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Preparing HACC for Exascale Cosmology on Aurora using SYCL

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Outline

- 1. Porting HACC to SYCL
 - 1. Migration Pipeline using SYCLomatic + LibTooling
 - 2. Abstracting Host-side Code
 - 3. Portability comparison with HIP, CUDA, and SYCL on AMD, NVDIA, and Intel GPUs
- 2. Optimizations for Intel[®] Data Center GPU Max 1550
- 3. Preliminary Results from Aurora
- 4. Preview of Ongoing eDSL Effort

Hardware/Hybrid Accelerated Cosmology Code

- New Conservative Reproducing Kernel (CRK) formulation of Smoothed Particle Hydrodynamics (SPH)
- Resolves some discrepancies with grid-based hydrodynamic schemes
- sub-grid models for radiative cooling, star formation, and feedback from supernovae and Active Galactic Nuclei (AGN)



CRK-HACC: N-body Cosmological Simulation

Porting HACC to SYCL



Kernel Launch Forms

CUDA/HIP

```
// CUDA kernel defined as a
function invoked by <<<>>>
__global__ void
cuda_kernel(float* a, int b) {
   /* kernel body */
}
...
cuda_kernel <<<...>>>(a, b);
```

SYCL lambda

```
// SYCL kernel defined as a
function invoked by a kernel
lambda
void sycl_kernel(float* a, int
b, sycl::nd_item<3> it) {
    /* kernel body */
}
...
q.parallel_for(sycl::nd_range
<3>(...),[=](sycl::nd_item <3>
it) {
    sycl_kernel(a, b, item_ct1);
});
```

The SYCL 2020 standard does not allow function pointers, however, a C++ class from a function object can be used as a template parameter and is how our kernel launch wrapper is defined. For CUDA, our wrapper is implemented as a macro.

SYCL function object

```
// SYCL kernel defined as a
function object invoked
directly
struct SYCLKernel {
void
operator()(sycl::nd_item<3> it)
{
    /* kernel body */
}
const float* a;
const int b;
};
...
q.parallel_for(sycl::nd_range
<3>(...), SYCLKernel(a, b));
```

Experimental Setup

System	CPU	GPU	FP32 (theoretical) Peak per GPU	
Aurora	2 x Intel® Xeon® CPU Max 9470C, 52 cores	6 x Intel® Data Center GPU Max 1550	45.9 TFLOPS	
Polaris	1 x AMD EPYC 7543P, 32 cores	4x NVIDIA A100-SXM4- 40GB	19.5 TFLOPS	
Frontier	1 x AMD EPYC 7A53, 64 cores	4 x AMD Instinct MI250X	53 TFLOPS	

Simulation

- 2x 512³ particles
- 5 timesteps (4 fixed sub-cycles)
- 8 MPI ranks

System Configurations (1-node)

- Aurora: 1 rank/stack
- Polaris: 2 ranks/GPU note: measured ~11% lower efficiency
- Frontier: 1 rank/GCD

Run Dates:

Aurora: September 15, 2023 : SYCL Compiler: Intel® oneAPI DPC++/C++ Compiler 2023.2.0 (2023.x.0.20230510) Polaris: August 3, 2023 : CUDA Compiler: cuda_11.8.r11.8/compiler.31833905_0

Frontier: September 25, 2023 : HIP Compiler: roc-5.3.0 22362 3cf23f77f8208174a2ee7c616f4be23674d7b081

Initial Comparison Results

Aggregate of all GPU Kernels



Performance results are based on testing as of dates shown in configurations and may not reflect all publicly available updates. See slide 6 for configuration details.

- The fast-math compiler option was not enabled by default with <u>nvcc</u> or <u>hipcc</u> but is with the <u>DPC++ compiler</u>.
- On Frontier, the HIP code uses wavefront size 64 and the SYCL code uses sub-group size 64.
- On Polaris, the CUDA code uses warp size 32 and the SYCL code uses sub-group size 32.
- On Aurora, the SYCL code uses sub-group size 32 or 16, whichever is optimal for the implementation.

Optimizations

Hotspot Kernels

- 1. Geometry: measures the volumes of gas particles
- 2. Corrections: computes the reproducing kernel coefficients of the higher order smoothed particle hydrodynamics (SPH) solver
- 3. *Extras*: evaluates the density and state gradients
- *4. Acceleration*: calculates the momentum derivative
- 5. *Energy*: solves the derivative of the internal energy.

The SIMD lane data layout of the "half-warp" algorithm, implemented in the hotspot kernels.



Particles are organized in a tree with "fat leaves" containing an SYCL work-group/CUDA block-sized number of particles.

Particle SIMD layout

Lane ID	0	1		15	16	17		31
Leaf	A	A	А	А	В	В	В	В

Lanes [0-15] load and update particles from leaf A, while lanes [16-31] operate on particles from leaf B.

Optimizing Communication

The communication pattern of the "half-warp" algorithm for interacting particles from leaves A and B within the same warp/sub-group.

Warp (32 CUDA threads)



Lower "Half Warp" Upper "Half Warp"

i and *j* particles are stored in different halves of the warp.



Each iteration swaps symmetric (i, j) pairs. Redundant compute in both directions.

- Notice the *pair-wise symmetry*, which is critically important for the correctness of the algorithm.
- XOR-based shuffle pattern implemented as:
 - _____shf1 intrinsic with CUDA
 - sycl::select_from_group with SYCL

The figure represents one of the total ($|LeafA| \times |LeafB|/warp_size$) instances required.

