International High-Performance Computing Summer School 2023

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**EPCC** 

July 17, 2023

ерсс

Acknowledgements

### Individuals

The material presented in this lecture is inspired from content developed by:

- John Urbanic
- Michael Klemm

Acknowledgements

### Contributors

#### In this material:

- The slides are created using LATEX Beamer available at https://ctan.org/pkg/beamer.
- The sequence diagrams are created using an extended version of the pgf-umlsd package available at https://ctan.org/pkg/pgf-umlsd.

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Preamble

# Your opinion matters



Figure: Link to surveys

Preamble

### **Timetable**

First session: Monday

■ Topic: CPUs

From 14:00 until 15:30

30-min break

From 16:00 until 17:30

### Timetable

First session: Monday

- Topic: CPUs
- From 14:00 until 15:30
- 30-min break
- From 16:00 until 17:30

Second session: Tuesday

- Topic: GPUs
- From 9:00 until 10:30
- 30-min break
- From 11:00 until 12:30

```
Introduction to OpenMP - GPUs
Preamble
```

#### Moment of truth

#### Before we start:

- Connect to bridges
  - ssh your\_username@bridges2.psc.edu
- Clone the repository used in this session

```
git clone https://github.com/capellil/
IHPCSS_Introduction_to_OpenMP_GPU_examples
```

Preamble

### Moment of truth

- You will see it contains multiple folders, numbered.
- Each of which will be an illustration to a concept we will see.

```
Introduction to OpenMP - GPUs
```

Preamble

#### Moment of truth

Load the modules needed for the GPU ecosystem.

```
module load nvhpc/22.9;
module load openmpi/4.0.5-nvhpc22.9;
```

Put these two lines in your .bashrc file.

```
Introduction to OpenMP - GPUs
```

Preamble

#### Moment of truth

- Go to the repository, inside the first folder.
  - 1 cd repository/<language>/1.Preamble
- Compile the source code in it
- 1 make
- Run the executable on the compute node
- 1 sbatch submit.slurm
- To see whether your code is executing, has finished, etc...
  - 1 watch squeue -u \$USER

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└ Motivation

### Table of Contents

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- Motivation

 $\sqsubseteq$  Motivation

#### What are GPUs?

Stands for **G**raphics **P**rocessing **U**nits.

 $\sqsubseteq$  Motivation

- Stands for **G**raphics **P**rocessing **U**nits.
- Devices specialised in... graphics processing.

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- Typically has a given operation to apply to a lot of inputs
  - points in a mesh
  - pixels in a picture

- Stands for Graphics Processing Units.
- Devices specialised in... graphics processing.
- Typically has a given operation to apply to a lot of inputs
  - points in a mesh
  - pixels in a picture
- The GPU architecture allows them to complete this task very effectively.

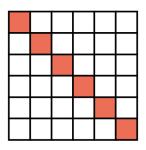
└ Motivation

#### Use case

- Imagine you are developing a graphic editing software.
- You are currently implementing a filter allowing to darken a picture by making every pixel 20% darker.
- The same calculation is applied to every pixel, regardless of its current colour.

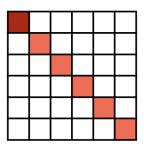
└ Motivation

### How CPUs would do it



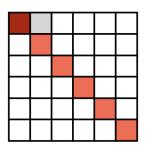
└ Motivation

### How CPUs would do it.



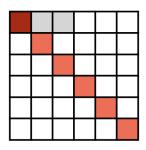
└ Motivation

### How CPUs would do it

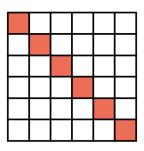


└ Motivation

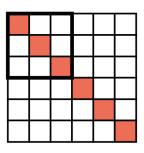
### How CPUs would do it



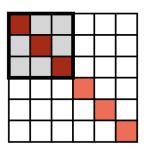
└ Motivation



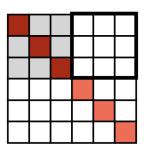
└ Motivation



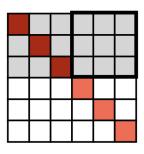
└ Motivation



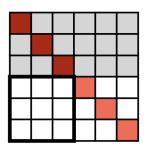
└ Motivation



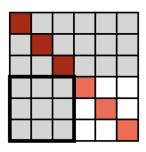
└ Motivation



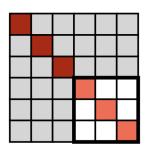
└ Motivation



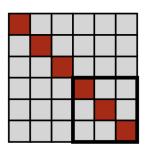
└ Motivation



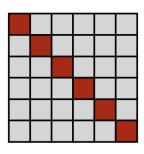
└ Motivation



└─ Motivation



└ Motivation



 $\sqsubseteq$  Motivation

### How GPUs work.

Demonstration by Mythbusters and NVIDIA at https://www.youtube.com/watch?v=-P28LKWTzrl.

