

# **Handling Asynchronicity**

COMPUTE | STOI

### **Exploiting overlap**

29. May 2017



#### If individual kernels are performing well

- Can you start overlapping different computational tasks?
  - kernels on the GPU
  - data transfers to/from the GPU
  - maybe even separate computation on the CPU

#### You can do this using the async features

- OpenACC: stream handles; very similar to streams in CUDA
- OpenMP: data dependencies; same as OpenMP tasks

COMPUTE | STORE | ANALYZE

Copyright 2017 Cray Inc.

### **Asynchronicity with Nvidia GPUs**



- accelerator operations are launched from the CPU
- they then execute asynchronously
  - control returns immediately to the host
  - host must then wait or test for completion
    - automatically (e.g. handled by compiler)
    - manually (e.g. via directives or API calls)
  - applies to:
    - computational kernels
    - data transfers: update directives
    - plus some other directives

COMPUTE

STORE

### **Automatic synchronisation**



- By default, the compiler will handle synchronisation:
  - may be conservative, and wait for every operation to complete; or
  - may be smarter, and reduce number of sync. points (consistent with correctness)

#### CCE-specific options:

- control behaviour with compiler option -hacc\_model=auto\_async\_\*
  - auto\_async\_none :
    - waits for every operation to separately complete
    - useful for debugging and generating profiles
    - but it may skew performance
  - auto\_async\_kernels:
    - may allow some computational kernels to overlap
    - the default behaviour
  - auto async all:
    - may allow kernels and data transfers to overlap
    - try this as a performance-tuning option

29. May 2017 Copyright 2017 Cray Inc.

# A stream of tasks in OpenMP 4.0 (and 4.5)



- Can simply add nowait clause to kernels, updates
  - Over-rides automatic synchronisation

#### Operations:

- guaranteed to execute sequentially in the order they were launched
- not guaranteed to execute immediately after each other; there could be delays
- this is known as a "stream" of tasks

#### Synchronisation

- taskwait or barrier directives ensure all asynchronous operations have completed
- If you've used CUDA streams, these concepts should be very familiar

COMPUTE | STORE | ANALYZE

Copyright 2017 Cray Inc.

# **Streams of tasks in OpenMP 4.5**



- Asynchronicity handled in the same way as tasks (4.0)
  - Each kernel or set of data transfers is a task
- depend clause gives much greater control than nowait
  - Over-rides automatic synchronisation
  - Allows user to specify dependencies between tasks
  - These control:
    - order of execution
    - whether tasks can overlap or not

6

#### **Dependency-types**



- Dependencies are expressed as data dependencies
  - depend(dependency-type:variable)
- Dependency types are:
  - in:
    - in:a task cannot start until all tasks with out:a or inout:a have finished
  - out:

29. May 2017

- out:a task must complete before any in:a or inout:a tasks can start
- inout: combination of in and out behaviour for same variable

COMPUTE | STORE | ANALYZE

Copyright 2017 Cray Inc.

# **Dependency variables**



### in/out/inout dependency types

- Simply used to define a dependency tree
- Do not imply any data movement to/from the accelerator
- Do not express how the variables is used in the task
  - in:a does not mean a is used in a read-only fashion in the task
    - a could be used in a write-only or read-write fashion, or
    - a might not be referenced at all in the task

29. May 2017 Copyright 2017 Cray Inc.

# **OpenMP depend first example**

#### a simple pipeline:

- processing an array, slice by slice
  - assume can be processed independently
- each slice requires 3 tasks:
  - copy data to GPU,
  - process on GPU,
  - bring back to CPU
- which must execute sequentially
- but we can overlap different slices

```
REAL :: a(Nvec, Nchunk), b(Nvec, Nchunk)
!$omp target data map(alloc:a,b)
DO i = 1.Nchunks
!$omp target update to(a(:,j)) &
      depend(out:a(:,j))
!$omp target teams distribute &
      depend(inout:a(:,j))
  DO i = 1, Nvec
    b(i,j) = \langle function \ of \ a(:,j) \rangle
  ENDDO
!$omp end target teams distribute
!$omp target update from(b(:,j)) &
      depend(in:a(:,j))
FNDDO
!$omp taskwait
!$omp end target data
```

#### Expect to see overlap of three streams at once

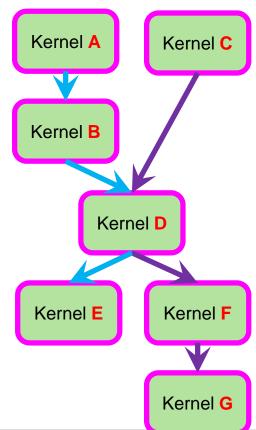
• one sending to the device; one processing the slice; one sending to host

COMPUTE

STORE

### **High-level example with OpenMP 4.5**

```
!$omp target depend(out:x)
<Kernel A>
!$omp target depend(inout:x)
<Kernel B>
!$omp target depend(out:y)
<Kernel C>
!$omp target depend(inout:x) &
             depend(inout:y)
<Kernel D>
!$omp target depend(in:x)
<Kernel E>
!$omp target depend(inout:y)
<Kernel F>
!$omp target depend(in:y)
<Kernel G>
!$omp taskwait ! all completed
```



