ECE408/CS483/CSE408 Fall 2022

Applied Parallel Programming

Lecture 23: Alternatives to CUDA

Accelerated Computing is no longer a question









GPU vendors include:

Nvidia

AMD

Intel

Samsung

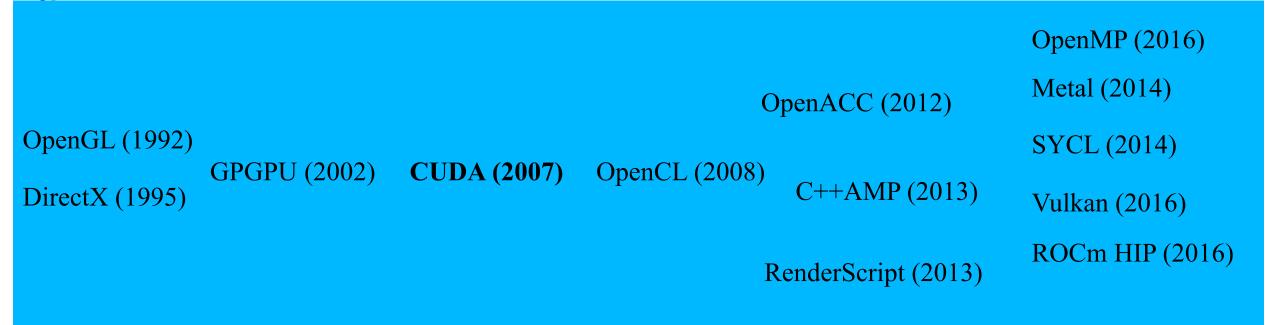
Apple

Qualcomm

ARM

Etc....

CUDA is just one model for Compute Acceleration



Existing frameworks such as MPI, TBB, OpenCV adapted to provide support. New frameworks such as Caffe, TensorFlow, R, PyCUDA natively support acceleration.

OpenCL, HIP, OpenACC, MPI

• OpenCL: An Open Standard Acceleration API

• Heterogeneous-Computing Interface for Portability (HIP)

• OpenACC: A "Low-Code" Acceleration API

• MPI: A Large Scale, Multi-Node Parallel API

Common Traits for Acceleration APIs

HARDWARE

- Hierarchy of lightweight cores
- Local scratchpad memories
- Lack of HW coherence
- Slow global atomics
- Threading

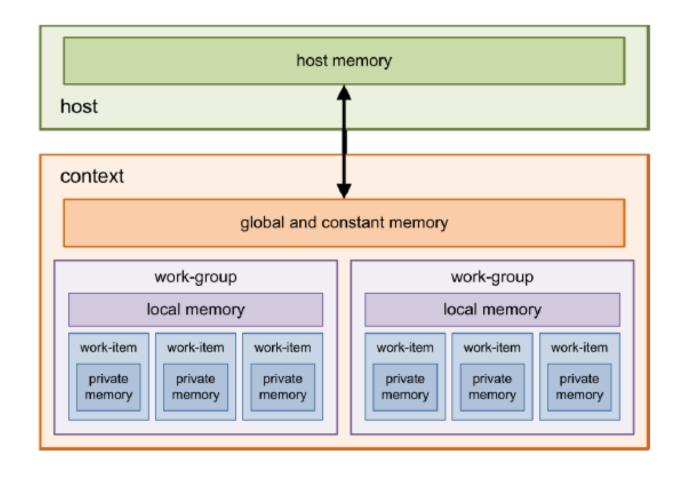
SOFTWARE

- Kernel oriented acceleration
- Device memory vs. Host memory
- Software managed memory
- Grids, Blocks, Threads
- Bulk Synchronous Parallelism

OpenCL

- Framework for CPUs, GPUs, DSPs, FPGAs, etc. (not just Nvidia GPUs)
- Initially developed by Apple with support from AMD, IBM, Qualcomm, Intel, and Nvidia. OpenCL 1.0 launched in 2008.
- OpenCL 2.2 launched in May 2017
- Apple announces dropping of OpenCL in 2018