 $^{ldsymbol{\sqcup}}$  Motivation

## **GPU** specs

On Bridges2, the GPU nodes have NVIDIA V100-32GB SXM2.

└ Motivation

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  - Core clock: 1.3GHz, boost at 1.5GHz.
  - 5,120 cores.
  - Bandwidth: 900GB/s
  - Max Thermal Design Power: 250W.
- Each GPU node on Bridges2 has 8 such GPUs...
  - Total of over 40K+ GPU cores per node
  - On top of the two Intel Xeon Gold 6248 "Cascade Lake" CPUs per node

— Motivation

 $^{ldsymbol{\sqcup}}$  Motivation

# Advantages of GPUs

Computation tanks: 15.7 TFLOPS per V100GPU, over 120 TFLOPS per GPU node.

 $\sqsubseteq$  Motivation

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## Advantages of GPUs

- Computation tanks: 15.7 TFLOPS per V100GPU, over 120 TFLOPS per GPU node.
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  - $\blacksquare$  £4,000 for 1 GPU that has 5,000 cores, so  $\sim\!\!£1$  per core.

Why even bothering with CPUs then?

<sup>&</sup>lt;sup>2</sup>though alleviated now with unified memory, which in turns raises new challenges as now the CPUs and the GPUs can have race conditions between them ◆□▶ ◆圖▶ ◆臺▶ ◆臺▶

L Motivation

#### Limitations

■ Memory transfers<sup>2</sup>

<sup>&</sup>lt;sup>2</sup>though alleviated now with unified memory, which in turns raises new challenges as now the CPUs and the GPUs can have race conditions between them

└ Motivation

- Memory transfers<sup>2</sup>
- Do not handle divergence well (needs to get cores idle)

<sup>&</sup>lt;sup>2</sup>though alleviated now with unified memory, which in turns raises new challenges as now the CPUs and the GPUs can have race conditions between them

└ Motivation

- Memory transfers<sup>2</sup>
- Do not handle divergence well (needs to get cores idle)
- Simpler Instruction Set Architecture (ISA): operations issued do not support as advanced calculations

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└ Motivation

- Memory transfers<sup>2</sup>
- Do not handle divergence well (needs to get cores idle)
- Simpler Instruction Set Architecture (ISA): operations issued do not support as advanced calculations
- Non-coalesced memory accesses

 $<sup>^2</sup>$ though alleviated now with unified memory, which in turns raises new challenges as now the CPUs and the GPUs can have race conditions between them

 $\sqsubseteq$  Motivation

# Opportunity to seize

Despite having limitations, GPUs still offer a highly attractive solution to certain types of workloads, especially in scientific simulations.

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- Decided to harness that graphical pipeline for more general computing tasks.

# Opportunity to seize

- Despite having limitations, GPUs still offer a highly attractive solution to certain types of workloads, especially in scientific simulations.
- Decided to harness that graphical pipeline for more general computing tasks.
- Birth of GPGPU: General Purpose Graphical Processing Unit.

— Motivation

### Solutions

CUDA First on it, free, amazing documentation, but only available for NVIDIA GPUs. Imposed their terms: CUDA cores, CUDA threads, warps etc...

### Solutions

- CUDA First on it, free, amazing documentation, but only available for NVIDIA GPUs. Imposed their terms: CUDA cores, CUDA threads, warps etc...
- Open Accelerators. Directive-based solution for GPGPU. Limited support in compilers.

└ Motivation

### Solutions

- CUDA First on it, free, amazing documentation, but only available for NVIDIA GPUs. Imposed their terms: CUDA cores, CUDA threads, warps etc...
- OpenACC **Open Acc**elerators. Directive-based solution for GPGPU. Limited support in compilers.
- OpenMP NVIDIA support on their GPUs noticeably improved recently with OpenMP 5.0, this is why we will use it, now that you already know how to use it for CPUs.

Motivation

#### Ideas

- declare target construct: associated variables and function are to be mapped onto the target devices (thus usable in device code).
- passing the device\_type(nohost) clause to a declare target construct forces the compiler to not produce host code for the associated structured block (pertaining to both variables and functions included)

Offloading

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Offloading

### Host vs device<sup>3</sup>

Host The device on which the OpenMP program begins execution.

<sup>&</sup>lt;sup>3</sup>See Section 1.2.1 in OpenMP standard version 5.2

└ Offloading

### Host vs device<sup>3</sup>

Host The device on which the OpenMP program begins execution.

Target A device onto which code and data may be offloaded from the host device.



