OpenMP 4.0 Accelerator Programming Model & the AClang Compiler

Marcio M Pereira IC/Unicamp

Agenda

What is an Accelerator in OpenMP?

Execution Model and Data Model

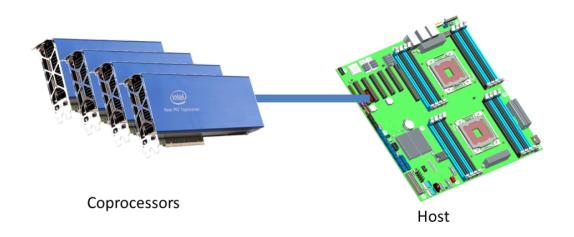
• target Construct and Accelerator-specific Constructs

Examples

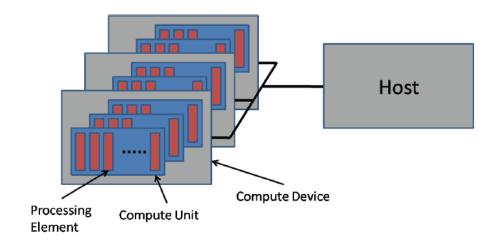
• The AClang Compiler

Accelerators

- In how differs an accelerator from just another core?
 - ➤ different functionality, i.e. optimized for something special
 - > different (possibly limited) instruction set
 - → heterogeneous device



Host and Co-processors



Host and GPUs

Heterogeneous device model

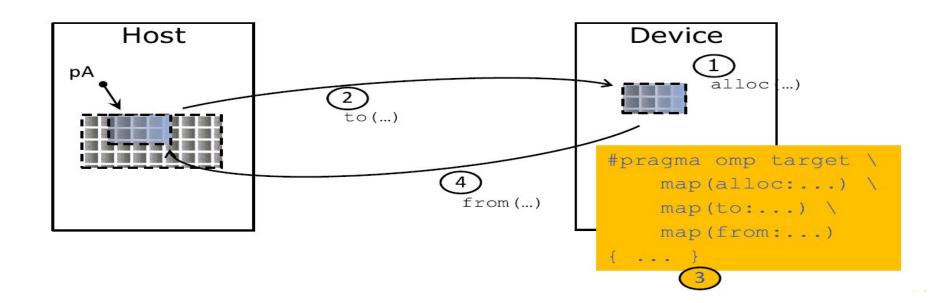
- Assumptions used as design goals for OpenMP 4.0:
 - right every accelerator device is attached to one host device
 - it may not be programmable in the same language as the host
 - >It may not implement all operations available on the host
 - >it may or may not share memory with the host device
 - some accelerators are specialized for loop nests

Execution Model

- Host-centric: the execution of an OpenMP program starts on the host device and it may offload target regions to target devices
- If a target device is not present, or not supported, or not available, the target region is executed by the host device
- If a construct creates a data environment, the data environment is created at the time the construct is encountered

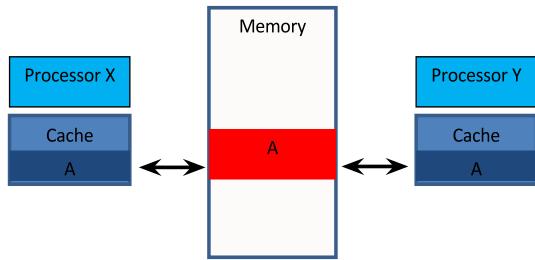
Data environment

- Data environment is lexically scoped
- Data environment is destroyed at closing curly brace
- Allocated buffers/data are automatically released



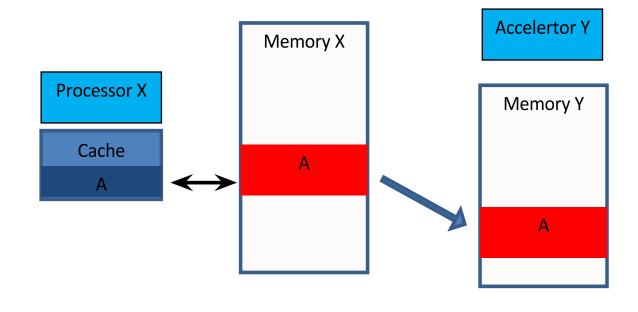
Data mapping: shared or distributed memory

Shared memory



- The corresponding variable in the device data environment may share storage with the original variable.
- Writes to the corresponding variable may alter the value of the original variable.