OpenCL Memory Model



OpenCL MatMult

- Notice similarity to CUDA
- WorkGroup similar to Block
- WorkItem similar to Thread
- __local similar to shared

```
1.// Tiled and coalesced version
2. kernel void myGEMM2(int M, int N, int K, global float* A, global float* B, global float*
     // Thread identifiers
     const int row = get local id(0); // Local row ID (max: TS)
     const int col = get local id(1); // Local col ID (max: TS)
     const int globalRow = TS*get group id(0) + row; // Row ID of C (0..M)
     const int globalCol = TS*get group id(1) + col; // Col ID of C (0..N)
9.
10.
      // Local memory to fit a tile of TS*TS elements of A and B
       local float Asub[TS][TS];
11.
12.
      local float Bsub[TS][TS];
13.
      // Initialise the accumulation register
14.
      float acc = 0.0f;
15.
     // Loop over all tiles
16.
17.
      const int numTiles = K/TS;
18.
      for (int t=0; t<numTiles; t++) {</pre>
19.
20.
         // Load one tile of A and B into local memory
21.
         const int tiledRow = TS*t + row;
22.
         const int tiledCol = TS*t + col;
23.
         Asub[col][row] = A[tiledCol*M + globalRow];
24.
         Bsub[col][row] = B[globalCol*K + tiledRow];
25.
26.
         // Synchronise to make sure the tile is loaded
27.
         barrier(CLK LOCAL MEM FENCE);
28.
29.
         // Perform the computation for a single tile
30.
         for (int k=0; k<TS; k++)</pre>
            acc += Asub[k][row] * Bsub[col][k];
31.
32.
33.
         // Synchronise before loading the next tile
34.
         barrier(CLK LOCAL MEM FENCE);
35.
36.
      // Store the final result in C
38.
      C[globalCol*M + globalRow] = acc;
39.}
```

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HIP

- Heterogeneous-Computing Interface for Portability (HIP)
 - C++ dialect designed to ease conversion of CUDA applications to portable C++ code.
 - Provides a C-style API and a C++ kernel language.
 - The C++ interface can use templates and classes across the host/kernel boundary.
- HIP code can run on AMD hardware (through the HCC compiler) or NVIDIA hardware (through the NVCC compiler).
- The HIPify tool automates much of the conversion work by performing a source-to-source transformation from CUDA to HIP.

vectorAdd with HIP

```
global void vecAdd(double *a, double *b, double *c, int n) {
  int id = blockIdx.x*blockDim.x+threadIdx.x;
  if (id < n) c[id] = a[id] + b[id];
  hipMalloc(&d a, nbytes);
  hipMalloc(&d b, nbytes);
  hipMalloc(&d c, nbytes);
  hipMemcpy(d a, h a, bytes, hipMemcpyHostToDevice);
  hipMemcpy(d b, h b, bytes, hipMemcpyHostToDevice);
  blockSize = 1024;
  gridSize = (int)ceil((float)n/blockSize);
  hipLaunchKernelGGL (vecAdd, dim3 (gridSize), dim3 (blockSize), 0, 0, d a, d b, d c, n);
  hipDeviceSynchronize();
  hipMemcpy(h c, d c, bytes, hipMemcpyDeviceToHost);
```

OpenACC

The OpenACC Application Programming Interface (API) provides a set of

- compiler directives (pragmas),
- library routines, and
- environment variables

that enable

- FORTRAN, C and C++ programs
- to execute on accelerator devices
- including GPUs and CPUs.

Pragmas Provide Extra Information

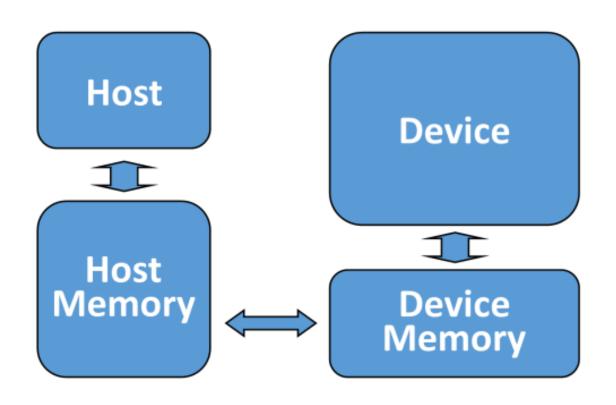
In C and C++,

- the #pragma directive
- provides the compiler with
- information not specified in the language.