Offloading

### target construct

■ Transfers the control flow to the target device

Offloading

- Transfers the control flow to the target device
- This transfer is:

└ Offloading

- **Transfers** the control flow to the target device
- This transfer is:
  - sequential

└ Offloading

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- Transfers the control flow to the target device
- This transfer is:
  - sequential
  - synchronous
- In theory, this can be combined with any OpenMP construct.
- In practice, there is only a useful subset of OpenMP features for a target device such as a GPU, e.g., no I/O, limited use of base language features.

```
Introduction to OpenMP - GPUs
```

# Offloading to a device - Example

```
1 // Block A: executed on host
2 // Block B: executed on host
3 // Block C: executed on host
```

```
! Block A: executed on host
! Block B: executed on host
! Block C: executed on host
```

```
Introduction to OpenMP - GPUs
Offloading
```

### Offloading to a device - Example

```
// Block A: executed on host
pragma omp target

{
    // Block B: executed on target
}

// Block C: executed on host
```

```
! Block A: executed on host
!$OMP TARGET
! Block B: executed on target
!$OMP END TARGET
! Block C: executed on host
```

Offloading

# Currently on the host or device?

- When you will start to have functions, and nested functions, you may not know at some point whether you are on the host or on a device.
- You can check by calling the function omp\_is\_initial\_device().

```
Introduction to OpenMP - GPUs
Offloading
```

## Currently on the host or device?

```
if(omp_is_initial_device())

printf("I am on the host\n");

pelse

printf("I am on the device.\n");

printf("I am on the device.\n");
```

```
IF (omp_is_initial_device())
WRITE(*, '(A)') 'I am on the host.'

ELSE
WRITE(*, '(A)') 'I am on the device.'

END IF
```

Offloading

### Time to practise: 2.HelloWorld

Update the source code provided such that the for loop marked is executed on the GPU.

#### Tips

You will need:

the target construct

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```
Introduction to OpenMP - GPUs
```

#### └─Data mapping

## Implicit mapping - CPUs

```
int a = 123;
int b[2] = {456, 789};

#pragma omp parallel

{
    a *= 2;
    b[0]++;
    b[1]++;
}
```

```
INTEGER :: a = 123
INTEGER, DIMENSION(0:1) :: b := (/456, 789/)

!$OMP PARALLEL

a *= 2
b(0) = b(0) + 1
b(1) = b(1) + 1

!$OMP END PARALLEL
```

```
Introduction to OpenMP - GPUs

Data mapping
```

## Implicit mapping - CPUs

```
int a = 123;
int b[2] = {456, 789};

#pragma omp parallel default(none) shared(a, b)

4 {
    a *= 2;
    b[0]++;
    b[1]++;
    b[1]++;
    }
```

```
INTEGER :: a = 123
INTEGER, DIMENSION(0:1) :: b = (/456, 789/)

!$OMP PARALLEL DEFAULT(NONE) SHARED(a, b)

a *= 2;

b(0) = b(0) + 1

b(1) = b(1) + 1

!$OMP END PARALLEL
```

```
Introduction to OpenMP - GPUs
```

#### └ Data mapping

# Implicit mapping - GPUs

```
int a = 123;
int b[2] = {456, 789};

#pragma omp target

{
    a *= 2;
    b[0]++;
    b[1]++;
}
```

```
INTEGER :: a = 123
INTEGER, DIMENSION(0:1) :: b := (/456, 789/)

!$OMP TARGET

a *= 2
b(0) = b(0) + 1
b(1) = b(1) + 1

!$OMP END TARGET
```

```
Introduction to OpenMP - GPUs

Data mapping
```

# Implicit mapping - GPUs

```
int a = 123;
int b[2] = {456, 789};

#pragma omp target firstprivate(a) shared(b)

{
    a *= 2;
    b[0]++;
    b[1]++;
}
```

```
INTEGER :: a = 123
INTEGER, DIMENSION(0:1) :: b := (/456, 789/)

!$OMP TARGET FIRSTPRIVATE(a) SHARED(b)

a *= 2
b(0) = b(0) + 1
b(1) = b(1) + 1

!$OMP END TARGET
```

LData mapping

# Implicit mapping

#### By default:

- scalar variables are implicitly mapped as firstprivate<sup>45</sup>.
- statically allocated array variables are implicitly mapped as shared.



<sup>&</sup>lt;sup>4</sup>See Section 2.19.7 in OpenMP standard version 5.0

<sup>&</sup>lt;sup>5</sup>Before OpenMP 4.5, it was shared.

LData mapping

#### Data movements

Passing copies back and forth requires data movements between the host and the target(s).

LData mapping

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- Transferring data to / from devices is slow, minimising them is therefore a good idea performance wise.

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private Uninitialised copies.

firstprivate Initialises copies upon entry.

LData mapping

#### Data movements

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- Transferring data to / from devices is slow, minimising them is therefore a good idea performance wise.

```
private Uninitialised copies.
```

firstprivate Initialises copies upon entry.

lastprivate Initialises original variable upon exit to semantically last value in construct.

