job-name = conda multi gpu
#SBATCH --partition=gpu-short
#SBATCH --gres=gpu:a100-80gb:2
#SBATCH --time=04:00:00
#SBATCH --ntasks=1
#SBATCH --cpus-per-task=16
#SBATCH --mem=64G
#SBATCH --output=multi gpu %j.out
#SBATCH --error=multi gpu %j.err
echo "=== Multi-GPU Job ==="
echo "GPUs allocated: $CUDA VISIBLE DEVICES"
# Setup conda with GPU support
source /nfs home/software/miniconda/etc/profile.d/conda.sh
conda activate multi gpu env
# Verify multi-GPU setup
python -c "
import torch
print(f'Available GPUs: {torch.cuda.device count()}')
for i in range(torch.cuda.device count()):
  print(f'GPU {i}: {torch.cuda.get device name(i)} -
{torch.cuda.get device properties(i).total memory/10243:.1f}GB')
# Multi-GPU training
python -m torch.distributed.launch --nproc_per_node=2 \
  train_multi_gpu.py --batch-size 64 --epochs 20
echo "=== Multi-GPU Job Completed ==="
```

#### **Key Features of These Included Templates:**

Users can copy the above templates and modify the required module environment name, Python/Bash script, and parameters as needed for their specific workloads.

- 1. Choose the right partition for their needs
- 2. Request appropriate resources (not too much, not too little)
- 3. **Stay within time limits** to avoid job termination
- 4. **Use GPUs efficiently** with proper memory/CPU ratios
- 5. **Leverage all available partitions** based on their account privileges
- 8) Getting Help and Support:
  - ✓ Users may email their queries or requests for additional module expansions to the support team.
  - ✓ In case of any technical error, users should include the corresponding job ID and the relevant section of the slurm-.out file in their email.
  - ✓ Support can be reached through the following channels:
- Email: <u>vincem@pilani.bits-pilani.ac.in</u>
   In-Person Assistance: CSIS Lab (Room 6017).