For OpenACC, they look like this:

#pragma acc [the information goes here]

The OpenACC Abstract Machine Model



The OpenACC Directives

```
Manage
              #pragma acc data copyin(x,y) copyout(z)
Data
Movement
                 #pragma acc parallel
Initiate
                 #pragma acc loop gang vector
Parallel
                     for (i = 0; i < n; ++i) {
Execution
                         z[i] = x[i] + y[i];
Optimize
Loop
Mappings
```

Simple Matrix-Matrix Multiplication in OpenACC

```
void computeAcc(float *P, const float *M, const float *N, int Mh, int Mw, int Nw)
2
    #pragma acc parallel loop copyin(M[0:Mh*Mw]) copyin(N[0:Nw*Mw]) copyout(P[0:Mh*Nw])
5
    for (int i=0; i<Mh; i++) {
6
        #pragma acc loop
        for (int j=0; j<Nw; j++) {
7
            float sum = 0;
            for (int k=0; k<Mw; k++) {
10
                float a = M[i*Mw+k];
11
                float b = N[k*Nw+j];
12
                sum += a*b;
13
14
            P[i*Nw+j] = sum;
15
16
17 }
```

Add Pragmas to Sequential Code

The code is

- identical to the sequential version
- except for the two pragmas
- at lines 2 and 4.

OpenACC uses the compiler directive mechanism to extend the base language.

Simple Matrix-Matrix Multiplication in OpenACC

```
void computeAcc(float *P, const float *M, const float *N, int Mh, int Mw, int Nw) {
  #pragma acc parallel loop copyin(M[0:Mh*Mw]) copyin(N[0:Nw*Mw]) copyout(P[0:Mh*Nw])
  for (int i=0; i<Mh; i++) {
     #pragma acc loop
     for (int j=0; j<Nw; j++) {
       float sum = 0;
6
       for (int k=0; k<Mw; k++) {
         float a = M[i*Mw+k];
9
         float b = N[k*Nw+j];
10
          sum += a*b;
11
        P[i*Nw+j] = sum;
12
13
14
15 }
```

tells compiler

- to execute 'i' loop
- (lines 3 through 14)
- in parallel on accelerator.

copyin/copyout specify

- how matrix data
- should be transferred between memories.

Simple Matrix-Matrix Multiplication in OpenACC

```
void computeAcc(float *P, const float *M, const float *N, int Mh, int Mw, int Nw) {
  #pragma acc parallel loop copyin(M[0:Mh*Mw]) copyin(N[0:Nw*Mw]) copyout(P[0:Mh*Nw])
  for (int i=0; i<Mh; i++) {
    #pragma acc loop
    for (int j=0; j<Nw; j++) {
       float sum = 0;
6
       for (int k=0; k<Mw; k++) {
         float a = M[i*Mw+k];
                                   tells compiler
         float b = N[k*Nw+j];
                                      to map 'j' loop
10
          sum += a*b;
                                     (lines 5 through 13)
11
                                     to second level
12
       P[i*Nw+j] = sum;
13
                                      of parallelism on accelerator.
14
15 }
```

Motivating Goal: One Version of Code

OpenACC programmers

- can often start with a sequential version,
- then annotate their program with directives,
- leaving most kernel details and data transfers
- to the OpenACC compiler.

OpenACC code can be compiled by non-OpenACC compilers by ignoring the pragmas.

Reality is More Complicated

Reality check:

- can be difficult to write code
- that works correctly and well
- with and without pragmas.

Some OpenACC programs

- behave differently or even incorrectly
- if pragmas are ignored.