```
Introduction to OpenMP - GPUs
```

LData mapping

## Explicit mapping: map clause

```
#pragma omp target map([[map-type-modifier[,][map-type-modifier[,]...]map-type : ]locator-list)
```

- to
- from
- tofrom

```
Introduction to OpenMP - GPUs
```

LData mapping

## Explicit mapping: map clause

```
#pragma omp target map([[map-type-modifier[,][map-type-
modifier[,]...]map-type : ]locator-list)
```

- to
- from
- tofrom
- alloc
- delete

LData mapping

LData mapping

# map(tofrom:my\_var) clause value

Value of data before target is sent to the device for the target region.

LData mapping

# map(tofrom:my\_var) clause value

- Value of data before target is sent to the device for the target region.
- Final value of the variable in target region brought back to the device.

LData mapping

# map(to:my\_var) clause value

- This clause value is used when passing variables to target constructs that need a access to a host-initialised variable, and/or whose updates done on the device are not needed back on the host.
- The variable my\_var begins the target region with the value it had before entering the region.
- The value of variable my\_var will be left unchanged on the host after the target region.

```
Introduction to OpenMP - GPUs

Data mapping
```

## map(to:my\_var) clause value

```
int a = 123;
int b;

#pragma omp target map(to:a,b)

{
    a *= 2;
    b = 100;
}
```

```
INTEGER :: a := 123
INTEGER :: b

!$OMP TARGET MAP(to:a,b)

a *= 2;
b = 100;
!$OMP END TARGET
```

# map(from:my\_var) clause value

- This clause value is used when passing variables to target constructs that will be initialised or update on the device, and the final value is needed back on the host.
- The variable my\_var begins the target region with an uninitialised value.
- The value of variable my\_var leaves the target region with the final value it had at the end of it.

```
Introduction to OpenMP - GPUs

Data mapping
```

## map(from:my\_var) clause value

```
int a;
int b = 123;
#pragma omp target map(from:a, b)
{
    a = 123;
}
```

```
Introduction to OpenMP - GPUs

Data mapping
```

## Mapping of arrays

```
int* a = (int*)malloc(sizeof(int) * 10);

pragma omp target map(tofrom:a[0:10])

{
    a[5] = 123;
}
```

```
INTEGER, ALLOCATABLE :: a

ALLOCATE (a(0:9))

!$OMP TARGET MAP(tofrom:a(0:10))

a(5) = 123

!$OMP END TARGET
```

LData mapping

# Mapping of arrays

- You can also map array variables.
- The notation requires to pass the interval of the array you want to map.<sup>6</sup>
- First, you pass the index of the first element to map.
- Second, you pass the **length** of the section to map.

#### Caution

You do not pass the start index and the end index, but the start index and the length of the array section to map.



<sup>&</sup>lt;sup>6</sup>Because you can map a part of the array if you so wish.

LData mapping

## Time to practise: 3.DataMapping

Update the source code provided to minimise the data movements of the different variables involved in calculations.

#### Tips

#### You will need:

- the target construct
- the to clause
- the tofrom clause
- the from clause

Data environment

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#### target data

- The target data construct creates a data environment, with variables that will persist throughout the target data region.
- In this data environment, we can declare the mapping of variables once, to avoid redundant data movements between target constructs.
- target constructs enclosed in a target data construct inherit their data environment.

└─ Data environment

#### target data construct

```
int a = 0;
   // Data moves to the device, stays until the end.
   #pragma omp target data map(tofrom:a)
4
5
       #pragma omp target
6
7
           a = 45;
8
9
       printf("a = %a \n", a);
10
       #pragma omp target
11
12
13
           a++;
14
       // Some host code
15
16
   printf("a = %a\n", a);
17
```

```
Introduction to OpenMP - GPUs
```

Data environment

#### target data construct

```
INTEGER :: a := 0
   ! Data moves to the device, stays until the end.
   !$OMP TARGET DATA map(tofrom:a)
       ! Some host code
4
       !SOMP TARGET
5
6
           a = 45
       !$OMP END TARGET
7
       WRITE (*, '(A, I0)') "a = ", a
8
       !$OMP TARGET
           a = a + 1
10
       !$OMP END TARGET
11
       ! Some host code
12
   !SOMP END TARGET DATA
13
   WRITE (*, '(A, I0)') "a = ", a
14
```

## Arbitrary data environment management points

- Sometimes, we may want to use data environment in a less structured way.
- Typically happens when ones starts to have function calls, and potential orphan constructs.
- We may have at some point one variable that we know we need to pass to the device.
- We may want to issue the data transfer asynchronously.

```
Introduction to OpenMP - GPUs
```

LData environment

#### Arbitrary data environment management points

```
#pragma omp target enter data map(alloc:a)
// Any code
#pragma omp target exit data map(delete:a)
```

```
1  !$OMP TARGET ENTER DARA MAP(alloc:a)
2  // Any code
3  !$OMP TARGET EXIT DATA MAP(delete:a)
```

#### Note:

- for target enter data, it maps data to the device, clause can be to or alloc only.
- for target exit data, it unmaps data from the device, clause can be from or delete only.
- alloc and delete allow to manipulate data on the device without requiring data movements.

Data environment

# map(alloc:my\_var) clause value

Allocates a variable on the device, does not issue any data movement between the host and the device.

