



# **GPU Computing on FASRC Cluster**

https://docs.rc.fas.harvard.edu/kb/gpgpu-computing-on-the-cluster

# **Cannon Cluster**

#### **Compute:**

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- 100,000 compute cores
- Cores/node: 8 to 64
- Memory/node: 12GB to 512GB (4GB/core)
- 2,500,000 NVIDIA GPU cores

#### Software:

- Operating System CentOS 7
- Slurm job manager
- 1,000+ scientific tools and programs
  - https://portal.rc.fas.harvard.edu/apps/modules

#### Interconnect:

- 2 underlying networks connecting 3 data centers
- TCP/IP network
- Low-latency 200 GB/s HDR InfiniBand (IB) and 56 GB/s FDR IB network:
  - inter-node parallel computing
  - fast access to Lustre mounted storage





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## What is GPGPU?

 General-Purpose Graphics Processing Unit (GPGPU) is a graphics processing unit (GPU) that is programmed for purposes beyond graphics processing, such as performing computations typically conducted by a Central Processing Unit (CPU).





### GPU vs CPU



CPU	GPU
Central Processing Unit	Graphics Processing Unit
Several cores	Many cores
Low latency	High throughput
Good for serial processing	Good for parallel processing
Can do a handful of operations at once	Can do thousands of operations at once

https://blogs.nvidia.com/blog/2009/12/16/whats-the-difference-between-a-cpu-and-a-gpu/





## **Heterogeneous Computing**

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#### **Application Code**





## Using GPGPUs

- GPU enabled applications requires a parallel computing platform and application programming interface (API) that allows software developers and software engineers to build algorithms to modify their application and map compute-intensive kernels to the GPU.
- GPGPU supports several types of memory in a memory hierarchy for designers to optimize their programs.
- GPGPU memory is used for transferring data between device and host -- shared memory is an efficient way for threads in the same block to share their runtime and data.



Ways to Accelerate your Applications

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# Drop-in Library: Cublas

CPU version

// define size

int N = 1 << 20;

// allocate cpu data

x = (float \*)malloc(N \* sizeof(float));

y = (float \*)malloc(N \* sizeof(float));

initData(x, y);

// Perform SAXPY on 1M elements: y[]=a\*x[]+y[]

saxpy(N, 2.0, x, 1, y, 1);

**GPU** Acceleration

int N = 1 << 20;

/ define size

// allocate GPU memory

cudaMalloc(&d\_x, N \* sizeof(float));

cudaMalloc(&d\_y, N \* sizeof(float));

initData(x, y);

// Copy working data from CPU->GPU

cublasSetVector(N, sizeof(x[0]), x, 1, d\_x, 1); cublasSetVector(N, sizeof(y[0]), y, 1, d\_y, 1);

// Perform SAXPY on 1M elements: y[]=a\*x[]+y[]

cublasSaxpy(N, 2.0, d\_x, 1, d\_y, 1);

// Bring the result back to the CPU

cublasGetVector(N, sizeof(y[0]), d\_y, 1, y, 1);





## OpenACC

- OpenACC (for Open Accelerators) is a programming standard for parallel computing on accelerators (mostly on NIVDIA GPU).
- It is designed to simplify GPU programming.
- The basic approach is to insert special comments (directives) into the code so as to offload computation onto GPUs and parallelize the code at the level of GPU (CUDA) cores.
- It is possible for programmers to create an efficient parallel OpenACC code with only minor modifications to a serial CPU code.







#### OpenACC COMPILER DIRECTIVES Parallel C Code Parallel Fortran Code

```
void saxpy(int n,
           float a,
           float *x,
           float *v)
#pragma acc kernels
  for (int i = 0; i < n; ++i)
   y[i] = a*x[i] + y[i];
// Perform SAXPY on 1M elements
saxpy(1<<20, 2.0, x, y);</pre>
```

```
subroutine saxpy(n, a, x, y)
  real :: x(:), y(:), a
  integer :: n, i
!$acc kernels
  do i=1,n
    y(i) = a*x(i)+y(i)
  enddo
!$acc end kernels
end subroutine saxpy
...
! Perform SAXPY on 1M elements
call saxpy(2**20, 2.0, x_d, y_d)
...
```