Distributed memory



OpenMP 4.0 device constructs

- the target construct transfers the control flow to the target device
 - the map clauses control direction of data flow
 - array notation is used to describe array length
- the target data construct creates a scoped device data environment
 - the map clauses control direction of data flow
 - the device data environment is valid through the lifetime of the target data region
- use target update to request data transfers from within a target data region

map clause

- Map a variable from the current task's data environment to the device data environment associated with the construct
 - ➤alloc-type: each new corresponding list item has an undefined initial value
 - >to-type: each new corresponding list item is initialized with the original list item's value
 - ▶from-type: declares that on exit from the region the corresponding list item's value is assigned to the original list item
 - >tofrom-type: the default, combination of to and from

target construct example

- Use **target** construct to:
 - Transfer control from the host to the device
 - Establish a device data environment (if not yet done)
- Host thread waits until offloaded region completed

```
void mvt_gpu(float* a, float* x1, float* x2,
                         float * y1, float * y2)
 #pragma omp target device (GPU) \
                      map(to: a[:N*N], y1[:N]) \setminus
                      map(tofrom: x1[:N])
  #pragma omp parallel for
  for (int i = 0; i < N; i + +)
    for (int j=0; j< N; j++)
      x1[i] = x1[i] + a[i*N + i] * y1[i];
  #pragma omp target device (GPU) \
                      map(to: a[:N*N], y2[:N]) \setminus
                      map(tofrom: x2[:N])
    #pragma omp parallel for
    for (int i = 0; i < N; i + +)
      for (int j=0; j< N; j++)
        x2[i] = x2[i] + a[j*N + i] * y2[j];
```

data environment example

- Create a data environment to keep data on devices
- Avoid frequent transfers to keep data on devices
- Pre-allocate temporary fields

```
#pragma omp target data device (GPU) \
                    map(alloc: tmp[:N]) \setminus
                    map(to: input[:N]) \
                    map(from: output[:N])
  #pragma omp target device (GPU) \
  #pragma omp parallel for
  for (i=0; i< N; i++)
    tmp[i] = some_computation(input[i], i);
  do_some_other_stuff_on_host();
  #pragma omp target device (GPU) \
  #pragma omp parallel for
  for (i=0; i< N; i++)
    output[i] = final_computation(tmp[i], i);
```

target update construct example

Sync input[N] between host and target

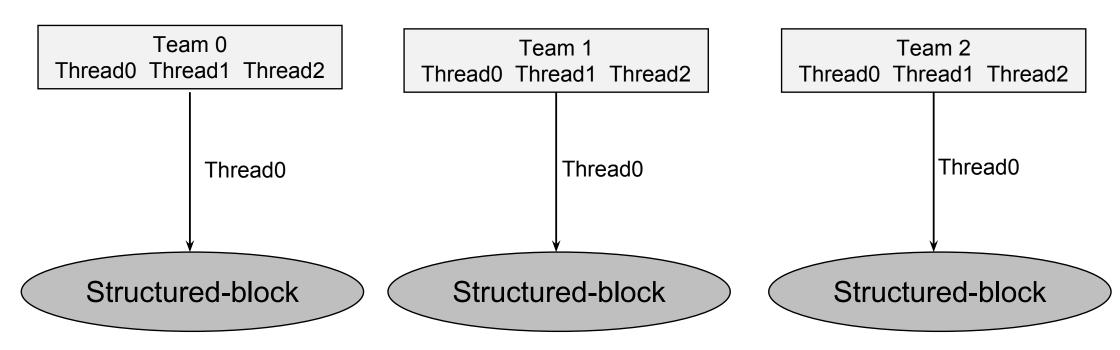
```
#pragma omp target data device (GPU) \
                   map(alloc: tmp[:N]) \
                   map(to: input[:N]) \
                   map(from: output[:N])
  #pragma omp target device (GPU)
  #pragma omp parallel for
  for (int i = 0; i < N; i + +)
    tmp[i] = some_computation(input[i], i);
  update_input_array_on_host() ;
  #pragma omp target update device(GPU) to(input[:N])
  #pragma omp target device (GPU)
  #pragma omp parallel for
  for (int i=0; i < N; i++)
    output[i] = final_computation(input[i], tmp[i]);
```