```
Introduction to OpenMP - GPUs
```

L Data environment

# map(alloc:my\_var) clause value

```
int a = 123;
#pragma omp target enter data map(alloc:a)
#pragma omp target
{
    a = 246;
    // Something with A.
}
```

```
INTEGER :: a := 123

! $OMP TARGET ENTER DATA MAP(alloc:a)

! $OMP TARGET

a = 246

! Something with A.

! $OMP END TARGET
```

Data environment

# map(delete:my\_var) clause value

Deletes a variable on the device, does not issue any data movement between the host and the device.

```
Introduction to OpenMP - GPUs
```

☐ Data environment

# map(delete:my\_var) clause value

```
1 // Some code...
2 #pragma omp target exit data map(delete:a)
```

```
! Some code...
! Some TARGET EXIT DATA MAP(delete:a)
```

```
Introduction to OpenMP - GPUs
```

LData environment

#### target update construct

Sometimes, we may want to get the current value of a variable mapped to a device.

```
int a:
   #pragma omp target data map(from:a)
3
       #pragma omp target
5
            a = 123:
       printf("a = %d\n", a);
       #pragma omp target
10
11
            a++;
12
13
```

```
Introduction to OpenMP - GPUs
```

L Data environment

### target update construct

Sometimes, we may want to get the current value of a variable mapped to a device.

```
int a:
   #pragma omp target data map(from:a)
3
       #pragma omp target
           a = 123:
6
       #pragma omp target update from(a)
8
       printf("a = %d\n", a);
10
       #pragma omp target
11
12
           a++;
13
14
```

```
Introduction to OpenMP - GPUs
```

### target update construct

Conversely, we may want to update the value of a variable mapped to a device.

```
int a;
   int b;
   #pragma omp target data map(to:a) map(from:b)
4
       #pragma omp target
5
6
            b = a;
7
8
       a++;
        #pragma omp target
10
11
            b = a;
12
13
       printf("a = %d\n", a);
14
15
```

Data environment

Data environment

### target update construct

```
int a = 123;
   #pragma omp target data map(to:a) map(from:result)
3
       #pragma omp target
5
           result += a * 100;
6
7
       a++;
8
       #pragma omp target update to(a)
       #pragma omp target
10
11
           result += a * 100;
12
13
14
15
```

### Time to practise: 4.DataEnvironment

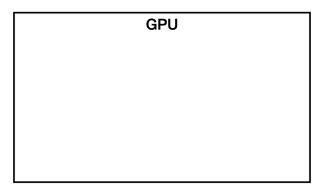
- Can ask for a target construct to be executed on a specific device.
- Try to get 2 GPUs, and get a target construct executed on each.
- Also try to get a target construct executed on the host device.
- Also try to get a target construct executed on a device that does not exist.

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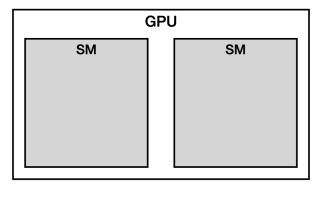
— Multi-level parallelism

### Hardware



Multi-level parallelism

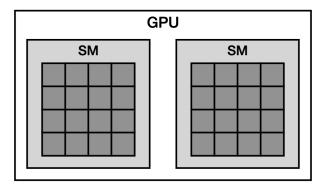
### Hardware



Streaming Multiprocessor (SM)

Multi-level parallelism

### Hardware





Multi-level parallelism

#### teams

The teams construct forks the execution into multiple teams of threads.

└─ Multi-level parallelism

- The teams construct forks the execution into multiple teams of threads.
- The code in the associated structured block will be executed by the master thread of each team.

Multi-level parallelism

- The teams construct forks the execution into multiple teams of threads.
- The code in the associated structured block will be executed by the master thread of each team.
- Each team will therefore has its own "thread 0".

└─ Multi-level parallelism

#### teams

#### Similarly to CPU OpenMP, you can:

- get your team number, using the omp\_get\_team\_num()
  function.
- get the total number of teams, using the omp\_get\_num\_teams() function.