Optimizing Communication

Intel[®] Data Center GPU Max 1550 assembly snippets for sycl::select-from-group

Elements are gathered from the registers specified in **a0** and written into **r2** using *indirect register* access

```
shl (16|M0)r24.0<1>:uwr82.0<2;1,0>:uw0x2:uwadd (16|M0)a0.0<1>:uwr24.0<1;1,0>:uw0x640:uwmov (16|M0)r2.0<1>:udr[a0.0]<1,0>:ud
```

alternative instruction sequence employing *register regioning* is more performant but not always achievable by the compiler

. . .

. . .

. . .

```
add (16|M0) r24.0<1>:f r68.0<1;1,0>:f -
r14.0<0;1,0>:f
add (16|M0) r26.0<1>:f r68.0<1;1,0>:f -
r14.1<0;1,0>:f
add (16|M0) r30.0<1>:f r68.0<1;1,0>:f -
r14.2<0;1,0>:f
```

Communication Strategies explored

- Broadcasts
 - Restructure loops so that sufficient information is known about the communication pattern at compile-time to generate more efficient assembly.
- Shared Local Memory
 - Uses sycl::local_accessor to reserve a small amount of work-group local memory per sub-group to communicate instead of via registers.
- Optimized Instruction Sequences using vISA¹
 - Explicitly code the assembly instructions for each communication step needed.

1 Rangel et al., A Performance-Portable SYCL Implementation of CRK-HACC for Exascale

Optimization Results

Aurora



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- Broadcast uses a sub-group size of 16, all other variants use a sub-group size of 32.
- Restructuring the loops to use broadcasts requires *fewer* atomic instructions, more noticeable in the <u>Extras</u> and <u>Corrections</u> kernels.
- Performance evaluation on NVIDIA and AMD architectures was also performed¹.

1 Rangel et al., A Performance-Portable SYCL Implementation of CRK-HACC for Exascale

Weak Scaling on Aurora

Figure-of-merit (FOM)

steps * sub-steps * number-of-particles / GPU time / 106

Disclaimer: Results were gathered on a pre-production state of Aurora with engineering versions of the SDK and system software. The two studies have different software versions and application configurations and are only intended to display a trend.



Weak Scaling on Aurora

Figure-of-merit (FOM)

steps * sub-steps * number-of-particles / total wall time / 106

Disclaimer: Results were gathered on a pre-production state of Aurora with engineering versions of the SDK and system software. The two studies have different software versions and application configurations and are only intended to display a trend.



Weak Scaling: Aurora Nodes

eDSL for Parallel Programming Model APIs

CRK-HACC is under active development, and since November 2023, the repository has had **377** files changed, **92,862** insertions(+), and **3,410** deletions(-) in **77** commits.

Goal: Design an embedded domain-specific language (eDSL) using C++ for HACC's supported programming models to have a single-source representation of kernels and host-side code.

- 1. Define the host-side abstractions, e.g., memory management, kernel invocation, data transfer, etc.
- 2. Define the kernel functionality abstractions, e.g., warp/sub-group functions, group functions, special math functions, etc.
- 3. Design the syntax for the eDSL.
- 4. Write the C++ wrappers that interface with the CUDA and SYCL APIs.



HACC eDSL Kernel Launcher

```
template<typename K>
INLINE_FUNCTION_ATTR_void_kernel_launch(K
kernel, kernel_Iaunch_params params)
{
    #ifdef __NVCC___
    cudaStream_t_stream = params.dispatcher;
    cuda_kernel_launch<<<params.num_blocks,
    params.block_size, params.local_memory_bytes,
    stream>>>(kernel);
    cudaStreamSynchronize(stream);
#endif
```

```
#ifdef __NVCC__
template<typename K>
__global__ void cuda_kernel_launch(K kernel)
{
    hacc::item<1> it = hacc::item<1>();
    kernel(it);
#endif
}
```

#ifdef SYCL_LANGUAGE_VERSION

• • •

```
// example kernel invocation (host-side)
hacc::kernel_launch(\
    updatePositions{m_devPosBuff[buff_indx], m_devVelBuff[buff_indx], prefactor},
    hacc::kernel_launch_params{posBlocksPerGrid, m_groupSize, 0, devManager}
);
```

HACC eDSL Kernel Example

```
#include "updatePositions.h"
```

```
#include "eDSL_common.h"
```

```
#include "eDSL_kernel_launch.h"
```

```
#include "eDSL_math.h"
```

```
#include "eDSL_utility.h"
```

The use of explicit instantiation of the launch wrapper in the kernel definition source file avoids the need to build CUDA relocatable device code.

```
template void hacc::kernel_launch<updatePositions>(updatePositions,
kernel_launch_params);
```

```
KERNEL_FUNCTION_ATTR void updatePositions::operator () (hacc::item<1>
item) const
{
    /* kernel body */
}
```

HACC eDSL Results

- HACC gravity-only simulation
 - 512^3 particles
 - 8 MPI ranks
- Test Systems
 - ANL JLSE, (4x) NVIDIA V100
 - Evaluate CUDA and eDSL with CUDA back-end
 - Sunspot, Intel (4x) Intel[®] Data Center GPU Max 1550
 - Evaluate SYCL and eDSL with SYCL back-end



Conclusion

- Described porting HACC from a CUDA codebase to SYCL.
- The "shuffle" operations used by CRK-HACC are not performance-portable on the Intel® Data Center GPU Max 1550.
- A straightforward workaround for *shuffles* using shared-local memory was proposed.

- Scaling on pre-production Aurora was demonstrated on up to 1792 nodes.
- Previewed our ongoing eDSL effort to support code maintainability.

Disclaimers

- Performance varies by use, configuration and other factors. Learn more at <u>https://www.intel.com/performanceindex</u>
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