# Synchronising subsets of tasks



- Dependencies ensure that tasks execute in order
- Can also globally synchronise
  - taskwait and barrier directives
- May also want to add CPU task into stream, e.g.
  - kernel to pack buffer on accelerator
  - update to move buffer back to host



# A stream of tasks in OpenACC



- Add async clause to make operation asynchronous
  - kernel or data transfer

#### Operations:

- guaranteed to execute sequentially in the order they were launched
- not guaranteed to execute immediately after each other; there could be delays
- this is known as a "stream" of tasks

#### Synchronisation

- wait directive ensures all asynchronous operations have completed
- can use an API call to do same thing (or to test for completion)
- If you've used CUDA streams, these concepts should be very familiar

### Multiple streams of tasks in OpenACC



#### The async clause can take a handle

- handle is an argument that is a positive- or zero-valued integer
- each handle value is a separate stream of tasks

#### Tasks within a given stream

guaranteed to execute sequentially in order they were issued

#### Tasks in different streams

- may overlap or serialise, as the hardware and runtime allows
- operations in different streams should be independent or we have a race condition

#### Synchronisation

- wait directive ensures all streams of tasks have completed
- wait(handle) directive ensures just the specified stream of tasks has completed
- can use an API call to do same thing (or to test for completion)
- If you've used CUDA streams, these concepts should be very familiar

29. May 2017 STORE | ANALYZE
Copyright 2017 Cray Inc.

### **OpenACC** async first example



#### a simple pipeline:

- processing an array, slice by slice
- assume slices can be processed independently
- each slice requires 3 tasks:
  - copy data to GPU,
  - process on GPU,
  - bring back to CPU
- which must execute sequentially
- but we can overlap different slices
- Use a different stream for each slice
  - use slice number as stream handle
  - don't worry if number gets too large

```
REAL :: a(Nvec, Nchunk), b(Nvec, Nchunk)
!$acc data create(a,b)
DO j = 1, Nchunks
!$acc update device(a(:,j)) async(j)
!$acc parallel loop async(j)
  DO i = 1, Nvec
    b(i,j) = \langle function \ of \ a(:,j) \rangle
  ENDDO
!$acc update host(b(:,j)) async(j)
FNDDO
!$acc wait
!$acc end data
```

#### Expect to see overlap of three streams at once

one sending to the device; one processing the slice; one sending to host

COMPUTE



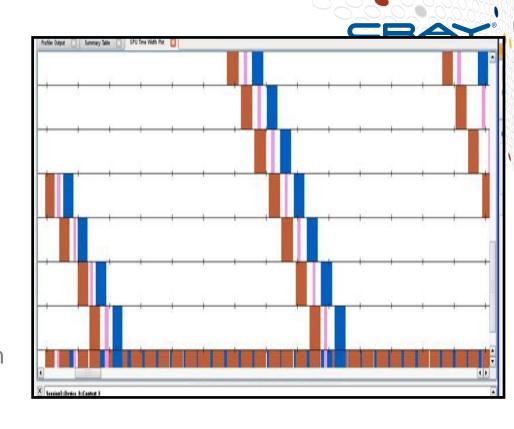
# **OpenACC** async results

#### • Execution times:

- CPU: 3.76s
- OpenACC: 1.10s
- OpenACC, async: 0.34s

### NVIDIA Visual profiler

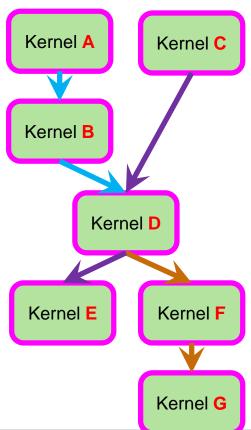
- only 7 of the 16 streams shown
- collapsed view at bottom



COMPUTE

# High-level example with OpenACC

```
!$acc parallel loop async(stream1)
<Kernel A>
!$acc parallel loop async(stream1)
<Kernel B>
!$acc parallel loop async(stream2)
<Kernel C>
!$acc parallel loop async(stream3) &
!$acc wait(stream1,stream2)
<Kernel D>
!$acc parallel loop async(stream4) &
!$acc
         wait(stream3)
<Kernel E>
!$acc parallel loop async(stream5) &
!$acc wait(stream3)
<Kernel F>
!$acc parallel loop async(stream5)
<Kernel G>
!$acc wait ! ensures all completed
```



### What's missing in OpenACC and OpenMP?

CRAY

- Deep copy is the elephant in the room
- Hierarchical data structures with pointers
  - C++ objects; C structs; Fortran derived types
  - On CPU, pointers point to CPU memory addresses
  - When "map" to GPU, pointers still point to CPU addresses
    - main impact: at best, inefficient; usually, just broken
  - User needs to explicitly remap all the pointers
    - this is not a very satisfactory solution



- CCE -hacc\_model=deep\_copy helps for Fortran only
- Proper solution deferred to OpenMP 5.0 (due Nov.17)
  - and OpenACC 3.0 (due ?)