teams construct

- Creates a league to threads teams
- The master thread of each team executes the teams region
- A **team** construct must be "perfectly" nested in a **target** construct
- Only special OpenMP construct can be nested inside a teams contruct:
 - **≻** distribute
 - **>** parallel for

teams execution model

#pragma omp teams num_teams(3), num_threads(3)structured-block



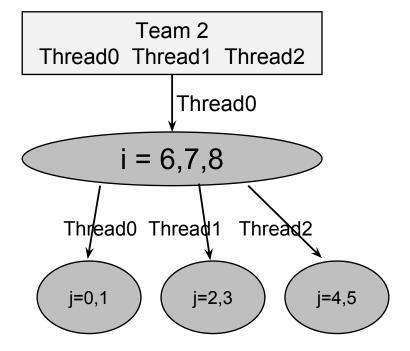
distribute construct

- Iterations distributed among master threads of all teams
- Specify to the loops only
- Must be closed nested to the teams construct
- Workshare among teams to exploit the parallelism on the target device

teams with distribute construct

#pragma omp teams num_teams(3), num_threads(3)

```
#pragma omp distribute
   for (int i=0; i<9; i++) {
    # pragma omp parallel for
         for (int j=0;j<6; j++) {
             Team 0
                                                Team 1
    Thread0 Thread1 Thread2
                                       Thread0 Thread1 Thread2
                  Thread0
                                                    Thread0
            i = 0,1,2
                                              = 3,4,5
   Thread0 Thread1 Thread2
                                      Thread0 Thread1
                                                       Thread2
                                                 j=2,3
 j=0,1
             j=2,3
                         j=4,5
                                      j=0,1
                                                            j=4,5
```



saxpy example

```
#pragma omp target data device (GPU) map(to: X[:N])
  #pragma omp target map(tofrom: Y[:N])
  #pragma omp teams num_teams(nblocks) \
                     num_threads (nthreads)
  #pragma omp distribute
  for (int i=0; i < N; i+=nblocks)
    #pragma omp parallel for
    for (int j=i; j < i+nblocks; j++)
     Y[j] = a X[j] + Y[j];
```

declare target directive

- Specifies that [static] variables and functions are mapped to a device
- if a list item is a function then a device-specific version of the routines is created that can be called from a target region
- if a list item is a variable then the original variable is mapped to a corresponding variable in the initial device data environment for all devices
- all declarations and definitions for a function must have a declare target directive

```
#pragma omp declare target
  float some_computation(float f, int i) {
    // ... some code ...
  }
  float final_computation(float f, int i) {
    // ... some code ...
  }
  #pragma omp end declare target
  ...
```

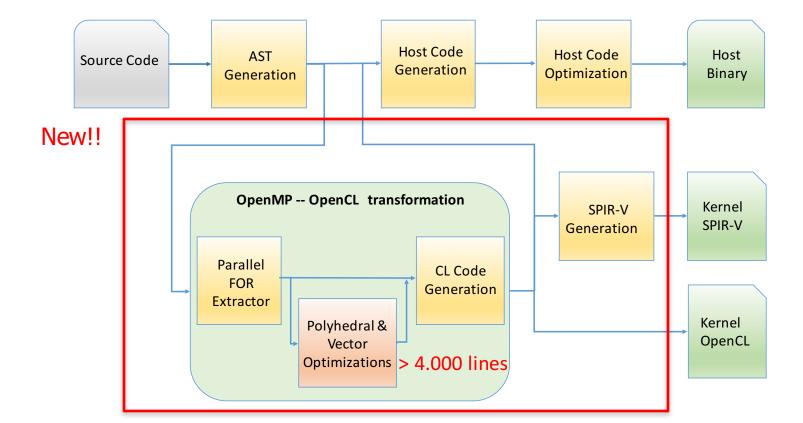
The AClang Compiler

- Started at march/2015
- Main focus: accelerate applications on mobile devices
 - ➤ OpenCL is the language to dispatch kernels on GPUs
- No support of OpenMP 4.0 on llvm/clang trunk
- OpenMP group on Ilvm with focus on NVIDIA/Cuda
- Solution
 - ➤ Map (part of) OpenMP 4.0 to OpenCL
 - but not source-to-source