Pitfall: Strong Dependence on Compiler

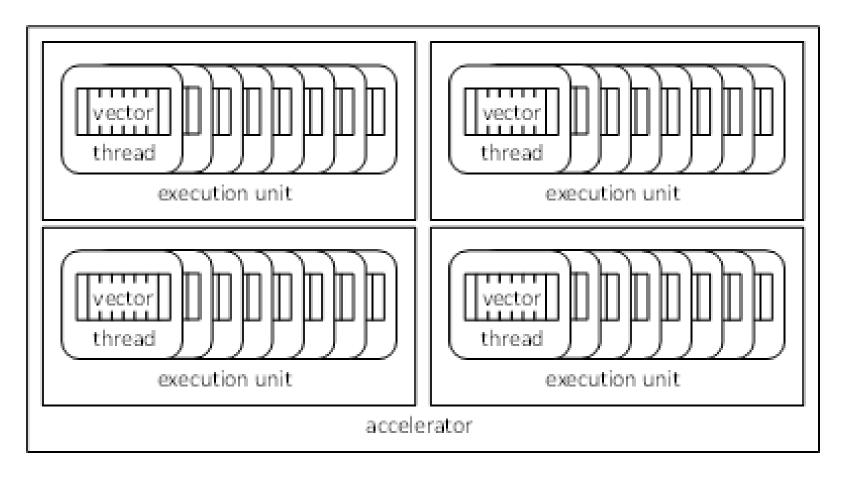
Some OpenACC pragmas

- are hints to the OpenACC compiler,
- which may or may not be able to act accordingly

Performance depends heavily

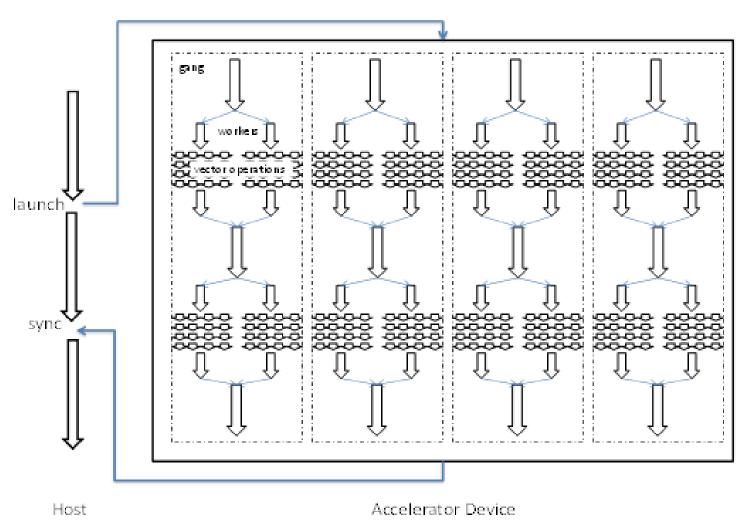
- on the quality of the compiler
- (more so than with CUDA or OpenCL).

OpenACC Device Model



Currently OpenACC does not allow user-specified synchronization across threads.

OpenACC Execution Model (Terminology: Gangs and Works)



Parallel vs. Loop Constructs

```
#pragma acc parallel loop copyin(M[0:Mh*Mw]) copyin(N[0:Nw*Mw]) copyout(P[0:Mh*Nw])
for (int i=0; i<Mh; i++) {
                                     is equivalent to:
#pragma acc parallel copyin(M[0:Mh*Mw]) copyin(N[0:Nw*Mw]) copyout(P[0:Mh*Nw])
   #pragma acc loop
   for (int i=0; i<Mh; i++) {
```

(a parallel region that consists of just a loop)

Parallel Construct

- A parallel construct is executed on an accelerator
- One can specify the number of gangs and number of works in each gang
- Programmer's directive

```
#pragma acc parallel copyout(a) num_gangs(1024) num_workers(32)
{
    a = 23;
}
```

1024*32 workers will be created. a=23 will be executed redundantly by all 1024 gang leads

What does each "Gang Loop" do?

execution

```
#pragma acc parallel num gangs(1024)
                                           #pragma acc parallel num gangs(1024)
  for (int i=0; i<2048; i++) {
                                           #pragma acc loop gang
                                              for (int i=0; i<2048; i++) {
                                                  The 2048 iterations of the
    The for-loop will be
    redundantly executed by
                                                  for-loop will be divided
                                                  among 1024 gangs for
    1024 gangs
```

Worker Loop

```
#pragma acc parallel num_gangs(1024) num_workers(32)
  #pragma acc loop gang
  for (int i=0; i<2048; i++) {
     #pragma acc loop worker
     for (int j=0; j<512; j++) {
        foo(i,j);
```

1024*32=32K workers will be created, each executing 1M/32K = 32 instance of foo()