```
Introduction to OpenMP - GPUs
```

```
#pragma omp target
{
    #pragma omp teams
    {
        int my_team_number = omp_get_team_num();
        printf("I am the master thread for team %d.\n",
        my_team_number);
    }
}
```

```
!$OMP TARGET
!$OMP TEAMS

int my_team_number = omp_get_team_num();

printf("I am the master thread for team %d.\n",
    my_team_number);
!$OMP END TEAMS
!$OMP END TARGET
```

```
Introduction to OpenMP - GPUs
```

# Composite target teams construct

Multi-level parallelism

#### teams

Similarly to the parallel construct, you can control the number of teams spawned.

└─ Multi-level parallelism

- Similarly to the parallel construct, you can control the number of teams spawned.
- Use the num\_teams clause.

Multi-level parallelism

- Similarly to the parallel construct, you can control the number of teams spawned.
- Use the num\_teams clause.
- When used, this clause indicates that **at most** <X> teams will be spawned.

```
Introduction to OpenMP - GPUs
```

└─ Multi-level parallelism

#### teams

Inside your teams construct, by default you just have the master thread running, just like in classic serial program execution.

└─ Multi-level parallelism

- Inside your teams construct, by default you just have the master thread running, just like in classic serial program execution.
- When wanting to fork threads in, you can use the parallel construct.

```
Introduction to OpenMP - GPUs
```

```
#pragma omp target
2
       // One team of one thread
3
       #pragma omp teams
           // Multiple teams of one thread
6
           #pragma omp parallel
7
8
                // Multiple teams of multiple threads
10
11
12
```

```
Introduction to OpenMP - GPUs
```

## Loops!

```
for(int i = 0; i < N; i++)
{
    a[i] = i + 123;
}</pre>
```

```
1 DO i = 0, N - 1
2 a(i) = i + 123
3 END DO
```

```
Introduction to OpenMP - GPUs
```

## Loops!

```
#pragma omp target teams
{
    for(int i = 0; i < N; i++)
    {
        a[i] = i + 123;
    }
}</pre>
```

└─ Multi-level parallelism

# Multi-level parallelism

Just like in a parallel construct, threads do not automatically split iterations unless they are asked to.

└─Multi-level parallelism

# Multi-level parallelism

- Just like in a parallel construct, threads do not automatically split iterations unless they are asked to.
- The distribute clause is equivalent to the for / do clause for CPUs.

- Just like in a parallel construct, threads do not automatically split iterations unless they are asked to.
- The distribute clause is equivalent to the for / do clause for CPUs.
- It allows the iteration set to be split across the master threads of all teams.

```
Introduction to OpenMP - GPUs
```

## Loops!

```
! $OMP TARGET TEAMS

! $OMP DISTRIBUTE

DO i = 0, N - 1

a(i) = i + 123

END DO

! $OMP END TARGET TEAMS
```

```
Introduction to OpenMP - GPUs
```

# Composite target teams distribute construct

```
#pragma omp target teams distribute
for(int i = 0; i < N; i++)
{
    a[i] = i + 123;
}</pre>
```

└─ Multi-level parallelism

# Scheduling

On CPUs, the schedule clause is available.

└─ Multi-level parallelism

- On CPUs, the schedule clause is available.
- On GPUs, the dist\_schedule clause is available.

Multi-level parallelism

- On CPUs, the schedule clause is available.
- On GPUs, the dist\_schedule clause is available.
- Only static schedule available however.

Multi-level parallelism

- On CPUs, the schedule clause is available.
- On GPUs, the dist\_schedule clause is available.
- Only static schedule available however.
- Can still specify a particular chunksize.

```
Introduction to OpenMP - GPUs
```

```
#pragma omp target teams distribute
for(int i = 0; i < N; i++)
{
    a[i] = i + 123;
}</pre>
```

```
Introduction to OpenMP - GPUs
```

```
#pragma omp target teams distribute schedule(static, 24)
for(int i = 0; i < N; i++)
{
    a[i] = i + 123;
}</pre>
```

```
1 | $\frac{1}{2} \text{DO i} = 0, N - 1 \\ a(i) = i + 123 \\ END DO
```

```
Introduction to OpenMP - GPUs
```

## Nested loops

```
for(int i = 0; i < N; i++)

for(int j = 0; j < N; j++)

a[i][j] = i * j + 123;

}

}</pre>
```

```
DO i = 0, N - 1
DO j = 0, j < N - 1
a(i,j) = i * j + 123
END DO
END DO
```

```
Introduction to OpenMP - GPUs
```

# Nested loops

```
#pragma omp target teams distribute
for(int i = 0; i < N; i++)

{
    for(int j = 0; j < N; j++)
    {
        a[i][j] = i * j + 123;
    }
}</pre>
```

```
Introduction to OpenMP - GPUs
```

## Nested loops

```
for(int i = 0; i < N; i++)

{
    #pragma omp target teams distribute
    for(int j = 0; j < N; j++)

{
        a[i][j] = i * j + 123;

}

8
}</pre>
```

```
DO i = 0, N - 1

! $OMP TARGET TEAMS DISTRIBUTE

DO j = 0, j < N - 1

a(i,j) = i * j + 123

END DO

END DO
```

```
Introduction to OpenMP - GPUs
```

## Nested loops

```
#pragma omp target teams distribute
  for(int i = 0; i < N; i++)</pre>
3
      #pragma omp parallel for
      for (int j = 0; j < N; j++)
           a[i][j] = i * j + 123;
8
9
```