Deliverable: AClang framework

Very advanced optimizing compiler

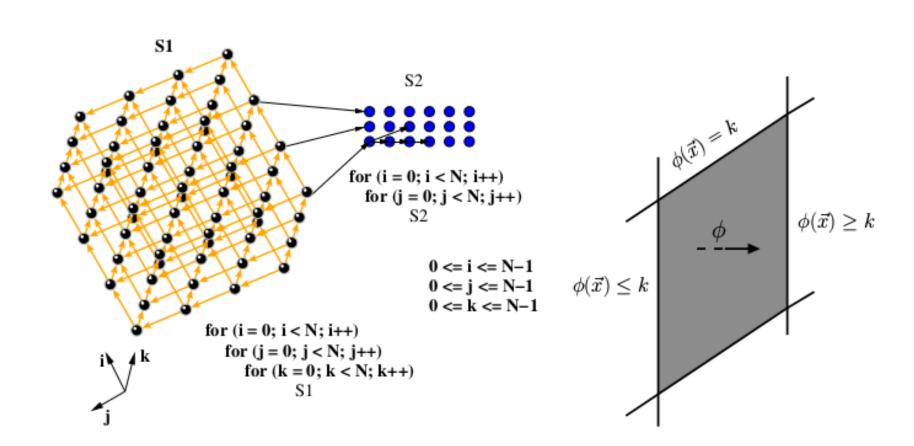
- > 6.000 lines just for the OpenMP OpenCL translation
- > 2.000 lines for the GPU runtime library



Deliverable: AClang optimizations

Use of polyhedral techniques for OpenCL kernels to do:

- loop tiling,
- vectorization



Case Study: Matrix multiplication

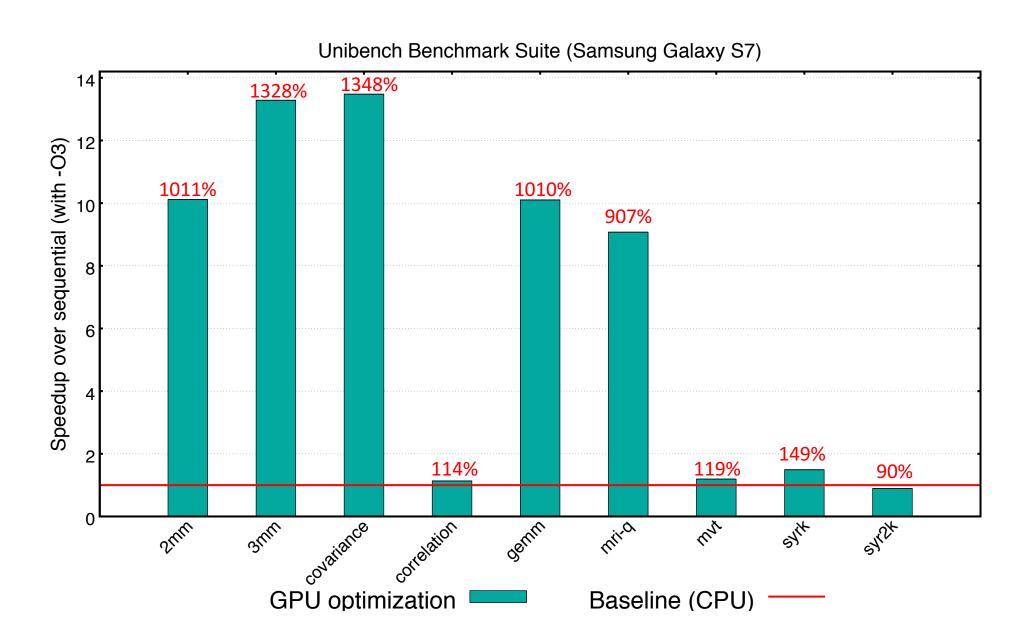
```
void matrix multiply (float *A, float *B, float *C) {
  for (int i = 0; i < 512; i++) {
     for (int j = 0; j < 512; j++) {
        for (int k = 0; k < 512; ++k) {
             C[i * 512 + i] += A[i * 512 + k] * B[k * 512 + i];
```