A More Complex Example

```
#pragma acc parallel num_gangs(32)
   Statement 1; Statement 2;
   #pragma acc loop gang
   for (int i=0; i< n; i++) {
     Statement 3; Statement 4;
   Statement 5; Statement 6;
   #pragma acc loop gang
   for (int i=0; i < m; i++) {
     Statement 7; Statement 8;
   Statement 9;
   if (condition)
     Statement 10;
```

- Statements 1 and 2 are redundantly executed by 32 gangs
- The n for-loop iterations are distributed to 32 gangs

Kernel Regions

```
#pragma acc kernels
   #pragma acc loop num_gangs(1024)
   for (int i=0; i<2048; i++) {
      a[i] = b[i];
   #pragma acc loop num_gangs(512)
   for (int j=0; j<2048; j++) {
      c[j] = a[j] *2;
   for (int k=0; k<2048; k++) {
      d[k] = c[k];
```

 Kernel constructs are descriptive of programmer intentions (suggestions)

Reduction

```
#pragma acc parallel loop
reduction(+:sum)
for(int i=0;i<n;i++) {
    sum +=
    xcoefs[i]*ycoefs[i];
}</pre>
```

- Because each iteration of the loop adds to the variable sum, we must declare a reduction.
- A parallel reduction may return a slightly different result than a sequential addition due to floating point limitations.

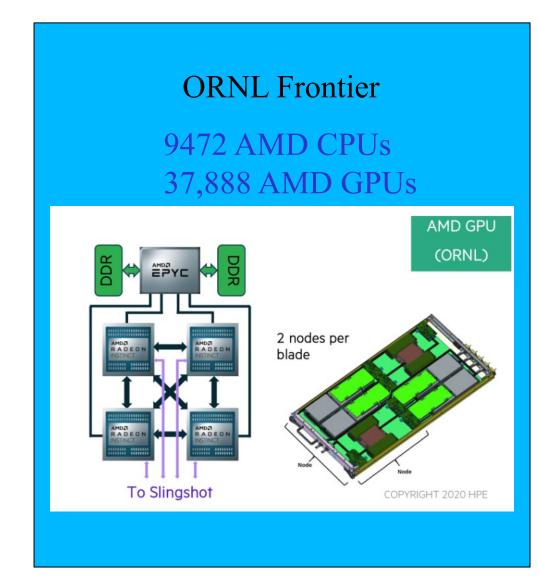
C/C++ vs. FORTRAN

```
// C or C++
#pragma acc <directive> <clauses>
{ ... }

! Fortran
!$acc <directive> <clauses>
...
!$acc end <directive>
```

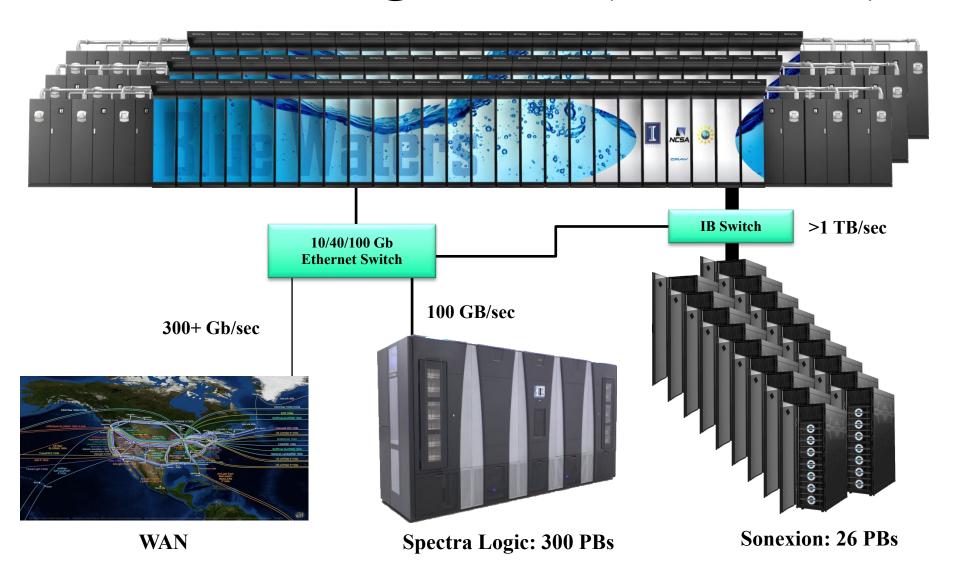
Top 5 Supercomputers (Fall 2022)