```
!SOMP TARGET TEAMS DISTRIBUTE
  DO i = 0, N - 1
      !SOMP PARALLEL DO
3
      DO j = 0, j < N - 1
          a(i,j) = i * j + 123
      END DO
  END DO
```

```
Introduction to OpenMP - GPUs
```

## Massive loop

```
for(int i = 0; i < SUPER_BIG_N; i++)

a[i = i + 123;
}</pre>
```

```
DO i = 0, SUPER_BIG_N - 1
a(i) = i + 123
END DO
```

```
Introduction to OpenMP - GPUs
```

## Massive loop

```
#pragma omp target teams distribute parallel for
for(int i = 0; i < SUPER_BIG_N; i++)

{
    a[i = i + 123;
}</pre>
```

```
1  !$OMP TARGET TEAMS DISTRIBUTE PARALLEL DO
2  DO i = 0, SUPER_BIG_N - 1
3  a(i) = i + 123
4  END DO
```

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└─ Multi-GPU

### Multi-GPU

Nowadays, GPU nodes contain more than 1 GPU.

└─ Multi-GPU

### Multi-GPU

- Nowadays, GPU nodes contain more than 1 GPU.
- On Bridges2, GPU nodes have 8 GPUs.

└─ Multi-GPU

### Multi-GPU

- Nowadays, GPU nodes contain more than 1 GPU.
- On Bridges2, GPU nodes have 8 GPUs.
- Fully using a GPU is great.

Multi-GPU

### Multi-GPU

- Nowadays, GPU nodes contain more than 1 GPU.
- On Bridges2, GPU nodes have 8 GPUs.
- Fully using a GPU is great.
- Fully using all GPUs is more fun.

- Multi-GPU

### Find our way in a multi-GPU environment.

└ Multi-GPU

### Find our way in a multi-GPU environment.

Implicitly, one of the devices is selected by default.

### Find our way in a multi-GPU environment.

- Implicitly, one of the devices is selected by default.
- You can know which one, by calling omp\_get\_default\_device().

#### └ Multi-GPU

### Find our way in a multi-GPU environment.

- Implicitly, one of the devices is selected by default.
- You can know which one, by calling omp\_get\_default\_device().
- You can change which one, by calling omp\_set\_default\_device (new\_device).

- Multi-GPU

#### Devices number and identifiers

∟ Multi-GPU

#### Devices number and identifiers

■ The host is a device too, you can get its number by calling omp\_get\_initial\_device().

└ Multi-GPU

#### Devices number and identifiers

- The host is a device too, you can get its number by calling omp\_get\_initial\_device().
- You can get the number of the device you are currently running on by calling omp\_get\_device\_num().

#### Devices number and identifiers

- The host is a device too, you can get its number by calling omp\_get\_initial\_device().
- You can get the number of the device you are currently running on by calling omp\_get\_device\_num().
- You can get the total number of target devices by calling omp\_get\_num\_devices().

#### Devices number and identifiers

- The host is a device too, you can get its number by calling omp\_get\_initial\_device().
- You can get the number of the device you are currently running on by calling omp\_get\_device\_num().<sup>7</sup>
- You can get the total number of target devices by calling omp\_get\_num\_devices().

#### Note

omp\_get\_device\_num() returns the number of target devices,
it does not include the host device.



<sup>&</sup>lt;sup>7</sup>Does not seem to be supported yet, as of nvhpc/22.2.

└─ Multi-GPU

## Splitting the work

- Send different parts of an array to different GPUs.
- Have multiple GPUs work in parallel on a given arrray.

```
Introduction to OpenMP - GPUs

Multi-GPU
```

## Addressing GPUs - The device clause

- Can ask for a target construct to be executed on a specific device.
- Just use the device clause on a target construct.

```
1  // This will execute on the first GPU, device #0.
2  #pragma omp target device(0)
3  {
4      // Code to execute on device 0
5  }
```

```
! This will execute on the first GPU, device #0.
! SOMP TARGET DEVICE(0)
! Code to execute on device 0
! SOMP END TARGET
```

## Time to practise: 6.MultiGPU

- Can ask for a target construct to be executed on a specific device.
- Try to get 2 GPUs, and get a target construct executed on each.
- Also try to get a target construct executed on the host device.
- Also try to get a target construct executed on a device that does not exist.

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## Asynchronous

- By default, target constructs are synchronous.
- You cannot process beyond it until it is finished.
- The nowait clause allows to execute a target construct in an asynchronous fashion.
- Execution does not wait for the end of the target construct to continue its progress.
- Dependencies between targets are expressed using the depend clause.
- If a target construct with a nowait clause does not have depend clauses, it is assumed to have no dependencies and can execute freely at any time.

L Asynchronous

## Asynchronous