Extracting Kernel: no-optimization

```
__kernel void kernel_mm_1 (__global float *A,
                                                                   work size = \{512\}
                                                                   block_size = {1};
                               __global float *B,
                                 global float *C) {
 int b0 = get_group_id(0);
  int t0 = get_local_id(0);
  for (int c2 = 0; c2 <= 511; c2 += 1)
    for(int c3 = 0; c3 \le 511; c3 += 1)
      C[512 * b0 + c2] += (A[512 * b0 + c3] * B[c2 + 512 * c3]);
          "kernel is just similar to C function"
```

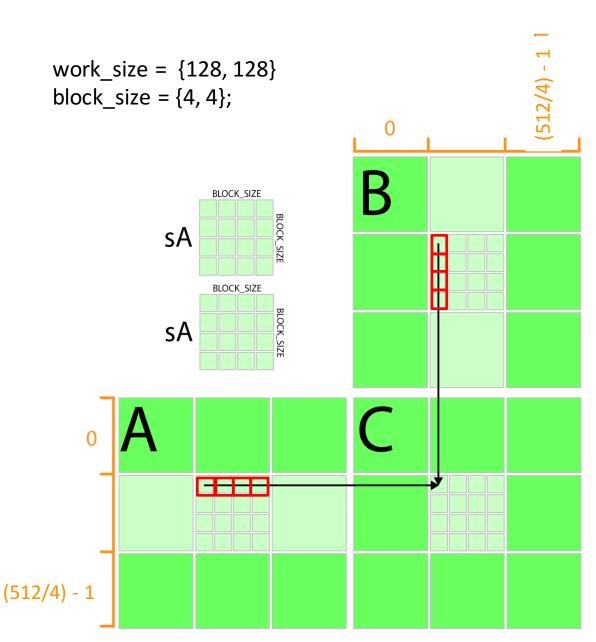
(512) - 1

GPU Optimization

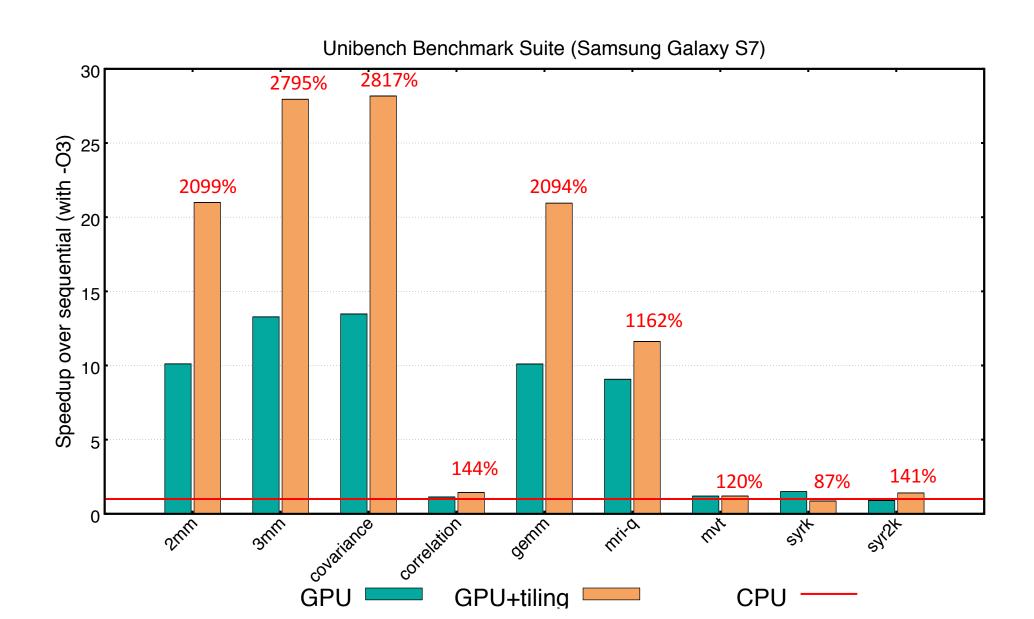


Applying tile optimization (tile-size=4)

```
_kernel void kernel_mm_2 (__global float *A,
                           global float *B,
                             global float *C) {
int b0 = get_group_id(0), b1 = get_group_id(1);
int t0 = get_local_id(0), t1 = get_local_id(1);
for (int c2 = 0; c2 < 512; c2 += 4) {
  for (int c5 = 0; c5 <= 3; c5 += 1)
   C[4 * (512 * b0 + b1) + 512 * t0 + t1] +=
      (A[512 * (4*b0 + t0) + (c2 + c5)] *
       B[4*b1+t1+512*(c2+c5)]);
```