Rank	System	Cores	Rmax (PFlop/s)	Rpeak (PFlop/s)	Power (kW)
1.	Frontier - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE DOE/SC/Oak Ridge National Laboratory United States	8,730,112	1,102.00	1,685.65	21,100
2	Supercomputer Fugaku - Supercomputer Fugaku, A64FX 48C 2.2GHz, Tofu interconnect D, Fujitsu RIKEN Center for Computational Science Japan	7,630,848	442.01	537.21	29,899
3	LUMI - HPE Cray EX235a, AMD Optimized 3rd Generation EPYC 64C 2GHz, AMD Instinct MI250X, Slingshot-11, HPE EuroHPC/CSC Finland	2,220,288	309.10	428.70	6,016
4	Leonardo - BullSequana XH2000, Xeon Platinum 8358 32C 2.6GHz, NVIDIA A100 SXM4 40 GB, Quad-rail NVIDIA HDR100 Infiniband, Atos EuroHPC/CINECA Italy	1,463,616	174.70	255.75	5,610
5	Summit - IBM Power System AC922, IBM POWER9 22C 3.07GHz, NVIDIA Volta GV100, Dual-rail Mellanox EDR Infiniband, IBM DOE/SC/Oak Ridge National Laboratory United States	2,414,592	148.60	200.79	10,096



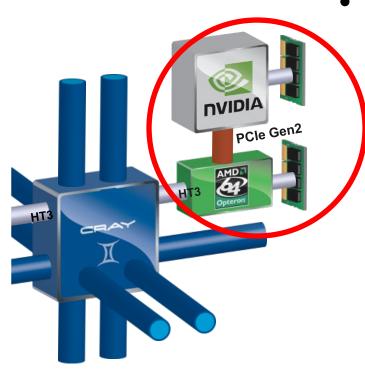
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Blue Waters @ UIUC (2013-2021)



Cray XK7 Nodes



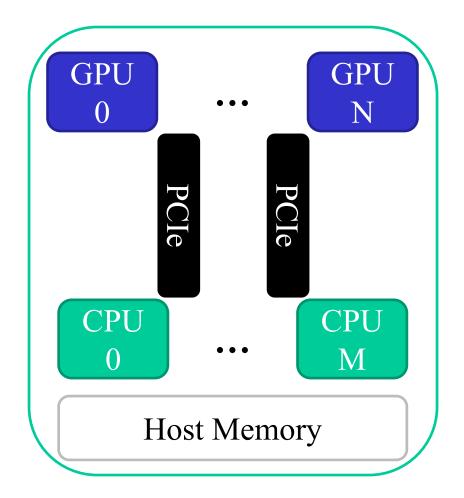


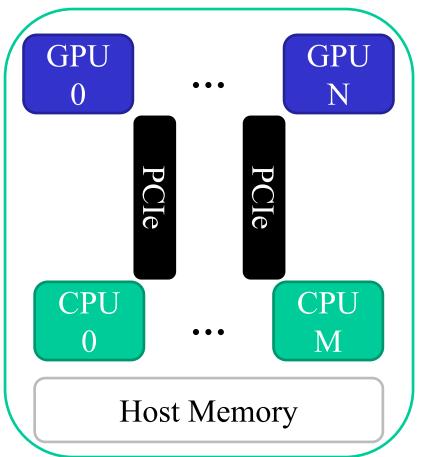
Blue Waters contains 4,224 Cray XK7 compute nodes.