Slide from Jeff Larkin - Nvidia For more information: https://www.bu.edu/tech/files/2017/04/OpenACC-2017Spring.pdf



## Compute Unified Device Architecture (CUDA)

- CUDA platform is a software layer that gives direct access to the GPU's virtual instruction set and parallel computational elements for the execution of compute kernels.
- Designed to work with programming languages such as C, C++, and Fortran
- CUDA is an accessible platform, requiring no advanced skills in graphics programming, and available to software developers through CUDA-accelerated libraries and compiler directives.
- CUDA-capable devices are typically connected with a host CPU and the host CPUs are used for data transmission and kernel invocation for CUDA devices.
- The CUDA Toolkit includes GPU-accelerated libraries, a compiler, programming guides, API references, and the CUDA runtime.

https://docs.nvidia.com/cuda/cuda-compiler-driver-nvcc/index.html



# Using Programming Languages

CUDA C	
<pre>void saxpy(int n, float a, float *x, float *y) { for (int i = 0; i &lt; n; ++i) y[i] = a*x[i] + y[i]; }</pre>	<pre>global void saxpy(int n, float a, float *x, float *y) { int i = blockIdx.x*blockDim.x + threadIdx.x if (i &lt; n) y[i] = a*x[i] + y[i]; }</pre>
int N = 1<<20;	<pre>int N = 1&lt;&lt;20; cudaMemcpy(d_x, x, N, cudaMemcpyHostToDevice) cudaMemcpy(d_y, y, N, cudaMemcpyHostToDevice)</pre>
<pre>// Perform SAXPY on 1M elements saxpy(N, 2.0, x, y);</pre>	<pre>// Perform SAXPY on 1M elements saxpy&lt;&lt;&lt;4096,256&gt;&gt;&gt;(N, 2.0, d_x, d_y);</pre>
	<pre>cudaMemcpy(y, d y, N, cudaMemcpyDeviceToHost)</pre>

Slide from Jeff Larkin - Nvidia https://docs.nvidia.com/cuda/cuda-compiler-driver-nvcc/index.html

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- NVIDIA Tesla general purpose graphics processing units (GPGPU).
- 15 nodes with 4 V100 per node is available for general use from the gpu partition
- Your lab may have access to other partitions with GPU's or bought few nodes
  - "grep -ri "idreos\_parkes" /etc/slurm/slurm.conf"

sinfo -p idreos\_parkes PARTITION AVAIL TIMELIMIT NODES STATE NODELIST idreos\_parkes up 7-00:00:00 1 mix holygpu2c1125

- Several other nodes with 4 V100 are available in gpu\_requeue. These nodes are owned by various research groups available and may be available when idle.
- FAS members have access to the fas\_gpu partition which has 34 nodes with 2xK80s.
- SEAS members has access to few other partitions so visit https://docs.rc.fas.harvard.edu/kb/seas-compute-resources/





## **Examples and Questions?**

https://github.com/fasrc/User\_Codes





VDI - Open OnDemand

For applications that need a GUI: <u>https://vdi.rc.fas.harvard.edu</u>

Supports:

- Remote Desktop
- Jupyter Notebooks
- Rstudio
- Matlab

Notes:

- Need to be on the RC VPN to use
- Sessions are submitted as jobs on the cluster and thus use fairshare but also can run on any partition





# Request Help - Resources

- <u>https://docs.rc.fas.harvard.edu/kb/support/</u>
  - Documentation
    - https://docs.rc.fas.harvard.edu/
  - Portal
    - http://portal.rc.fas.harvard.edu/rcrt/submit\_ticket
  - Email
    - rchelp@rc.fas.harvard.edu
  - Office Hours
    - Wednesday noon-3pm https://harvard.zoom.us/j/255102481
  - Consulting Calendar
    - https://www.rc.fas.harvard.edu/consulting-calendar/
  - Training
    - https://www.rc.fas.harvard.edu/upcoming-training/