GPU + tile optimization Results

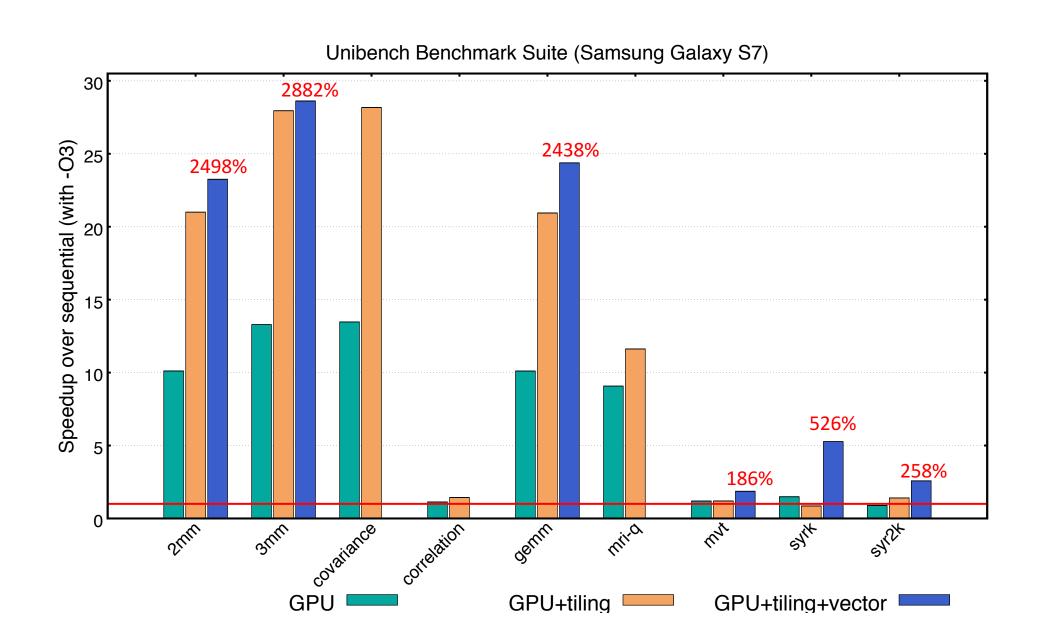


Applying vector optimization

```
_kernel void kernel _mm_3 (__global float *A,
                              __global float *B,
                              global float *C) {
 int b0 = get_group_id(0), b1 = get_group_id(1);
 int t0 = get_local_id(0), t1 = get_local_id(1);
 __private float4_ft0 = {0., 0., 0., 0.};
 __private float4 ft1, ft2;
 for (int c2 = 0; c2 < 512; c2 += 4) {
  _{\text{ft1}} = \text{vload4}(0, \&A[512 * (4 * b0 + t0) + c2]);
  _{\text{ft2.x}} = B[4 * b1 + t1 + 512 * c2];
  _{ft2.y} = B[4 * b1 + t1 + 512 * (c2 + 1)];
  _{\text{ft2.z}} = B[4 * b1 + t1 + 512 * (c2 + 2)];
  _{\text{ft2.w}} = B[4 * b1 + t1 + 512 * (c2 + 3)];
  ft0 += ft1 * ft2;
 C[4 * (512 * b0 + b1) + 512 * t0 + t1] =
   ft0.x + ft0.y + ft0.z + ft0.w;
```

```
work size = \{128,128\}
   block size = \{4, 4\};
            sA
                  BLOCK_SIZE
            sA
127
```

GPU +vector optimization Results



How to use the AClang Compiler?

Some examples

- > clang-fopenmp-omptargets=opencl-unknown-unknown-rtl-mode=verbose-o test test.c
- > clang -fopenmp=lgomp -omptargets=opencl-unknown-unknown -opt-poly=vectorize -o test test.c
- > clang –fopenmp=liomp5 -omptargets=spir64-unknown-unknown -opt-poly=tile -tile-size=32 -o test test.c

Flavors

rtl-mode: none (default), profile, verbose, all

opt-poly: node (default), tile, vectorize

tile-size: default=16

Questions?