- Dual-socket Node
 - One AMD Interlagos chip
 - 8 core modules, 32 threads
 - 156.5 GFs peak performance
 - 32 GBs memory
 - 51 GB/s bandwidth
 - One NVIDIA Kepler chip
 - 1.3 TFs peak performance
 - 6 GBs GDDR5 memory
 - 250 GB/sec bandwidth
 - Gemini Interconnect
 - Same as XE6 nodes

Abstract CUDA-based Node

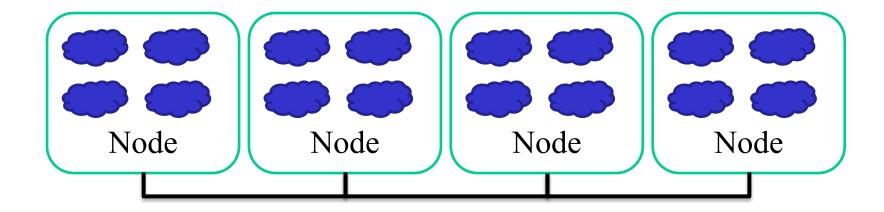
• Each node contains N GPUs





MPI Model

Many processes distributed in a cluster



- Each process computes part of the output
- Processes communicate with each other through message passing (not global memory)
- Processes can synchronize through messages

MPI Initialization, Info

- User launches an MPI job with X processes by executing in the command shell
 MPIrun -np X
- int MPI_Init(int *argc, char ***argv)
 - Initialize MPI
- MPI_COMM_WORLD
 - MPI group formed with all allocated nodes
- int MPI_Comm_rank(MPI_Comm comm, int *rank)
 - Rank of the calling process in group of comm
- int MPI_Comm_size(MPI_Comm comm, int *size)
 - Number of processes in the group of comm

Vector Addition: Main Process

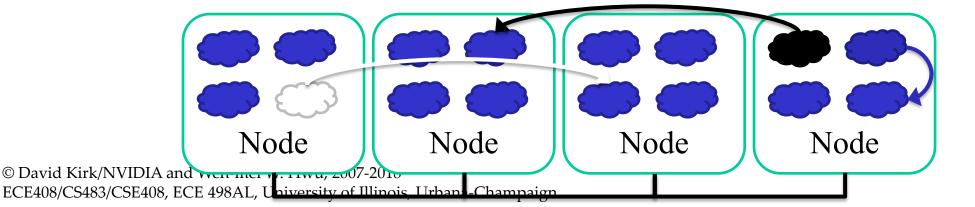
```
int main(int argc, char *argv[]) {
    int vector size = 1024 * 1024 * 1024;
    int pid=-1, np=-1;
   MPI Init(&argc, &argv);
    MPI_Comm_rank(MPI_COMM_WORLD, &pid);
    MPI Comm size(MPI COMM WORLD, &np);
    if(np < 3) {
        if(0 == pid) printf("Nedded 3 or more processes.\n");
        MPI Abort( MPI COMM WORLD, 1 ); return 1;
    if(pid < np - 1)
        compute node(vector size / (np - 1));
    else
        data_server(vector_size);
    MPI_Finalize();
    return 0;
```

MPI Sending Data

- int MPI_Send(void *buf, int count, MPI_Datatype datatype, int dest, int tag, MPI_Comm comm)
 - buf: Starting address of send buffer
 - count: Number of elements in send buffer (nonnegative integer)
 - datatype: Datatype of each send buffer element
 - dest: Rank of destination (integer)
 - tag: Message tag (integer)
 - comm: Communicator (handle)

MPI Sending Data

- int MPI_Send(void *buf, int count, MPI_Datatype datatype, int dest, int tag, MPI_Comm comm)
 - Buf: Initial address of send buffer
 - Count: Number of elements in send buffer (nonnegative integer)
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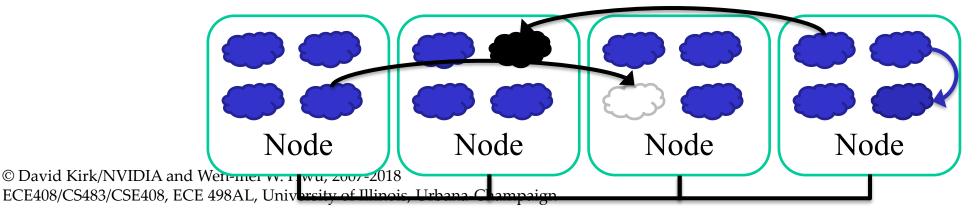