```
depend(in:a) I will read from variable a.
depend(out:a) I will write to variable a.
depend(inout:a) I will read from, and write to, variable a.
```

```
Introduction to OpenMP - GPUs
```

└ Asynchronous

## Structured synchronisation

```
#pragma omp target nowait map(tofrom:a) depend(inout:a)
2
      a = 456;
3
4
   #pragma omp target nowait map(to:a) depend(in:a)
6
      b = a + 3;
7
8
   #pragma omp target map(to:a) map(from:c) \
                       depend(in:a) depend(out:c)
10
11
       c = a + 10
12
13
```

```
Introduction to OpenMP - GPUs
```

## Structured synchronisation

```
!$OMP TARGET NOWAIT MAP(TOFROM:a) DEPEND(inout:a)
       a = 456;
2
  !$OMP END TARGET
   !$OMP TARGET NOWAIT MAP(TO:a) MAP(FROM:b) &
                        DEPEND (IN:a) DEPEND (OUT:b)
5
       b = a + 3;
6
  !SOMP END TARGET
   !$OMP TARGET MAP(FROM:a) MAP(TO:c) &
                DEPEND(IN:a) DEPEND(OUT:c)
g
       c = a + 10
10
11
   !SOMP END TARGET
```

Asynchronous

```
Introduction to OpenMP - GPUs
```

Asynchronous

# Construct synchronisation<sup>8</sup>

```
#pragma omp single
2
       #pragma omp target nowait map(tofrom:a) depend(inout
       :a)
4
           a = 456;
6
       #pragma omp target nowait map(to:a) depend(in:a)
7
8
           b = a + 3;
10
       #pragma omp target nowait map(to:a) map(from:c) \
11
12
                            depend(in:a) depend(out:c)
13
           c = a + 10
14
15
16
```

```
Introduction to OpenMP - GPUs
```

Asynchronous

# Construct synchronisation<sup>9</sup>

```
!$OMP SINGLE
   !$OMP TARGET NOWAIT MAP(TOFROM:a) DEPEND(inout:a)
       a = 456:
3
   !SOMP END TARGET
   !$OMP TARGET NOWAIT MAP(TO:a) MAP(FROM:b) &
                        DEPEND (IN:a) DEPEND (OUT:b)
6
       b = a + 3;
7
   !SOMP END TARGET
   !$OMP TARGET NOWAIT MAP(FROM:a) MAP(TO:c) &
10
                 DEPEND(IN:a) DEPEND(OUT:c)
11
       c = a + 10
   !SOMP END TARGET
12
   !SOMP END SINGLE
13
```

<sup>&</sup>lt;sup>9</sup>Example assumes everything in parallel construct

Asynchronous

## **Explicit** synchronisation

```
#pragma omp target nowait map(tofrom:a) depend(inout:a)
2
      a = 456;
3
4
   #pragma omp target nowait map(to:a) depend(in:a)
6
      b = a + 3;
7
8
   #pragma omp target nowait map(to:a) map(from:c) \
                       depend(in:a) depend(out:c)
10
11
       c = a + 10
12
13
   #pragma omp taskwait
14
```

```
Introduction to OpenMP - GPUs
```

# Construct synchronisation 10

```
!$OMP TARGET NOWAIT MAP(TOFROM:a) DEPEND(inout:a)
       a = 456;
2
   !SOMP END TARGET
   !$OMP TARGET NOWAIT MAP(TO:a) MAP(FROM:b) &
                        DEPEND (IN:a) DEPEND (OUT:b)
5
       b = a + 3;
6
   !SOMP END TARGET
   !$OMP TARGET NOWAIT MAP(FROM:a) MAP(TO:c) &
                DEPEND(IN:a) DEPEND(OUT:c)
g
       c = a + 10
10
   !SOMP END TARGET
11
   !SOMP TASKWAIT
12
```

<sup>&</sup>lt;sup>10</sup>Example assumes everything in parallel construct

L Asynchronous

## Time to practise: 7. Asynchronous

Execute the different statements shown in different target constructs, using asynchronous execution.

#### Tips

#### You will need:

- target construct
- nowait clause
- depend clause
- map clause

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## Summary

#### You now know how to:

- offload your workload to GPUs.
- control the mapping of data between host and targets.
- handle multiple GPUs.
- issue kernels in asynchronous fashion.

#### To be seen

- Reverse offloading
- Coalescing memory accesses.
- Metadirectives
- Directive cancellation
- Conditional execution
- And a lot, lot more...

#### To be seen

- Reverse offloading
- Coalescing memory accesses.
- Metadirectives
- Directive cancellation
- Conditional execution
- And a lot, lot more...

The best place to learn more about OpenMP and how it works, to get the specifications and so on is the OpenMP forum website.

## Your opinion matters



Figure: Link to surveys