MPI Receiving Data

- int MPI_Recv(void *buf, int count, MPI_Datatype datatype, int source, int tag, MPI_Comm comm, MPI_Status *status)
 - Buf: Starting address of receive buffer
 - Count: Maximum number of elements in receive buffer (non-negative integer)
 - Datatype: Datatype of each receive buffer element
 - Source: Rank of source (integer)
 - Tag: Message tag (integer)
 - Comm: Communicator (handle)
 - Status: Status object

MPI Receiving Data

- int MPI_Recv(void *buf, int count, MPI_Datatype datatype, int source, int tag, MPI_Comm comm, MPI_Status *status)
 - Buf: Initial address of receive buffer
 - Count: Maximum number of elements in receive buffer (non-negative integer)
 - Datatype: Datatype of each receive buffer element
 - Source: Rank of source (integer)
 - Tag: Message tag (integer)
 - Comm: Communicator (handle)
 - Status: Status object (Status)



Vector Addition: Server Process (I)

```
void data_server(unsigned int vector_size) {
    int np, num nodes = np - 1, first node = 0, last node = np - 2;
    unsigned int num_bytes = vector_size * sizeof(float);
    float *input a = 0, *input b = 0, *output = 0;
   /* Set MPI Communication Size */
   MPI Comm size(MPI COMM WORLD, &np);
   /* Allocate input data */
    input a = (float *)malloc(num bytes);
    input b = (float *)malloc(num bytes);
    output = (float *)malloc(num_bytes);
    if(input_a == NULL || input_b == NULL || output == NULL) {
        printf("Server couldn't allocate memory\n");
        MPI Abort( MPI COMM WORLD, 1 );
    /* Initialize input data */
    random_data(input_a, vector_size , 1, 10);
    random data(input b, vector size ,
```

Vector Addition: Server Process (II)

```
/* Send data to compute nodes */
float *ptr a = input a;
float *ptr b = input b;
for(int process = 1; process < last node; process++) {</pre>
    MPI_Send(ptr_a, vector_size / num_nodes, MPI_FLOAT,
            process, DATA DISTRIBUTE, MPI COMM WORLD);
    ptr a += vector size / num nodes;
    MPI_Send(ptr_b, vector_size / num_nodes, MPI_FLOAT,
            process, DATA_DISTRIBUTE, MPI_COMM_WORLD);
    ptr b += vector size / num nodes;
```

Vector Addition: Server Process (III)

```
/* Wait for compute to complete*/
MPI Barrier(MPI COMM WORLD);
/* Collect output data */
MPI Status status;
for(int process = 0; process < num nodes; process++) {</pre>
    MPI Recv(output + process * num points / num nodes,
        num_points / num_comp_nodes, MPI_REAL, process,
        DATA COLLECT, MPI COMM WORLD, &status );
/* Store output data */
store output(output, dimx, dimy, dimz);
/* Release resources */
free(input);
free(output);
```

Vector Addition: Compute Process (I)

```
void compute_node(unsigned int vector_size ) {
    int np;
    unsigned int num_bytes = vector_size * sizeof(float);
    float *input a, *input b, *output;
   MPI Status status;
   MPI Comm size(MPI COMM WORLD, &np);
    int server process = np - 1;
    /* Alloc host memory */
    input a = (float *)malloc(num bytes);
    input b = (float *)malloc(num bytes);
    output = (float *)malloc(num bytes);
    /* Get the input data from server process */
    MPI_Recv(input_a, vector_size, MPI_FLOAT, server_process,
            DATA DISTRIBUTE, MPI COMM WORLD, &status);
   MPI_Recv(input_b, vector_size, MPI_FLOAT, server_process,
            DATA DISTRIBUTE, MPI COMM WORLD, &status);
```

Vector Addition: Compute Process (II)

```
/* Compute the partial vector addition */
for(int i = 0; i < vector_size; ++i) {</pre>
    output[i] = input a[i] + input b[i];
/* Or, can offload to GPU here */
/* cudaMalloc(), cudaMemcpy(), kernel launch, etc. */
MPI Barrier(MPI COMM WORLD);
/* Send the output */
MPI_Send(output, vector_size, MPI_FLOAT,
        server process, DATA COLLECT, MPI COMM WORLD);
/* Release memory */
free(input_a);
free(input_b);
free(output);
```

ANY MORE QUESTIONS? READ CHAPTER 15

Also see https://developer.nvidia.com/intro-to-openacc-course-2016